Topics in
Real and Functional Analysis

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Abstract. This manuscript provides a brief introduction to Real and (linear and nonlinear) Functional Analysis. It covers basic Hilbert and Banach space theory as well as basic measure theory including Lebesgue spaces and the Fourier transform.

Keywords and phrases. Functional Analysis, Banach space, Hilbert space, Measure theory, Lebesgue spaces, Fourier transform, Mapping degree, fixed-point theorems, differential equations, Navier–Stokes equation.
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Preface

The present manuscript was written for my course Functional Analysis given at the University of Vienna in winter 2004 and 2009. It was adapted and extended for a course Real Analysis given in summer 2011. The last part are the notes for my course Nonlinear Functional Analysis held at the University of Vienna in Summer 1998 and 2001. The three parts are essentially independent. In particular, the first part does not assume any knowledge from measure theory (at the expense of hardly mentioning $L^p$ spaces).

It is updated whenever I find some errors and extended from time to time. Hence you might want to make sure that you have the most recent version, which is available from


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Goals

The main goal of the present book is to give students a concise introduction which gets to some interesting results without much ado while using a sufficiently general approach suitable for later extensions. Still I have tried to always start with some interesting special cases and then work my way up to the general theory. While this unavoidably leads to some duplications, it usually provides much better motivation and implies that the core material always comes first (while the more general results are then optional). Nevertheless, my aim is not to present an encyclopedic treatment but to provide
a student with a versatile toolbox for further study. Moreover, in contradis-
tinction to many other books, I do not have a particular direction in mind
and hence I am trying to give a broad introduction which should prepare
you for diverse fields such as spectral theory, partial differential equations,
or probability theory. This is related to the fact that I am working in math-
ematical physics, an area where you never know what mathematical theory
you will need next.

I have tried to keep a balance between verbosity and clarity in the sense
that I have tried to provide sufficient detail for being able to follow the argu-
ments but without drowning the key ideas in boring details. In particular,
you will find a show this from time to time encouraging the reader to check
the claims made (these tasks typically involve only simple routine calcula-
tions). Moreover, to make the presentation student friendly, I have tried
to include many worked out examples within the main text. Some of them
are standard counterexamples pointing out the limitations of theorems (and
explaining why the assumptions are important). Others show how to use
the theory in the investigation of practical examples.

Preliminaries

The present manuscript is intended to be gentle when it comes to re-
quired background. Of course I assume basic familiarity with analysis (real
and complex numbers, limits, differentiation, basic integration, open sets)
and linear algebra (vector spaces). Apart from this natural assumptions
I also expect basic familiarity with metric spaces and elementary concepts
from point set topology. As this might not always be the case, I have re-
viewed all the necessary facts in a preliminary chapter. For convenience this
chapter contains full proofs in case one needs to fill some gaps. As some
things are only outlined (or outsourced to exercises) it will require extra
effort in case you see all this for the first time. Moreover, only a part is
required for the core results. On the other hand I do not assume familiarity
with Lebesgue integration and consequently $L^p$ spaces will only be briefly
mentioned as the completion of continuous functions with respect to the
corresponding integral norms in the first part. At a few places I also assume
some basic results from complex analysis but it will be sufficient to just
believe them.

Similarly, the second part on real analysis only requires a similar back-
ground and is essentially independent on the first part. Of course here you
should already know what a Banach/Hilbert space is, but Chapter I will be
sufficient to get you started.
Finaly, the last part of course requires a basic familiarity with functional analysis and measure theory. But apart from this it is again independent form the first two parts.

Content

The idea is that you start reading in Chapter 1. In particular, I do not expect you to know everything from Chapter 0 but I will refer you there from time to time such that you can refresh your memory should need arise. Moreover, you can always go there if you are unsure about a certain term (using the extensive index) or if there should be a need to clarify notation or conventions. I prefer this over referring you to several other books which might in the worst case not be readily available.

The first part starts with Fourier’s treatment of the heat equation which lead to the theory of Fourier analysis as well as the development of spectral theory which drove much of the development of functional analysis around the turn of the last century. In particular, the first chapter tries to introduce and motivate some of the key concepts, the second chapter discusses basic Hilbert space theory with applications to Fourier series, and the third chapter develops basic spectral theory for compact self-adjoint operators with applications to Sturm–Liouville problems. The fourth chapter discusses what is typically considered as the core results from Banach space theory. Some further results, in particular those requiring advanced topological concepts, are moved to an extra chapter such that they can be can be skipped. Finally, spectral theory for bounded self-adjoint operators is developed via the framework of $C^*$ algebras. Again a bottom-up approach is used such that the core results are in the first two sections and the rest is optional. I think that this gives a well-balanced introduction to functional analysis which contains several optional topics to choose from depending on personal preferences and time constraints. The main topic missing from my point of view is spectral theory for unbounded operators. However, this is beyond a first course and I refer you to my book [35] instead.

In a similar vein, the second part tries to give a succinct introduction to measure theory. I have chosen the Carathéodory approach because I feel that it is the most versatile one. Again the first two chapters contain the core material about measure theory and integration. Measures on $\mathbb{R}^n$ are introduced via distribution (the case of $n = 1$ is done first) which should meet the needs of partial differential equations, spectral theory, and probability theory. There is also an appendix on transforming one-dimensional measures which should be useful in both spectral theory and probability theory. Then there is a chapter with core material on $L^p$ spaces. Next there
are two chapters on advanced topics which are to a large extend independent of each other (again leaving several options to choose from). Finally there is a chapter on the Fourier transform (including a discussion of Sobolev spaces in $\mathbb{R}^n$) with some basic applications to linear partial differential equations and a brief chapter on interpolation.

Finally, there is a part on nonlinear functional analysis. The first chapter discusses analysis in Banach spaces (with a view towards applications in the calculus of variations and infinite dimensional dynamical systems). Then there are two chapters on degree theory and fixed-point theorems in finite and infinite dimensional spaces. These are then applied to the stationary Navier–Stokes equation and we close with a brief chapter on monotone maps.

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Finally, no book is free of errors. So if you find one, or if you have comments or suggestions (no matter how small), please let me know.

Gerald Teschl

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Part 1

Functional Analysis
Chapter 0

Warm up: Metric and topological spaces

0.1. Basics

Before we begin, I want to recall some basic facts from metric and topological spaces. I presume that you are familiar with these topics from your calculus course. As a general reference I can warmly recommend Kelly's classical book [19] or the nice book by Kaplansky [18]. As always such a brief compilation introduces a zoo of properties. While sometimes the connection between these properties are straightforward, oftentimes they might be quite tricky. So if at some point you are wondering if there exists an infinite multivariable sub-polynomial Woffle which does not satisfy the lower regular $Q$-property, start searching in the book by Steen and Seebach [34].

One of the key concepts in analysis is convergence. To define convergence requires the notion of distance. Motivated by the Euclidean distance one is lead to the following definition:

A **metric space** is a space $X$ together with a distance function $d : X \times X \rightarrow \mathbb{R}$ such that for arbitrary points $x, y, z \in X$ we have

1. $d(x, y) \geq 0$,
2. $d(x, y) = 0$ if and only if $x = y$,
3. $d(x, y) = d(y, x)$,
4. $d(x, z) \leq d(x, y) + d(y, z)$ (**triangle inequality**).


If (ii) does not hold, $d$ is called a pseudometric. As a straightforward consequence we record the inverse triangle inequality (Problem 0.1)

$$|d(x, y) - d(z, y)| \leq d(x, z). \quad (0.1)$$

**Example.** The role model for a metric space is of course Euclidean space $\mathbb{R}^n$ together with $d(x, y) := \left(\sum_{k=1}^{n}(x_k - y_k)^2\right)^{1/2}$ as well as $\mathbb{C}^n$ together with $d(x, y) := \left(\sum_{k=1}^{n}|x_k - y_k|^2\right)^{1/2}$.

Several notions from $\mathbb{R}^n$ carry over to metric spaces in a straightforward way. The set $B_r(x) := \{y \in X|d(x, y) < r\}$ (0.2)
is called an open ball around $x$ with radius $r > 0$. We will write $B_r^X(x)$ in case we want to emphasize the corresponding space. A point $x$ of some set $U \subseteq X$ is called an interior point of $U$ if $U$ contains some ball around $x$. If $x$ is an interior point of $U$, then $U$ is also called a neighborhood of $x$. A point $x$ is called a limit point of $U$ (also accumulation or cluster point) if $B_r(x) \cap (U \setminus \{x\}) \neq \emptyset$ for every ball around $x$. Note that a limit point $x$ need not lie in $U$, but $U$ must contain points arbitrarily close to $x$. A point $x$ is called an isolated point of $U$ if there exists a neighborhood of $x$ not containing any other points of $U$. A set which consists only of isolated points is called a discrete set. If any neighborhood of $x$ contains at least one point in $U$ and at least one point not in $U$, then $x$ is called a boundary point of $U$. The set of all boundary points of $U$ is called the boundary of $U$ and denoted by $\partial U$.

**Example.** Consider $\mathbb{R}$ with the usual metric and let $U := (-1, 1)$. Then every point $x \in U$ is an interior point of $U$. The points $[-1, 1]$ are limit points of $U$, and the points $\{-1, +1\}$ are boundary points of $U$.

Let $U := \mathbb{Q}$, the set of rational numbers. Then $U$ has no interior points and $\partial U = \mathbb{R}$.

A set all of whose points are interior points is called open. The family of open sets $\mathcal{O}$ satisfies the properties

(i) $\emptyset, X \in \mathcal{O}$,

(ii) $O_1, O_2 \in \mathcal{O}$ implies $O_1 \cap O_2 \in \mathcal{O}$,

(iii) $\{O_\alpha\} \subseteq \mathcal{O}$ implies $\bigcup_\alpha O_\alpha \in \mathcal{O}$.

That is, $\mathcal{O}$ is closed under finite intersections and arbitrary unions. Indeed, (i) is obvious, (ii) follows since the intersection of two open balls centered at $x$ is again an open ball centered at $x$ (explicitly $B_{r_1}(x) \cap B_{r_2}(x) = B_{\min(r_1, r_2)}(x)$), and (iii) follows since every ball contained in one of the sets is also contained in the union.
Now it turns out that for defining convergence, a distance is slightly more than what is actually needed. In fact, it suffices to know when a point is in the neighborhood of another point. And if we adapt the definition of a neighborhood by requiring it to contain an open set around $x$, then we see that it suffices to know when a set is open. This motivates the following definition:

A space $X$ together with a family of sets $\mathcal{O}$, the open sets, satisfying (i)–(iii), is called a **topological space**. The notions of interior point, limit point, and neighborhood carry over to topological spaces if we replace open ball around $x$ by open set containing $x$.

There are usually different choices for the topology. Two not too interesting examples are the **trivial topology** $\mathcal{O} = \{\emptyset, X\}$ (the power set of $X$). Given two topologies $\mathcal{O}_1$ and $\mathcal{O}_2$ on $X$, $\mathcal{O}_1$ is called **weaker** (or **coarser**) than $\mathcal{O}_2$ if $\mathcal{O}_1 \subseteq \mathcal{O}_2$. Conversely, $\mathcal{O}_1$ is called **stronger** (or **finer**) than $\mathcal{O}_2$ if $\mathcal{O}_2 \subseteq \mathcal{O}_1$.

**Example.** Note that different metrics can give rise to the same topology. For example, we can equip $\mathbb{R}^n$ (or $\mathbb{C}^n$) with the Euclidean distance $d(x,y)$ as before or we could also use

$$
\tilde{d}(x,y) := \sum_{k=1}^{n} |x_k - y_k|.
$$

Then

$$
\frac{1}{\sqrt{n}} \sum_{k=1}^{n} |x_k| \leq \sqrt{\sum_{k=1}^{n} |x_k|^2} \leq \sum_{k=1}^{n} |x_k|
$$

shows $B_{\sqrt{n}}(x) \subseteq \tilde{B}_r(x) \subseteq B_r(x)$, where $B, \tilde{B}$ are balls computed using $d$, $\tilde{d}$, respectively. In particular, both distances will lead to the same notion of convergence.

**Example.** We can always replace a metric $d$ by the bounded metric

$$
\tilde{d}(x,y) := \frac{d(x,y)}{1 + d(x,y)}
$$

without changing the topology (since the family of open balls does not change: $B_{\delta}(x) = \tilde{B}_{\delta/(1+\delta)}(x)$). To see that $\tilde{d}$ is again a metric, observe that $f(r) = \frac{r}{1+r}$ is monotone as well as concave and hence subadditive, $f(r+s) \leq f(r) + f(s)$ (Problem 0.3).

Every subspace $Y$ of a topological space $X$ becomes a topological space of its own if we call $O \subseteq Y$ open if there is some open set $\tilde{O} \subseteq X$ such that $O = \tilde{O} \cap Y$. This natural topology $\mathcal{O} \cap Y$ is known as the **relative topology** (also **subspace**, **trace** or **induced topology**).
Example. The set $(0,1] \subseteq \mathbb{R}$ is not open in the topology of $X := \mathbb{R}$, but it is open in the relative topology when considered as a subset of $Y := [-1,1]$.

A family of open sets $\mathcal{B} \subseteq \mathcal{O}$ is called a base for the topology if for each $x$ and each neighborhood $U(x)$, there is some set $O \in \mathcal{B}$ with $x \in O \subseteq U(x)$. Since an open set $O$ is a neighborhood of every one of its points, it can be written as $O = \bigcup_{O \supseteq \tilde{O} \in \mathcal{B}} \tilde{O}$ and we have

**Lemma 0.1.** A family of open sets $\mathcal{B} \subseteq \mathcal{O}$ is a base for the topology if and only if every open set can be written as a union of elements from $\mathcal{B}$.

**Proof.** To see the converse let $x$ and $U(x)$ be given. Then $U(x)$ contains an open set $O$ containing $x$ which can be written as a union of elements from $\mathcal{B}$. One of the elements in this union must contain $x$ and this is the set we are looking for. □

There is also a local version of the previous notions. A neighborhood base for a point $x$ is a collection of neighborhoods $\mathcal{B}(x)$ of $x$ such that for each neighborhood $U(x)$, there is some set $B \in \mathcal{B}(x)$ with $B \subseteq U(x)$. Note that the sets in a neighborhood base are not required to be open.

If every point has a countable neighborhood base, then $X$ is called first countable. If there exists a countable base, then $X$ is called second countable. Note that a second countable space is in particular first countable since for every base $\mathcal{B}$ the subset $\mathcal{B}(x) := \{O \in \mathcal{B} | x \in O\}$ is a neighborhood base for $x$.

**Example.** By construction, in a metric space the open balls $\{B_{1/m}(x)\}_{m \in \mathbb{N}}$ are a neighborhood base for $x$. Hence every metric space is first countable.

Taking the union over all $x$, we obtain a base. In the case of $\mathbb{R}^n$ (or $\mathbb{C}^n$) it even suffices to take balls with rational center, and hence $\mathbb{R}^n$ (as well as $\mathbb{C}^n$) is second countable. □

Given two topologies on $X$ their intersection will again be a topology on $X$. In fact, the intersection of an arbitrary collection of topologies is again a topology and hence given a collection $\mathcal{M}$ of subsets of $X$ we can define the topology generated by $\mathcal{M}$ as the smallest topology (i.e., the intersection of all topologies) containing $\mathcal{M}$. Note that if $\mathcal{M}$ is closed under finite intersections and $\emptyset, X \in \mathcal{M}$, then it will be a base for the topology generated by $\mathcal{M}$.

Given two bases we can use them to check if the corresponding topologies are equal.

**Lemma 0.2.** Let $\mathcal{O}_j$, $j = 1,2$ be two topologies for $X$ with corresponding bases $\mathcal{B}_j$. Then $\mathcal{O}_1 \subseteq \mathcal{O}_2$ if and only if for every $x \in X$ and every $B_1 \in \mathcal{B}_1$ with $x \in B_1$ there is some $B_2 \in \mathcal{B}_2$ such that $x \in B_2 \subseteq B_1$. 

0.1. Basics

Proof. Suppose $\mathcal{O}_1 \subseteq \mathcal{O}_2$, then $B_1 \in \mathcal{O}_2$ and there is a corresponding $B_2$ by the very definition of a base. Conversely, let $\mathcal{O}_1 \in \mathcal{O}_1$ and pick some $x \in \mathcal{O}_1$. Then there is some $B_1 \in \mathcal{B}_1$ with $x \in B_1 \subseteq \mathcal{O}_1$ and by assumption some $B_2 \in \mathcal{B}_2$ such that $x \in B_2 \subseteq B_1 \subseteq \mathcal{O}_1$ which shows that $x$ is an interior point with respect to $\mathcal{O}_2$. Since $x$ was arbitrary we conclude $\mathcal{O}_1 \in \mathcal{O}_2$. □

The next definition will ensure that limits are unique: A topological space is called a Hausdorff space if for any two different points there are always two disjoint neighborhoods.

Example. Any metric space is a Hausdorff space: Given two different points $x$ and $y$, the balls $B_{d/2}(x)$ and $B_{d/2}(y)$, where $d = d(x, y) > 0$, are disjoint neighborhoods. A pseudometric space will in general not be Hausdorff since two points of distance 0 cannot be separated by open balls.

The complement of an open set is called a closed set. It follows from De Morgan’s laws

$$X \setminus \left( \bigcup \alpha U_\alpha \right) = \bigcap \alpha (X \setminus U_\alpha), \quad X \setminus \left( \bigcap \alpha U_\alpha \right) = \bigcup \alpha (X \setminus U_\alpha)$$

that the family of closed sets $\mathcal{C}$ satisfies

(i) $\emptyset, X \in \mathcal{C}$,

(ii) $C_1, C_2 \in \mathcal{C}$ implies $C_1 \cup C_2 \in \mathcal{C}$,

(iii) $\{C_\alpha\} \subseteq \mathcal{C}$ implies $\bigcap \alpha C_\alpha \in \mathcal{C}$.

That is, closed sets are closed under finite unions and arbitrary intersections.

The smallest closed set containing a given set $U$ is called the closure

$$\overline{U} := \bigcap \{C \in \mathcal{C} : U \subseteq C\}$$

and the largest open set contained in a given set $U$ is called the interior

$$U^o := \bigcup \{O \in \mathcal{O} : O \subseteq U\}$$

It is not hard to see that the closure satisfies the following axioms (Kuratowski closure axioms):

(i) $\emptyset = \emptyset$,

(ii) $U \subseteq \overline{U}$,

(iii) $\overline{\overline{U}} = \overline{U}$,

(iv) $\overline{U \cup V} = \overline{U} \cup \overline{V}$.

In fact, one can show that these axioms can equivalently be used to define the topology by observing that the closed sets are precisely those which satisfy $\overline{A} = A$. 
**Lemma 0.3.** Let $X$ be a topological space. Then the interior of $U$ is the set of all interior points of $U$, and the closure of $U$ is the union of $U$ with all limit points of $U$. Moreover, $\partial U = \overline{U} \setminus U^\circ$.

**Proof.** The first claim is straightforward. For the second claim observe that by Problem 0.6 we have that $\overline{U} = (X \setminus (X \setminus U)^\circ)$, that is, the closure is the set of all points which are not interior points of the complement. That is, $x \not\in \overline{U}$ iff there is some open set $O$ containing $x$ with $O \subseteq X \setminus U$. Hence, $x \in \overline{U}$ iff for all open sets $O$ containing $x$ we have $O \nsubseteq X \setminus U$, that is, $\overline{U} = X \setminus U$. The last claim is left as Problem 0.7. \[ \square \]

**Example.** For any $x \in X$ the **closed ball**

$$B_r(x) := \{ y \in X | d(x, y) \leq r \}$$

(0.9)

is a closed set (check that its complement is open). But in general we have only

$$\overline{B_r(x)} \subseteq B_r(x)$$

(0.10)

since an isolated point $y$ with $d(x, y) = r$ will not be a limit point. In $\mathbb{R}^n$ (or $\mathbb{C}^n$) we have of course equality. \[ \diamond \]

**Problem 0.1.** Show that $|d(x, y) - d(z, y)| \leq d(x, z)$.

**Problem 0.2.** Show the **quadrangle inequality** $|d(x, y) - d(x', y')| \leq d(x, x') + d(y, y')$.

**Problem 0.3.** Show that if $f : [0, \infty) \to \mathbb{R}$ is concave, $f(\lambda x + (1-\lambda)y) \geq \lambda f(x) + (1-\lambda)f(y)$ for $\lambda \in [0, 1]$, and satisfies $f(0) = 0$, then it is subadditive, $f(x+y) \leq f(x)+f(y)$. If in addition $f$ is increasing and $d$ is a pseudometric, then so is $f(d)$. (Hint: Begin by showing $f(\lambda x) \geq \lambda f(x)$.)

**Problem 0.4.** Show De Morgan’s laws.

**Problem 0.5.** Show that the closure satisfies the Kuratowski closure axioms.

**Problem 0.6.** Show that the closure and interior operators are dual in the sense that

$$X \setminus \overline{U} = (X \setminus U)^\circ \quad \text{and} \quad X \setminus U^\circ = \overline{(X \setminus U)}.$$  

In particular, the closure is the set of all points which are not interior points of the complement. (Hint: De Morgan’s laws.)

**Problem 0.7.** Show that the boundary of $U$ is given by $\partial U = \overline{U} \setminus U^\circ$. 

- **Warm up: Metric and topological spaces**
0.2. Convergence and completeness

A sequence \((x_n)_{n=1}^{\infty} \in X^N\) is said to converge to some point \(x \in X\) if \(\lim_{n \to \infty} d(x, x_n) = 0\). We write \(\lim_{n \to \infty} x_n = x\) or \(x_n \to x\) as usual in this case. Clearly the limit \(x\) is unique if it exists (this is not true for a pseudometric). We will also frequently identify the sequence with its values \(x_n\) for simplicity of notation.

Note that convergent sequences are bounded. Here a set \(U \subseteq X\) is called bounded if it is contained within a ball, that is \(U \subseteq B_r(x)\) for some \(x \in X\) and \(r > 0\).

Note that convergence can also be equivalently formulated in topological terms: A sequence \(x_n\) converges to \(x\) if and only if for every neighborhood \(U(x)\) of \(x\) there is some \(N \in \mathbb{N}\) such that \(x_n \in U(x)\) for \(n \geq N\). In a Hausdorff space the limit is unique. However, sequences usually do not suffice to describe a topology and in general definitions in terms of sequences are weaker (see the example below). This can be avoided by using generalized sequences, so-called nets, where the index set \(\mathbb{N}\) is replaced by arbitrary directed sets.

**Example.** For example, we can call a set \(U\) sequentially closed if every convergent sequence from \(U\) also has its limit in \(U\). If \(U\) is closed, then every point in the complement is an inner point of the complement, thus no sequence from \(U\) can converge to such a point. Hence every closed set is sequentially closed. In a metric space (or more generally in a first countable space) we can find a sequence for every limit point \(x\) by choosing a point (different from \(x\)) from every set in a neighborhood base. Hence the converse is also true in this case. \(\diamond\)

Note that the argument from the previous example shows that in a first countable space sequentially closed is the same as closed. In particular, in this case the family of closed sets is uniquely determined by the convergent sequences:

**Lemma 0.4.** Two first countable topologies agree if and only if they give rise to the same convergent sequences.

Of course every subsequence of a convergent sequence will converge to the same limit and we have the following converse:

**Lemma 0.5.** Let \(X\) be a topological space, \((x_n)_{n=1}^{\infty} \in X^N\) a sequence and \(x \in X\). If every subsequence has a further subsequence which converges to \(x\), then \(x_n\) converges to \(x\).
**Proof.** We argue by contradiction: If $x_n \not\to x$ we can find a neighborhood $U(x)$ and a subsequence $x_{n_k} \notin U(x)$. But then no subsequence of $x_{n_k}$ can converge to $x$. □

This innocent observation is often useful to establish convergence in situations where one knows that the limit of a subsequence solves a given problem together with uniqueness of solutions for this problem. It can also be used to show that a notion of convergence does not stem from a topology (cf. Problem 9.6).

In summary: A metric induces a natural topology and a topology induces a natural notion of convergence. However, a notion of convergence might not stem form a topology (or different topologies might give rise to the same notion of convergence) and a topology might not stem from a metric.

A sequence $(x_n)_{n=1}^\infty \in X^\mathbb{N}$ is called a **Cauchy sequence** if for every $\varepsilon > 0$ there is some $N \in \mathbb{N}$ such that

$$d(x_n, x_m) \leq \varepsilon, \quad n, m \geq N. \quad (0.11)$$

Every convergent sequence is a Cauchy sequence. If the converse is also true, that is, if every Cauchy sequence has a limit, then $X$ is called **complete**. It is easy to see that a Cauchy sequence converges if and only if it has a convergent subsequence.

**Example.** Both $\mathbb{R}^n$ and $\mathbb{C}^n$ are complete metric spaces. ◦

**Example.** The metric

$$d(x, y) := |\arctan(x) - \arctan(y)| \quad (0.12)$$

gives rise to the standard topology on $\mathbb{R}$ (since arctan is bi-Lipschitz on every compact interval). However, $x_n = n$ is a Cauchy sequence with respect to this metric but not with respect to the usual metric. Moreover, any sequence with $x_n \to \infty$ or $x_n \to -\infty$ will be Cauchy with respect to this metric and hence (show this) for the completion of $\mathbb{R}$ precisely the two new points $-\infty$ and $+\infty$ have to be added. ◦

As noted before, in a metric space $x$ is a limit point of $U$ if and only if there exists a sequence $(x_n)_{n=1}^\infty \subseteq U \setminus \{x\}$ with $\lim_{n \to \infty} x_n = x$. Hence $U$ is closed if and only if for every convergent sequence the limit is in $U$. In particular,

**Lemma 0.6.** A subset of a complete metric space is again a complete metric space if and only if it is closed.

A set $U \subseteq X$ is called **dense** if its closure is all of $X$, that is, if $\overline{U} = X$. A space is called **separable** if it contains a countable dense set.
Lemma 0.7. A metric space is separable if and only if it is second countable as a topological space.

Proof. From every dense set we get a countable base by considering open balls with rational radii and centers in the dense set. Conversely, from every countable base we obtain a dense set by choosing an element from each set in the base. □

Lemma 0.8. Let $X$ be a separable metric space. Every subset $Y$ of $X$ is again separable.

Proof. Let $A = \{ x_n \}_{n \in \mathbb{N}}$ be a dense set in $X$. The only problem is that $A \cap Y$ might contain no elements at all. However, some elements of $A$ must be at least arbitrarily close: Let $J \subseteq \mathbb{N}^2$ be the set of all pairs $(n,m)$ for which $B_{1/m}(x_n) \cap Y \neq \emptyset$ and choose some $y_{n,m} \in B_{1/m}(x_n) \cap Y$ for all $(n,m) \in J$. Then $B = \{ y_{n,m} \}_{(n,m) \in J} \subseteq Y$ is countable. To see that $B$ is dense, choose $y \in Y$. Then there is some sequence $x_{nk}$ with $d(x_{nk}, y) < 1/k$. Hence $(n_k, k) \in J$ and $d(y_{nk,k}, y) \leq d(y_{nk,k}, x_{nk}) + d(x_{nk}, y) \leq 2/k \to 0$. □

If $X$ is an (incomplete) metric space, consider the set of all Cauchy sequences $X$. Call two Cauchy sequences equivalent if their difference converges to zero and denote by $\bar{X}$ the set of all equivalence classes. Moreover, the quadrangle inequality (Problem 0.2) shows that if $x = (x_n)_{n \in \mathbb{N}}$ and $y = (y_n)_{n \in \mathbb{N}}$ are Cauchy sequences, so is $d(x_n, y_n)$ and hence we can define a metric on $\bar{X}$ via

$$d_{\bar{X}}([x], [y]) = \lim_{n \to \infty} d_X(x_n, y_n). \quad (0.13)$$

Indeed, it is straightforward to check that $d_{\bar{X}}$ is well defined (independent of the representative) and inherits all properties of a metric from $d_X$. Moreover, $d_{\bar{X}}$ agrees with $d_X$ whenever both Cauchy sequences converge in $X$.

Theorem 0.9. The space $\bar{X}$ is a metric space containing $X$ as a dense subspace if we identify $x \in X$ with the equivalence class of all sequences converging to $x$. Moreover, this embedding is isometric.

Proof. The map $J : X \to \bar{X}, x_0 \mapsto [(x_0, x_0, \ldots)]$ is an isometric embedding (i.e., it is injective and preserves the metric). Moreover, for a Cauchy sequence $x = (x_n)_{n \in \mathbb{N}}$ the sequence $J(x_n) \to x$ and hence $J(X)$ is dense in $\bar{X}$. It remains to show that $\bar{X}$ is complete. Let $\xi_n = [(x_{nj})_{j \in \mathbb{N}}]$ be a Cauchy sequence in $\bar{X}$. Then it is not hard to see that $\xi = [(x_{j,j})_{j \in \mathbb{N}}]$ is its limit. □

Let me remark that the completion $\bar{X}$ is unique. More precisely, suppose $\tilde{X}$ is another complete metric space which contains $X$ as a dense subset such that the embedding $\tilde{J} : X \hookrightarrow \bar{X}$ is isometric. Then $I = \tilde{J} \circ J^{-1} : J(X) \to \tilde{J}(X)$ has a unique isometric extension $\tilde{I} : \tilde{X} \to \bar{X}$ (compare Theorem 0.36
below). In particular, it is no restriction to assume that a metric space is complete.

**Problem 0.8.** Let $U \subseteq V$ be subsets of a metric space $X$. Show that if $U$ is dense in $V$ and $V$ is dense in $X$, then $U$ is dense in $X$.

**Problem 0.9.** Let $X$ be a metric space and denote by $B(X)$ the set of all bounded functions $X \to \mathbb{C}$. Introduce the metric 

$$d(f, g) = \sup_{x \in X} |f(x) - g(x)|.$$ 

Show that $B(X)$ is complete.

**Problem 0.10.** Let $X$ be a metric space and $B(X)$ as in the previous problem. Consider the embedding $J : X \hookrightarrow B(X)$ defined via 

$$y \mapsto J(x)(y) = d(x, y) - d(x_0, y)$$

for some fixed $x_0 \in X$. Show that this embedding is isometric. Hence $\overline{J(X)}$ is another (equivalent) completion of $X$.

### 0.3. Functions

Next, we come to functions $f : X \to Y$, $x \mapsto f(x)$. We use the usual conventions $f(U) := \{f(x) | x \in U\}$ for $U \subseteq X$ and $f^{-1}(V) := \{x | f(x) \in V\}$ for $V \subseteq Y$. Note 

$$U \subseteq f^{-1}(f(U)), \quad f(f^{-1}(V)) \subseteq V. \quad (0.14)$$

Recall that we always have 

$$f^{-1}(\bigcup_{\alpha} V_\alpha) = \bigcup_{\alpha} f^{-1}(V_\alpha), \quad f^{-1}(\bigcap_{\alpha} V_\alpha) = \bigcap_{\alpha} f^{-1}(V_\alpha),$$

$$f^{-1}(Y \setminus V) = X \setminus f^{-1}(V) \quad (0.15)$$

as well as 

$$f(\bigcup_{\alpha} U_\alpha) = \bigcup_{\alpha} f(U_\alpha), \quad f(\bigcap_{\alpha} U_\alpha) \subseteq \bigcap_{\alpha} f(U_\alpha),$$

$$f(X) \setminus f(U) \subseteq f(X \setminus U) \quad (0.16)$$

with equality if $f$ is injective. The set $\text{Ran}(f) := f(X)$ is called the range of $f$, and $X$ is called the domain of $f$. A function $f$ is called injective if for each $y \in Y$ there is at most one $x \in X$ with $f(x) = y$ (i.e., $f^{-1}(\{y\})$ contains at most one point) and surjective or onto if $\text{Ran}(f) = Y$. A function $f$ which is both injective and surjective is called one-to-one or bijective.

A function $f$ between metric spaces $X$ and $Y$ is called continuous at a point $x \in X$ if for every $\varepsilon > 0$ we can find a $\delta > 0$ such that 

$$d_Y(f(x), f(y)) \leq \varepsilon \quad \text{if} \quad d_X(x, y) < \delta. \quad (0.17)$$
If \( f \) is continuous at every point, it is called \textit{continuous}. In the case \( d_Y(f(x), f(y)) = d_X(x, y) \) we call \( f \) \textit{isometric} and every isometry is of course continuous.

**Lemma 0.10.** Let \( X, Y \) be metric spaces. The following are equivalent:

(i) \( f \) is continuous at \( x \) (i.e., \((0.17)\) holds).

(ii) \( f(x_n) \to f(x) \) whenever \( x_n \to x \).

(iii) For every neighborhood \( V \) of \( f(x) \) the preimage \( f^{-1}(V) \) is a neighborhood of \( x \).

**Proof.** (i) \( \Rightarrow \) (ii) is obvious. (ii) \( \Rightarrow \) (iii): If (iii) does not hold, there is a neighborhood \( V \) of \( f(x) \) such that \( B_\delta(x) \not\subseteq f^{-1}(V) \) for every \( \delta \). Hence we can choose a sequence \( x_n \in B_{1/n}(x) \) such that \( f(x_n) \not\in f^{-1}(V) \). Thus \( x_n \to x \) but \( f(x_n) \not\to f(x) \). (iii) \( \Rightarrow \) (i): Choose \( V = B_\varepsilon(f(x)) \) and observe that by (iii), \( B_\delta(x) \subseteq f^{-1}(V) \) for some \( \delta \). \( \square \)

The last item serves as a definition for topological spaces. In particular, it implies that \( f \) is continuous if and only if the inverse image of every open set is again open (equivalently, the inverse image of every closed set is closed). If the image of every open set is open, then \( f \) is called \textit{open}. A bijection \( f \) is called a \textit{homeomorphism} if both \( f \) and its inverse \( f^{-1} \) are continuous. Note that if \( f \) is a bijection, then \( f^{-1} \) is continuous if and only if \( f \) is open. Two topological spaces are called \textit{homeomorphic} if there is a homeomorphism between them.

In a general topological space we use (iii) as the definition of continuity and (ii) is called \textit{sequential continuity}. Then continuity will imply sequential continuity but the converse will not be true unless we assume (e.g.) that \( X \) is first countable (Problem 0.11).

The \textbf{support} of a function \( f : X \to \mathbb{C}^n \) is the closure of all points \( x \) for which \( f(x) \) does not vanish; that is,

\[
\text{supp}(f) := \{ x \in X | f(x) \neq 0 \}.
\]  
(0.18)

**Problem 0.11.** Let \( X, Y \) be topological spaces. Show that if \( f : X \to Y \) is continuous at \( x \in X \) then it is also sequential continuous. Show that the converse holds if \( X \) is first countable.

**Problem 0.12.** Let \( f : X \to Y \) be continuous. Then \( f(A) \subseteq f(\overline{A}) \).

**Problem 0.13.** Let \( X, Y \) be topological spaces and let \( f : X \to Y \) be continuous. Show that if \( X \) is separable, then so is \( f(Y) \).
Problem 0.14. Let $X$ be a topological space and $f : X \to \mathbb{R}$. Let $x_0 \in X$ and let $B(x_0)$ be a neighborhood base for $x_0$. Define

$$
\liminf_{x \to x_0} f(x) := \sup_{U \in B(x_0)} \inf_{U(x_0)} f,
\limsup_{x \to x_0} f(x) := \inf_{U \in B(x_0)} \sup_{U(x_0)} f.
$$

Show that both are independent of the neighborhood base and satisfy

(i) $\liminf_{x \to x_0} (-f(x)) = -\limsup_{x \to x_0} f(x)$.
(ii) $\liminf_{x \to x_0} (\alpha f(x)) = \alpha \liminf_{x \to x_0} f(x), \ \alpha \geq 0$.
(iii) $\liminf_{x \to x_0} (f(x) + g(x)) \geq \liminf_{x \to x_0} f(x) + \liminf_{x \to x_0} g(x)$.

Moreover, show that

$$
\liminf_{n \to \infty} f(x_n) \geq \liminf_{x \to x_0} f(x), \quad \limsup_{n \to \infty} f(x_n) \leq \limsup_{x \to x_0} f(x)
$$

for every sequence $x_n \to x_0$ and there exists a sequence attaining equality if $X$ is a metric space.

0.4. Product topologies

If $X$ and $Y$ are metric spaces, then $X \times Y$ together with

$$
d((x_1, y_1), (x_2, y_2)) := d_X(x_1, x_2) + d_Y(y_1, y_2)
$$

(0.19)

is a metric space. A sequence $(x_n, y_n)$ converges to $(x, y)$ if and only if $x_n \to x$ and $y_n \to y$. In particular, the projections onto the first $(x, y) \mapsto x$, respectively, onto the second $(x, y) \mapsto y$, coordinate are continuous. Moreover, if $X$ and $Y$ are complete, so is $X \times Y$.

In particular, by the inverse triangle inequality (0.1),

$$
|d(x_n, y_n) - d(x, y)| \leq d(x_n, x) + d(y_n, y),
$$

(0.20)

we see that $d : X \times X \to \mathbb{R}$ is continuous.

Example. If we consider $\mathbb{R} \times \mathbb{R}$, we do not get the Euclidean distance of $\mathbb{R}^2$ unless we modify (0.19) as follows:

$$
\tilde{d}((x_1, y_1), (x_2, y_2)) := \sqrt{d_X(x_1, x_2)^2 + d_Y(y_1, y_2)^2}.
$$

(0.21)

As noted in our previous example, the topology (and thus also convergence/continuity) is independent of this choice.

If $X$ and $Y$ are just topological spaces, the **product topology** is defined by calling $O \subseteq X \times Y$ open if for every point $(x, y) \in O$ there are open neighborhoods $U$ of $x$ and $V$ of $y$ such that $U \times V \subseteq O$. In other words, the products of open sets form a base of the product topology. Again the projections onto the first and second component are continuous. In the case of metric spaces this clearly agrees with the topology defined via the product metric (0.19). There is also another way of constructing the product
0.4. Product topologies

topology, namely, as the weakest topology which makes the projections continuous. In fact, this topology must contain all sets which are inverse images of open sets \( U \subseteq X \), that is all sets of the form \( U \times Y \) as well as all inverse images of open sets \( V \subseteq Y \), that is all sets of the form \( X \times V \). Adding finite intersections we obtain all sets of the form \( U \times V \) and hence the same base as before. In particular, a sequence \((x_n, y_n)\) will converge if and only if both components converge.

Note that the product topology immediately extends to the product of an arbitrary number of spaces \( X = \prod_{\alpha \in \mathcal{A}} X_\alpha \) by defining it as the weakest topology which makes all projections \( \pi_\alpha : X \to X_\alpha \) continuous.

**Example.** Let \( X \) be a topological space and \( A \) an index set. Then \( X^A = \prod_{\alpha \in \mathcal{A}} X_\alpha \) is the set of all functions \( x : A \to X \) and a neighborhood base at \( x \) are sets of functions which coincide with \( x \) at a given finite number of points. Convergence with respect to the product topology corresponds to pointwise convergence (note that the projection \( \pi_\alpha \) is the point evaluation at \( \alpha : \pi_\alpha(x) = x(\alpha) \)). If \( A \) is uncountable (and \( X \) is not equipped with the trivial topology), then there is no countable neighborhood base (if there were such a base, it would involve only a countable number of points, now choose a point from the complement...). In particular, there is no corresponding metric even if \( X \) has one. Moreover, this topology cannot be characterized with sequences alone. For example, let \( X = \{0, 1\} \) (with the discrete topology) and \( A = \mathbb{R} \). Then the set \( F = \{x | x^{-1}(1) \text{ is countable}\} \) is sequentially closed but its closure is all of \( \{0, 1\}^\mathbb{R} \) (every set from our neighborhood base contains an element which vanishes except at finitely many points).

In fact this is a special case of a more general construction which is often used. Let \( \{f_\alpha\}_{\alpha \in A} \) be a collection of functions \( f_\alpha : X \to Y_\alpha \), where \( Y_\alpha \) are some topological spaces. Then we can equip \( X \) with the weakest topology (known as the initial topology) which makes all \( f_\alpha \) continuous. That is, we take the topology generated by sets of the forms \( f_\alpha^{-1}(O_\alpha) \), where \( O_\alpha \subseteq Y_\alpha \) is open, known as open cylinders. Finite intersections of such sets, known as open cylinder sets, are hence a base for the topology and a sequence \( x_n \) will converge to \( x \) if and only if \( f_\alpha(x_n) \to f_\alpha(x) \) for all \( \alpha \in A \). In particular, if the collection is countable, then \( X \) will be first (or second) countable if all \( Y_\alpha \) are.

The initial topology has the following characteristic property:

**Lemma 0.11.** Let \( X \) have the initial topology from a collection of functions \( \{f_\alpha\}_{\alpha \in A} \). A function \( f : Z \to X \) is continuous (at \( z \)) if and only if \( f_\alpha \circ f \) is continuous (at \( z \)) for all \( \alpha \in A \).

**Proof.** If \( f \) is continuous at \( z \), then so is the composition \( f_\alpha \circ f \). Conversely, let \( U \subseteq X \) be a neighborhood of \( f(z) \). Then \( \bigcap_{j=1}^n f_{\alpha_j}^{-1}(O_{\alpha_j}) \subseteq U \) for some \( \alpha_j \).
and some open neighborhoods \(O_{\alpha_j}\) of \(f_{\alpha_j}(f(z))\). But then \(f^{-1}(U)\) contains the neighborhood \(f^{-1}((\bigcap_{j=1}^n f_{\alpha_j}^{-1}(O_{\alpha_j})) = \bigcap_{j=1}^n (f_{\alpha_j} \circ f)^{-1}(O_{\alpha_j})\) of \(z\). \(\square\)

If all \(X_\alpha\) are Hausdorff and if the collection \(\{f_\alpha\}_{\alpha \in A}\) separates points, that is for every \(x \neq y\) there is some \(\alpha\) with \(f_\alpha(x) \neq f_\alpha(y)\), then \(X\) will again be Hausdorff. Indeed for \(x \neq y\) choose \(\alpha\) such that \(f_\alpha(x) \neq f_\alpha(y)\) and let \(U_\alpha, V_\alpha\) be two disjoint neighborhoods separating \(f_\alpha(x), f_\alpha(y)\). Then \(f_\alpha^{-1}(U_\alpha), f_\alpha^{-1}(V_\alpha)\) are two disjoint neighborhoods separating \(x, y\). In particular \(X = \times_{\alpha \in A} X_\alpha\) is Hausdorff if all \(X_\alpha\) are.

Note that a similar construction works in the other direction. Let \(\{f_\alpha\}_{\alpha \in A}\) be a collection of functions \(f_\alpha : X_\alpha \to Y\), where \(X_\alpha\) are some topological spaces. Then we can equip \(Y\) with the strongest topology (known as the final topology) which makes all \(f_\alpha\) continuous. That is, we take as open sets those for which \(f_\alpha^{-1}(O)\) is open for all \(\alpha \in A\).

The prototypical example being the quotient topology: Let \(\sim\) be an equivalence relation on \(X\) with equivalence classes \([x] = \{y \in X | x \sim y\}\). Then the quotient topology on the set of equivalence classes \(X/\sim\) is the final topology of the projection map \(\pi : X \to X/\sim\).

**Lemma 0.12.** Let \(Y\) have the final topology from a collection of functions \(\{f_\alpha\}_{\alpha \in A}\) and let \(Z\) be another topological space. A function \(f : Y \to Z\) is continuous if and only if \(f \circ f_\alpha\) is continuous for all \(\alpha \in A\).

**Proof.** If \(f\) is continuous, then so is the composition \(f \circ f_\alpha\). Conversely, let \(V \subseteq Z\) be open. Then \(f \circ f_\alpha\) implies \((f \circ f_\alpha)^{-1}(V) = f_\alpha^{-1}(f^{-1}(V))\) open for all \(\alpha\) and hence \(f^{-1}(V)\) open. \(\square\)

**Problem 0.15.** Let \(X = \times_{\alpha \in A} X_\alpha\) with the product topology. Show that the projection maps are open.

**Problem 0.16** (Gluing lemma). Suppose \(X, Y\) are topological spaces and \(f_\alpha : A_\alpha \to Y\) are continuous functions defined on \(A_\alpha \subseteq X\). Suppose \(f_\alpha = f_\beta\) on \(A_\alpha \cap A_\beta\) such that \(f : A := \bigcup_{\alpha} A_\alpha \to Y\) is well defined by \(f(x) = f_\alpha(x)\) if \(x \in A_\alpha\). Show that \(f\) is continuous if either all sets \(A_\alpha\) are open or if the collection \(A_\alpha\) is finite and all are closed.

**Problem 0.17.** Let \(\{(X_j, d_j)\}_{j \in \mathbb{N}}\) be a sequence of metric spaces. Show that

\[
    d(x, y) = \sum_{j \in \mathbb{N}} \frac{1}{2^j} \frac{d_j(x_j, y_j)}{1 + d_j(x_j, y_j)} \quad \text{or} \quad d(x, y) = \max_{j \in \mathbb{N}} \frac{1}{2^j} \frac{d_j(x_j, y_j)}{1 + d_j(x_j, y_j)}
\]

is a metric on \(X = \times_{n \in \mathbb{N}} X_n\) which generates the product topology. Show that \(X\) is complete if all \(X_n\) are.
0.5. Compactness

A **cover** of a set $Y \subseteq X$ is a family of sets $\{U_\alpha\}$ such that $Y \subseteq \bigcup_\alpha U_\alpha$. A cover is called open if all $U_\alpha$ are open. Any subset of $\{U_\alpha\}$ which still covers $Y$ is called a **subcover**.

**Lemma 0.13** (Lindelöf). If $X$ is second countable, then every open cover has a countable subcover.

**Proof.** Let $\{U_\alpha\}$ be an open cover for $Y$, and let $\mathcal{B}$ be a countable base. Since every $U_\alpha$ can be written as a union of elements from $\mathcal{B}$, the set of all $B \in \mathcal{B}$ which satisfy $B \subseteq U_\alpha$ for some $\alpha$ form a countable open cover for $Y$. Moreover, for every $B_n$ in this set we can find an $\alpha_n$ such that $B_n \subseteq U_{\alpha_n}$. By construction, $\{U_{\alpha_n}\}$ is a countable subcover. $\Box$

A **refinement** $\{V_\beta\}$ of a cover $\{U_\alpha\}$ is a cover such that for every $\beta$ there is some $\alpha$ with $V_\beta \subseteq U_\alpha$. A cover is called **locally finite** if every point has a neighborhood that intersects only finitely many sets in the cover.

**Lemma 0.14** (Stone). In a metric space every countable open cover has a locally finite open refinement.

**Proof.** Denote the cover by $\{O_j\}_{j \in \mathbb{N}}$ and introduce the sets

$$\hat{O}_{j,n} := \bigcup_{x \in A_{j,n}} B_{2^{-n}}(x),$$

$$A_{j,n} := \{x \in O_j \setminus (O_1 \cup \cdots \cup O_{j-1}) | x \notin \bigcup_{k \in \mathbb{N}, 1 \leq l < n} \hat{O}_{k,l} \text{ and } B_{3 \cdot 2^{-n}}(x) \subseteq O_j\}.$$

Then, by construction, $\hat{O}_{j,n}$ is open, $\hat{O}_{j,n} \subseteq O_j$, and it is a cover since for every $x$ there is a smallest $j$ such that $x \in O_j$ and a smallest $n$ such that $B_{3 \cdot 2^{-n}}(x) \subseteq O_j$ implying $x \in \hat{O}_{k,l}$ for some $l \leq n$.

To show that $\hat{O}_{j,n}$ is locally finite fix some $x$ and let $j$ be the smallest integer such that $x \in \hat{O}_{j,n}$ for some $n$. Moreover, choose $m$ such that $B_{2^{-m}}(x) \subseteq \hat{O}_{j,n}$. It suffices to show that:

(i) If $i \geq n + m$ then $B_{2^{-n-m}}(x)$ is disjoint from $\hat{O}_{k,i}$ for all $k$.

(ii) If $i < n + m$ then $B_{2^{-n-m}}(x)$ intersects $\hat{O}_{k,i}$ for at most one $k$.

To show (i) observe that since $i > n$ every ball $B_{2^{-i}}(y)$ used in the definition of $\hat{O}_{k,i}$ has its center outside of $\hat{O}_{j,n}$. Hence $d(x,y) \geq 2^{-m}$ and $B_{2^{-n-m}}(x) \cap B_{2^{-i}}(x) = \emptyset$ since $i \geq m + 1$ as well as $n + m \geq m + 1$.

To show (ii) let $y \in \hat{O}_{j,i}$ and $z \in \hat{O}_{k,i}$ with $j < k$. We will show $d(y,z) \geq 2^{-n-m+1}$. There are points $r$ and $s$ such that $y \in B_{2^{-i}}(r) \subseteq \hat{O}_{j,i}$ and $z \in B_{2^{-i}}(s) \subseteq \hat{O}_{k,i}$. Then by definition $B_{3 \cdot 2^{-i}}(r) \subseteq O_j$ but $s \notin O_j$. So $d(r,s) \geq 3 \cdot 2^{-i}$ and $d(y,z) > 2^{-i} \geq 2^{-n-m+1}$. $\Box$
A subset \( K \subset X \) is called **compact** if every open cover of \( K \) has a finite subcover. A set is called **relatively compact** if its closure is compact.

**Lemma 0.15.** A topological space is compact if and only if it has the **finite intersection property**: The intersection of a family of closed sets is empty if and only if the intersection of some finite subfamily is empty.

**Proof.** By taking complements, to every family of open sets there is a corresponding family of closed sets and vice versa. Moreover, the open sets are a cover if and only if the corresponding closed sets have empty intersection. \( \square \)

**Lemma 0.16.** Let \( X \) be a topological space.

(i) The continuous image of a compact set is compact.

(ii) Every closed subset of a compact set is compact.

(iii) If \( X \) is Hausdorff, every compact set is closed.

(iv) The product of finitely many compact sets is compact.

(v) The finite union of compact sets is again compact.

(vi) If \( X \) is Hausdorff, any intersection of compact sets is again compact.

**Proof.**

(i) Observe that if \( \{ O_\alpha \} \) is an open cover for \( f(Y) \), then \( \{ f^{-1}(O_\alpha) \} \) is one for \( Y \).

(ii) Let \( \{ O_\alpha \} \) be an open cover for the closed subset \( Y \) (in the induced topology). Then there are open sets \( \bar{O}_\alpha \) with \( O_\alpha = \bar{O}_\alpha \cap Y \) and \( \{ \bar{O}_\alpha \} \cup \{ X \setminus Y \} \) is an open cover for \( X \) which has a finite subcover. This subcover induces a finite subcover for \( Y \).

(iii) Let \( Y \subseteq X \) be compact. We show that \( X \setminus Y \) is open. Fix \( x \in X \setminus Y \) (if \( Y = X \), there is nothing to do). By the definition of Hausdorff, for every \( y \in Y \) there are disjoint neighborhoods \( V(y) \) of \( y \) and \( U_y(x) \) of \( x \). By compactness of \( Y \), there are \( y_1, \ldots, y_n \) such that the \( V(y_j) \) cover \( Y \). But then \( U(x) = \bigcap_{j=1}^n U_{y_j}(x) \) is a neighborhood of \( x \) which does not intersect \( Y \).

(iv) Let \( \{ O_\alpha \} \) be an open cover for \( X \times Y \). For every \( (x, y) \in X \times Y \) there is some \( \alpha(x, y) \) such that \( (x, y) \in O_{\alpha(x, y)} \). By definition of the product topology there is some open rectangle \( U(x, y) \times V(x, y) \subseteq O_{\alpha(x, y)} \). Hence for fixed \( x \), \( \{ V(x, y) \}_{y \in Y} \) is an open cover of \( Y \). Hence there are finitely many points \( y_k(x) \) such that the \( V(x, y_k(x)) \) cover \( Y \). Set \( U(x) = \bigcap_k U(x, y_k(x)) \). Since finite intersections of open sets are open, \( \{ U(x) \}_{x \in X} \) is an open cover and there are finitely many points \( x_j \) such that the \( U(x_j) \) cover \( X \). By construction, the \( U(x_j) \times V(x_j, y_k(x_j)) \subseteq O_{\alpha(x_j, y_k(x_j))} \) cover \( X \times Y \).
(v) Note that a cover of the union is a cover for each individual set and the union of the individual subcovers is the subcover we are looking for.

(vi) Follows from (ii) and (iii) since an intersection of closed sets is closed.

□

As a consequence we obtain a simple criterion when a continuous function is a homeomorphism.

Corollary 0.17. Let $X$ and $Y$ be topological spaces with $X$ compact and $Y$ Hausdorff. Then every continuous bijection $f : X \to Y$ is a homeomorphism.

Proof. It suffices to show that $f$ maps closed sets to closed sets. By (ii) every closed set is compact, by (i) its image is also compact, and by (iii) it is also closed.

□

Moreover, item (iv) generalizes to arbitrary products:

Theorem 0.18 (Tychonoff). The product $\prod_{\alpha \in A} K_\alpha$ of an arbitrary collection of compact topological spaces $\{K_\alpha\}_{\alpha \in A}$ is compact with respect to the product topology.

Proof. We say that a family $\mathcal{F}$ of closed subsets of $K$ has the finite intersection property if the intersection of every finite subfamily has nonempty intersection. The collection of all such families which contain $\mathcal{F}$ is partially ordered by inclusion and every chain has an upper bound (the union of all sets in the chain). Hence, by Zorn’s lemma, there is a maximal family $\mathcal{F}_M$ (note that this family is closed under finite intersections).

Denote by $\pi_\alpha : K \to K_\alpha$ the projection onto the $\alpha$ component. Then the closed sets $\{\pi_\alpha(F)\}_{F \in \mathcal{F}_M}$ also have the finite intersection property and since $K_\alpha$ is compact, there is some $x_\alpha \in \bigcap_{F \in \mathcal{F}_M} \overline{\pi_\alpha(F)}$. Consequently, if $F_\alpha$ is a closed neighborhood of $x_\alpha$, then $\pi_\alpha^{-1}(F_\alpha) \in \mathcal{F}_M$ (otherwise there would be some $F \in \mathcal{F}_M$ with $F \cap \pi_\alpha^{-1}(F_\alpha) = \emptyset$ contradicting $F_\alpha \cap \pi_\alpha(F) \neq \emptyset$). Furthermore, for every finite subset $A_0 \subseteq A$ we have $\bigcap_{\alpha \in A_0} \pi_\alpha^{-1}(F_\alpha) \in \mathcal{F}_M$ and so every neighborhood of $x = (x_\alpha)_{\alpha \in A}$ intersects $F$. Since $F$ is closed, $x \in F$ and hence $x \in \bigcap_{\mathcal{F}_M} F$.

□

A subset $K \subset X$ is called sequentially compact if every sequence from $K$ has a convergent subsequence whose limit is in $K$. In a metric space, compact and sequentially compact are equivalent.

Lemma 0.19. Let $X$ be a metric space. Then a subset is compact if and only if it is sequentially compact.
Proof. Without loss of generality we can assume the subset to be all of $X$. Suppose $X$ is compact and let $x_n$ be a sequence which has no convergent subsequence. Then $K := \{x_n\}$ has no limit points and is hence compact by Lemma 0.16 (ii). For every $n$ there is a ball $B_{\varepsilon_n}(x_n)$ which contains only finitely many elements of $K$. However, finitely many suffice to cover $K$, a contradiction.

Conversely, suppose $X$ is sequentially compact and let $\{O_\alpha\}$ be some open cover which has no finite subcover. For every $x \in X$ we can choose some $\alpha(x)$ such that if $B_r(x)$ is the largest ball contained in $O_\alpha(x)$, then either $r \geq 1$ or there is no $\beta$ with $B_{2r}(x) \subseteq O_\beta$ (show that this is possible). Now choose a sequence $x_n$ such that $x_n \not\in \bigcup_{m<n} O_{\alpha(x_m)}$. Note that by construction the distance $d = d(x_m, x_n)$ to every successor of $x_m$ is either larger than 1 or the ball $B_{2d}(x_m)$ will not fit into any of the $O_\alpha$.

Now let $y$ be the limit of some convergent subsequence and fix some $r \in (0, 1)$ such that $B_r(y) \subseteq O_\alpha(y)$. Then this subsequence must eventually be in $B_{r/5}(y)$, but this is impossible since if $d := d(x_{n_1}, x_{n_2})$ is the distance between two consecutive elements of this subsequence, then $B_{2d}(x_{n_1})$ cannot fit into $O_{\alpha(y)}$ by construction whereas on the other hand $B_{2d}(x_{n_1}) \subseteq B_{4r/5}(a) \subseteq O_{\alpha(y)}$. □

If we drop the requirement that the limit must be in $K$, we obtain relatively compact sets:

Corollary 0.20. Let $X$ be a metric space and $K \subset X$. Then $K$ is relatively compact if and only if every sequence from $K$ has a convergent subsequence (the limit must not be in $K$).

Proof. For any sequence $x_n \in \overline{K}$ we can find a nearby sequence $y_n \in K$ with $x_n - y_n \to 0$. If we can find a convergent subsequence of $y_n$ then the corresponding subsequence of $x_n$ will also converge (to the same limit) and $K$ is (sequentially) compact in this case. The converse is trivial. □

As another simple consequence observe that

Corollary 0.21. A compact metric space $X$ is complete and separable.

Proof. Completeness is immediate from the previous lemma. To see that $X$ is separable note that, by compactness, for every $n \in \mathbb{N}$ there is a finite set $S_n \subseteq X$ such that the balls $\{B_{1/n}(x)\}_{x \in S_n}$ cover $X$. Then $\bigcup_{n \in \mathbb{N}} S_n$ is a countable dense set. □

In a metric space, a set is called bounded if it is contained inside some ball. Note that compact sets are always bounded since the can be covered by finitely many balls. In $\mathbb{R}^n$ (or $\mathbb{C}^n$) the converse also holds.
Theorem 0.22 (Heine–Borel). In $\mathbb{R}^n$ (or $\mathbb{C}^n$) a set is compact if and only if it is bounded and closed.

Proof. By Lemma 0.16 (ii), (iii), and (iv) it suffices to show that a closed interval in $I \subseteq \mathbb{R}$ is compact. Moreover, by Lemma 0.19, it suffices to show that every sequence in $I = [a, b]$ has a convergent subsequence. Let $x_n$ be our sequence and divide $I = [a, a + b/2] \cup [a + b/2, b]$. Then at least one of these two intervals, call it $I_1$, contains infinitely many elements of our sequence. Let $y_1 = x_{n_1}$ be the first one. Subdivide $I_1$ and pick $y_2 = x_{n_2}$, with $n_2 > n_1$ as before. Proceeding like this, we obtain a Cauchy sequence $y_n$ (note that by construction $I_{n+1} \subseteq I_n$ and hence $|y_n - y_m| \leq \frac{b-a}{2^n}$ for $m \geq n$).

By Lemma 0.19 this is equivalent to

Theorem 0.23 (Bolzano–Weierstraß). Every bounded infinite subset of $\mathbb{R}^n$ (or $\mathbb{C}^n$) has at least one limit point.

Combining Theorem 0.22 with Lemma 0.16 (i) we also obtain the extreme value theorem.

Theorem 0.24 (Weierstraß). Let $X$ be compact. Every continuous function $f : X \rightarrow \mathbb{R}$ attains its maximum and minimum.

A metric space for which the Heine–Borel theorem holds is called proper. Lemma 0.16 (ii) shows that $X$ is proper if and only if every closed ball is compact. Note that a proper metric space must be complete (since every Cauchy sequence is bounded). A topological space is called locally compact if every point has a compact neighborhood. Clearly a proper metric space is locally compact. It is called $\sigma$-compact, if it can be written as a countable union of compact sets. Again a proper space is $\sigma$-compact.

Lemma 0.25. For a metric space $X$ the following are equivalent:

(i) $X$ is separable and locally compact.

(ii) $X$ contains a countable base consisting of relatively compact sets.

(iii) $X$ is locally compact and $\sigma$-compact.

(iv) $X$ can be written as the union of an increasing sequence $U_n$ of relatively compact open sets satisfying $U_n \subseteq U_{n+1}$ for all $n$.

Proof. (i) $\Rightarrow$ (ii): Let $\{x_n\}$ be a dense set. Then the balls $B_{n,m} = B_{1/m}(x_n)$ form a base. Moreover, for every $n$ there is some $m_n$ such that $B_{n,m}$ is relatively compact for $m \leq m_n$. Since those balls are still a base we are done. (ii) $\Rightarrow$ (iii): Take the union over the closures of all sets in the base. (iii) $\Rightarrow$ (vi): Let $X = \bigcup_n K_n$ with $K_n$ compact. Without loss $K_n \subseteq K_{n+1}$. For a given set compact set $K$ we can find an relatively compact open set
V(K) such that K ⊆ V(K) (cover K by relatively compact open balls and choose a finite subcover). Now define $U_n = V(U_n)$. (vi) $\Rightarrow$ (i): Each of the sets $U_n$ has a countable dense subset by Corollary 0.21. The union gives a countable dense set for $X$. Since every $x \in U_n$ for some $n$, $X$ is also locally compact.

Note that since a sequence as in (iv) is a cover for $X$, every compact set is contained in $U_n$ for $n$ sufficiently large.

**Example.** $X = (0,1)$ with the usual metric is locally compact and $\sigma$-compact but not proper.

**Example.** Consider $\ell^2(\mathbb{N})$ with the standard basis $\delta^j$. Let $X_j := \{\lambda \delta^j | \lambda \in [0,1]\}$ and note that the metric on $X_j$ inherited from $\ell^2(\mathbb{Z})$ is the same as the usual metric from $\mathbb{R}$. Then $X := \bigcup_{j \in \mathbb{N}} X_j$ is a complete separable $\sigma$-compact space, which is however not locally compact. In fact, consider a ball of radius $\varepsilon$ around zero. Then $(\varepsilon/2) \delta^j \in B_\varepsilon(0)$ is a bounded sequence which has no convergent subsequence since $d((\varepsilon/2) \delta^j, (\varepsilon/2) \delta^k) = \varepsilon/\sqrt{2}$ for $k \neq j$.

However, under the assumptions of the previous lemma we can always switch to a new metric which generates the same topology and for which $X$ is proper. To this end recall that a function $f : X \to Y$ between topological spaces is called **proper** if the inverse image of a compact set is again compact. Now given a proper function (Problem 0.25) there is a new metric with the claimed properties (Problem 0.21).

**Problem 0.18** (One point compactification). Suppose $X$ is a locally compact Hausdorff space which is not compact. Introduce a new point $\infty$, set $\hat{X} = X \cup \{\infty\}$ and make it into a topological space by calling $O \subseteq \hat{X}$ open if either $\infty \notin O$ and $O$ is open in $X$ or if $\infty \in O$ and $\hat{X} \setminus O$ is compact. Show that $\hat{X}$ is a compact Hausdorff space which contains $X$ as a dense subset.

**Problem 0.19.** Show that every open set $O \subseteq \mathbb{R}$ can be written as a countable union of disjoint intervals. (Hint: Consider the set $\{I_\alpha\}$ of all maximal open subintervals of $O$; that is, $I_\alpha \subseteq O$ and there is no other subinterval of $O$ which contains $I_\alpha$.

**Problem 0.20.** A subset $U$ of a complete metric space $X$ is called **totally bounded** if for every $\varepsilon > 0$ it can be covered with a finite number of balls of radius $\varepsilon$. Show that a set is compact if and only it is totally bounded and closed. What are the totally bounded sets in $\mathbb{R}^n$?

**Problem 0.21.** Let $(X,d)$ be a metric space. Show that if there is a proper function $f : X \to \mathbb{R}$, then

$$\hat{d}(x,y) = d(x,y) + |f(x) - f(y)|$$
is a metric which generates the same topology and for which \((X, \tilde{d})\) is proper.

0.6. Separation

The **distance** between a point \(x \in X\) and a subset \(Y \subseteq X\) is

\[
\text{dist}(x, Y) := \inf_{y \in Y} d(x, y).
\]

Note that \(x\) is a limit point of \(Y\) if and only if \(\text{dist}(x, Y) = 0\) (Problem 0.22).

**Lemma 0.26.** Let \(X\) be a metric space and \(Y \subseteq X\) nonempty. Then

\[
|\text{dist}(x, Y) - \text{dist}(z, Y)| \leq d(x, z).
\]

In particular, \(x \mapsto \text{dist}(x, Y)\) is continuous.

**Proof.** Taking the infimum in the triangle inequality \(d(x, y) \leq d(x, z) + d(z, y)\) shows \(\text{dist}(x, Y) \leq d(x, z) + \text{dist}(z, Y)\). Hence \(\text{dist}(x, Y) - \text{dist}(z, Y) \leq d(x, z)\). Interchanging \(x\) and \(z\) shows \(\text{dist}(z, Y) - \text{dist}(x, Y) \leq d(x, z)\). □

**Lemma 0.27** (Urysohn). Suppose \(C_1\) and \(C_2\) are disjoint closed subsets of a metric space \(X\). Then there is a continuous function \(f : X \to [0, 1]\) such that \(f\) is zero on \(C_2\) and one on \(C_1\).

If \(X\) is locally compact and \(C_1\) is compact, there is a continuous function \(f : X \to [0, 1]\) with compact support which is one on \(C_1\).

**Proof.** To prove the first claim, set \(f(x) := \frac{\text{dist}(x, C_2)}{\text{dist}(x, C_1) + \text{dist}(x, C_2)}\). For the second claim, observe that there is an open set \(O\) such that \(\overline{O}\) is compact and \(C_1 \subseteq \overline{O} \subseteq X \setminus C_2\). In fact, for every \(x \in C_1\), there is a ball \(B_\varepsilon(x)\) such that \(\overline{B_\varepsilon(x)}\) is compact and \(\overline{B_\varepsilon(x)} \subseteq X \setminus C_2\). Since \(C_1\) is compact, finitely many of them cover \(C_1\) and we can choose the union of those balls to be \(O\). Now replace \(C_2\) by \(X \setminus O\). □

Note that Urysohn’s lemma implies that a metric space is **normal**; that is, for any two disjoint closed sets \(C_1\) and \(C_2\), there are disjoint open sets \(O_1\) and \(O_2\) such that \(C_j \subseteq O_j, j = 1, 2\). In fact, choose \(f\) as in Urysohn’s lemma and set \(O_1 := f^{-1}((0, 1/2))\), respectively, \(O_2 := f^{-1}((1/2, 1])\).

**Lemma 0.28.** Let \(X\) be a metric space and \(\{O_j\}\) a countable open cover. Then there is a continuous **partition of unity** subordinate to this cover; that is, there are continuous functions \(h_j : X \to [0, 1]\) such that \(h_j\) has support contained in \(O_j\) and

\[
\sum_j h_j(x) = 1.
\]
Moreover, the partition of unity can be chosen locally finite; that is, every \( x \) has a neighborhood where all but a finite number of the functions \( h_j \) vanish.

**Proof.** For notational simplicity we assume \( j \in \mathbb{N} \). Now introduce \( f_n(x) := \min(1, \sup_{j \leq n} d(x, X \setminus O_j)) \) and \( g_n = f_n - f_{n-1} \) (with the convention \( f_0(x) := 0 \)). Since \( f_n \) is increasing we have \( 0 \leq g_n \leq 1 \). Moreover, \( g_n(x) > 0 \) implies \( d(x, X \setminus O_n)) > 0 \) and thus \( \text{supp}(g_n) \subset O_n \). Next, by monotonicity \( f_\infty := \lim_{n \to \infty} f_n = \sum_n g_n \) exists and \( f_\infty(x) = 0 \) implies \( d(x, X \setminus O_j) = 0 \) for all \( j \), that is, \( x \notin O_j \) for all \( j \). Hence \( f_\infty \) is everywhere positive since \( \{O_j\} \) is a cover. Finally, by

\[
|f_n(x) - f_n(y)| \leq |\sup_{j \leq n} d(x, X \setminus O_j) - \sup_{j \leq n} d(y, X \setminus O_j)| \\
\leq \sup_{j \leq n} |d(x, X \setminus O_j) - d(y, X \setminus O_j)| \leq d(x, y)
\]

we see that all \( f_n \) (and hence all \( g_n \)) are continuous. Moreover, the very same argument shows that \( f_\infty \) is continuous and thus we have found the required partition of unity \( h_j = g_j / f_\infty \).

Finally, by Lemma 0.14 we can replace the cover by a locally finite refinement and the resulting partition of unity for this refinement will be locally finite. \( \square \)

Another important result is the **Tietze extension theorem**:

**Theorem 0.29** (Tietze). Suppose \( C \) is a closed subset of a metric space \( X \). For every continuous function \( f : C \to [-1, 1] \) there is a continuous extension \( \overline{f} : X \to [-1, 1] \).

**Proof.** The idea is to construct a rough approximation using Urysohn’s lemma and then iteratively improve this approximation. To this end we set \( C_1 := f^{-1}([\frac{1}{3}, 1]) \) and \( C_2 := f^{-1}([-1, -\frac{1}{3}]) \) and let \( g \) be the function from Urysohn’s lemma. Then \( f_1 := \frac{2g - 1}{3} \) satisfies \( |f(x) - f_1(x)| \leq \frac{2}{3} \) for \( x \in C \) as well as \( |f_1(x)| \leq \frac{1}{3} \) for all \( x \in X \). Applying this same procedure to \( f - f_1 \) we obtain a function \( f_2 \) such that \( |f(x) - f_1(x) - f_2(x)| \leq \left(\frac{2}{3}\right)^2 \) for \( x \in C \) and \( |f_2(x)| \leq \frac{1}{3} \left(\frac{2}{3}\right) \). Continuing this process we arrive at a sequence of functions \( f_n \) such that \( |f(x) - \sum_{j=1}^n f_j(x)| \leq \left(\frac{2}{3}\right)^n \) for \( x \in C \) and \( |f_n(x)| \leq \frac{1}{3} \left(\frac{2}{3}\right)^{n-1} \). By construction the corresponding series converges uniformly to the desired extension \( \overline{f} := \sum_{j=1}^\infty f_j \). \( \square \)

**Problem 0.22.** Show \( \text{dist}(x, Y) = \text{dist}(x, \overline{Y}) \). Moreover, show \( x \in \overline{Y} \) if and only if \( \text{dist}(x, Y) = 0 \).

**Problem 0.23.** Let \( Y, Z \subseteq X \) and define

\[
\text{dist}(Y, Z) := \inf_{y \in Y, z \in Z} d(y, z).
\]
Show \( \text{dist}(Y, Z) = \text{dist}(\overline{Y}, \overline{Z}) \). Moreover, show that if \( K \) is compact, then \( \text{dist}(K, Y) > 0 \) if and only if \( K \cap Y = \emptyset \).

**Problem 0.24.** Let \( K \subseteq U \) with \( K \) compact and \( U \) open. Show that there is some \( \varepsilon > 0 \) such that \( K_\varepsilon := \{ x \in X | \text{dist}(x, K) < \varepsilon \} \subseteq U \).

**Problem 0.25.** Let \( (X, d) \) be a locally compact metric space. Then \( X \) is \( \sigma \)-compact if and only if there exists a proper function \( f : X \to [0, \infty) \). (Hint: Let \( U_n \) be as in item (iv) of Lemma 0.25 and use Uryson’s lemma to find functions \( f_n : X \to [0, 1] \) such that \( f(x) = 0 \) for \( x \in \overline{U_n} \) and \( f(x) = 1 \) for \( x \in X \setminus U_{n+1} \). Now consider \( f = \sum_{n=1}^{\infty} f_n \)).

### 0.7. Connectedness

Roughly speaking a topological space \( X \) is disconnected if it can be split into two (nonempty) separated sets. This of course raises the question what should be meant by separated. Evidently it should be more than just disjoint since otherwise we could split any space containing more than one point. Hence we will consider two sets separated if each is disjoint from the closure of the other. Note that if we can split \( X \) into two separated sets \( X = U \cup V \) then \( U \cap V = \emptyset \) implies \( U = U \) (and similarly \( V = V \)). Hence both sets must be closed and thus also open (being complements of each other). This brings us to the following definition:

A topological space \( X \) is called **disconnected** if one of the following equivalent conditions holds

- \( X \) is the union of two nonempty separated sets.
- \( X \) is the union of two nonempty disjoint open sets.
- \( X \) is the union of two nonempty disjoint closed sets.

In this case the sets from the splitting are both open and closed. A topological space \( X \) is called **connected** if it cannot be split as above. That is, in a connected space \( X \) the only sets which are both open and closed are \( \emptyset \) and \( X \). This last observation is frequently used in proofs: If the set where a property holds is both open and closed it must either hold nowhere or everywhere. In particular, any mapping from a connected to a discrete space must be constant since the inverse image of a point is both open and closed.

A subset of \( X \) is called (dis-)connected if it is (dis-)connected with respect to the relative topology. In other words, a subset \( A \subseteq X \) is disconnected if there are disjoint nonempty open sets \( U \) and \( V \) which split \( A \) according to \( A = (U \cap A) \cup (V \cap A) \).

**Example.** In \( \mathbb{R} \) the nonempty connected sets are precisely the intervals (Problem 0.26). Consequently \( A = [0, 1] \cup [2, 3] \) is disconnected with \( [0, 1] \)
and $[2, 3]$ being its components (to be defined precisely below). While you might be reluctant to consider the closed interval $[0, 1]$ as open, it is important to observe that it is the relative topology which is relevant here.

The maximal connected subsets (ordered by inclusion) of a nonempty topological space $X$ are called the **connected components** of $X$.

**Example.** Consider $\mathbb{Q} \subseteq \mathbb{R}$. Then every rational point is its own component (if a set of rational points contains more than one point there would be an irrational point in between which can be used to split the set).

In many applications one also needs the following stronger concept. A space $X$ is called **path-connected** if any two points $x, y \in X$ can be joined by a path, that is a continuous map $\gamma : [0, 1] \to X$ with $\gamma(0) = x$ and $\gamma(1) = y$. A space is called **locally path-connected** if every point has a path-connected neighborhood.

Every path-connected space is connected. In fact, if $X = U \cup V$ were disconnected but path-connected we could choose $x \in U$ and $y \in V$ plus a path $\gamma$ joining them. But this would give a splitting $[0, 1] = \gamma^{-1}(U) \cup \gamma^{-1}(V)$ contradicting our assumption. The converse however is not true in general as a space might be impassable (an example will follow).

**Example.** The spaces $\mathbb{R}$ and $\mathbb{R}^n$, $n > 1$, are not homeomorphic. In fact, removing any point form $\mathbb{R}$ gives a disconnected space while removing a point form $\mathbb{R}^n$ still leaves it (path-)connected.

We collect a few simple but useful properties below.

**Lemma 0.30.** Suppose $X$ and $Y$ are topological spaces.

(i) Suppose $f : X \to Y$ is continuous. Then if $X$ is (path-)connected so is the image $f(X)$.

(ii) Suppose $A_\alpha \subseteq X$ are (path-)connected and $\bigcap_\alpha A_\alpha \neq \emptyset$. Then $\bigcup_\alpha A_\alpha$ is (path-)connected.

(iii) $A \subseteq X$ is (path-)connected if and only if any two points $x, y \in A$ are contained in a (path-)connected set $B \subseteq A$.

(iv) Suppose $X_1, \ldots, X_n$ are (path-)connected then so is $\bigtimes_{j=1}^n X_j$.

(v) Suppose $A \subseteq X$ is connected, then $\overline{A}$ is connected.

(vi) A locally path-connected space is path-connected if and only if it is connected.

**Proof.** (i) Suppose we have a splitting $f(X) = U \cup V$ into nonempty disjoint sets which are open in the relative topology. Hence, there are open sets $U'$ and $V'$ such that $U = U' \cap f(X)$ and $V = V' \cap f(X)$ implying that the sets $f^{-1}(U) = f^{-1}(U')$ and $f^{-1}(V) = f^{-1}(V')$ are open. Thus we get a
corresponding splitting \( X = f^{-1}(U) \cup f^{-1}(V) \) into nonempty disjoint open sets contradicting connectedness of \( X \).

If \( X \) is path connected, let \( y_1 = f(x_1) \) and \( y_2 = f(x_2) \) be given. If \( \gamma \) is a path connecting \( x_1 \) and \( x_2 \), then \( f \circ \gamma \) is a path connecting \( y_1 \) and \( y_2 \).

(ii). Let \( A = \bigcup_{\alpha} A_\alpha \) and suppose there is a splitting \( A = (U \cap A) \cup (V \cap A) \). Since there is some \( x \in \bigcap_{\alpha} A_\alpha \) we can assume \( x \in U \) w.l.o.g. Hence there is a splitting \( A_\alpha = (U \cap A_\alpha) \cup (V \cap A_\alpha) \) and since \( A_\alpha \) is connected and \( U \cap A_\alpha \) is nonempty we must have \( V \cap A_\alpha = \emptyset \). Hence \( V \cap A = \emptyset \) and \( A \) is connected.

If the \( x \in A_\alpha \) and \( y \in A_\beta \) then choose a point \( z \in A_\alpha \cap A_\beta \) and paths \( \gamma_\alpha \) from \( x \) to \( z \) and \( \gamma_\beta \) from \( z \) to \( y \), then \( \gamma_\alpha \circ \gamma_\beta \) is a path from \( x \) to \( y \), where \( \gamma_\alpha \circ \gamma_\beta(t) = \gamma_\alpha(2t) \) for \( 0 \leq t \leq \frac{1}{2} \) and \( \gamma_\alpha \circ \gamma_\beta(t) = \gamma_\beta(2t - 1) \) for \( \frac{1}{2} \leq t \leq 1 \) (cf. Problem 0.16).

(iii). If \( X \) is connected we can choose \( B = A \). Conversely, fix some \( x \in A \) and let \( B_x \) be the corresponding set for the pair \( x,y \). Then \( A = \bigcup_{y \in A} B_y \) is (path-)connected by the previous item.

(iv). We first consider two spaces \( X = X_1 \times X_2 \). Let \( x,y \in X \). Then \( \{x_1\} \times X_2 \) is homeomorphic to \( X_2 \) and hence (path-)connected. Similarly \( X_1 \times \{y_2\} \) is (path-)connected as well as \( \{x_1\} \times X_2 \cup X_1 \times \{y_2\} \) by (ii) since both sets contain \( \{x_1,y_2\} \in X \). But this last set contains both \( x,y \) and hence the claim follows from (iii). The general case follows by iterating this result.

(v). Let \( x \in \overline{A} \). Then \( \{x\} \) and \( A \) cannot be separated and hence \( \{x\} \cup A \) is connected. The rest follows from (ii).

(vi). Consider the set \( U(x) \) of all points connected to a fixed point \( x \) (via paths). If \( y \in U(x) \) then so is any path-connected neighborhood of \( y \) by gluing paths (as in item (ii)). Hence \( U(x) \) is open. Similarly, if \( y \in \overline{U(x)} \) then any path-connected neighborhood of \( y \) will intersect \( U(y) \) and hence \( y \in U(x) \). Thus \( U(x) \) is also closed and hence must be all of \( X \) by connectedness. The converse is trivial. \( \square \)

A few simple consequences are also worth while noting: If two different components contain a common point, their union is again connected contradicting maximality. Hence two different components are always disjoint. Moreover, every point is contained in a component, namely the union of all connected sets containing this point. In other words, the components of any topological space \( X \) form a partition of \( X \) (i.e., they are disjoint, nonempty, and their union is \( X \)). Moreover, every component is a closed subset of the original space. In the case where their number is finite we can take complements and each component is also an open subset (the rational numbers from our first example show that components are not open in general).
Example. Consider the graph of the function $f : (0,1] \to \mathbb{R}$, $x \mapsto \sin(\frac{1}{x})$. Then $\Gamma(f) \subseteq \mathbb{R}^2$ is path-connected and its closure $\bar{\Gamma}(f) = \Gamma(f) \cup \{0\} \times [-1,1]$ is connected. However, $\Gamma(f)$ is not path-connected as there is no path from $(1,0)$ to $(0,0)$. Indeed, suppose $\gamma$ were such a path. Then, since $\gamma_1$ covers $[0,1]$ by the intermediate value theorem (see below), there is a sequence $t_n \to 1$ such that $\gamma_1(t_n) = \frac{2}{(2n+1)\pi}$. But then $\gamma_2(t_n) = (-1)^n \not\to 0$ contradicting continuity.

Theorem 0.31 (Intermediate Value Theorem). Let $X$ be a connected topological space and $f : X \to \mathbb{R}$ be continuous. For any $x,y \in X$ the function $f$ attains every value between $f(x)$ and $f(y)$.

Proof. The image $f(X)$ is connected and hence an interval.$\square$

Problem 0.26. A nonempty subset of $\mathbb{R}$ is connected if and only if it is an interval.

0.8. Continuous functions on metric spaces

Let $X,Y$ be topological spaces and let $C(X,Y)$ be the set of all continuous functions $f : X \to Y$. In particular, we will set $C(X) := C(X,\mathbb{C})$. Moreover, if $Y$ is a metric space then $C_b(X,Y)$ will denote the set of all bounded continuous functions, that is, those continuous functions for which $\sup_{x \in X} d_Y(f(x),y)$ is finite for some (and hence for all) $y \in Y$. Note that by the extreme value theorem $C_b(X,Y) = C(X,Y)$ if $X$ is compact. For these functions we can introduce a metric via

$$d(f,g) := \sup_{x \in X} d_Y(f(x),g(x)). \quad (0.25)$$

In fact, the requirements for a metric are readily checked. Of course convergence with respect to this metric implies pointwise convergence but not the other way round.

Example. Consider $X := [0,1]$, then $f_n(x) := \max(1-|nx-1|,0)$ converges pointwise to 0 (in fact, $f_n(0) = 0$ and $f_n(x) = 0$ on $[\frac{2}{n},1]$) but not with respect to the above metric since $f_n(\frac{1}{n}) = 1$.$\diamond$

This kind of convergence is known as uniform convergence since for every $\varepsilon > 0$ there is some index $N$ (independent of $x$) such that $d_Y(f_n(x),f(x)) < \varepsilon$ for $n \geq N$. In contradistinction, in the case of pointwise convergence, $N$ is allowed to depend on $x$. One advantage is that continuity of the limit function comes for free.

Theorem 0.32. Let $X$ be a topological space and $Y$ a metric space. Suppose $f_n \in C(X,Y)$ converges uniformly to some function $f : X \to Y$. Then $f$ is continuous.
Proof. Let \( x \in X \) be given and write \( y := f(x) \). We need to show that 
\( f^{-1}(B_{\varepsilon}(y)) \) is a neighborhood of \( x \) for every \( \varepsilon > 0 \). So fix \( \varepsilon \). Then we can find an \( N \) such that 
\[ d(f_n, f) < \frac{\varepsilon}{2} \] 
for \( n \geq N \) implying 
\[ f_N^{-1}(B_{\varepsilon/2}(y)) \subseteq f^{-1}(B_{\varepsilon}(y)) \] 
since 
\[ d(f_n(z), y) < \frac{\varepsilon}{2} \] 
implies 
\[ d(f(z), y) \leq d(f(z), f_n(z)) + d(f_n(z), y) \leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon \] 
for \( n \geq N \). \( \square \)

Corollary 0.33. Let \( X \) be a topological space and \( Y \) a complete metric space. The space \( C_b(X, Y) \) together with the metric \( d \) is complete.

Proof. Suppose \( f_n \) is a Cauchy sequence with respect to \( d \), then \( f_n(x) \) is a Cauchy sequence for fixed \( x \) and has a limit since \( Y \) is complete. Call this limit \( f(x) \). Then 
\[ d_Y(f(x), f_n(x)) = \lim_{m \to \infty} d_Y(f_m(x), f_n(x)) \leq \sup_{m \geq n} d(f_m, f_n) \] 
and since this last expression goes to 0 as \( n \to \infty \), we see that \( f_n \) converges uniformly to \( f \). Moreover, \( f \in C(X, Y) \) by the previous theorem so we are done. \( \square \)

Let \( Y \) be a vector space. By \( C_c(X, Y) \subseteq C_b(X, Y) \) we will denote the set of continuous functions with compact support. Its closure will be denoted by \( C_0(X, Y) := \overline{C_0(X, Y)} \subseteq C_b(X, Y) \). Of course if \( X \) is compact all these spaces agree \( C_c(X, Y) = C_0(X, Y) = C_b(X, Y) = C(X, Y) \). In the general case one assumes \( X \) to be locally compact since if we take a closed neighborhood \( V \) of \( f(x) \neq 0 \) which does not contain 0, then \( f^{-1}(U) \) will be a compact neighborhood of \( x \). Hence without this assumption \( f \) must vanish on every point which does not have a compact neighborhood and \( C_c(X, Y) \) will not be sufficiently rich.

Example. Let \( X \) be a separable and locally compact metric space and \( Y = \mathbb{C}^n \). Then
\[
C_0(X, \mathbb{C}^n) = \left\{ f \in C_b(X, \mathbb{C}^n) \mid \forall \varepsilon > 0, \exists K \subseteq X \text{ compact :} \begin{aligned}
\forall x \in X \setminus K, |f(x)| < \varepsilon,
\forall x \in X, |f(x)| < \varepsilon
\end{aligned} \right\}. \tag{0.26}
\]

Denote the set on the right-hand side by \( C \). Let \( K_m \) be an increasing sequence of compact sets with \( K_m \nearrow X \) (Lemma 0.25) and let \( \varphi_m \) be a corresponding sequence as in Urysohn’s lemma (Lemma 0.27). Then for \( f \in C \) the sequence \( f_m = \varphi_m f \in C_b(X, \mathbb{C}^n) \) will converge to \( f \). Conversely, if \( f_n \in C_c(X, \mathbb{C}^n) \) converges to \( f \in C_b(X, \mathbb{C}^n) \), then given \( \varepsilon > 0 \) choose \( K = \text{supp}(f_m) \) for some \( m \) with 
\[ d(f_m, f) < \varepsilon. \]

In the case where \( X \) is an open subset of \( \mathbb{R}^n \) this says that \( C_0(X, Y) \) are those which vanish at the boundary (including the case as \( |x| \to \infty \) if \( U \) is unbounded).

Lemma 0.34. If \( X \) is a separable and locally compact space then \( C_0(X, \mathbb{C}^n) \) is separable.
Proof. Choose a countable base $\mathcal{B}$ for $X$ and let $I$ the collection of all balls in $\mathbb{C}^n$ with rational radius and center. Given $O_1,\ldots,O_m \in \mathcal{B}$ and $I_1,\ldots,I_m \in I$ we say that $f \in C_c(X,\mathbb{C}^n)$ is adapted to these sets if $\text{supp}(f) \subseteq \bigcup_{j=1}^m O_j$ and $f(O_j) \subseteq I_j$. The set of all tuples $(O_j,I_j)_{1 \leq j \leq m}$ is countable and for each tuple we choose a corresponding adapted function (if there exists one at all). Then the set of these functions $\mathcal{F}$ is dense. It suffices to show that the closure of $\mathcal{F}$ contains $C_c(X,\mathbb{C}^n)$. So let $f \in C_c(X,\mathbb{C}^n)$ and let $\varepsilon > 0$ be given. Then for every $x \in X$ there is some neighborhood $O(x) \in \mathcal{B}$ such that $|f(x) - f(y)| < \varepsilon$ for $y \in O(x)$. Since $\text{supp}(f)$ is compact, it can be covered by $O(x_1),\ldots,O(x_m)$. In particular $f(O(x_j)) \subseteq B_{\varepsilon}(f(x_j))$ and we can find a ball $I_j$ of radius at most $2\varepsilon$ with $f(O(x_j)) \subseteq I_j$. Now let $g$ be the function from $\mathcal{F}$ which is adapted to $(O(x_j),I_j)_{1 \leq j \leq m}$ and observe that $|f(x) - g(x)| < 4\varepsilon$ since $x \in O(x_j)$ implies $f(x),g(x) \in I_j$. \hfill \Box

Let $X,Y$ be metric spaces. A function $f \in C(X,Y)$ is called uniformly continuous if for every $\varepsilon > 0$ there is a $\delta > 0$ such that
\[d_Y(f(x), f(y)) \leq \varepsilon \quad \text{whenever} \quad d_X(x, y) < \delta. \quad (0.27)\]

Note that with the usual definition of continuity on fixes $x$ and then chooses $\delta$ depending on $x$. Here $\delta$ has to be independent of $x$. If the domain is compact, this extra condition comes for free.

**Theorem 0.35.** Let $X$ be a compact metric space and $Y$ a metric space. Then every $f \in C(X,Y)$ is uniformly continuous.

**Proof.** Suppose the claim were wrong. Fix $\varepsilon > 0$. Then for every $\delta_n = \frac{1}{n}$ we can find $x_n, y_n$ with $d_X(x_n, y_n) < \delta_n$ but $d_Y(f(x_n), f(y_n)) \geq \varepsilon$. Since $X$ is compact we can assume that $x_n$ converges to some $x \in X$ (after passing to a subsequence if necessary). Then we also have $y_n \to x$ implying $d_Y(f(x_n), f(y_n)) \to 0$, a contradiction. \hfill \Box

Note that a uniformly continuous function maps Cauchy sequences to Cauchy sequences. This fact can be used to extend a uniformly continuous function to boundary points.

**Theorem 0.36.** Let $X$ be a metric space and $Y$ a complete metric space. A uniformly continuous function $f : A \subseteq X \to Y$ has a unique continuous extension $\bar{f} : \overline{A} \to Y$. This extension is again uniformly continuous.

**Proof.** If there is an extension at all, it must be given by $\bar{f}(x) = \lim_{n \to \infty} f(x_n)$, where $x_n \in A$ is some sequence converging to $x \in \overline{A}$. Indeed, since $x_n$ converges, $f(x_n)$ is Cauchy and hence has a limit since $Y$ is assumed complete. Moreover, uniqueness of limits shows that $\bar{f}(x)$ is independent of the sequence chosen. Also $\bar{f}(x) = f(x)$ for $x \in A$ by continuity. To see that $\bar{f}$ is uniformly continuous, let $\varepsilon > 0$ be given and choose a $\delta$ which works for $f$. 


Then for given $x, y$ with $d_X(x, y) < \frac{\delta}{3}$ we can find $\tilde{x}, \tilde{y} \in A$ with $d_X(\tilde{x}, x) < \frac{\delta}{3}$ and $d_Y(f(\tilde{x}), f(x)) \leq \varepsilon$ as well as $d_X(\tilde{y}, y) < \frac{\delta}{3}$ and $d_Y(f(\tilde{y}), f(y)) \leq \varepsilon$. Hence $d_Y(f(x), f(y)) \leq d_Y(f(x), f(\tilde{x})) + d_Y(f(\tilde{x}), f(\tilde{y})) + d_Y(f(\tilde{y}), f(y)) \leq 3\varepsilon$.

Next we want to identify relatively compact subsets in $C(X, Y)$. A family of functions $F \subset C(X, Y)$ is called (uniformly) equicontinuous if for every $\varepsilon > 0$ there is a $\delta > 0$ such that

$$d_Y(f(x), f(y)) \leq \varepsilon \quad \text{whenever} \quad d_X(x, y) < \delta, \quad \forall f \in F. \quad (0.28)$$

That is, in this case $\delta$ is required to not only be independent of $x \in X$ but also independent of the function $f \in F$.

**Theorem 0.37** (Arzelà–Ascoli). Let $X$ be a compact metric space and $Y$ a complete metric space. Let $F \subset C(X, Y)$ be a family of continuous functions. Then every sequence from $F$ has a uniformly convergent subsequence if and only if $F$ is equicontinuous and the set $\{f(x) | f \in F\}$ is bounded for every $x \in X$. In this case $F$ is even bounded.

**Proof.** First of all note that if $F$ is equicontinuous and the set $\{f(x) | f \in F\}$ is bounded for all $x$ then $F$ is bounded. To see this fix $\varepsilon$ and cover $X$ by balls $B_\delta(x_j)$. Then $|f(x)| \leq \max_j \sup_{f \in F} |f(x_j)| + \varepsilon$ for every $x \in X$ and every $f \in F$.

Now let $f_n$ be a sequence from $F$. Let $\{x_j\}_{j=1}^\infty$ be a dense subset of $X$ (cf. Corollary 0.21). We will first construct a subsequence which converges on this dense subset using a diagonal series argument: Since $f_n(x_1)$ is bounded, we can choose a subsequence $f_n^{(1)}(x)$ such that $f_n^{(1)}(x_1)$ converges (Bolzano–Weierstraß). Similarly we can extract a subsequence $f_n^{(2)}(x)$ from $f_n^{(1)}(x)$ which converges at $x_2$ (and hence also at $x_1$ since it is a subsequence of $f_n^{(1)}(x)$). By induction we get a sequence $f_n^{(j)}(x)$ converging at $x_1, \ldots, x_j$. The diagonal sequence $\tilde{f}_n(x) = f_n^{(m)}(x)$ will hence converge for all $x = x_j$ (why?). We will show that it converges uniformly for all $x$:

Fix $\varepsilon > 0$ and chose $\delta$ such that $d_Y(f_n(x), f_n(y)) \leq \frac{\varepsilon}{3}$ for $d_X(x, y) < \delta$. The balls $B_\delta(x_j)$ cover $X$ and by compactness even finitely many, say $1 \leq j \leq p$, suffice. Furthermore, choose $N_\varepsilon$ such that $d_Y(\tilde{f}_n(x_j), \tilde{f}_n(x_j)) \leq \frac{\varepsilon}{3}$ for $n, m \geq N_\varepsilon$ and $1 \leq j \leq p$.

Now pick $x$ and note that $x \in B_\delta(x_j)$ for some $j$. Thus

$$d_Y(\tilde{f}_m(x), \tilde{f}_n(x)) \leq d_Y(\tilde{f}_m(x_j), \tilde{f}_n(x_j)) + d_Y(\tilde{f}_m(x_j), \tilde{f}_n(x_j)) + d_Y(\tilde{f}_n(x_j), \tilde{f}_n(x)) \leq \varepsilon$$

for $n, m \geq N_\varepsilon$, which shows that $\tilde{f}_n$ is Cauchy with respect to the maximum norm. By completeness of $C(K)$ it has a limit.
To see the converse first note that if \( \{ f(x) | f \in F \} \) were unbounded for some \( x \), then there would be a sequence of functions \( f_n \) such that \( d_Y(f_n(x), f_0) \to \infty \). A contradiction. Similarly, if \( F \) were not equicontinuous, there must be an \( \varepsilon_0 > 0 \) such that for every \( n \in \mathbb{N} \) there is a function \( f_n \in F \) and points \( x_n, y_n \) with \( d_X(x_n, y_n) < \frac{1}{n} \) and \( d_Y(f_n(x_n), f_n(y_n)) \geq \varepsilon_0 \). By passing to a subsequence we can assume \( x_n \to x \) and hence also \( y_n \to x \). Moreover, passing to yet another subsequence we can assume that \( f_n \to f \) uniformly. But then \( 0 = d_Y(f(x), f(x)) = \lim_{n \to \infty} d_Y(f_n(x_n), f_n(y_n)) \geq \varepsilon_0 \), a contradiction. \( \square \)

**Example.** Consider the solution \( y_n(x) \) of the initial value problem

\[
y' = \sin(ny), \quad y(0) = 1.
\]

(Assuming this solution exists — it can in principle be found using separation of variables.) Then \( |y'_n(x)| \leq 1 \) hence the mean value theorem shows that the family \( \{ y_n \} \subseteq C([0, 1]) \) is equicontinuous. Hence there is a uniformly convergent subsequence. \( \diamond \)

In many situations a certain property can be seen for a class of *nice* functions and then extended to a more general class of functions by approximation. In this respect it is important to identify classes of functions which allow to approximate *all* functions. That is, in our present situation we are looking for functions which are dense in \( C(X, Y) \). For example, the classical Weierstraß approximation theorem (see Theorem 1.3 below for an elementary approach) says that the polynomials are dense in \( C([a, b]) \) for any compact interval. Here we will present a generalization of this result. For its formulation observe that \( C(X) \) is not only a vector space but also comes with a natural product, given by pointwise multiplication of functions, which turns it into an algebra over \( \mathbb{C} \). By a subalgebra we will mean a subspace which is closed under multiplication and by a \( * \)-subalgebra we will mean a subalgebra which is also closed under complex conjugation. The \( * \)-subalgebra generated by a set is of course the smallest \( * \)-subalgebra containing this set.

The proof will use the fact that the absolute value can be approximated by polynomials on \([-1, 1]\). This of course follows from the Weierstraß approximation theorem but can also be seen directly by defining the sequence of polynomials \( p_n \) via

\[
p_1(t) := 0, \quad p_{n+1}(t) := p_n(t) + \frac{t^2 - p_n(t)^2}{2}.
\]
Then this sequence of polynomials satisfies \( p_n(t) \leq p_{n+1}(t) \leq |t| \) and converges pointwise to \(|t|\) for \( t \in [-1, 1] \). Hence by Dini’s theorem (Problem 0.28) it converges uniformly. By scaling we get the corresponding result for arbitrary compact subsets of the real line.

**Theorem 0.38 (Stone–Weierstraß, real version).** Suppose \( K \) is a compact topological space and consider \( C(K, \mathbb{R}) \). If \( F \subset C(K, \mathbb{R}) \) contains the identity 1 and separates points (i.e., for every \( x_1 \neq x_2 \) there is some function \( f \in F \) such that \( f(x_1) \neq f(x_2) \)), then the subalgebra generated by \( F \) is dense.

**Proof.** Denote by \( A \) the subalgebra generated by \( F \). Note that if \( f \in A \), we have \( |f| \in \overline{A} \): Choose a polynomial \( p_n(t) \) such that \( |t| - p_n(t) < \frac{1}{n} \) for \( t \in f(K) \) and hence \( p_n(f) \to |f| \).

In particular, if \( f, g \) are in \( \overline{A} \), we also have
\[
\max\{f, g\} = \frac{(f + g) + |f - g|}{2}, \quad \min\{f, g\} = \frac{(f + g) - |f - g|}{2}
\]
in \( \overline{A} \).

Now fix \( f \in C(K, \mathbb{R}) \). We need to find some \( f^\varepsilon \in \overline{A} \) with \( \|f - f^\varepsilon\|_{\infty} < \varepsilon \).

First of all, since \( A \) separates points, observe that for given \( y, z \in K \) there is a function \( f_{y,z} \in A \) such that \( f_{y,z}(y) = f(y) \) and \( f_{y,z}(z) = f(z) \) (show this). Next, for every \( y \in K \) there is a neighborhood \( U(y) \) such that
\[
f_{y,z}(x) > f(x) - \varepsilon, \quad x \in U(y),
\]
and since \( K \) is compact, finitely many, say \( U(y_1), \ldots, U(y_j) \), cover \( K \). Then
\[
f_z = \max\{f_{y_1,z}, \ldots, f_{y_j,z}\} \in \overline{A}
\]
and satisfies \( f_z > f - \varepsilon \) by construction. Since \( f_z(z) = f(z) \) for every \( z \in K \), there is a neighborhood \( V(z) \) such that
\[
f_z(x) < f(x) + \varepsilon, \quad x \in V(z),
\]
and a corresponding finite cover \( V(z_1), \ldots, V(z_k) \). Now
\[
f^\varepsilon = \min\{f_{z_1}, \ldots, f_{z_k}\} \in \overline{A}
\]
satisfies \( f^\varepsilon < f + \varepsilon \). Since \( f - \varepsilon < f_{z_i} \) for all \( z_i \) we have \( f - \varepsilon < f^\varepsilon \) and we have found a required function. \( \square \)

**Theorem 0.39 (Stone–Weierstraß).** Suppose \( K \) is a compact topological space and consider \( C(K) \). If \( F \subset C(K) \) contains the identity 1 and separates points, then the \(*\)-subalgebra generated by \( F \) is dense.

**Proof.** Just observe that \( \tilde{F} = \{\text{Re}(f), \text{Im}(f) | f \in F\} \) satisfies the assumption of the real version. Hence every real-valued continuous function can be
approximated by elements from the subalgebra generated by $\tilde{F}$; in particular, this holds for the real and imaginary parts for every given complex-valued function. Finally, note that the subalgebra spanned by $\tilde{F}$ contains the $*$-subalgebra spanned by $F$. \hfill \square

Note that the additional requirement of being closed under complex conjugation is crucial: The functions holomorphic on the unit disc and continuous on the boundary separate points, but they are not dense (since the uniform limit of holomorphic functions is again holomorphic).

**Corollary 0.40.** Suppose $K$ is a compact topological space and consider $C(K)$. If $F \subseteq C(K)$ separates points, then the closure of the $*$-subalgebra generated by $F$ is either $C(K)$ or $\{f \in C(K) | f(t_0) = 0\}$ for some $t_0 \in K$.

**Proof.** There are two possibilities: either all $f \in F$ vanish at one point $t_0 \in K$ (there can be at most one such point since $F$ separates points) or there is no such point.

If there is no such point, then the identity can be approximated by elements in $\tilde{A}$: First of all note that $|f| \in \tilde{A}$ if $f \in \mathcal{A}$, since the polynomials $p_n(t)$ used to prove this fact can be replaced by $p_n(t) - p_n(0)$ which contain no constant term. Hence for every point $y$ we can find a nonnegative function in $\mathcal{A}$ which is positive at $y$ and by compactness we can find a finite sum of such functions which is positive everywhere, say $m \leq f(t) \leq M$. Now approximate $\min(m^{-1}t, t^{-1})$ by polynomials $q_n(t)$ (again a constant term is not needed) to conclude that $q_n(f) \to f^{-1} \in \mathcal{A}$. Hence $1 = f \cdot f^{-1} \in \mathcal{A}$ as claimed and so $\mathcal{A} = C(K)$ by the Stone–Weierstraß theorem.

If there is such a $t_0$ we have $\mathcal{A} \subseteq \{f \in C(K) | f(t_0) = 0\}$ and the identity is clearly missing from $\mathcal{A}$. However, adding the identity to $\mathcal{A}$ we get $\mathcal{A} + \mathbb{C} = C(K)$ by the Stone–Weierstraß theorem. Moreover, if $f \in C(K)$ with $f(t_0) = 0$ we get $f = \tilde{f} + \alpha$ with $\tilde{f} \in \mathcal{A}$ and $\alpha \in \mathbb{C}$. But $0 = f(t_0) = \tilde{f}(t_0) + \alpha$, hence $\alpha$ is a constant term. That is, $\mathcal{A} = \{f \in C(K) | f(t_0) = 0\}$.

**Problem 0.27.** Suppose $X$ is compact and connected and let $F \subseteq C(X,Y)$ be a family of equicontinuous functions. Then $\{f(x) | f \in F\}$ bounded for one $x$ implies $F$ bounded.

**Problem 0.28** (Dini’s theorem). Suppose $X$ is compact and let $f_n \in C(X)$ be a sequence of decreasing (or increasing) functions converging pointwise $f_n(x) \searrow f(x)$ to some function $f \in C(X)$. Then $f_n \to f$ uniformly. (Hint: Reduce it to the case $f_n \searrow 0$ and apply the finite intersection property to $f_n^{-1}([\varepsilon, \infty))$.)

**Problem 0.29.** Let $k \in \mathbb{N}$ and $I \subseteq \mathbb{R}$. Show that the $*$-subalgebra generated by $f_{z_0}(t) = \frac{1}{(t - z_0)^k}$ for one $z_0 \in \mathbb{C}$ and $k \in \mathbb{N}$ is dense in the set $C_0(I)$ of continuous functions vanishing at infinity:
0.8. Continuous functions on metric spaces

- for \( I = \mathbb{R} \) if \( z_0 \in \mathbb{C} \setminus \mathbb{R} \) and \( k = 1 \) or \( k = 2 \),
- for \( I = [a, \infty) \) if \( z_0 \in (\infty, a) \) and \( k \) arbitrary,
- for \( I = (-\infty, a] \cup [b, \infty) \) if \( z_0 \in (a, b) \) and \( k \) odd.

(Hint: Add \( \infty \) to \( \mathbb{R} \) to make it compact.)

Problem 0.30. Let \( U \subseteq \mathbb{C} \setminus \mathbb{R} \) be a set which has a limit point and is symmetric under complex conjugation. Show that the span of \( \{(t-z)^{-1} | z \in U\} \) is dense in the set \( C_0(\mathbb{R}) \) of continuous functions vanishing at infinity. (Hint: The product of two such functions is in the span provided they are different.)

Problem 0.31. Let \( K \subseteq \mathbb{C} \) be a compact set. Show that the set of all functions \( f(z) = p(x, y) \), where \( p : \mathbb{R}^2 \rightarrow \mathbb{C} \) is polynomial and \( z = x + iy \), is dense in \( C(K) \).
Chapter 1

A first look at Banach and Hilbert spaces

Functional analysis is an important tool in the investigation of all kind of problems in pure mathematics, physics, biology, economics, etc.. In fact, it is hard to find a branch in science where functional analysis is not used.

The main objects are (infinite dimensional) vector spaces with different concepts of convergence. The classical theory focuses on linear operators (i.e., functions) between these spaces but nonlinear operators are of course equally important. However, since one of the most important tools in investigating nonlinear mappings is linearization (differentiation), linear functional analysis will be our first topic in any case.

1.1. Introduction: Linear partial differential equations

Rather than listing an overwhelming number of classical examples I want to focus on one: linear partial differential equations. We will use this example as a guide throughout our first chapter and will develop all necessary tools for a successful treatment of our particular problem.

In his investigation of heat conduction Fourier was lead to the (one dimensional) heat or diffusion equation

$$\frac{\partial}{\partial t} u(t, x) = \frac{\partial^2}{\partial x^2} u(t, x).$$

Here $u(t, x)$ is the temperature distribution in a thin rod at time $t$ at the point $x$. It is usually assumed, that the temperature at $x = 0$ and $x = 1$ is fixed, say $u(t, 0) = a$ and $u(t, 1) = b$. By considering $u(t, x) \rightarrow u(t, x) - a - (b - a)x$ it is clearly no restriction to assume $a = b = 0$. Moreover, the
initial temperature distribution $u(0,x) = u_0(x)$ is assumed to be known as well.

Since finding the solution seems at first sight unfeasable, we could try to find at least some solutions of (1.1). For example, we could make an ansatz for $u(t,x)$ as a product of two functions, each of which depends on only one variable, that is,

$$u(t,x) = w(t)y(x).$$

Plugging this ansatz into the heat equation we arrive at

$$\dot{w}(t)y(x) = y''(x)w(t),$$

where the dot refers to differentiation with respect to $t$ and the prime to differentiation with respect to $x$. Bringing all $t$, $x$ dependent terms to the left, right side, respectively, we obtain

$$\frac{\dot{w}(t)}{w(t)} = \frac{y''(x)}{y(x)}.$$ 

Accordingly, this ansatz is called separation of variables.

Now if this equation should hold for all $t$ and $x$, the quotients must be equal to a constant $-\lambda$ (we choose $-\lambda$ instead of $\lambda$ for convenience later on). That is, we are led to the equations

$$-\dot{w}(t) = \lambda w(t)$$

and

$$-y''(x) = \lambda y(x), \quad y(0) = y(1) = 0,$$

which can easily be solved. The first one gives

$$w(t) = c_1 e^{-\lambda t}$$

and the second one

$$y(x) = c_2 \cos(\sqrt{\lambda} x) + c_3 \sin(\sqrt{\lambda} x).$$

However, $y(x)$ must also satisfy the boundary conditions $y(0) = y(1) = 0$. The first one $y(0) = 0$ is satisfied if $c_2 = 0$ and the second one yields ($c_3$ can be absorbed by $w(t)$)

$$\sin(\sqrt{\lambda}) = 0,$$

which holds if $\lambda = (\pi n)^2$, $n \in \mathbb{N}$ (in the case $\lambda < 0$ we get $\sinh(\sqrt{-\lambda}) = 0$, which cannot be satisfied and explains our choice of sign above). In summary, we obtain the solutions

$$u_n(t,x) := c_n e^{-(\pi n)^2 t} \sin(n \pi x), \quad n \in \mathbb{N}.$$ 

So we have found a large number of solutions, but we still have not dealt with our initial condition $u(0,x) = u_0(x)$. This can be done using the superposition principle which holds since our equation is linear: Any finite linear combination of the above solutions will again be a solution. Moreover,
under suitable conditions on the coefficients we can even consider infinite linear combinations. In fact, choosing
\[ u(t, x) = \sum_{n=1}^{\infty} c_n e^{-\frac{(\pi n)^2 t}{2}} \sin(n\pi x), \]  
(1.11)
where the coefficients \( c_n \) decay sufficiently fast, we obtain further solutions of our equation. Moreover, these solutions satisfy
\[ u(0, x) = \sum_{n=1}^{\infty} c_n \sin(n\pi x) \]  
(1.12)
and expanding the initial conditions into a Fourier series
\[ u_0(x) = \sum_{n=1}^{\infty} u_{0,n} \sin(n\pi x), \]  
(1.13)
we see that the solution of our original problem is given by (1.11) if we choose \( c_n = u_{0,n} \).

Of course for this last statement to hold we need to ensure that the series in (1.11) converges and that we can interchange summation and differentiation. You are asked to do so in Problem 1.1.

In fact, many equations in physics can be solved in a similar way:

- **Reaction-Diffusion equation:**
  \[
  \frac{\partial}{\partial t} u(t, x) - \frac{\partial^2}{\partial x^2} u(t, x) + q(x)u(t, x) = 0,
  \]
  \[
  u(0, x) = u_0(x),
  \]
  \[
  u(t, 0) = u(t, 1) = 0.
  \]  
(1.14)
Here \( u(t, x) \) could be the density of some gas in a pipe and \( q(x) > 0 \) describes that a certain amount per time is removed (e.g., by a chemical reaction).

- **Wave equation:**
  \[
  \frac{\partial^2}{\partial t^2} u(t, x) - \frac{\partial^2}{\partial x^2} u(t, x) = 0,
  \]
  \[
  u(0, x) = u_0(x), \quad \frac{\partial u}{\partial t}(0, x) = v_0(x)
  \]
  \[
  u(t, 0) = u(t, 1) = 0.
  \]  
(1.15)
Here \( u(t, x) \) is the displacement of a vibrating string which is fixed at \( x = 0 \) and \( x = 1 \). Since the equation is of second order in time, both the initial displacement \( u_0(x) \) and the initial velocity \( v_0(x) \) of the string need to be known.
• Schrödinger equation:

\[ i \frac{\partial}{\partial t} u(t, x) = -\frac{\partial^2}{\partial x^2} u(t, x) + q(x) u(t, x), \]
\[ u(0, x) = u_0(x), \]
\[ u(t, 0) = u(t, 1) = 0. \]  \hspace{1cm} (1.16)

Here \(|u(t, x)|^2\) is the probability distribution of a particle trapped in a box \(x \in [0, 1]\) and \(q(x)\) is a given external potential which describes the forces acting on the particle.

All these problems (and many others) lead to the investigation of the following problem

\[ L y(x) = \lambda y(x), \quad L := -\frac{d^2}{dx^2} + q(x), \]  \hspace{1cm} (1.17)

subject to the boundary conditions

\[ y(a) = y(b) = 0. \]  \hspace{1cm} (1.18)

Such a problem is called a Sturm–Liouville boundary value problem. Our example shows that we should prove the following facts about our Sturm–Liouville problems:

(i) The Sturm–Liouville problem has a countable number of eigenvalues \(E_n\) with corresponding eigenfunctions \(u_n(x)\), that is, \(u_n(x)\) satisfies the boundary conditions and \(L u_n(x) = E_n u_n(x)\).

(ii) The eigenfunctions \(u_n\) are complete, that is, any nice function \(u(x)\) can be expanded into a generalized Fourier series

\[ u(x) = \sum_{n=1}^{\infty} c_n u_n(x). \]

This problem is very similar to the eigenvalue problem of a matrix and we are looking for a generalization of the well-known fact that every symmetric matrix has an orthonormal basis of eigenvectors. However, our linear operator \(L\) is now acting on some space of functions which is not finite dimensional and it is not at all clear what (e.g.) orthogonal should mean in this context. Moreover, since we need to handle infinite series, we need convergence and hence we need to define the distance of two functions as well.

Hence our program looks as follows:

• What is the distance of two functions? This automatically leads us to the problem of convergence and completeness.
• If we additionally require the concept of orthogonality, we are lead to Hilbert spaces which are the proper setting for our eigenvalue problem.

• Finally, the spectral theorem for compact symmetric operators will be the solution of our above problem.

**Problem 1.1.** Suppose \( \sum_{n=1}^{\infty} |c_n| < \infty \). Show that (1.11) is continuous for \((t, x) \in [0, \infty) \times [0, 1]\) and solves the heat equation for \((t, x) \in (0, \infty) \times [0, 1]\). (Hint: Weierstraß M-test. When can you interchange the order of summation and differentiation?)

### 1.2. The Banach space of continuous functions

Our point of departure will be the set of continuous functions \( C(I) \) on a compact interval \( I := [a, b] \subset \mathbb{R} \). Since we want to handle both real and complex models, we will formulate most results for the more general complex case only. In fact, most of the time there will be no difference but we will add a remark in the rare case where the real and complex case do indeed differ.

One way of declaring a distance, well-known from calculus, is the maximum norm:

\[
\|f\|_{\infty} := \max_{x \in I} |f(x)|. \tag{1.19}
\]

It is not hard to see that with this definition \( C(I) \) becomes a normed vector space:

A normed vector space \( X \) is a vector space \( X \) over \( \mathbb{C} \) (or \( \mathbb{R} \)) with a nonnegative function (the norm) \( \|\cdot\| \) such that

- \( \|f\| > 0 \) for \( f \neq 0 \) (positive definiteness),
- \( \|\alpha f\| = |\alpha| \|f\| \) for all \( \alpha \in \mathbb{C}, f \in X \) (positive homogeneity), and
- \( \|f + g\| \leq \|f\| + \|g\| \) for all \( f, g \in X \) (triangle inequality).

If positive definiteness is dropped from the requirements, one calls \( \|\cdot\| \) a seminorm.

From the triangle inequality we also get the inverse triangle inequality (Problem 1.2)

\[
\|\|f\| - \|g\|\| \leq \|f - g\|, \tag{1.20}
\]

which shows that the norm is continuous.

Let me also briefly mention that norms are closely related to convexity. To this end recall that a subset \( C \subseteq X \) is called convex if for every \( x, y \in C \) we also have \( \lambda x + (1 - \lambda)y \in C \) whenever \( \lambda \in (0, 1) \). Moreover, a mapping \( f : C \to \mathbb{R} \) is called convex if \( f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y) \)
whenever $\lambda \in (0, 1)$ and in our case the triangle inequality plus homogeneity imply that every norm is convex:

$$\|\lambda x + (1 - \lambda)y\| \leq \lambda \|x\| + (1 - \lambda)\|y\|, \quad \lambda \in [0, 1]. \tag{1.21}$$

Moreover, choosing $\lambda = \frac{1}{2}$ we get back the triangle inequality upon using homogeneity. In particular, the triangle inequality could be replaced by convexity in the definition.

Once we have a norm, we have a distance $d(f, g) := \|f - g\|$ and hence we know when a sequence of vectors $f_n$ converges to a vector $f$ (namely if $d(f_n, f) \to 0$). We will write $f_n \to f$ or $\lim_{n \to \infty} f_n = f$, as usual, in this case. Moreover, a mapping $F : X \to Y$ between two normed spaces is called continuous if $f_n \to f$ implies $F(f_n) \to F(f)$. In fact, the norm, vector addition, and multiplication by scalars are continuous (Problem 1.3).

In addition to the concept of convergence, we have also the concept of a Cauchy sequence and hence the concept of completeness: A normed space is called complete if every Cauchy sequence has a limit. A complete normed space is called a Banach space.

**Example.** By completeness of the real numbers, $\mathbb{R}$ as well as $\mathbb{C}$ with the absolute value as norm are Banach spaces.

**Example.** The space $\ell^1(\mathbb{N})$ of all complex-valued sequences $a = (a_j)_{j=1}^\infty$ for which the norm

$$\|a\|_1 := \sum_{j=1}^\infty |a_j| \tag{1.22}$$

is finite is a Banach space.

To show this, we need to verify three things: (i) $\ell^1(\mathbb{N})$ is a vector space that is closed under addition and scalar multiplication, (ii) $\|\cdot\|_1$ satisfies the three requirements for a norm, and (iii) $\ell^1(\mathbb{N})$ is complete.

First of all, observe

$$\sum_{j=1}^k |a_j + b_j| \leq \sum_{j=1}^k |a_j| + \sum_{j=1}^k |b_j| \leq \|a\|_1 + \|b\|_1 \tag{1.23}$$

for every finite $k$. Letting $k \to \infty$, we conclude that $\ell^1(\mathbb{N})$ is closed under addition and that the triangle inequality holds. That $\ell^1(\mathbb{N})$ is closed under scalar multiplication together with homogeneity as well as definiteness are straightforward. It remains to show that $\ell^1(\mathbb{N})$ is complete. Let $a^n = (a^n_j)_{j=1}^\infty$ be a Cauchy sequence; that is, for given $\varepsilon > 0$ we can find an $N_\varepsilon$ such that $\|a^m - a^n\|_1 \leq \varepsilon$ for $m, n \geq N_\varepsilon$. This implies, in particular, $|a^m_j - a^n_j| \leq \varepsilon$ for every fixed $j$. Thus $a^n_j$ is a Cauchy sequence for fixed $j$ and, by completeness
of \( C \), it has a limit: \( \lim_{n \to \infty} a^n_j = a_j \). Now consider
\[
\sum_{j=1}^{k} |a^n_j - a^n_j| \leq \varepsilon
\]
and take \( m \to \infty \):
\[
\sum_{j=1}^{k} |a_j - a^n_j| \leq \varepsilon.
\]
Since this holds for all finite \( k \), we even have \( \|a - a^n\|_1 \leq \varepsilon \). Hence \( (a - a^n) \in \ell^1(\mathbb{N}) \) and since \( a^n \in \ell^1(\mathbb{N}) \), we finally conclude \( a = a^n + (a - a^n) \in \ell^1(\mathbb{N}) \).

**Example.** The previous example can be generalized by considering the space \( \ell^p(\mathbb{N}) \) of all complex-valued sequences \( a = (a_j)_{j=1}^{\infty} \) for which the norm \( \|a\|_p := \left( \sum_{j=1}^{\infty} |a_j|^p \right)^{1/p}, \quad p \in [1, \infty) \), is finite. By \( |a_j + b_j|^p \leq 2^p \max(|a_j|, |b_j|)^p = 2^p \max(|a_j|^p, |b_j|^p) \leq 2^p (|a_j|^p + |b_j|^p) \) it is a vector space, but the triangle inequality is only easy to see in the case \( p = 1 \). (It is also not hard to see that it fails for \( p < 1 \), which explains our requirement \( p \geq 1 \). See also Problem 1.14.)

To prove it we need the elementary inequality (Problem 1.7)
\[
\alpha^{1/p} \beta^{1/q} \leq \frac{1}{p} \alpha + \frac{1}{q} \beta, \quad \frac{1}{p} + \frac{1}{q} = 1, \quad \alpha, \beta \geq 0,
\]
which implies **Hölder’s inequality**
\[
\|ab\|_1 \leq \|a\|_p \|b\|_q
\]
for \( a \in \ell^p(\mathbb{N}), b \in \ell^q(\mathbb{N}) \). In fact, by homogeneity of the norm it suffices to prove the case \( \|a\|_p = \|b\|_q = 1 \). But this case follows by choosing \( \alpha = |a_j|^p \) and \( \beta = |b_j|^q \) in (1.27) and summing over all \( j \). (A different proof based on convexity will be given in Section 10.2.)

Now using \( |a_j + b_j|^p \leq |a_j| |a_j + b_j|^{p-1} + |b_j| |a_j + b_j|^{p-1} \), we obtain from Hölder’s inequality (note \( (p - 1)q = p \))
\[
\|a + b\|_p \leq \|a\|_p \|(a + b)^{p-1}\|_q + \|b\|_p \|(a + b)^{p-1}\|_q
= (\|a\|_p + \|b\|_p)\|a + b\|_p^{p-1}.
\]
Hence \( \ell^p \) is a normed space. That it is complete can be shown as in the case \( p = 1 \) (Problem 1.8).

The unit ball with respect to these norms in \( \mathbb{R}^2 \) is depicted in Figure 1. One sees that for \( p < 1 \) the unit ball is not convex (explaining once more our restriction \( p \geq 1 \)). Moreover, for \( 1 < p < \infty \) it is even strictly convex (that
is, the line segment joining two distinct points is always in the interior). This is related to the question of equality in the triangle inequality and will be discussed in Problems 1.11 and 1.12.

**Example.** The space $\ell^\infty(\mathbb{N})$ of all complex-valued bounded sequences $a = (a_j)_{j=1}^\infty$ together with the norm

$$
\|a\|_\infty := \sup_{j \in \mathbb{N}} |a_j|
$$

is a Banach space (Problem 1.9). Note that with this definition, Hölder’s inequality (1.28) remains true for the cases $p = 1$, $q = \infty$ and $p = \infty$, $q = 1$. The reason for the notation is explained in Problem 1.13.

**Example.** Every closed subspace of a Banach space is again a Banach space. For example, the space $c_0(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ of all sequences converging to zero is a closed subspace. In fact, if $a \in \ell^\infty(\mathbb{N}) \setminus c_0(\mathbb{N})$, then $\limsup_{j \to \infty} |a_j| = \epsilon > 0$ and thus $\|a - b\|_\infty \geq \epsilon$ for every $b \in c_0(\mathbb{N})$.

Now what about completeness of $C(I)$? A sequence of functions $f_n(x)$ converges to $f$ if and only if

$$
\lim_{n \to \infty} \|f - f_n\|_\infty = \lim_{n \to \infty} \sup_{x \in I} |f(x) - f_n(x)| = 0.
$$

That is, in the language of real analysis, $f_n$ converges uniformly to $f$. Now let us look at the case where $f_n$ is only a Cauchy sequence. Then $f_n(x)$ is clearly a Cauchy sequence of complex numbers for every fixed $x \in I$. In
1.2. The Banach space of continuous functions

particular, by completeness of $\mathbb{C}$, there is a limit $f(x)$ for each $x$. Thus we get a limiting function $f(x)$. Moreover, letting $m \to \infty$ in

$$|f_m(x) - f_n(x)| \leq \varepsilon \quad \forall m, n > N_\varepsilon, \, x \in I,$$

we see

$$|f(x) - f_n(x)| \leq \varepsilon \quad \forall n > N_\varepsilon, \, x \in I;$$

that is, $f_n(x)$ converges uniformly to $f(x)$. However, up to this point we do not know whether it is in our vector space $C(I)$, that is, whether it is continuous. Fortunately, there is a well-known result from real analysis which tells us that the uniform limit of continuous functions is again continuous: Fix $x \in I$ and $\varepsilon > 0$. To show that $f(x) \in C(I)$ and thus every Cauchy sequence in $C(I)$ converges. Or, in other words,

**Theorem 1.1.** $C(I)$ with the maximum norm is a Banach space.

For finite dimensional vector spaces the concept of a basis plays a crucial role. In the case of infinite dimensional vector spaces one could define a basis as a maximal set of linearly independent vectors (known as a Hamel basis, Problem 1.6). Such a basis has the advantage that it only requires finite linear combinations. However, the price one has to pay is that such a basis will be way too large (typically uncountable, cf. Problems 1.5 and 4.1). Since we have the notion of convergence, we can handle countable linear combinations and try to look for countable bases. We start with a few definitions.

The set of all finite linear combinations of a set of vectors $\{u_n\}_{n \in \mathbb{N}} \subset X$ is called the span of $\{u_n\}_{n \in \mathbb{N}}$ and denoted by

$$\text{span}\{u_n\}_{n \in \mathbb{N}} := \left\{ \sum_{j=1}^{m} \alpha_j u_{n_j} \mid n_j \in \mathcal{N}, \alpha_j \in \mathbb{C}, m \in \mathbb{N} \right\}.$$  \hfill (1.33)

A set of vectors $\{u_n\}_{n \in \mathcal{N}} \subset X$ is called linearly independent if every finite subset is. If $\{u_n\}_{n=1}^{N} \subset X, \, N \in \mathbb{N} \cup \{\infty\}$, is countable, we can throw away all elements which can be expressed as linear combinations of the previous ones to obtain a subset of linearly independent vectors which have the same span.

We will call a countable set of vectors $\{u_n\}_{n=1}^{N} \subset X$ a Schauder basis if every element $f \in X$ can be uniquely written as a countable linear
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combination of the basis elements:
\[ f = \sum_{n=1}^{N} \alpha_n u_n, \quad \alpha_n = \alpha_n(f) \in \mathbb{C}, \quad (1.34) \]
where the sum has to be understood as a limit if \( N = \infty \) (the sum is not required to converge unconditionally and hence the order of the basis elements is important). Since we have assumed the coefficients \( \alpha_n(f) \) to be uniquely determined, the vectors are necessarily linearly independent. Moreover, one can show that the coordinate functionals \( f \mapsto \alpha_n(f) \) are continuous (cf. Problem 4.5). A Schauder basis and its corresponding coordinate functionals \( u^*_n: X \to \mathbb{C}, f \mapsto \alpha_n(f) \) form a so-called biorthogonal system:
\[ u^*_m(u_n) = \delta_{m,n}, \quad (1.35) \]
is the Kronecker delta.

Example. The set of vectors \( \delta^n = (\delta^m_m)_{m \in \mathbb{N}} \) is a Schauder basis for the Banach space \( \ell^p(\mathbb{N}) \), \( 1 \leq p < \infty \).

Let \( a = (a_j)_{j=1}^{\infty} \in \ell^p(\mathbb{N}) \) be given and set \( a^m := \sum_{n=1}^{m} a_n \delta^n \). Then
\[ \|a - a^m\|_p = \left( \sum_{j=m+1}^{\infty} |a_j|^p \right)^{1/p} \to 0 \]
since \( a_j = a_j \) for \( 1 \leq j \leq m \) and \( a_j^m = 0 \) for \( j > m \). Hence
\[ a = \sum_{n=1}^{\infty} a_n \delta^n \]
and \( (\delta^n)_{n=1}^{\infty} \) is a Schauder basis (uniqueness of the coefficients is left as an exercise).

Note that \( (\delta^n)_{n=1}^{\infty} \) is also Schauder basis for \( c_0(\mathbb{N}) \) but not for \( \ell^\infty(\mathbb{N}) \) (try to approximate a constant sequence).

A set whose span is dense is called total, and if we have a countable total set, we also have a countable dense set (consider only linear combinations with rational coefficients — show this). A normed vector space containing a countable dense set is called separable.

Warning: Some authors use the term total in a slightly different way — see the warning on page 146.

Example. Every Schauder basis is total and thus every Banach space with a Schauder basis is separable (the converse puzzled mathematicians for quite some time and was eventually shown to be false by Per Enflo). In particular, the Banach space \( \ell^p(\mathbb{N}) \) is separable for \( 1 \leq p < \infty \).
While we will not give a Schauder basis for \( C(I) \) (Problem 1.15), we will at least show that it is separable. We will do this by showing that every continuous function can be approximated by polynomials, a result which is of independent interest. But first we need a lemma.

**Lemma 1.2 (Smoothing).** Let \( u_n(x) \) be a sequence of nonnegative continuous functions on \([-1, 1]\) such that

\[
\int_{|x| \leq 1} u_n(x) \, dx = 1 \quad \text{and} \quad \int_{\delta \leq |x| \leq 1} u_n(x) \, dx \to 0, \quad \delta > 0. \tag{1.36}
\]

(In other words, \( u_n \) has mass one and concentrates near \( x = 0 \) as \( n \to \infty \).)

Then for every \( f \in C[-\frac{1}{2}, \frac{1}{2}] \) which vanishes at the endpoints, \( f(-\frac{1}{2}) = f(\frac{1}{2}) = 0 \), we have that

\[
f_n(x) := \int_{-\frac{1}{2}}^{1/2} u_n(x-y) f(y) \, dy \tag{1.37}
\]

converges uniformly to \( f(x) \).

**Proof.** Since \( f \) is uniformly continuous, for given \( \varepsilon \) we can find a \( \delta < 1/2 \) (independent of \( x \)) such that \( |f(x) - f(y)| \leq \varepsilon \) whenever \( |x - y| \leq \delta \). Moreover, we can choose \( n \) such that \( \int_{\delta \leq |y| \leq 1} u_n(y) \, dy \leq \varepsilon \). Now abbreviate \( M = \max_{x \in [-1/2, 1/2]} \{1, |f(x)|\} \) and note

\[
|f(x) - \int_{-1/2}^{1/2} u_n(x-y) f(y) \, dy| = |f(x)| \left| 1 - \int_{-1/2}^{1/2} u_n(x-y) \, dy \right| \leq M \varepsilon.
\]

In fact, either the distance of \( x \) to one of the boundary points \( \pm \frac{1}{2} \) is smaller than \( \delta \) and hence \( |f(x)| \leq \varepsilon \) or otherwise \( [-\delta, \delta] \subset [x-1/2, x+1/2] \) and the difference between one and the integral is smaller than \( \varepsilon \).

Using this, we have

\[
|f_n(x) - f(x)| \leq \int_{-1/2}^{1/2} u_n(x-y) |f(y) - f(x)| \, dy + M \varepsilon
\]

\[
= \int_{|y| \leq 1/2, |x-y| \leq \delta} u_n(x-y) |f(y) - f(x)| \, dy
\]

\[
+ \int_{|y| \leq 1/2, |x-y| \geq \delta} u_n(x-y) |f(y) - f(x)| \, dy + M \varepsilon
\]

\[
\leq \varepsilon + 2M \varepsilon + M \varepsilon = (1 + 3M) \varepsilon, \tag{1.38}
\]

which proves the claim. \( \square \)
Note that \( f_n \) will be as smooth as \( u_n \), hence the title smoothing lemma. Moreover, \( f_n \) will be a polynomial if \( u_n \) is. The same idea is used to approximate noncontinuous functions by smooth ones (of course the convergence will no longer be uniform in this case).

Now we are ready to show:

**Theorem 1.3 (Weierstraß).** Let \( I \) be a compact interval. Then the set of polynomials is dense in \( C(I) \).

**Proof.** Let \( f(x) \in C(I) \) be given. By considering \( f(x) - f(a) - \frac{f(b) - f(a)}{b-a}(x - a) \) it is no loss to assume that \( f \) vanishes at the boundary points. Moreover, without restriction, we only consider \( I = [-\frac{1}{2}, \frac{1}{2}] \) (why?).

Now the claim follows from Lemma 1.2 using the Landau kernel

\[
  u_n(x) = \frac{1}{I^n}(1-x^2)^n,
\]

where (using integration by parts)

\[
  I_n := \int_{-1}^{1} (1-x^2)^n dx = \frac{n}{n+1} \int_{-1}^{1} (1-x)^{n-1}(1+x)^{n+1} dx
  = \cdots = \frac{n!}{(n+1)\cdots(2n+1)} 2^{2n+1} = \frac{(n!)^2 2^{2n+1}}{(2n+1)!} = \frac{n!}{\frac{1}{2}(\frac{1}{2} + 1)\cdots(\frac{1}{2} + n)}.
\]

Indeed, the first part of (1.36) holds by construction, and the second part follows from the elementary estimate

\[
  \frac{1}{2^n + 1} < I_n < 2,
\]

which shows \( \int_{|x| \leq 1} u_n(x) dx \leq 2u_n(\delta) < (2n+1)(1-\delta^2)^n \to 0 \). \( \square \)

**Corollary 1.4.** The monomials are total and hence \( C(I) \) is separable.

However, \( \ell^\infty(\mathbb{N}) \) is not separable (Problem 1.10)!

Note that while the proof of Theorem 1.3 provides an explicit way of constructing a sequence of polynomials \( f_n(x) \) which will converge uniformly to \( f(x) \), this method still has a few drawbacks from a practical point of view: Suppose we have approximated \( f \) by a polynomial of degree \( n \) but our approximation turns out to be insufficient for a certain purpose. First of all, since our polynomial will not be optimal in general, we could try to find another polynomial of the same degree giving a better approximation. However, as this is by no means straightforward, it seems more feasible to simply increase the degree. However, if we do this all coefficients will change and we need to start from scratch. This is in contradistinction to a Schauder basis where we could just add one new element from the basis (and where it suffices to compute one new coefficient).
In particular, note that this shows that the monomials are no Schauder basis for $C(I)$ since adding monomials incrementally to the expansion gives a convergent power series whose limit must be analytic.

We will see in the next section that the concept of orthogonality resolves these problems.

**Problem 1.2.** Show that $\|f\| - \|g\| \leq \|f - g\|$.

**Problem 1.3.** Let $X$ be a Banach space. Show that the norm, vector addition, and multiplication by scalars are continuous. That is, if $f_n \to f$, $g_n \to g$, and $\alpha_n \to \alpha$, then $\|f_n\| \to \|f\|$, $f_n + g_n \to f + g$, and $\alpha_n g_n \to \alpha g$.

**Problem 1.4.** Let $X$ be a Banach space. Show that $\sum_{j=1}^{\infty} \|f_j\| < \infty$ implies that
\[
\sum_{j=1}^{\infty} f_j = \lim_{n \to \infty} \sum_{j=1}^{n} f_j
\]
exists. The series is called **absolutely convergent** in this case.

**Problem 1.5.** While $\ell^1(\mathbb{N})$ is separable, it still has room for an uncountable set of linearly independent vectors. Show this by considering vectors of the form
\[
a^\alpha = (1, \alpha, \alpha^2, \ldots), \quad \alpha \in (0, 1).
\]
(Hint: Recall the Vandermonde determinant. See Problem 4.1 for a generalization.)

**Problem 1.6.** A **Hamel basis** is a maximal set of linearly independent vectors. Show that every vector space $X$ has a Hamel basis $\{u_\alpha\}_{\alpha \in \Lambda}$. Show that given a Hamel basis, every $x \in X$ can be written as a finite linear combination $x = \sum_{j=1}^{n} c_j u_{\alpha_j}$, where the vectors $u_{\alpha_j}$ and the constants $c_j$ are uniquely determined. (Hint: Use Zorn’s lemma, see Appendix A, to show existence.)

**Problem 1.7.** Prove (1.27). Show that equality occurs precisely if $\alpha = \beta$. (Hint: Take logarithms on both sides.)

**Problem 1.8.** Show that $\ell^p(\mathbb{N})$ is complete.

**Problem 1.9.** Show that $\ell^\infty(\mathbb{N})$ is a Banach space.

**Problem 1.10.** Show that $\ell^\infty(\mathbb{N})$ is not separable. (Hint: Consider sequences which take only the value one and zero. How many are there? What is the distance between two such sequences?)

**Problem 1.11.** Show that there is equality in the Hölder inequality (1.28) for $1 < p < \infty$ if and only if either $a = 0$ or $|b_j|^p = \alpha |a_j|^q$ for all $j \in \mathbb{N}$. Show that we have equality in the triangle inequality for $\ell^1(\mathbb{N})$ if and only if $a_j b_j^* \geq 0$ for all $j \in \mathbb{N}$. Show that we have equality in the triangle inequality for $\ell^p(\mathbb{N})$ for $1 < p < \infty$ if and only if $a = 0$ or $b = \alpha a$ with $\alpha \geq 0$. 
Problem 1.12. Let $X$ be a normed space. Show that the following conditions are equivalent.

(i) If $\|x + y\| = \|x\| + \|y\|$ then $y = \alpha x$ for some $\alpha \geq 0$ or $x = 0$.

(ii) If $\|x\| = \|y\| = 1$ and $x \neq y$ then $\|\lambda x + (1 - \lambda)y\| < 1$ for all $0 < \lambda < 1$.

(iii) If $\|x\| = \|y\| = 1$ and $x \neq y$ then $\|\frac{1}{2}\|x + y\| < 1$.

(iv) The function $x \mapsto \|x\|^2$ is strictly convex.

A norm satisfying one of them is called strictly convex.

Show that $\ell^p(N)$ is strictly convex for $1 < p < \infty$ but not for $p = 1, \infty$.

Problem 1.13. Show that $p_0 \leq p$ implies $\ell^{p_0}(N) \subseteq \ell^p(N)$ and $\|a\|_p \leq \|a\|_{p_0}$.

Moreover, show

$$\lim_{p \to \infty} \|a\|_p = \|a\|_\infty.$$ 

Problem 1.14. Formally extend the definition of $\ell^p(N)$ to $p \in (0, 1)$. Show that $\|\cdot\|_p$ does not satisfy the triangle inequality. However, show that it is a quasinormed space, that is, it satisfies all requirements for a normed space except for the triangle inequality which is replaced by

$$\|a + b\| \leq K(\|a\| + \|b\|)$$

with some constant $K \geq 1$. Show, in fact,

$$\|a + b\|_p \leq 2^{1/p - 1}(\|a\|_p + \|b\|_p), \quad p \in (0, 1).$$

Moreover, show that $\|\cdot\|_p^p$ satisfies the triangle inequality in this case, but of course it is no longer homogeneous (but at least you can get an honest metric $d(a, b) = \|a - b\|_p^p$ which gives rise to the same topology). (Hint: Show $\alpha + \beta \leq (\alpha^p + \beta^p)^{1/p} \leq 2^{1/p - 1}(\alpha + \beta)$ for $0 < p < 1$ and $\alpha, \beta \geq 0$.)

Problem 1.15. Show that the following set of functions is a Schauder basis for $C[0, 1]$: We start with $u_1(t) = t$, $u_2(t) = 1 - t$ and then split $[0, 1]$ into $2^n$ intervals of equal length and let $u_{2^n + k + 1}(t)$, for $1 \leq k \leq 2^n$, be a piecewise linear peak of height 1 supported in the $k$’th subinterval: $u_{2^n + k + 1}(t) = \max(0, 1 - |2^n + 1 - 2k + 1|)$ for $n \in \mathbb{N}_0$ and $1 \leq k \leq 2^n$.

1.3. The geometry of Hilbert spaces

So it looks like $C(I)$ has all the properties we want. However, there is still one thing missing: How should we define orthogonality in $C(I)$? In Euclidean space, two vectors are called orthogonal if their scalar product vanishes, so we would need a scalar product:
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Suppose \( \mathcal{H} \) is a vector space. A map \( \langle \cdot, \cdot \rangle : \mathcal{H} \times \mathcal{H} \to \mathbb{C} \) is called a sesquilinear form if it is conjugate linear in the first argument and linear in the second; that is,

\[
\begin{align*}
\langle \alpha_1 f_1 + \alpha_2 f_2, g \rangle &= \alpha_1^* \langle f_1, g \rangle + \alpha_2^* \langle f_2, g \rangle, \\
\langle f, \alpha_1 g_1 + \alpha_2 g_2 \rangle &= \alpha_1 \langle f, g_1 \rangle + \alpha_2 \langle f, g_2 \rangle,
\end{align*}
\]

(1.39)

where ‘*’ denotes complex conjugation. A symmetric sesquilinear form is also called a Hermitian form and a positive definite

\( \langle f, f \rangle > 0 \) for \( f \neq 0 \) (positive definite),

Hermitian form is called an inner product or scalar product. Note that (ii) follows in fact from (i) (Problem 1.19). Associated with every scalar product is a norm

\[
\|f\| := \sqrt{\langle f, f \rangle}.
\]

(1.40)

Only the triangle inequality is nontrivial. It will follow from the Cauchy–Schwarz inequality below. Until then, just regard (1.40) as a convenient short hand notation.

Warning: There is no common agreement whether a sesquilinear form (scalar product) should be linear in the first or in the second argument and different authors use different conventions.

The pair \((\mathcal{H}, \langle \cdot, \cdot \rangle)\) is called an inner product space. If \( \mathcal{H} \) is complete (with respect to the norm (1.40)), it is called a Hilbert space.

Example. Clearly, \( \mathbb{C}^n \) with the usual scalar product

\[
\langle a, b \rangle := \sum_{j=1}^{n} a_j^* b_j
\]

(1.41)

is a (finite dimensional) Hilbert space.

Example. A somewhat more interesting example is the Hilbert space \( \ell^2(\mathbb{N}) \), that is, the set of all complex-valued sequences

\[
\left\{ \{a_j\}_{j=1}^{\infty} \mid \sum_{j=1}^{\infty} |a_j|^2 < \infty \right\}
\]

(1.42)

with scalar product

\[
\langle a, b \rangle := \sum_{j=1}^{\infty} a_j^* b_j.
\]

(1.43)
By the Cauchy–Schwarz inequality for $\mathbb{C}^n$ we infer
\[
\left| \sum_{j=1}^{n} a_j^* b_j \right|^2 \leq \left( \sum_{j=1}^{n} |a_j^* b_j| \right)^2 \leq \sum_{j=1}^{n} |a_j|^2 \sum_{j=1}^{n} |b_j|^2 \leq \sum_{j=1}^{\infty} |a_j|^2 \sum_{j=1}^{\infty} |b_j|^2
\]
that the sum in the definition of the scalar product is absolutely convergent (and thus well-defined) for $a, b \in \ell^2(\mathbb{N})$. Observe that the norm $\|a\| = \sqrt{\langle a, a \rangle}$ is identical to the norm $\|a\|_2$ defined in the previous section. In particular, $\ell^2(\mathbb{N})$ is complete and thus indeed a Hilbert space.

A vector $f \in H$ is called normalized or a unit vector if $\|f\| = 1$. Two vectors $f, g \in H$ are called orthogonal or perpendicular ($f \perp g$) if $\langle f, g \rangle = 0$ and parallel if one is a multiple of the other.

If $f$ and $g$ are orthogonal, we have the Pythagorean theorem:
\[
\|f + g\|^2 = \|f\|^2 + \|g\|^2, \quad f \perp g, \quad (1.44)
\]
which is one line of computation (do it!).

Suppose $u$ is a unit vector. Then the projection of $f$ in the direction of $u$ is given by
\[
f_\parallel := \langle u, f \rangle u, \quad (1.45)
\]
and $f_\perp$, defined via
\[
f_\perp := f - \langle u, f \rangle u, \quad (1.46)
\]
is perpendicular to $u$ since $\langle u, f_\perp \rangle = \langle u, f - \langle u, f \rangle u \rangle = \langle u, f \rangle - \langle u, f \rangle \langle u, u \rangle = 0$.

Taking any other vector parallel to $u$, we obtain from (1.44)
\[
\|f - \alpha u\|^2 = \|f_\perp + (f_\parallel - \alpha u)\|^2 = \|f_\perp\|^2 + |\langle u, f \rangle - \alpha|^2 \quad (1.47)
\]
and hence $f_\parallel$ is the unique vector parallel to $u$ which is closest to $f$.

As a first consequence we obtain the Cauchy–Bunyakovsky–Schwarz inequality:

**Theorem 1.5** (Cauchy–Schwarz–Bunyakovsky). Let $H$ be an inner product space. Then for every $f, g \in H$ we have
\[
|\langle f, g \rangle| \leq \|f\| \|g\| \quad (1.48)
\]
with equality if and only if $f$ and $g$ are parallel.
Proof. It suffices to prove the case \( \|g\| = 1 \). But then the claim follows from \( \|f\|^2 = |\langle g, f \rangle|^2 + \|f_\perp\|^2 \).

We will follow common practice and refer to (1.48) simply as Cauchy–Schwarz inequality. Note that the Cauchy–Schwarz inequality implies that the scalar product is continuous in both variables; that is, if \( f_n \to f \) and \( g_n \to g \), we have \( \langle f_n, g_n \rangle \to \langle f, g \rangle \).

As another consequence we infer that the map \( \|\cdot\| \) is indeed a norm. In fact,

\[
\|f + g\|^2 = \|f\|^2 + \langle f, g \rangle + \langle g, f \rangle + \|g\|^2 \leq (\|f\| + \|g\|)^2.
\]  
(1.49)

But let us return to \( C(I) \). Can we find a scalar product which has the maximum norm as associated norm? Unfortunately the answer is no! The reason is that the maximum norm does not satisfy the parallelogram law (Problem 1.18).

**Theorem 1.6** (Jordan–von Neumann). A norm is associated with a scalar product if and only if the parallelogram law

\[
\|f + g\|^2 + \|f - g\|^2 = 2\|f\|^2 + 2\|g\|^2
\]

(1.50)

holds.

In this case the scalar product can be recovered from its norm by virtue of the polarization identity

\[
\langle f, g \rangle = \frac{1}{4} \left( \|f + g\|^2 - \|f - g\|^2 + i\|f - ig\|^2 - i\|f + ig\|^2 \right).
\]  
(1.51)

Proof. If an inner product space is given, verification of the parallelogram law and the polarization identity is straightforward (Problem 1.19).

To show the converse, we define

\[
s(f, g) := \frac{1}{4} \left( \|f + g\|^2 - \|f - g\|^2 + i\|f - ig\|^2 - i\|f + ig\|^2 \right).
\]

Then \( s(f, f) = \|f\|^2 \) and \( s(f, g) = s(g, f)^* \) are straightforward to check. Moreover, another straightforward computation using the parallelogram law shows

\[
s(f, g) + s(f, h) = 2s(f, \frac{g + h}{2}).
\]

Now choosing \( h = 0 \) (and using \( s(f, 0) = 0 \)) shows \( s(f, g) = 2s(f, \frac{g}{2}) \) and thus \( s(f, g) + s(f, h) = s(f, g + h) \). Furthermore, by induction we infer \( \frac{m}{2^n} s(f, g) = s(f, \frac{mg}{2^n} g) \); that is, \( \alpha s(f, g) = s(f, \alpha g) \) for a dense set of positive rational numbers \( \alpha \). By continuity (which follows from continuity of the norm) this holds for all \( \alpha \geq 0 \) and \( s(f, -g) = -s(f, g) \), respectively, \( s(f, ig) = is(f, g) \), finishes the proof. □
In the case of a real Hilbert space, the polarization identity of course simplifies to $(f, g) = \frac{1}{4}(\|f + g\|^2 - \|f - g\|^2)$.

Note that the parallelogram law and the polarization identity even hold for sesquilinear forms (Problem 1.19).

But how do we define a scalar product on $C(I)$? One possibility is

$$(f, g) := \int_a^b f^*(x)g(x)dx. \quad (1.52)$$

The corresponding inner product space is denoted by $L^2_{\text{cont}}(I)$. Note that we have

$$\|f\| \leq \sqrt{b-a}\|f\|_{\infty} \quad (1.53)$$

and hence the maximum norm is stronger than the $L^2_{\text{cont}}$ norm.

Suppose we have two norms $\|\cdot\|_1$ and $\|\cdot\|_2$ on a vector space $X$. Then $\|\cdot\|_2$ is said to be stronger than $\|\cdot\|_1$ if there is a constant $m > 0$ such that

$$\|f\|_1 \leq m\|f\|_2. \quad (1.54)$$

It is straightforward to check the following.

**Lemma 1.7.** If $\|\cdot\|_2$ is stronger than $\|\cdot\|_1$, then every $\|\cdot\|_2$ Cauchy sequence is also a $\|\cdot\|_1$ Cauchy sequence.

Hence if a function $F : X \to Y$ is continuous in $(X, \|\cdot\|_1)$, it is also continuous in $(X, \|\cdot\|_2)$, and if a set is dense in $(X, \|\cdot\|_2)$, it is also dense in $(X, \|\cdot\|_1)$.

In particular, $L^2_{\text{cont}}$ is separable since the polynomials are dense. But is it also complete? Unfortunately the answer is no:

**Example.** Take $I = [0, 2]$ and define

$$f_n(x) := \begin{cases} 0, & 0 \leq x < 1 - \frac{1}{n}, \\ 1 + n(x - 1), & 1 - \frac{1}{n} \leq x \leq 1, \\ 1, & 1 \leq x \leq 2. \end{cases} \quad (1.55)$$

Then $f_n(x)$ is a Cauchy sequence in $L^2_{\text{cont}}$, but there is no limit in $L^2_{\text{cont}}$! Clearly, the limit should be the step function which is 0 for $0 \leq x < 1$ and 1 for $1 \leq x \leq 2$, but this step function is discontinuous (Problem 1.22)! \(\diamond\)

**Example.** The previous example indicates that we should consider (1.52) on a larger class of functions, for example on the class of Riemann integrable functions

$$\mathcal{R}(I) := \{ f : I \to \mathbb{C} | f \text{ is Riemann integrable} \}$$

such that the integral makes sense. While this seems natural it implies another problem: Any function which vanishes outside a set which is negligible for the integral (e.g. finitely many points) has norm zero! That is,
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\[ \|f\|_2 := (\int_I |f(x)|^2 \, dx)^{1/2} \] is only a seminorm on \( \mathcal{R}(I) \) (Problem 1.21). To get a norm we consider \( \mathcal{N}(I) := \{ f \in \mathcal{R}(I) | \|f\|_2 = 0 \} \). By homogeneity and the triangle inequality \( \mathcal{N}(I) \) is a subspace and we can consider equivalence classes of functions which differ by a negligible function from \( \mathcal{N}(I) \):

\[ \mathcal{L}^2_{R^I} := \mathcal{R}(I)/\mathcal{N}(I). \]

Since \( \|f\|_2 = \|g\|_2 \) for \( f - g \in \mathcal{N}(I) \) we have a norm on \( \mathcal{L}^2_{R^I} \). Moreover, since this norm inherits the parallelogram law we even have an inner product space. However, this space will not be complete unless we replace the Riemann by the Lebesgue integral. Hence will not pursue this further until we have the Lebesgue integral at our disposal.

This shows that in infinite dimensional vector spaces, different norms will give rise to different convergent sequences! In fact, the key to solving problems in infinite dimensional spaces is often finding the right norm! This is something which cannot happen in the finite dimensional case.

**Theorem 1.8.** If \( X \) is a finite dimensional vector space, then all norms are equivalent. That is, for any two given norms \( \|\cdot\|_1 \) and \( \|\cdot\|_2 \), there are positive constants \( m_1 \) and \( m_2 \) such that

\[ \frac{1}{m_2} \|f\|_1 \leq \|f\|_2 \leq m_1 \|f\|_1. \quad (1.56) \]

**Proof.** Choose a basis \( \{u_j\}_{1 \leq j \leq n} \) such that every \( f \in X \) can be written as \( f = \sum_j \alpha_j u_j \). Since equivalence of norms is an equivalence relation (check this!), we can assume that \( \|\cdot\|_2 \) is the usual Euclidean norm: \( \|f\|_2 = \|\sum_j \alpha_j u_j\|_2 = (\sum_j |\alpha_j|^2)^{1/2} \). Then by the triangle and Cauchy–Schwarz inequalities,

\[ \|f\|_1 \leq \sum_j |\alpha_j||u_j|_1 \leq \sqrt{\sum_j \|u_j\|_1^2} \|f\|_2 \]

and we can choose \( m_2 = \sqrt{\sum_j \|u_j\|_1^2} \).

In particular, if \( f_n \) is convergent with respect to \( \|\cdot\|_2 \), it is also convergent with respect to \( \|\cdot\|_1 \). Thus \( \|\cdot\|_1 \) is continuous with respect to \( \|\cdot\|_2 \) and attains its minimum \( m > 0 \) on the unit sphere \( S := \{ u \|u\|_2 = 1 \} \) (which is compact by the Heine–Borel theorem, Theorem 0.22). Now choose \( m_1 = 1/m \). \( \Box \)

Finally, we remark that a real Hilbert space can always be embedded into a complex Hilbert space. In fact, if \( \mathcal{H} \) is a real Hilbert space, then \( \mathcal{H} \times \mathcal{H} \) is a complex Hilbert space if we define

\[ (f_1, f_2) + (g_1, g_2) = (f_1 + g_1, f_2 + g_2), \quad (\alpha + i\beta)(f_1, f_2) = (\alpha f_1 - \beta f_2, \alpha f_2 + \beta f_1) \quad (1.57) \]
and
\[
\langle (f_1, f_2), (g_1, g_2) \rangle = \langle f_1, f_2 \rangle + \langle g_1, g_2 \rangle + i(\langle f_1, g_2 \rangle - \langle f_2, g_1 \rangle).
\] (1.58)

Here you should think of \((f_1, f_2)\) as \(f_1 + if_2\). Note that we have a conjugate linear map \(C : H \times H \rightarrow H \times H\), \((f_1, f_2) \mapsto (f_1, -f_2)\) which satisfies \(C^2 = I\) and \(\langle Cf, Cg \rangle = \langle g, f \rangle\). In particular, we can get our original Hilbert space back if we consider \(\text{Re}(f) = \frac{1}{2}(f + Cf) = (f_1, 0)\).

**Problem 1.16.** Show that the norm in a Hilbert space satisfies \(\|f + g\| = \|f\| + \|g\|\) if and only if \(f = \alpha g, \alpha \geq 0\), or \(g = 0\). Hence Hilbert spaces are strictly convex (cf. Problem 1.12).

**Problem 1.17** (Generalized parallelogram law). Show that, in a Hilbert space,
\[
\sum_{1 \leq j < k \leq n} \|x_j - x_k\|^2 + \| \sum_{1 \leq j \leq n} x_j \|^2 = n \sum_{1 \leq j \leq n} \|x_j\|^2.
\]
The case \(n = 2\) is (1.50).

**Problem 1.18.** Show that the maximum norm on \(C[0, 1]\) does not satisfy the parallelogram law.

**Problem 1.19.** Suppose \(Q\) is a complex vector space. Let \(s(f, g)\) be a sesquilinear form on \(Q\) and \(q(f) := s(f, f)\) the associated quadratic form. Prove the parallelogram law
\[
q(f + g) + q(f - g) = 2q(f) + 2q(g)
\] (1.59)
and the polarization identity
\[
s(f, g) = \frac{1}{4}(q(f + g) - q(f - g) + i(q(f - ig) - i(q(f + ig))).
\] (1.60)

Show that \(s(f, g)\) is symmetric if and only if \(q(f)\) is real-valued.

Note, that if \(Q\) is a real vector space, then the parallelogram law is unchanged but the polarization identity in the form \(s(f, g) = \frac{1}{4}(q(f + g) - q(f - g))\) will only hold if \(s(f, g)\) is symmetric.

**Problem 1.20.** A sesquilinear form on a complex inner product space is called **bounded** if
\[
\|s\| := \sup_{\|f\| = 1 \|g\| = 1} |s(f, g)|
\]
is finite. Similarly, the associated quadratic form \(q\) is **bounded** if
\[
\|q\| := \sup_{\|f\| = 1} |q(f)|
\]
is finite. Show
\[
\|q\| \leq \|s\| \leq 2\|q\|.
\]
(Hint: Use the polarization identity from the previous problem.)
Problem 1.21. Suppose $Ω$ is a vector space. Let $s(f, g)$ be a sesquilinear form on $Ω$ and $q(f) := s(f, f)$ the associated quadratic form. Show that the Cauchy–Schwarz inequality

$$|s(f, g)| \leq q(f)^{1/2}q(g)^{1/2}$$

holds if $q(f) \geq 0$. In this case $q(.)^{1/2}$ satisfies the triangle inequality and hence is a seminorm.

(Hint: Consider $0 \leq q(f + αg) = q(f) + 2\text{Re}(α s(f, g)) + |α|^2q(g)$ and choose $α = ts(f, g)^*/|s(f, g)|$ with $t \in \mathbb{R}$.)

Problem 1.22. Prove the claims made about $f_n$, defined in (1.55), in this example.

1.4. Completeness

Since $L^2_{\text{cont}}$ is not complete, how can we obtain a Hilbert space from it? Well, the answer is simple: take the completion.

If $X$ is an (incomplete) normed space, consider the set of all Cauchy sequences $X$. Call two Cauchy sequences equivalent if their difference converges to zero and denote by $\tilde{X}$ the set of all equivalence classes. It is easy to see that $\tilde{X}$ (and $X$) inherit the vector space structure from $X$. Moreover,

Lemma 1.9. If $x_n$ is a Cauchy sequence in $X$, then $\|x_n\|$ is also a Cauchy sequence and thus converges.

Consequently, the norm of an equivalence class $[(x_n)]_{n=1}^\infty$ can be defined by $\|(x_n)_{n=1}^\infty\| := \lim_{n \to \infty} \|x_n\|$ and is independent of the representative (show this!). Thus $\tilde{X}$ is a normed space.

Theorem 1.10. $\tilde{X}$ is a Banach space containing $X$ as a dense subspace if we identify $x \in X$ with the equivalence class of all sequences converging to $x$.

Proof. (Outline) It remains to show that $\tilde{X}$ is complete. Let $ξ_n = [(x_{n,j})_{j=1}^\infty]$ be a Cauchy sequence in $\tilde{X}$. Then it is not hard to see that $ξ = [(x_{j,j})_{j=1}^\infty]$ is its limit.

Let me remark that the completion $\tilde{X}$ is unique. More precisely, every other complete space which contains $X$ as a dense subset is isomorphic to $\tilde{X}$. This can for example be seen by showing that the identity map on $X$ has a unique extension to $\tilde{X}$ (compare Theorem 1.12 below).

In particular, it is no restriction to assume that a normed vector space or an inner product space is complete (note that by continuity of the norm the parallelogram law holds for $\tilde{X}$ if it holds for $X$).
Example. The completion of the space $L^2_{\text{cont}}(I)$ is denoted by $L^2(I)$. While this defines $L^2(I)$ uniquely (up to isomorphisms) it is often inconvenient to work with equivalence classes of Cauchy sequences. Hence we will give a different characterization as equivalence classes of square integrable (in the sense of Lebesgue) functions later.

Similarly, we define $L^p(I)$, $1 \leq p < \infty$, as the completion of $C(I)$ with respect to the norm
$$\|f\|_p := \left(\int_a^b |f(x)|^p\right)^{1/p}.$$The only requirement for a norm which is not immediate is the triangle inequality (except for $p = 1, 2$) but this can be shown as for $\ell^p$ (cf. Problem 1.25).

Problem 1.23. Provide a detailed proof of Theorem 1.10.

Problem 1.24. For every $f \in L^1(I)$ we can define its integral
$$\int_c^d f(x)dx$$as the (unique) extension of the corresponding linear functional from $C(I)$ to $L^1(I)$ (by Theorem 1.12 below). Show that this integral is linear and satisfies
$$\int_c^e f(x)dx = \int_c^d f(x)dx + \int_d^e f(x)dx, \quad \left| \int_c^d f(x)dx \right| \leq \int_c^d |f(x)|dx.$$

Problem 1.25. Show the Hölder inequality
$$\|fg\|_1 \leq \|f\|_p \|g\|_q, \quad \frac{1}{p} + \frac{1}{q} = 1, \quad 1 \leq p, q \leq \infty,$$and conclude that $\|\cdot\|_p$ is a norm on $C(I)$. Also conclude that $L^p(I) \subseteq L^1(I)$.

1.5. Bounded operators

A linear map $A$ between two normed spaces $X$ and $Y$ will be called a (linear) operator
$$A : \mathcal{D}(A) \subseteq X \to Y.$$The linear subspace $\mathcal{D}(A)$ on which $A$ is defined is called the domain of $A$ and is frequently required to be dense. The kernel (also null space)
$$\text{Ker}(A) := \{f \in \mathcal{D}(A) | Af = 0\} \subseteq X$$and range
$$\text{Ran}(A) := \{Af | f \in \mathcal{D}(A)\} = A\mathcal{D}(A) \subseteq Y$$are defined.
1.5. Bounded operators

are defined as usual. Note that a linear map \( A \) will be continuous if and only if it is continuous at 0, that is, \( x_n \in \mathcal{D}(A) \to 0 \) implies \( Ax_n \to 0 \).

The operator \( A \) is called \textbf{bounded} if the operator norm
\[
\|A\| := \sup_{f \in \mathcal{D}(A), \|f\|_X = 1} \|Af\|_Y
\]
is finite. This says that \( A \) is bounded if the image of the closed unit ball \( \bar{B}_1(0) \subset X \) is contained in some closed ball \( \bar{B}_r(0) \subset Y \) of finite radius \( r \) (with the smallest radius being the operator norm). Hence \( A \) is bounded if and only if it maps bounded sets to bounded sets.

Note that if you replace the norm on \( X \) or \( Y \) then the operator norm will of course also change in general. However, if the norms are equivalent so will be the operator norms.

By construction, a bounded operator is Lipschitz continuous,
\[
\|Af\|_Y \leq \|A\| \|f\|_X, \quad f \in \mathcal{D}(A),
\]
and hence continuous. The converse is also true:

**Theorem 1.11.** An operator \( A \) is bounded if and only if it is continuous.

**Proof.** Suppose \( A \) is continuous but not bounded. Then there is a sequence of unit vectors \( u_n \) such that \( \|Au_n\| \geq n \). Then \( f_n := \frac{1}{n} u_n \) converges to 0 but \( \|Af_n\| \geq 1 \) does not converge to 0. \( \square \)

In particular, if \( X \) is finite dimensional, then every operator is bounded. In the infinite dimensional an operator can be unbounded. Moreover, one and the same operation might be bounded (i.e. continuous) or unbounded, depending on the norm chosen.

**Example.** Let \( X = \mathcal{B}^p(\mathbb{N}) \) and \( a \in \ell^\infty(\mathbb{N}) \). Consider the multiplication operator \( A : X \to X \) defined by
\[
(AB)_j := a_j b_j.
\]
Then \( |(AB)_j| \leq \|a\|_\infty |b_j| \) shows \( \|A\| \leq \|a\|_\infty \). In fact, we even have \( \|A\| = \|a\|_\infty \) (show this).

**Example.** Consider the vector space of differentiable functions \( X = C^1[0,1] \) and equip it with the norm (cf. Problem 1.28)
\[
\|f\|_{\infty,1} := \max_{x \in [0,1]} |f(x)| + \max_{x \in [0,1]} |f'(x)|.
\]
Let \( Y = C[0,1] \) and observe that the differential operator \( A = \frac{d}{dx} : X \to Y \) is bounded since
\[
\|Af\|_\infty = \max_{x \in [0,1]} |f'(x)| \leq \max_{x \in [0,1]} |f(x)| + \max_{x \in [0,1]} |f'(x)| = \|f\|_{\infty,1}.
\]
However, if we consider $A = \frac{d}{dx} : \mathcal{D}(A) \subseteq Y \to Y$ defined on $\mathcal{D}(A) = C^1[0, 1]$, then we have an unbounded operator. Indeed, choose 

$$u_n(x) = \sin(n\pi x)$$

which is normalized, $\|u_n\|_\infty = 1$, and observe that 

$$Au_n(x) = u'_n(x) = n\pi \cos(n\pi x)$$

is unbounded, $\|Au_n\|_\infty = n\pi$. Note that $\mathcal{D}(A)$ contains the set of polynomials and thus is dense by the Weierstraß approximation theorem (Theorem 1.3).

If $A$ is bounded and densely defined, it is no restriction to assume that it is defined on all of $X$.

**Theorem 1.12.** Let $A : \mathcal{D}(A) \subseteq X \to Y$ be a bounded linear operator between a normed space $X$ and a Banach space $Y$. If $\mathcal{D}(A)$ is dense, there is a unique (continuous) extension of $A$ to $X$ which has the same operator norm.

**Proof.** Since a bounded operator maps Cauchy sequences to Cauchy sequences, this extension can only be given by 

$$\overline{A}f := \lim_{n \to \infty} Af_n, \quad f_n \in \mathcal{D}(A), \quad f \in X.$$ 

To show that this definition is independent of the sequence $f_n \to f$, let $g_n \to f$ be a second sequence and observe 

$$\|Af_n - Ag_n\| = \|A(f_n - g_n)\| \leq \|A\|\|f_n - g_n\| \to 0.$$ 

Since for $f \in \mathcal{D}(A)$ we can choose $f_n = f$, we see that $\overline{A}f = Af$ in this case, that is, $\overline{A}$ is indeed an extension. From continuity of vector addition and scalar multiplication it follows that $\overline{A}$ is linear. Finally, from continuity of the norm we conclude that the operator norm does not increase.

The set of all bounded linear operators from $X$ to $Y$ is denoted by $\mathcal{L}(X, Y)$. If $X = Y$, we write $\mathcal{L}(X) := \mathcal{L}(X, X)$. An operator in $\mathcal{L}(X, \mathbb{C})$ is called a **bounded linear functional**, and the space $X^* := \mathcal{L}(X, \mathbb{C})$ is called the **dual space** of $X$. The dual space takes the role of coordinate functions in a Banach space.

**Example.** Let $X = \ell^p(\mathbb{N})$. Then the coordinate functions 

$$\ell_j(a) := a_j$$

are bounded linear functionals: $|\ell_j(a)| = |a_j| \leq \|a\|_p$ and hence $\|\ell_j\| = 1$. More general, let $b \in \ell^q(\mathbb{N})$ where $\frac{1}{p} + \frac{1}{q} = 1$. Then 

$$\ell_b(a) := \sum_{j=1}^n b_j a_j$$
1.5. Bounded operators

is a bounded linear functional satisfying \( \| \ell_b \| \leq \| b \|_q \) by Hölder’s inequality. In fact, we even have \( \| \ell_b \| = \| b \|_q \) (Problem 4.15). Note that the first example is a special case of the second one upon choosing \( b = \delta^j \).

**Example.** Consider \( X := C(I) \). Then for every \( x_0 \in I \) the point evaluation \( \ell_{x_0}(f) := f(x_0) \) is a bounded linear functional. In fact, \( \| \ell_{x_0} \| = 1 \) (show this).

However, note that \( \ell_{x_0} \) is unbounded on \( L^2_{\text{cont}}(I) \)! To see this take \( f_n(x) = \sqrt{\frac{3n}{2}} \max(0, 1 - n|x - x_0|) \) which is a triangle shaped peak supported on \([x_0 - n^{-1}, x_0 + n^{-1}]\) and normalized according to \( \| f_n \|_2 = 1 \) for \( n \) sufficiently large such that the support is contained in \( I \). Then \( \ell_{x_0}(f) = f_n(x_0) = \sqrt{\frac{3n}{2}} \to \infty \). This implies that \( \ell_{x_0} \) cannot be extended to the completion of \( L^2_{\text{cont}}(I) \) in a natural way and reflects the fact that the integral cannot see individual points (changing the value of a function at one point does not change its integral).

**Example.** Consider \( X = C(I) \) and let \( g \) be some (Riemann or Lebesgue) integrable function. Then

\[
\ell_g(f) := \int_a^b g(x) f(x) \, dx
\]

is a linear functional with norm

\[
\| \ell_g \| = \| g \|_1.
\]

Indeed, first of all note that

\[
|\ell_g(f)| \leq \int_a^b |g(x) f(x)| \, dx \leq \| f \|_\infty \int_a^b |g(x)| \, dx
\]

shows \( \| \ell_g \| \leq \| g \|_1 \). To see that we have equality consider \( f_\varepsilon = g^*/(|g| + \varepsilon) \) and note

\[
|\ell_g(f_\varepsilon)| = \int_a^b \frac{|g(x)|^2}{1 + \varepsilon |g(x)|^2} \, dx \geq \int_a^b \frac{|g(x)|^2 - \varepsilon^2}{|g(x)|^2} \, dx = \| g \|_1 - (b - a)\varepsilon.
\]

Since \( \| f_\varepsilon \| \leq 1 \) and \( \varepsilon > 0 \) is arbitrary this establishes the claim.

**Theorem 1.13.** The space \( \mathcal{L}(X,Y) \) together with the operator norm \((1.65)\) is a normed space. It is a Banach space if \( Y \) is.

**Proof.** That \((1.65)\) is indeed a norm is straightforward. If \( Y \) is complete and \( A_n \) is a Cauchy sequence of operators, then \( A_nf \) converges to an element \( g \) for every \( f \). Define a new operator \( A \) via \( Af = g \). By continuity of the vector operations, \( A \) is linear and by continuity of the norm \( \| Af \| = \lim \| A_n f \| \leq (\lim \| A_n \|) \| f \| \), it is bounded. Furthermore, given \( \varepsilon > 0 \), there is some \( N \) such that \( \| A_n - A_m \| \leq \varepsilon \) for \( n, m \geq N \) and thus
\[ \|A_n f - A_m f\| \leq \varepsilon \|f\|. \] Taking the limit \( m \to \infty \), we see \( \|A_n f - Af\| \leq \varepsilon \|f\| \); that is, \( A_n \to A \). \qed

The Banach space of bounded linear operators \( \mathcal{L}(X) \) even has a multiplication given by composition. Clearly, this multiplication satisfies

\[ (A + B)C = AC + BC, \quad A(B + C) = AB + BC, \quad A, B, C \in \mathcal{L}(X) \] (1.67)

and

\[ (AB)C = A(BC), \quad \alpha (AB) = (\alpha A)B = A (\alpha B), \quad \alpha \in \mathbb{C}. \] (1.68)

Moreover, it is easy to see that we have

\[ \|AB\| \leq \|A\| \|B\|. \] (1.69)

In other words, \( \mathcal{L}(X) \) is a so-called Banach algebra. However, note that our multiplication is not commutative (unless \( X \) is one-dimensional). We even have an identity, the identity operator \( I \), satisfying \( \|I\| = 1 \).

**Problem 1.26.** Consider \( X = \mathbb{C}^n \) and let \( A \in \mathcal{L}(X) \) be a matrix. Equip \( X \) with the norm (show that this is a norm)

\[ \|x\|_\infty := \max_{1 \leq j \leq n} |x_j| \]

and compute the operator norm \( \|A\| \) with respect to this norm in terms of the matrix entries. Do the same with respect to the norm

\[ \|x\|_1 := \sum_{1 \leq j \leq n} |x_j|. \]

**Problem 1.27.** Show that the integral operator

\[ (Kf)(x) := \int_0^1 K(x, y)f(y)dy, \]

where \( K(x, y) \in C([0, 1] \times [0, 1]) \), defined on \( \mathcal{D}(K) := C[0, 1] \), is a bounded operator both in \( X := C[0, 1] \) (max norm) and \( X := \mathcal{L}^2_{\text{cont}}(0, 1) \). Show that the norm in the \( X = C[0, 1] \) case is given by

\[ \|K\| = \max_{x \in [0, 1]} \int_0^1 |K(x, y)|dy. \]

**Problem 1.28.** Let \( I \) be a compact interval. Show that the set of differentiable functions \( C^1(I) \) becomes a Banach space if we set \( \|f\|_{\infty, 1} := \max_{x \in I} |f(x)| + \max_{x \in I} |f'(x)| \).

**Problem 1.29.** Show that \( \|AB\| \leq \|A\| \|B\| \) for every \( A, B \in \mathcal{L}(X) \). Conclude that the multiplication is continuous: \( A_n \to A \) and \( B_n \to B \) imply \( A_n B_n \to AB \).
**Problem 1.30.** Let \( A \in \mathcal{L}(X) \) be a bijection. Show 
\[
\|A^{-1}\|^{-1} = \inf_{f \in X, \|f\|=1} \|Af\|.
\]

**Problem 1.31.** Suppose \( B \in \mathcal{L}(X) \) with \( \|B\| < 1 \). Then \( I + B \) is invertible with 
\[
(I + B)^{-1} = \sum_{n=0}^{\infty} (-1)^n B^n.
\]
Consequently for \( A, B \in \mathcal{L}(X,Y) \), \( A + B \) is invertible if \( A \) is invertible and \( \|B\| < \|A^{-1}\|^{-1} \).

**Problem 1.32.** Let 
\[
f(z) := \sum_{j=0}^{\infty} f_j z^j, \quad |z| < R,
\]
be a convergent power series with radius of convergence \( R > 0 \). Suppose \( X \) is a Banach space and \( A \in \mathcal{L}(X) \) is a bounded operator with \( \limsup_n \|A^n\|^{1/n} < R \) (note that by \( \|A^n\| \leq \|A\|^n \) the limsup is finite). Show that 
\[
f(A) := \sum_{j=0}^{\infty} f_j A^j
\]
exists and defines a bounded linear operator. Moreover, if \( f \) and \( g \) are two such functions and \( \alpha \in \mathbb{C} \), then 
\[
(f + g)(A) = f(A) + g(A), \quad (\alpha f)(A) = \alpha f(A), \quad (fg)(A) = f(A)g(A).
\]
(Hint: Problem 1.4.)

**Problem 1.33.** Show that a linear map \( \ell : X \to \mathbb{C} \) is continuous if and only if its kernel is closed. (Hint: If \( \ell \) is not continuous, we can find a sequence of normalized vectors \( x_n \) with \( |\ell(x_n)| \to \infty \) and a vector \( y \) with \( \ell(y) = 1 \).)

### 1.6. Sums and quotients of Banach spaces

Given two Banach spaces \( X_1 \) and \( X_2 \) we can define their (direct) **sum** \( X := X_1 \oplus X_2 \) as the Cartesian product \( X_1 \times X_2 \) together with the norm \( \|(x_1, x_2)\| := \|x_1\| + \|x_2\| \). In fact, since all norms on \( \mathbb{R}^2 \) are equivalent (Theorem 1.8), we could as well take \( \|(x_1, x_2)\| := (\|x_1\|^p + \|x_2\|^p)^{1/p} \) or \( \|(x_1, x_2)\| := \max(\|x_1\|, \|x_2\|) \). In particular, in the case of Hilbert spaces the choice \( p = 2 \) will ensure that \( X \) is again a Hilbert space.

Note that \( X_1 \) and \( X_2 \) can be regarded as closed subspaces of \( X_1 \times X_2 \) by virtue of the obvious embeddings \( x_1 \mapsto (x_1,0) \) and \( x_2 \mapsto (0,x_2) \). It is straightforward to show that \( X \) is again a Banach space and to generalize this concept to finitely many spaces (Problem 1.34).
Given two subspaces $M, N \subseteq X$ of a vector space, we can define their sum as usual: $M + N := \{x + y | x \in M, y \in N\}$. In particular, the decomposition $x + y$ with $x \in M, y \in N$ is unique iff $M \cap N = \{0\}$ and we will write $M + N$ in this case. It is important to observe, that $M + N$ is in general different from $M \oplus N$ since both have different norms. In fact, $M + N$ might not even be closed.

**Example.** Consider $X := \ell^p(\mathbb{N})$. Let $M = \{a \in X | a_{2n} = 0\}$ and $N = \{a \in X | a_{2n+1} = n^3 a_{2n}\}$. Then both subspaces are closed and $M \cap N = \{0\}$. Moreover, $M + N$ is dense since it contains all sequences with finite support. However, it is not all of $X$ since $a_n = \frac{1}{n^3} \not\in M + N$. Indeed, if we could write $a = b + c \in M + N$, then $c_{2n} = \frac{1}{n^3}$ and hence $c_{2n+1} = 0$ contradicting $c \in N \subseteq X$.

**Example.** Given a real normed space $X$ its **complexification** is given by $X_C := X \times X$ together with the (complex) scalar multiplication $\alpha(x, y) = (\text{Re}(\alpha)x - \text{Im}(\alpha)y, \text{Re}(\alpha)y + \text{Im}(\alpha)x)$. By virtue of our embedding $x \mapsto (x, 0)$ you should of course think of $(x, y)$ as $x + iy$. As a norm one can take (show this)

$$
\|x + iy\|_C := \max_{0 \leq t \leq 2\pi} \|\cos(t)x + \sin(t)y\|,
$$

which satisfies $\|x\|_C = \|x\|$ and $\|x + iy\|_C = \|x - iy\|_C$. Given two real normed spaces $X_1, X_2$, every linear operator $A : X_1 \to X_2$ gives rise to a linear operator $A_C : X_1,C \to X_2,C$ via $A_C(x + iy) = Ax + iAy$. Similarly, a bilinear form $s : X \times X \to \mathbb{R}$ gives rise to a sesquilinear form $s_C(x_1 + iy_1, x_2 + iy_2) := s(x_1, x_2) + s(y_1, y_2) + i(s(x_1, y_2) - s(y_1, x_2))$. In particular, if $X$ is a Hilbert space, so will be $X_C$.

Note that if you start with a complex normed space and regard it as a real normed space (by restricting scalar multiplication to real numbers), complexification will give you a larger space. If you want to get back your original space, you need to observe that you have an automorphism $I : X \to X$ of real spaces satisfying $I^2x = -x$ given by multiplication with $i$. Given such an automorphism you can define the complex scalar multiplication via

$$
\alpha x := \text{Re}(\alpha)x + \text{Im}(\alpha)Ix.
$$

We will show below that this cannot happen if one of the spaces is finite dimensional.

A closed subspace $M$ is called **complemented** if we can find another closed subspace $N$ with $M \cap N = \{0\}$ and $M \dot{+} N = X$. In this case every $x \in X$ can be uniquely written as $x = x_1 + x_2$ with $x_1 \in M, x_2 \in N$ and we can define a projection $P : X \to M, x \mapsto x_1$. By definition $P^2 = P$ and we have a complementary projection $Q := I - P$ with $Q : X \to N, x \mapsto x_2$. Moreover, it is straightforward to check $M = \text{Ker}(Q) = \text{Ran}(P)$.
and $N = \text{Ker}(P) = \text{Ran}(Q)$. Of course one would like $P$ (and hence also $Q$) to be continuous. If we consider the map $\phi : M \oplus N \to X$, $(x_1, x_2) \to x_1 + x_2$ then this is equivalent to the question if $\phi^{-1}$ is continuous. By the triangle inequality $\phi$ is continuous with $\|\phi\| \leq 1$ and the inverse mapping theorem (Theorem 4.6) will answer this question affirmative.

It is important to emphasize, that it is precisely the requirement that $N$ is closed which makes $P$ continuous (conversely observe that $N = \text{Ker}(P)$ is closed if $P$ is continuous). Without this requirement we can always find $N$ by a simple application of Zorn’s lemma (order the subspaces which have trivial intersection with $M$ by inclusion and note that a maximal element has the required properties). Moreover, the question which closed subspaces can be complemented is a highly nontrivial one. If $M$ is finite (co)dimensional, then it can be complemented (see Problems 1.39 and 4.21).

Given a subspace $M$ of a linear space $X$ we can define the quotient space $X/M$ as the set of all equivalence classes $[x] = x + M$ with respect to the equivalence relation $x \equiv y$ if $x - y \in M$. It is straightforward to see that $X/M$ is a vector space when defining $[x] + [y] = [x + y]$ and $\alpha[x] = [\alpha x]$ (show that these definitions are independent of the representative of the equivalence class). In particular, for a linear operator $A : X \to Y$ the linear space $\text{Coker}(A) := Y/\text{Ran}(A)$ is know as the cokernel of $A$. The dimension of $X/M$ is known as the codimension of $M$.

**Lemma 1.14.** Let $M$ be a closed subspace of a Banach space $X$. Then $X/M$ together with the norm

\[ \| [x] \| := \text{dist}(x, M) = \inf_{y \in M} \| x + y \| \]  

(1.70)

is a Banach space.

**Proof.** First of all we need to show that (1.70) is indeed a norm. If $\| [x] \| = 0$ we must have a sequence $y_j \in M$ with $y_j \to -x$ and since $M$ is closed we conclude $x \in M$, that is $[x] = [0]$ as required. To see $\| \alpha [x] \| = |\alpha| \|[x] \|$ we use again the definition

\[ \| \alpha [x] \| = \|[\alpha x] \| = \inf_{y \in M} \| \alpha x + y \| = \inf_{y \in M} \| \alpha x + \alpha y \| = |\alpha| \inf_{y \in M} \| x + y \| = |\alpha| \|[x] \|. \]

The triangle inequality follows with a similar argument and is left as an exercise.

Thus (1.70) is a norm and it remains to show that $X/M$ is complete. To this end let $[x_n]$ be a Cauchy sequence. Since it suffices to show that some subsequence has a limit, we can assume $\|[x_{n+1}] - [x_n]\| < 2^{-n}$ without loss of generality. Moreover, by definition of (1.70) we can chose the representatives
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Let \( x_n \) such that \( \|x_{n+1} - x_n\| < 2^{-n} \) (start with \( x_1 \) and then chose the remaining ones inductively). By construction \( x_n \) is a Cauchy sequence which has a limit \( x \in X \) since \( X \) is complete. Moreover, by \( \|[x_n] - [x]\| = \|[x_n - x]\| \leq \|x_n - x\| \) we see that \([x] \) is the limit of \([x_n] \). □

Observe that \( \|[x]\| = \text{dist}(x, M) = 0 \) whenever \( x \in M \) and hence we only get a semi-norm if \( M \) is not closed.

**Example.** If \( X := C[0,1] \) and \( M := \{ f \in X | f(0) = 0 \} \) then \( X/M \cong \mathbb{C} \). □

**Example.** If \( X := c(\mathbb{N}) \), the convergent sequences and \( M := c_0(\mathbb{N}) \) the sequences converging to 0, then \( X/M \cong \mathbb{C} \). In fact, note that every sequence \( x \in c(\mathbb{N}) \) can be written as \( x = y + \alpha e \) with \( y \in c_0(\mathbb{N}), e = (1, 1, 1, \ldots) \), and \( \alpha \in \mathbb{C} \) its limit. □

Note that by \( \|[x]\| \leq \|x\| \) the quotient map \( \pi : X \to X/M, x \mapsto [x] \) is bounded with norm at most one. As a small application we note:

**Corollary 1.15.** Let \( X \) be a Banach space and let \( M, N \subseteq X \) be two closed subspaces with one of them, say \( N \), finite dimensional. Then \( M + N \) is also closed.

**Proof.** If \( \pi : X \to X/M \) denotes the quotient map, then \( M + N = \pi^{-1}(\pi(N)) \). Moreover, since \( \pi(N) \) is finite dimensional it is closed and hence \( \pi^{-1}(\pi(N)) \) is closed by continuity. □

**Problem 1.34.** Let \( X_j, j = 1, \ldots, n \), be Banach spaces. Let \( X \) be the Cartesian product \( X_1 \times \cdots \times X_n \) together with the norm

\[
\|(x_1, \ldots, x_n)\|_p := \begin{cases} 
(\sum_{j=1}^n \|x_j\|^p)^{1/p}, & 1 \leq p < \infty, \\
\max_{j=1,\ldots,n} \|x_j\|, & p = \infty.
\end{cases}
\]

Show that \( X \) is a Banach space. Show that all norms are equivalent.

**Problem 1.35.** Let \( X_j, j \in \mathbb{N} \), be Banach spaces. Let \( X = \bigoplus_{p,j \in \mathbb{N}} X_j \) be the set of all elements \( x = (x_j)_{j \in \mathbb{N}} \) of the Cartesian product for which the norm

\[
\|x\|_p := \begin{cases} 
(\sum_{j \in \mathbb{N}} \|x_j\|^p)^{1/p}, & 1 \leq p < \infty, \\
\max_{j \in \mathbb{N}} \|x_j\|, & p = \infty,
\end{cases}
\]

is finite. Show that \( X \) is a Banach space. Show that for \( 1 \leq p < \infty \) the elements with finitely many nonzero terms are dense and conclude that \( X \) is separable if all \( X_j \) are.

**Problem 1.36.** Let \( \ell \) be a nontrivial linear functional. Then its kernel has codimension one.

**Problem 1.37.** Compute \( \|[e]\| \) in \( \ell^\infty(\mathbb{N})/c_0(\mathbb{N}) \), where \( e = (1, 1, 1, \ldots) \).
Problem 1.38. Suppose $A \in \mathcal{L}(X,Y)$. Show that $\text{Ker}(A)$ is closed. Suppose $M \subseteq \text{Ker}(A)$ is a closed subspace. Show that the induced map $\tilde{A} : X/M \to Y, \ [x] \mapsto Ax$ is a well-defined operator satisfying $\|\tilde{A}\| = \|A\|$ and $\text{Ker}(\tilde{A}) = \text{Ker}(A)/M$. In particular, $A$ is injective for $M = \text{Ker}(A)$.

Problem 1.39. Show that if a closed subspace $M$ of a Banach space $X$ has finite codimension, then it can be complemented. (Hint: Start with a basis $\{[x_j]\}$ for $X/M$ and choose a corresponding dual basis $\{\ell_k\}$ with $\ell_k([x_j]) = \delta_{j,k}$.)

1.7. Spaces of continuous and differentiable functions

In this section we introduce a few further sets of continuous and differentiable functions which are of interest in applications.

First, for any set $U \subseteq \mathbb{R}^m$ the set of all bounded continuous functions $C_b(U)$ together with the sup norm

$$\|f\|_\infty := \sup_{x \in U} |f(x)|$$

is a Banach space as can be shown as in Section 1.2 (or use Corollary 0.33). The space of continuous functions with compact support $C_c(U) \subseteq C_b(U)$ is in general not dense and its closure will be denoted by $C_0(U)$. If $U$ is open it can be interpreted as the functions in $C_b(U)$ which vanish at the boundary

$$C_0(U) := \{ f \in C(U) | \forall \varepsilon > 0, \exists K \subseteq U \text{ compact} : |f(x)| < \varepsilon, \ x \in U \setminus K \}. \quad (1.72)$$

Of course $\mathbb{R}^m$ could be replaced by any topological space up to this point.

Moreover, the above norm can be augmented to handle differentiable functions by considering the space $C^1_b(U)$ of all continuously differentiable functions for which the following norm

$$\|f\|_{\infty,1} := \|f\|_\infty + \sum_{j=1}^m \|\partial_j f\|_\infty$$

is finite, where $\partial_j = \frac{\partial}{\partial x_j}$. Note that $\|\partial_j f\|$ for one $j$ (or all $j$) is not sufficient as it is only a seminorm (it vanishes for every constant function). However, since the sum of seminorms is again a seminorm (Problem 1.41) the above expression defines indeed a norm. It is also not hard to see that $C^1_b(U)$ is complete. In fact, let $f^k$ be a Cauchy sequence, then $f^k(x)$ converges uniformly to some continuous function $f(x)$ and the same is true for the partial derivatives $\partial_j f^k(x) \to g_j(x)$. Moreover, since $f^k(x) = f^k(t, x_2, \ldots, x_m) + \int_c x^1 \partial_j f^k(t, x_2, \ldots, x_m)dt \to f(x) = f(t, x_2, \ldots, x_m) + \int_c x^1 g_j(t, x_2, \ldots, x_m)$ we obtain $\partial_j f(x) = g_j(x)$. The remaining derivatives follow analogously and thus $f^k \to f$ in $C^1_b(U)$. 
To extend this approach to higher derivatives let $C^k(U)$ be the set of all complex-valued functions which have partial derivatives of order up to $k$. For $f \in C^k(U)$ and $\alpha \in \mathbb{N}_0^n$ we set

$$\partial^\alpha f := \frac{\partial^{|\alpha|} f}{\partial x_1^{\alpha_1} \cdots \partial x_n^{\alpha_n}}, \quad |\alpha| = \alpha_1 + \cdots + \alpha_n.$$ (1.74)

An element $\alpha \in \mathbb{N}_0^n$ is called a multi-index and $|\alpha|$ is called its order. With this notation the above considerations can be easily generalized to higher order derivatives:

**Theorem 1.16.** Let $U \subseteq \mathbb{R}^m$ be open. The space $C^k_b(U)$ of all functions whose partial derivatives up to order $k$ are bounded and continuous form a Banach space with norm

$$\|f\|_{\infty,k} := \sum_{|\alpha| \leq k} \sup_{x \in U} |\partial^\alpha f(x)|.$$ (1.75)

An important subspace is $C^k_b(\mathbb{R}^m)$, the set of all functions in $C^k_b(\mathbb{R}^m)$ for which $\lim_{|x| \to \infty} |\partial^\alpha f(x)| = 0$ for all $|\alpha| \leq k$. For any function $f$ not in $C^k_b(\mathbb{R}^m)$ there must be a sequence $|x_j| \to \infty$ and some $\alpha$ such that $|\partial^\alpha f(x_j)| \geq \varepsilon > 0$. But then $\|f - g\|_{\infty,k} \geq \varepsilon$ for every $g$ in $C^k_b(\mathbb{R}^m)$ and thus $C^k_b(\mathbb{R}^m)$ is a closed subspace. In particular, it is a Banach space of its own.

Note that the space $C^k_b(U)$ could be further refined by requiring the highest derivatives to be Hölder continuous. Recall that a function $f : U \to \mathbb{C}$ is called uniformly Hölder continuous with exponent $\gamma \in (0, 1]$ if

$$[f]_{\gamma} := \sup_{x \neq y \in U} \frac{|f(x) - f(y)|}{|x - y|^\gamma}$$ (1.76)

is finite. Clearly, any Hölder continuous function is uniformly continuous and, in the special case $\gamma = 1$, we obtain the Lipschitz continuous functions.

**Example.** By the mean value theorem every function $f \in C^1_b(U)$ is Lipschitz continuous with $[f]_1 \leq \|\partial f\|_\infty$, where $\partial f = (\partial_1 f, \ldots, \partial_m f)$ denotes the gradient.

**Example.** The prototypical example of a Hölder continuous function is of course $f(x) = x^\gamma$ on $[0, \infty)$ with $\gamma \in (0, 1]$. In fact, without loss of generality we can assume $0 \leq x < y$ and set $t = \frac{x}{y} \in [0, 1)$. Then we have

$$\frac{y^\gamma - x^\gamma}{(y - x)^\gamma} \leq \frac{1 - t^\gamma}{(1 - t)^\gamma} \leq \frac{1 - t}{1 - t} = 1.$$

From this one easily gets further examples since the composition of two Hölder continuous functions is again Hölder continuous (the exponent being the product).
It is easy to verify that this is a seminorm and that the corresponding space is complete.

**Theorem 1.17.** The space $C^{k,\gamma}_b(U)$ of all functions whose partial derivatives up to order $k$ are bounded and Hölder continuous with exponent $\gamma \in (0, 1]$ form a Banach space with norm

$$
\|f\|_{\infty,k,\gamma} := \|f\|_{\infty,k} + \sum_{|\alpha|=k} |\partial_\alpha f| \gamma.
$$

Note that by the mean value theorem all derivatives up to order lower than $k$ are automatically Lipschitz continuous. Moreover, every Hölder continuous function is uniformly continuous and hence has a unique extension to the closure $\overline{U}$ (cf. Theorem 0.36). In this sense, the spaces $C^{0,\gamma}_b(U)$ and $C^{k,\gamma}_b(U)$ are naturally isomorphic. Consequently, we can also understand $C^{k,\gamma}_b(U)$ in this fashion since for a function from $C^{k,\gamma}_b(U)$ all derivatives have a continuous extension to $\overline{U}$. For a function in $C^k_b(U)$ this only works for the derivatives of order up to $k - 1$ and hence we define $C^k_b(U)$ as the functions from $C^k_b(U)$ for which all derivatives have a continuous extensions to $\overline{U}$. Note that with this definition $C^k_b(U)$ is still a Banach space (since $C^0_b(U)$ is a closed subspace of $C^0_b(U)$).

While the above spaces are able to cover a wide variety of situations, there are still cases where the above definitions are not suitable. In fact, for some of these cases one cannot define a suitable norm and we will postpone this to Section 5.4.

Note that in all the above spaces we could replace complex-valued by $\mathbb{C}^n$-valued functions.

**Problem 1.40.** Suppose $f : [a, b] \to \mathbb{C}$ is Hölder continuous with exponent $\gamma > 1$. Show that $f$ is constant.

**Problem 1.41.** Suppose $X$ is a vector space and $\|\cdot\|_j$, $1 \leq j \leq n$, is a finite family of seminorms. Show that $\|x\| := \sum_{j=1}^n \|x\|_j$ is a seminorm. It is a norm if and only if $\|x\|_j = 0$ for all $j$ implies $x = 0$.

**Problem 1.42.** Show that $C_0(U)$ is a Banach space when equipped with the sup norm. Show that $C^k(U) = C^k_0(U)$. (Hint: The function $m_\varepsilon(z) = \text{sign}(z) \max(0, |z| - \varepsilon) \in C(\mathbb{C})$ might be useful.)

**Problem 1.43.** Suppose $U$ is bounded. Show $C^{k,\gamma_2}_b(U) \subseteq C^{k,\gamma_1}_b(U) \subseteq C^k_b(U)$ for $0 < \gamma_1 < \gamma_2 \leq 1$.

**Problem 1.44.** Show that the product of two bounded Hölder continuous functions is again Hölder continuous with

$$
[f g]_\gamma \leq \|f\|_\infty [g]_\gamma + [f]_\gamma \|g\|_\infty.
$$
1. A first look at Banach and Hilbert spaces
Chapter 2

Hilbert spaces

2.1. Orthonormal bases

In this section we will investigate orthonormal series and you will notice hardly any difference between the finite and infinite dimensional cases. Throughout this chapter \( \mathcal{H} \) will be a (complex) Hilbert space.

As our first task, let us generalize the projection into the direction of one vector:

A set of vectors \( \{u_j\} \) is called an **orthonormal set** if \( \langle u_j, u_k \rangle = 0 \) for \( j \neq k \) and \( \langle u_j, u_j \rangle = 1 \). Note that every orthonormal set is linearly independent (show this).

**Lemma 2.1.** Suppose \( \{u_j\}_{j=1}^n \) is an orthonormal set. Then every \( f \in \mathcal{H} \) can be written as

\[
f = f_\parallel + f_\perp, \quad f_\parallel = \sum_{j=1}^n \langle u_j, f \rangle u_j, \tag{2.1}
\]

where \( f_\parallel \) and \( f_\perp \) are orthogonal. Moreover, \( \langle u_j, f_\perp \rangle = 0 \) for all \( 1 \leq j \leq n \). In particular,

\[
\|f\|^2 = \sum_{j=1}^n |\langle u_j, f \rangle|^2 + \|f_\perp\|^2. \tag{2.2}
\]

Moreover, every \( \hat{f} \) in the span of \( \{u_j\}_{j=1}^n \) satisfies

\[
\|f - \hat{f}\| \geq \|f_\perp\| \quad \tag{2.3}
\]

with equality holding if and only if \( \hat{f} = f_\parallel \). In other words, \( f_\parallel \) is uniquely characterized as the vector in the span of \( \{u_j\}_{j=1}^n \) closest to \( f \).
Proof. A straightforward calculation shows that $\langle u_j, f - f\rangle = 0$ and hence $f\|$ and $f_\perp = f - f\|$ are orthogonal. The formula for the norm follows by applying (1.44) iteratively.

Now, fix a vector
$$\hat{f} = \sum_{j=1}^{n} \alpha_j u_j$$
in the span of $\{u_j\}_{j=1}^{n}$. Then one computes
$$\|f - \hat{f}\|^2 = \|f\| + \|f_\perp - \hat{f}\|^2 = \|f\|^2 + \|f_\perp - \hat{f}\|^2$$
$$= \|f_\perp\|^2 + \sum_{j=1}^{n} |\alpha_j - \langle u_j, f \rangle|^2$$
from which the last claim follows. \qed

From (2.2) we obtain Bessel’s inequality
$$\sum_{j=1}^{n} |\langle u_j, f \rangle|^2 \leq \|f\|^2$$
with equality holding if and only if $f$ lies in the span of $\{u_j\}_{j=1}^{n}$.

Of course, since we cannot assume $\mathcal{H}$ to be a finite dimensional vector space, we need to generalize Lemma 2.1 to arbitrary orthonormal sets $\{u_j\}_{j \in J}$. We start by assuming that $J$ is countable. Then Bessel’s inequality (2.4) shows that
$$\sum_{j \in J} |\langle u_j, f \rangle|^2$$
converges absolutely. Moreover, for any finite subset $K \subset J$ we have
$$\|\sum_{j \in K} \langle u_j, f \rangle u_j\|^2 = \sum_{j \in K} |\langle u_j, f \rangle|^2$$
by the Pythagorean theorem and thus $\sum_{j \in J} \langle u_j, f \rangle u_j$ is a Cauchy sequence if and only if $\sum_{j \in J} |\langle u_j, f \rangle|^2$ is. Now let $J$ be arbitrary. Again, Bessel’s inequality shows that for any given $\varepsilon > 0$ there are at most finitely many $j$ for which $|\langle u_j, f \rangle| \geq \varepsilon$ (namely at most $\|f\|/\varepsilon$). Hence there are at most countably many $j$ for which $|\langle u_j, f \rangle| > 0$. Thus it follows that
$$\sum_{j \in J} |\langle u_j, f \rangle|^2$$
is well defined (as a countable sum over the nonzero terms) and (by completeness) so is
$$\sum_{j \in J} \langle u_j, f \rangle u_j.$$ Furthermore, it is also independent of the order of summation.
In particular, by continuity of the scalar product we see that Lemma 2.1 can be generalized to arbitrary orthonormal sets.

**Theorem 2.2.** Suppose \( \{u_j\}_{j \in J} \) is an orthonormal set in a Hilbert space \( \mathcal{H} \). Then every \( f \in \mathcal{H} \) can be written as

\[
f = f_\parallel + f_\perp = \sum_{j \in J} \langle u_j, f \rangle u_j,
\]

where \( f_\parallel \) and \( f_\perp \) are orthogonal. Moreover, \( \langle u_j, f_\perp \rangle = 0 \) for all \( j \in J \). In particular,

\[
\|f\|^2 = \sum_{j \in J} |\langle u_j, f \rangle|^2 + \|f_\perp\|^2.
\]

Furthermore, every \( \hat{f} \in \text{span}\{u_j\}_{j \in J} \) satisfies

\[
\|f - \hat{f}\| \geq \|f_\parallel\|
\]

with equality holding if and only if \( \hat{f} = f_\parallel \). In other words, \( f_\parallel \) is uniquely characterized as the vector in \( \text{span}\{u_j\}_{j \in J} \) closest to \( f \).

**Proof.** The first part follows as in Lemma 2.1 using continuity of the scalar product. The same is true for the last part except for the fact that every \( f \in \text{span}\{u_j\}_{j \in J} \) can be written as \( f = \sum_{j \in J} \alpha_j u_j \) (i.e., \( f = f_\parallel \)). To see this, let \( f_n \in \text{span}\{u_j\}_{j \in J} \) converge to \( f \). Then \( \|f - f_n\|^2 = \|f_\parallel - f_n\|^2 + \|f_\perp\|^2 \to 0 \) implies \( f_n \to f_\parallel \) and \( f_\perp = 0 \).

Note that from Bessel’s inequality (which of course still holds), it follows that the map \( f \to f_\parallel \) is continuous.

Of course we are particularly interested in the case where every \( f \in \mathcal{H} \) can be written as \( \sum_{j \in J} \langle u_j, f \rangle u_j \). In this case we will call the orthonormal set \( \{u_j\}_{j \in J} \) an **orthonormal basis** (ONB).

If \( \mathcal{H} \) is separable it is easy to construct an orthonormal basis. In fact, if \( \mathcal{H} \) is separable, then there exists a countable total set \( \{f_j\}_{j=1}^N \). Here \( N \in \mathbb{N} \) if \( \mathcal{H} \) is finite dimensional and \( N = \infty \) otherwise. After throwing away some vectors, we can assume that \( f_{n+1} \) cannot be expressed as a linear combination of the vectors \( f_1, \ldots, f_n \). Now we can construct an orthonormal set as follows: We begin by normalizing \( f_1 \):

\[
u_1 := \frac{f_1}{\|f_1\|}.
\]

Next we take \( f_2 \) and remove the component parallel to \( u_1 \) and normalize again:

\[
u_2 := \frac{f_2 - \langle u_1, f_2 \rangle u_1}{\|f_2 - \langle u_1, f_2 \rangle u_1\|}.
\]
Proceeding like this, we define recursively
\[ u_n := \frac{f_n - \sum_{j=1}^{n-1} (u_j, f_n) u_j}{\|f_n - \sum_{j=1}^{n-1} (u_j, f_n) u_j\|.} \] (2.14)

This procedure is known as \textbf{Gram–Schmidt orthogonalization}. Hence we obtain an orthonormal set \( \{u_j\}_{j=1}^N \) such that \( \text{span}\{u_j\}_{j=1}^n = \text{span}\{f_j\}_{j=1}^n \) for any finite \( n \) and thus also for \( n = N \) (if \( N = \infty \)). Since \( \{f_j\}_{j=1}^N \) is total, so is \( \{u_j\}_{j=1}^N \). Now suppose there is some \( f = f_\parallel + f_\perp \in \mathcal{H} \) for which \( f_\perp \neq 0 \). Since \( \{u_j\}_{j=1}^N \) is total, we can find a \( \hat{f} \) in its span such that \( \|f - \hat{f}\| < \|f_\perp\| \), contradicting (2.11). Hence we infer that \( \{u_j\}_{j=1}^N \) is an orthonormal basis.

\textbf{Theorem 2.3.} Every separable Hilbert space has a countable orthonormal basis.

\textbf{Example.} In \( L^2_{\text{cont}}(-1,1) \), we can orthogonalize the monomials \( f_n(x) = x^n \) (which are total by the Weierstrass approximation theorem — Theorem 1.3). The resulting polynomials are up to a normalization equal to the Legendre polynomials
\[ P_0(x) = 1, \quad P_1(x) = x, \quad P_2(x) = \frac{3x^2 - 1}{2}, \quad \ldots \] (2.15)
(which are normalized such that \( P_n(1) = 1 \)).

\textbf{Example.} The set of functions
\[ u_n(x) = \frac{1}{\sqrt{2\pi}} e^{inx}, \quad n \in \mathbb{Z}, \] (2.16)
forms an orthonormal basis for \( \mathcal{H} = L^2_{\text{cont}}(0,2\pi) \). The corresponding orthogonal expansion is just the ordinary Fourier series. We will discuss this example in detail in Section 2.5.

The following equivalent properties also characterize a basis.

\textbf{Theorem 2.4.} For an orthonormal set \( \{u_j\}_{j \in J} \) in a Hilbert space \( \mathcal{H} \), the following conditions are equivalent:

(i) \( \{u_j\}_{j \in J} \) is a maximal orthogonal set.

(ii) For every vector \( f \in \mathcal{H} \) we have
\[ f = \sum_{j \in J} (u_j, f) u_j. \] (2.17)

(iii) For every vector \( f \in \mathcal{H} \) we have Parseval’s relation
\[ \|f\|^2 = \sum_{j \in J} |(u_j, f)|^2. \] (2.18)

(iv) \( (u_j, f) = 0 \) for all \( j \in J \) implies \( f = 0 \).
Proof. We will use the notation from Theorem 2.2.

(i) ⇒ (ii): If \( f_\perp \neq 0 \), then we can normalize \( f_\perp \) to obtain a unit vector \( \tilde{f}_\perp \) which is orthogonal to all vectors \( u_j \). But then \( \{u_j\}_{j \in J} \cup \{\tilde{f}_\perp\} \) would be a larger orthonormal set, contradicting the maximality of \( \{u_j\}_{j \in J} \).

(ii) ⇒ (iii): This follows since (ii) implies \( f_\perp = 0 \).

(iii) ⇒ (iv): If \( \langle f, u_j \rangle = 0 \) for all \( j \in J \), we conclude \( \|f\|^2 = 0 \) and hence \( f = 0 \).

(iv) ⇒ (i): If \( \{u_j\}_{j \in J} \) were not maximal, there would be a unit vector \( g \) such that \( \{u_j\}_{j \in J} \cup \{g\} \) is a larger orthonormal set. But \( \langle u_j, g \rangle = 0 \) for all \( j \in J \) implies \( g = 0 \) by (iv), a contradiction. \( \Box \)

By continuity of the norm it suffices to check (iii), and hence also (ii), for \( f \) in a dense set. In fact, by the inverse triangle inequality for \( \ell^2(\mathbb{N}) \) and the Bessel inequality we have

\[
\sum_{j \in J} |\langle u_j, f \rangle|^2 - \sum_{j \in J} |\langle u_j, g \rangle|^2 \leq \sqrt{\sum_{j \in J} |\langle u_j, f - g \rangle|^2} \sqrt{\sum_{j \in J} |\langle u_j, f + g \rangle|^2} \\
\leq \|f - g\| \|f + g\|
\]

implying \( \sum_{j \in J} |\langle u_j, f_n \rangle|^2 \to \sum_{j \in J} |\langle u_j, f \rangle|^2 \) if \( f_n \to f \).

It is not surprising that if there is one countable basis, then it follows that every other basis is countable as well.

**Theorem 2.5.** In a Hilbert space \( \mathcal{H} \) every orthonormal basis has the same cardinality.

Proof. Let \( \{u_j\}_{j \in J} \) and \( \{v_k\}_{k \in K} \) be two orthonormal bases. We first look at the case where one of them, say the first, is finite dimensional: \( J = \{1, \ldots, n\} \). Suppose the other basis has at least \( n \) elements \( \{1, \ldots, n\} \subseteq K \). Then \( v_k = \sum_{j=1}^n U_{k,j} u_j \), where \( U_{k,j} = \langle u_j, v_k \rangle \). By \( \delta_{j,k} = \langle v_j, v_k \rangle = \sum_{l=1}^n U_{j,l} U_{k,l} \) we see \( u_j = \sum_{k=1}^n U_{k,j}^* v_k \) showing that \( K \) cannot have more than \( n \) elements.

Now let us turn to the case where both \( J \) and \( K \) are infinite. Set \( K_j = \{k \in K | \langle v_k, u_j \rangle \neq 0 \} \). Since these are the expansion coefficients of \( u_j \) with respect to \( \{v_k\}_{k \in K} \), this set is countable (and nonempty). Hence the set \( \tilde{K} = \bigcup_{j \in J} K_j \) satisfies \( |\tilde{K}| \leq |J \times \mathbb{N}| = |J| \) (Theorem A.9) But \( k \in K \setminus \tilde{K} \) implies \( v_k = 0 \) and hence \( \tilde{K} = K \). So \( |K| \leq |J| \) and reversing the roles of \( J \) and \( K \) shows \( |K| = |J| \). \( \Box \)

The cardinality of an orthonormal basis is also called the Hilbert space dimension of \( \mathcal{H} \).

It even turns out that, up to unitary equivalence, there is only one separable infinite dimensional Hilbert space:
A bijective linear operator $U \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ is called \textbf{unitary} if $U$ preserves scalar products:
\begin{equation}
\langle Ug, Uf \rangle_2 = \langle g, f \rangle_1, \quad g, f \in \mathcal{H}_1.
\end{equation}
By the polarization identity, (1.51) this is the case if and only if $U$ preserves norms: $\|Uf\|_2 = \|f\|_1$ for all $f \in \mathcal{H}_1$ (note the a norm preserving linear operator is automatically injective). The two Hilbert spaces $\mathcal{H}_1$ and $\mathcal{H}_2$ are called \textbf{unitarily equivalent} in this case.

Let $\mathcal{H}$ be a separable infinite dimensional Hilbert space and let $\{u_j\}_{j \in \mathbb{N}}$ be any orthogonal basis. Then the map $U : \mathcal{H} \to \ell^2(\mathbb{N})$, $f \mapsto (\langle u_j, f \rangle)_{j \in \mathbb{N}}$ is unitary. Indeed by Theorem 2.4 (iii) it is norm preserving and hence injective. To see that it is onto, let $a \in \ell^2(\mathbb{N})$ and observe that by $\| \sum_{j=m}^{n} a_j u_j \|_2^2 = \sum_{j=m}^{n} |a_j|^2$ the vector $f := \sum_{j \in \mathbb{N}} a_j u_j$ is well defined and satisfies $a_j = \langle u_j, f \rangle$. In particular,

\textbf{Theorem 2.6.} Any separable infinite dimensional Hilbert space is unitarily equivalent to $\ell^2(\mathbb{N})$.

Of course the same argument shows that every finite dimensional Hilbert space of dimension $n$ is unitarily equivalent to $\mathbb{C}^n$ with the usual scalar product.

Finally we briefly turn to the case where $\mathcal{H}$ is not separable.

\textbf{Theorem 2.7.} Every Hilbert space has an orthonormal basis.

\textbf{Proof.} To prove this we need to resort to Zorn’s lemma (see Appendix A): The collection of all orthonormal sets in $\mathcal{H}$ can be partially ordered by inclusion. Moreover, every linearly ordered chain has an upper bound (the union of all sets in the chain). Hence Zorn’s lemma implies the existence of a maximal element, that is, an orthonormal set which is not a proper subset of every other orthonormal set.

Hence, if $\{u_j\}_{j \in J}$ is an orthogonal basis, we can show that $\mathcal{H}$ is unitarily equivalent to $\ell^2(J)$ and, by prescribing $J$, we can find a Hilbert space of any given dimension. Here $\ell^2(J)$ is the set of all complex valued functions $(a_j)_{j \in J}$ where at most countably many values are nonzero and $\sum_{j \in J} |a_j|^2 < \infty$.

\textbf{Example.} Define the set of \textbf{almost periodic functions} $\text{AP}(\mathbb{R})$ as the closure of the set of trigonometric polynomials
\[ f(t) = \sum_{k=1}^{n} \alpha_k e^{i \theta_k t}, \quad \alpha_k \in \mathbb{C}, \ \theta_k \in \mathbb{R}, \]
with respect to the sup norm. In particular $\text{AP}(\mathbb{R}) \subset C_b(\mathbb{R})$ is a Banach space when equipped with the sup norm. Since the trigonometric polynomials form an algebra, it is even a Banach algebra. Using the Stone–Weierstraß...
theorem one can verify that every periodic function is almost periodic (make
the approximation on one period and note that you get the rest of $\mathbb{R}$ for free
from periodicity) but the converse is not true (e.g. $e^{it} + e^{i\sqrt{2}t}$ is not periodic).

It is not difficult to show that every almost periodic function has a mean value

$$M(f) := \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} f(t) dt$$

and one can show that

$$\langle f, g \rangle := M(f^* g)$$

defines a scalar product on $AP(\mathbb{R})$ (only positivity is nontrivial and it will
not be shown here). Note that $\|f\| \leq \|f\|_\infty$. Abbreviating $e_\theta(t) = e^{i\theta t}$ one
computes $M(e_\theta) = 0$ if $\theta \neq 0$ and $M(e_0) = 1$. In particular, \{e_\theta\}_{\theta \in \mathbb{R}} is an
uncountable orthonormal set and

$$f(t) \mapsto \hat{f}(\theta) := \langle e_\theta, f \rangle = M(e_{-\theta} f)$$

maps $AP(\mathbb{R})$ isometrically (with respect to $\|\cdot\|$) into $\ell^2(\mathbb{R})$. This map is
however not surjective (take e.g. a Fourier series which converges in mean square but not uniformly — see later).

Problem 2.1. Given some vectors $f_1, \ldots, f_n$ we define their Gram deter-
minant as

$$\Gamma(f_1, \ldots, f_n) := \det(\langle f_j, f_k \rangle)_{1 \leq j, k \leq n}.$$ 

Show that the Gram determinant is nonzero if and only if the vectors are
linearly independent. Moreover, show that in this case

$$\text{dist}(g, \text{span}\{f_1, \ldots, f_n\})^2 = \frac{\Gamma(f_1, \ldots, f_n, g)}{\Gamma(f_1, \ldots, f_n)}$$

and

$$\Gamma(f_1, \ldots, f_n) \leq \prod_{j=1}^{n} \|f_j\|^2.$$ 

with equality if the vectors are orthogonal. (Hint: How does $\Gamma$ change when
you apply the Gram–Schmidt procedure?)

Problem 2.2. Let \{u_j\} be some orthonormal basis. Show that a bounded
linear operator $A$ is uniquely determined by its matrix elements $A_{jk} :=
\langle u_j, Au_k \rangle$ with respect to this basis.

Problem 2.3. Give an example of a nonempty closed bounded subset of a
Hilbert space which does not contain an element with minimal norm.
2.2. The projection theorem and the Riesz lemma

Let $M \subseteq \mathcal{H}$ be a subset. Then $M^\perp = \{ f | \langle g, f \rangle = 0, \forall g \in M \}$ is called the **orthogonal complement** of $M$. By continuity of the scalar product it follows that $M^\perp$ is a closed linear subspace and by linearity that $(\text{span}(M))^\perp = M^\perp$. For example, we have $\mathcal{H}^\perp = \{ 0 \}$ since any vector in $\mathcal{H}^\perp$ must be in particular orthogonal to all vectors in some orthonormal basis.

**Theorem 2.8** (Projection theorem). Let $M$ be a closed linear subspace of a Hilbert space $\mathcal{H}$. Then every $f \in \mathcal{H}$ can be uniquely written as $f = f_\parallel + f_\perp$ with $f_\parallel \in M$ and $f_\perp \in M^\perp$. One writes

$$M \oplus M^\perp = \mathcal{H}$$

in this situation.

**Proof.** Since $M$ is closed, it is a Hilbert space and has an orthonormal basis $\{u_j\}_{j \in J}$. Hence the existence part follows from Theorem 2.2. To see uniqueness, suppose there is another decomposition $f = \tilde{f}_\parallel + \tilde{f}_\perp$. Then $f_\parallel - \tilde{f}_\parallel = \tilde{f}_\perp - f_\perp \in M \cap M^\perp = \{ 0 \}$. \hfill $\square$

**Corollary 2.9.** Every orthogonal set $\{u_j\}_{j \in J}$ can be extended to an orthonormal basis.

**Proof.** Just add an orthogonal basis for $\left(\{u_j\}_{j \in J}\right)^\perp$. \hfill $\square$

Moreover, Theorem 2.8 implies that to every $f \in \mathcal{H}$ we can assign a unique vector $f_\parallel$ which is the vector in $M$ closest to $f$. The rest, $f - f_\parallel$, lies in $M^\perp$. The operator $P_M f := f_\parallel$ is called the **orthogonal projection** corresponding to $M$. Note that we have

$$P_M^2 = P_M$$

and

$$\langle P_M g, f \rangle = \langle g, P_M f \rangle$$

(2.22)

since $\langle P_M g, f \rangle = \langle g_\parallel, f_\parallel \rangle = \langle g, P_M f \rangle$. Clearly we have $P_M f = f - P_M^\perp f = f_\perp$. Furthermore, (2.22) uniquely characterizes orthogonal projections (Problem 2.6).

Moreover, if $M$ is a closed subspace, we have $P_{M^\perp} = I - P_M = I - (I - P_M) = P_M$; that is, $M^{\perp \perp} = M$. If $M$ is an arbitrary subset, we have at least

$$M^{\perp \perp} = \text{span}(M).$$

(2.23)

Note that by $\mathcal{H}^\perp = \{ 0 \}$ we see that $M^\perp = \{ 0 \}$ if and only if $M$ is total.

Finally we turn to **linear functionals**, that is, to operators $\ell : \mathcal{H} \to \mathbb{C}$. By the Cauchy–Schwarz inequality we know that $\ell_g : f \mapsto \langle g, f \rangle$ is a bounded linear functional (with norm $\|g\|$). In turns out that, in a Hilbert space, every bounded linear functional can be written in this way.
2.2. The projection theorem and the Riesz lemma

Theorem 2.10 (Riesz lemma). Suppose \( \ell \) is a bounded linear functional on a Hilbert space \( \mathcal{H} \). Then there is a unique vector \( g \in \mathcal{H} \) such that \( \ell(f) = \langle g, f \rangle \) for all \( f \in \mathcal{H} \).

In other words, a Hilbert space is equivalent to its own dual space \( \mathcal{H}^* \cong \mathcal{H} \) via the map \( f \mapsto \langle \cdot, f \rangle \) which is a conjugate linear isometric bijection between \( \mathcal{H} \) and \( \mathcal{H}^* \).

Proof. If \( \ell \equiv 0 \), we can choose \( g = 0 \). Otherwise \( \text{Ker}(\ell) = \{ f | \ell(f) = 0 \} \) is a proper subspace and we can find a unit vector \( \tilde{g} \in \text{Ker}(\ell)^\perp \). For every \( f \in \mathcal{H} \), we have \( \ell(f)\tilde{g} - \ell(\tilde{g})f \in \text{Ker}(\ell) \) and hence
\[
0 = \langle \tilde{g}, \ell(f)\tilde{g} - \ell(\tilde{g})f \rangle = \ell(f) - \ell(\tilde{g})\langle \tilde{g}, f \rangle.
\]
In other words, we can choose \( g = \ell(\tilde{g})^*\tilde{g} \). To see uniqueness, let \( g_1 \), \( g_2 \) be two such vectors. Then \( \langle g_1 - g_2, f \rangle = \langle g_1, f \rangle - \langle g_2, f \rangle = \ell(f) - \ell(f) = 0 \) for every \( f \in \mathcal{H} \), which shows \( g_1 - g_2 \in \mathcal{H}^\perp = \{ 0 \} \).

In particular, this shows that \( \mathcal{H}^* \) is again a Hilbert space whose scalar product (in terms of the above identification) is given by \( \langle \langle f, \cdot \rangle, \langle g, \cdot \rangle \rangle_{\mathcal{H}^*} = \langle f, g \rangle^* \).

We can even get a unitary map between \( \mathcal{H} \) and \( \mathcal{H}^* \) but such a map is not unique. To this end note that every Hilbert space has a conjugation \( C \) which generalizes taking the complex conjugate of every coordinate. In fact, choosing an orthonormal basis (and different choices will produce different maps in general) we can set
\[
Cf := \sum_{j \in J} \langle u_j, f \rangle^*u_j = \sum_{j \in J} \langle f, u_j \rangle u_j.
\]
Then \( C \) is conjugate linear, isometric \( \| Cf \| = \| f \| \), and idempotent \( C^2 = I \). Note also \( \langle Cf, Cg \rangle = \langle f, g \rangle^* \). As promised, the map \( f \rightarrow \langle Cf, \cdot \rangle \) is a unitary map from \( \mathcal{H} \) to \( \mathcal{H}^* \).

Problem 2.4. Suppose \( U : \mathcal{H} \rightarrow \mathcal{H} \) is unitary and \( M \subseteq \mathcal{H} \). Show that \( UM^\perp = (UM)^\perp \).

Problem 2.5. Show that an orthogonal projection \( P_M \neq 0 \) has norm one.

Problem 2.6. Suppose \( P \in \mathcal{L}(\mathcal{H}) \) satisfies
\[
P^2 = P \quad \text{and} \quad \langle Pf, g \rangle = \langle f, Pg \rangle
\]
and set \( M = \text{Ran}(P) \). Show
\begin{itemize}
  \item \( Pf = f \) for \( f \in M \) and \( M \) is closed,
  \item \( g \in M^\perp \) implies \( Pg \in M^\perp \) and thus \( Pg = 0 \),
\end{itemize}
and conclude \( P = P_M \). In particular

\[
\mathcal{H} = \ker(P) \oplus \operatorname{ran}(P), \quad \ker(P) = (I - P)\mathcal{H}, \ \operatorname{ran}(P) = P\mathcal{H}.
\]

2.3. Operators defined via forms

One of the key results about linear maps is that they are uniquely determined once we know the images of some basis vectors. In fact, the matrix elements with respect to some basis uniquely determine a linear map. Clearly this raises the question how this results extends to the infinite dimensional setting. As a first result we show that the Riesz lemma, Theorem 2.10, implies that a bounded operator \( A \) is uniquely determined by its associated sesquilinear form \( \langle g, Af \rangle \). In fact, there is a one-to-one correspondence between bounded operators and bounded sesquilinear forms:

**Lemma 2.11.** Suppose \( s : \mathcal{H}_1 \times \mathcal{H}_2 \to \mathbb{C} \) is a bounded sesquilinear form; that is,

\[
|s(g, f)| \leq C \|g\|_{\mathcal{H}_2} \|f\|_{\mathcal{H}_1}.
\]  

Then there is a unique bounded operator \( A \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2) \) such that

\[
s(g, f) = \langle g, Af \rangle_{\mathcal{H}_2}.
\]

Moreover, the norm of \( A \) is given by

\[
\|A\| = \sup_{\|g\|_{\mathcal{H}_2} = 1} \|\langle g, Af \rangle_{\mathcal{H}_2}\| \leq C.
\]  

**Proof.** For every \( f \in \mathcal{H}_1 \) we have an associated bounded linear functional \( \ell_f(g) := s(g, f)^* \) on \( \mathcal{H}_2 \). By Theorem 2.10 there is a corresponding \( h \in \mathcal{H}_2 \) (depending on \( f \)) such that \( \ell_f(g) = \langle h, g \rangle_{\mathcal{H}_2} \), that is \( s(g, f) = \langle g, h \rangle_{\mathcal{H}_2} \) and we can define \( A \) via \( Af := h \). It is not hard to check that \( A \) is linear and from

\[
\|Af\|_{\mathcal{H}_2}^2 = \langle Af, Af \rangle_{\mathcal{H}_2} = s(Af, f) \leq C \|Af\|_{\mathcal{H}_2} \|f\|_{\mathcal{H}_1}
\]

we infer \( \|Af\|_{\mathcal{H}_2} \leq C \|f\|_{\mathcal{H}_1} \), which shows that \( A \) is bounded with \( \|A\| \leq C \). Equation (2.26) is left as an exercise (Problem 2.9). \( \square \)

Note that if \( \{u_k\}_{k \in K} \subseteq \mathcal{H}_1 \) and \( \{v_j\}_{j \in J} \subseteq \mathcal{H}_2 \) are some orthogonal bases, then the matrix elements \( A_{j,k} := \langle v_j, Au_k \rangle_{\mathcal{H}_2} \) for all \( (j, k) \in J \times K \) uniquely determine \( \langle g, Af \rangle_{\mathcal{H}_2} \) for arbitrary \( f \in \mathcal{H}_1, g \in \mathcal{H}_2 \) (just expand \( f, g \) with respect to these bases) and thus \( A \) by our theorem.

**Example.** Consider \( \ell^2(\mathbb{N}) \) and let \( A \in \mathcal{L}(\ell^2(\mathbb{N})) \) be some bounded operator. Let \( A_{j,k} = \langle \delta^j, A\delta^k \rangle \) be its matrix elements such that

\[
(Aa)_j = \sum_{k=1}^\infty A_{j,k} a_k.
\]
Here the sum converges in $\ell^2(\mathbb{N})$ and hence, in particular, for every fixed $j$. Moreover, choosing $a_k^n = \alpha_n A_{jk}$ for $k \leq n$ and $a_k^n = 0$ for $k > n$ with $\alpha_n = (\sum_{j=1}^n |A_{jk}|^2)^{1/2}$ we see $\alpha_n = |(Aa)^n|_j \leq \|A\|\|a^n\| = \|A\|$. Thus
\[
\sum_{j=1}^\infty |A_{jk}|^2 \leq \|A\|^2
\]
and the sum is even absolutely convergent. \hfill \checkmark

Moreover, for $A \in \mathcal{L}(\mathfrak{H})$ the polarization identity (Problem 1.19) implies that $A$ is already uniquely determined by its quadratic form $q_A(f) := \langle f, Af \rangle$.

As a first application we introduce the adjoint operator via Lemma 2.11 as the operator associated with the sesquilinear form $s(f, g) := \langle Af, g \rangle_{\mathcal{H}^2}$.

**Theorem 2.12.** For every bounded operator $A \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ there is a unique bounded operator $A^* \in \mathcal{L}(\mathcal{H}_2, \mathcal{H}_1)$ defined via
\[
\langle f, A^* g \rangle_{\mathcal{H}_1} = \langle Af, g \rangle_{\mathcal{H}_2}.
\] (2.27)

A bounded operator $A \in \mathcal{L}(\mathfrak{H})$ satisfying $A^* = A$ is called self-adjoint. Note that $q_{A^*}(f) = \langle Af, f \rangle = q_A(f)^*$ and hence a bounded operator is self-adjoint if and only if its quadratic form is real-valued.

**Example.** If $\mathfrak{H} := \mathbb{C}^n$ and $A := (a_{jk})_{1 \leq j,k \leq n}$, then $A^* = (a^*_{kj})_{1 \leq j,k \leq n}$. \hfill \checkmark

**Example.** If $\mathbb{I} \in \mathcal{L}(\mathfrak{H})$ is the identity, then $\mathbb{I}^* = \mathbb{I}$. \hfill \checkmark

**Example.** Consider the linear functional $\ell : \mathfrak{H} \to \mathbb{C}$, $f \mapsto \langle g, f \rangle$. Then by the definition $\langle f, \ell^* \alpha \rangle = \ell(f)^* \alpha = \langle f, \alpha g \rangle$ we obtain $\ell^* : \mathbb{C} \to \mathfrak{H}$, $\alpha \mapsto \alpha g$. \hfill \checkmark

**Example.** Let $\mathfrak{H} := \ell^2(\mathbb{N})$, $a \in \ell^\infty(\mathbb{N})$ and consider the multiplication operator
\[
(Ab)_{j} := a_j b_j.
\]
Then
\[
\langle Ab, c \rangle = \sum_{j=1}^\infty (a_j b_j)^* c_j = \sum_{j=1}^\infty b_j^* (a^*_j c_j) = \langle b, A^* c \rangle
\]
with $(A^* c)_j = a^*_j c_j$, that is, $A^*$ is the multiplication operator with $a^*$. \hfill \checkmark

**Example.** Let $\mathfrak{H} := \ell^2(\mathbb{N})$ and consider the shift operators defined via
\[
(S^\pm a)_{j} := a_{j \pm 1}
\]
with the convention that $a_0 = 0$. That is, $S^-$ shifts a sequence to the right and fills up the left most place by zero and $S^+$ shifts a sequence to the left dropping the left most place:
\[
S^-(a_1, a_2, a_3, \cdots) = (0, a_1, a_2, \cdots), \quad S^+(a_1, a_2, a_3, \cdots) = (a_2, a_3, a_4, \cdots).
\]
Then
\[ \langle S^- a, b \rangle = \sum_{j=1}^{\infty} a_j^* b_{j+1} = \sum_{j=2}^{\infty} a_{j-1}^* b_j = \langle a, S^+ b \rangle, \]
which shows that \((S^-)^* = S^+\). Using symmetry of the scalar product we also get \(\langle b, S^- a \rangle = \langle S^+ b, a \rangle\), that is, \((S^+)^* = S^-\).

Note that \(S^+\) is a left inverse of \(S^-\), \(S^+ S^- = I\), but not a right inverse as \(S^- S^+ \neq I\). This is different from the finite dimensional case, where a left inverse is also a right inverse and vice versa.

\[ \Box \]

**Example.** Suppose \(U \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)\) is unitary. Then \(U^* = U^{-1}\). This follows from Lemma 2.11 since \(\langle f, g \rangle_{\mathcal{H}_1} = \langle Uf, Ug \rangle_{\mathcal{H}_2} = \langle f, U^* Ug \rangle_{\mathcal{H}_1}\) implies \(U^* U = I_{\mathcal{H}_1}\). Since \(U\) is bijective we can multiply this last equation from the right with \(U^{-1}\) to obtain the claim. Of course this calculation shows that the converse is also true, that is \(U \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)\) is unitary if and only if \(U^* = U^{-1}\).

\[ \Box \]

A few simple properties of taking adjoints are listed below.

**Lemma 2.13.** Let \(A, B \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)\), \(C \in \mathcal{L}(\mathcal{H}_2, \mathcal{H}_3)\), and \(\alpha \in \mathbb{C}\). Then

(i) \((A + B)^* = A^* + B^*\), \((\alpha A)^* = \alpha^* A^*\),

(ii) \(A^{**} = A\),

(iii) \((CA)^* = A^* C^*\),

(iv) \(\|A^*\| = \|A\|\) and \(\|A^*\|^2 = \|A^* A\| = \|AA^*\|\).

**Proof.** (i) is obvious. (ii) follows from \(\langle g, A^{**} f \rangle_{\mathcal{H}_2} = \langle A^* g, f \rangle_{\mathcal{H}_1} = \langle g, A f \rangle_{\mathcal{H}_2}\). (iii) follows from \(\langle g, (CA) f \rangle_{\mathcal{H}_3} = \langle C^* g, Af \rangle_{\mathcal{H}_2} = \langle A^* C^* g, f \rangle_{\mathcal{H}_1}\). (iv) follows using (2.26) from

\[
\|A^*\| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle f, A^* g \rangle_{\mathcal{H}_1}| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle A f, g \rangle_{\mathcal{H}_2}| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle g, A f \rangle_{\mathcal{H}_2}| = \|A\|
\]

and

\[
\|A^* A\| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle f, A^* A g \rangle_{\mathcal{H}_1}| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle A f, Ag \rangle_{\mathcal{H}_2}| = \sup_{\|f\|_{\mathcal{H}_1} = \|g\|_{\mathcal{H}_2} = 1} |\langle A f, Ag \rangle_{\mathcal{H}_2}| = \sup_{\|f\|_{\mathcal{H}_1} = 1} \|A f\|^2 = \|A\|^2,
\]

where we have used that \(|\langle A f, Ag \rangle_{\mathcal{H}_2}|\) attains its maximum when \(Af\) and \(Ag\) are parallel (compare Theorem 1.5). \(\square\)
Note that \( \|A\| = \|A^*\| \) implies that taking adjoints is a continuous operation. For later use also note that (Problem 2.11)

\[
\text{Ker}(A^*) = \text{Ran}(A)^\perp. \tag{2.28}
\]

For the remainder of this section we restrict to the case of one Hilbert space. A sesquilinear form \( s : \mathcal{H} \times \mathcal{H} \to \mathbb{C} \) is called nonnegative if \( s(f, f) \geq 0 \) and we will call \( A \in \mathcal{L}(\mathcal{H}) \) nonnegative, \( A \geq 0 \), if its associated sesquilinear form is. We will write \( A \geq B \) if \( A - B \geq 0 \). Observe that nonnegative operators are self-adjoint (as their quadratic forms are real-valued — here it is important that the underlying space is complex).

**Example.** For any operator \( A \) the operators \( A^*A \) and \( AA^* \) are both nonnegative. In fact \( \langle f, A^*Af \rangle = \langle Af, Af \rangle = \|Af\|^2 \geq 0 \) and similarly \( \langle f, AA^*f \rangle = \|A^*f\|^2 \geq 0 \). \( \diamond \)

**Lemma 2.14.** Suppose \( A \in \mathcal{L}(\mathcal{H}) \) satisfies \( A \geq \varepsilon I \) for some \( \varepsilon > 0 \). Then \( A \) is a bijection with bounded inverse, \( \|A^{-1}\| \leq \frac{1}{\varepsilon} \).

**Proof.** By definition \( \varepsilon \|f\|^2 \leq \langle f, Af \rangle \leq \|f\|\|Af\| \) and thus \( \varepsilon \|f\| \leq \|Af\| \). In particular, \( Af = 0 \) implies \( f = 0 \) and thus for every \( g \in \text{Ran}(A) \) there is a unique \( f = A^{-1}g \). Moreover, by \( \|A^{-1}g\| = \|f\| \leq \varepsilon^{-1}\|Af\| = \varepsilon^{-1}\|g\| \) the operator \( A^{-1} \) is bounded. So if \( g_n \in \text{Ran}(A) \) converges to some \( g \in \mathcal{H} \), then \( f_n = A^{-1}g_n \) converges to some \( f \). Taking limits in \( g_n = Af_n \) shows that \( g = Af \) is in the range of \( A \), that is, the range of \( A \) is closed. To show that \( \text{Ran}(A) = \mathcal{H} \) we pick \( h \in \text{Ran}(A)^\perp \). Then \( 0 = \langle h, Ah \rangle \geq \varepsilon \|h\|^2 \) shows \( h = 0 \) and thus \( \text{Ran}(A)^\perp = \{0\} \). \( \square \)

Combining the last two results we obtain the famous Lax–Milgram theorem which plays an important role in theory of elliptic partial differential equations.

**Theorem 2.15** (Lax–Milgram). Let \( s : \mathcal{H} \times \mathcal{H} \to \mathbb{C} \) be a sesquilinear form which is

- **bounded**, \( |s(f, g)| \leq C \|f\| \|g\| \), and
- **coercive**, \( s(f, f) \geq \varepsilon \|f\|^2 \) for some \( \varepsilon > 0 \).

Then for every \( g \in \mathcal{H} \) there is a unique \( f \in \mathcal{H} \) such that

\[
s(h, f) = \langle h, g \rangle, \quad \forall h \in \mathcal{H}. \tag{2.29}
\]

**Proof.** Let \( A \) be the operator associated with \( s \). Then \( A \geq \varepsilon \) and \( f = A^{-1}g \). \( \square \)

**Example.** Consider \( \mathcal{H} = \ell^2(\mathbb{N}) \) and introduce the operator

\[
(Aa)_j := -a_{j+1} + 2a_j - a_{j-1}
\]
which is a discrete version of a second derivative (discrete one-dimensional Laplace operator). Here we use the convention $a_0 = 0$, that is, $(Aa)_1 = -a_2 + 2a_1$. In terms of the shift operators $S^\pm$ we can write
\[ A = -S^+ + 2 - S^- = (S^+ - 1)(S^- - 1) \]
and using $(S^\pm)^* = S^{\mp}$ we obtain
\[ s_A(a, b) = ((S^- - 1)a, (S^- - 1)b) = \sum_{j=1}^{\infty} (a_{j-1} - a_j)^*(b_{j-1} - b_j). \]
In particular, this shows $A \geq 0$. Moreover, we have $|s_A(a, b)| \leq 4\|a\|_2\|b\|_2$ or equivalently $\|A\| \leq 4$.

Next, let
\[(Qa)_j = q_j a_j \]
for some sequence $q \in \ell^\infty(N)$. Then
\[ s_Q(a, b) = \sum_{j=1}^{\infty} q_j a_j^* b_j \]
and $|s_Q(a, b)| \leq \|q\|_\infty\|a\|_2\|b\|_2$ or equivalently $\|Q\| \leq \|q\|_\infty$. If in addition $q_j \geq \varepsilon > 0$, then $s_{A+Q}(a, b) = s_A(a, b) + s_Q(a, b)$ satisfies the assumptions of the Lax–Milgram theorem and
\[(A + Q)a = b \]
has a unique solution $a = (A + Q)^{-1}b$ for every given $b \in \ell^2(Z)$. Moreover, since $(A + Q)^{-1}$ is bounded, this solution depends continuously on $b$. \hfill \blacklozenge

**Problem 2.7.** Let $\mathcal{H}_1, \mathcal{H}_2$ be Hilbert spaces and let $u \in \mathcal{H}_1, v \in \mathcal{H}_2$. Show that the operator
\[ Af := \langle u, f \rangle v \]
is bounded and compute its norm. Compute the adjoint of $A$.

**Problem 2.8.** Show that under the assumptions of Problem 1.32 one has $f(A)^* = f^\#(A^*)$ where $f^\#(z) = f(z^*)^*$.

**Problem 2.9.** Prove (2.26). (Hint: Use $\|f\| = \sup_{\|g\| = 1} |\langle g, f \rangle|$ — compare Theorem 1.5.)

**Problem 2.10.** Suppose $A \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ has a bounded inverse $A^{-1} \in \mathcal{L}(\mathcal{H}_2, \mathcal{H}_1)$. Show $(A^{-1})^* = (A^*)^{-1}$.

**Problem 2.11.** Show (2.28).

**Problem 2.12.** Show that every operator $A \in \mathcal{L}(\mathcal{H})$ can be written as the linear combination of two self-adjoint operators $\text{Re}(A) := \frac{1}{2}(A + A^*)$ and $\text{Im}(A) := \frac{1}{2i}(A - A^*)$. Moreover, every self-adjoint operator can be written
2.4. Orthogonal sums and tensor products

as a linear combination of two unitary operators. (Hint: For the last part consider \( f_\pm(z) = z \pm i\sqrt{1 - z^2} \) and Problems 1.32, 2.8.)

**Problem 2.13** (Abstract Dirichlet problem). Show that the solution of (2.29) is also the unique minimizer of

\[
\text{Re} \left( \frac{1}{2} s(h, h) - (h, g) \right).
\]

2.4. Orthogonal sums and tensor products

Given two Hilbert spaces \( \mathcal{H}_1 \) and \( \mathcal{H}_2 \), we define their **orthogonal sum** \( \mathcal{H}_1 \oplus \mathcal{H}_2 \) to be the set of all pairs \((f_1, f_2) \in \mathcal{H}_1 \times \mathcal{H}_2\) together with the scalar product

\[
\langle (g_1, g_2), (f_1, f_2) \rangle := \langle g_1, f_1 \rangle_{\mathcal{H}_1} + \langle g_2, f_2 \rangle_{\mathcal{H}_2}.
\]

(2.30)

It is left as an exercise to verify that \( \mathcal{H}_1 \oplus \mathcal{H}_2 \) is again a Hilbert space. Moreover, \( \mathcal{H}_1 \) can be identified with \( \{ (f_1, 0) | f_1 \in \mathcal{H}_1 \} \), and we can regard \( \mathcal{H}_1 \) as a subspace of \( \mathcal{H}_1 \oplus \mathcal{H}_2 \), and similarly for \( \mathcal{H}_2 \). With this convention we have \( \mathcal{H}_1^\perp = \mathcal{H}_2 \). It is also customary to write \( f_1 \oplus f_2 \) instead of \( (f_1, f_2) \).

**Example.** For example we have \( \bigoplus_{j=1}^n \mathbb{C} = \mathbb{C}^n \) and hence we will write \( \bigoplus_{j=1}^n \mathcal{H} = \mathcal{H}^n \).

More generally, let \( \mathcal{H}_j, j \in \mathbb{N} \), be a countable collection of Hilbert spaces and define

\[
\bigoplus_{j=1}^{\infty} \mathcal{H}_j := \{ \sum_{j=1}^{\infty} \alpha_j (f_j, \tilde{f}_j) | (f_j, \tilde{f}_j) \in \mathcal{H}_j, \sum_{j=1}^{\infty} \|f_j\|_{\mathcal{H}_j}^2 < \infty \},
\]

(2.31)

which becomes a Hilbert space with the scalar product

\[
\langle \sum_{j=1}^{\infty} g_j, \sum_{j=1}^{\infty} f_j \rangle := \sum_{j=1}^{\infty} \langle g_j, f_j \rangle_{\mathcal{H}_j},
\]

(2.32)

**Example.** \( \bigoplus_{j=1}^{\infty} \mathbb{C} = \ell^2(\mathbb{N}) \).

Similarly, if \( \tilde{\mathcal{H}} \) and \( \check{\mathcal{H}} \) are two Hilbert spaces, we define their tensor product as follows: The elements should be products \( f \otimes \tilde{f} \) of elements \( f \in \mathcal{H} \) and \( \tilde{f} \in \tilde{\mathcal{H}} \). Hence we start with the set of all finite linear combinations of elements of \( \mathcal{H} \times \check{\mathcal{H}} \)

\[
\mathcal{F}(\mathcal{H}, \check{\mathcal{H}}) := \{ \sum_{j=1}^{n} \alpha_j (f_j, \tilde{f}_j) | (f_j, \tilde{f}_j) \in \mathcal{H} \times \check{\mathcal{H}}, \alpha_j \in \mathbb{C} \},
\]

(2.33)
Since we want \((f_1 + f_2) \otimes \tilde{f} = f_1 \otimes \tilde{f} + f_2 \otimes \tilde{f}, \ f \otimes (\tilde{f}_1 + \tilde{f}_2) = f \otimes \tilde{f}_1 + f \otimes \tilde{f}_2, \) and \((\alpha f) \otimes \tilde{f} = f \otimes (\alpha f) = \alpha(f \otimes \tilde{f})\) we consider \(\mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}})/\mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}}),\) where

\[
\mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}}) := \text{span}\{ \sum_{j,k=1}^{n} \alpha_j \beta_k (f_j, \tilde{f}_k) - \sum_{j,k=1}^{n} \alpha_j f_j, \sum_{k=1}^{n} \beta_k \tilde{f}_k) \}
\]

(2.34)

and write \(f \otimes \tilde{f}\) for the equivalence class of \((f, \tilde{f}).\) By construction, every element in this quotient space is a linear combination of elements of the type \(f \otimes \tilde{f}.)\)

Next, we want to define a scalar product such that

\[
\langle f \otimes \tilde{f}, g \otimes \tilde{g}\rangle = \langle f, g\rangle_{\mathfrak{J}} \langle \tilde{f}, \tilde{g}\rangle_{\tilde{\mathfrak{J}}}
\]

(2.35)

holds. To this end we set

\[
s(\sum_{j=1}^{n} \alpha_j (f_j, \tilde{f}_j), \sum_{k=1}^{n} \beta_k (g_k, \tilde{g}_k)) = \sum_{j,k=1}^{n} \alpha_j \beta_k \langle f_j, g_k\rangle_{\mathfrak{J}} \langle \tilde{f}_j, \tilde{g}_k\rangle_{\tilde{\mathfrak{J}}},
\]

(2.36)

which is a symmetric sesquilinear form on \(\mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}}).\) Moreover, one verifies that \(s(f, g) = 0\) for arbitrary \(f \in \mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}})\) and \(g \in \mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}})\) and thus

\[
\langle \sum_{j=1}^{n} \alpha_j f_j \otimes \tilde{f}_j, \sum_{k=1}^{n} \beta_k g_k \otimes \tilde{g}_k\rangle = \sum_{j,k=1}^{n} \alpha_j \beta_k \langle f_j, g_k\rangle_{\mathfrak{J}} \langle \tilde{f}_j, \tilde{g}_k\rangle_{\tilde{\mathfrak{J}}}
\]

(2.37)

is a symmetric sesquilinear form on \(\mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}})/\mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}}).\) To show that this is in fact a scalar product, we need to ensure positivity. Let \(f = \sum_{i} \alpha_i f_i \otimes \tilde{f}_i \neq 0\) and pick orthonormal bases \(u_j, \tilde{u}_k\) for \(\text{span}\{f_i\}, \text{span}\{\tilde{f}_i\},\) respectively. Then

\[
f = \sum_{j,k} \alpha_{jk} u_j \otimes \tilde{u}_k, \quad \alpha_{jk} = \sum_{i} \alpha_i \langle u_j, f_i\rangle_{\mathfrak{J}} \langle \tilde{u}_k, \tilde{f}_i\rangle_{\tilde{\mathfrak{J}}}
\]

(2.38)

and we compute

\[
\langle f, f\rangle = \sum_{j,k} |\alpha_{jk}|^2 > 0.
\]

(2.39)

The completion of \(\mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}})/\mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}})\) with respect to the induced norm is called the tensor product \(\mathfrak{J} \otimes \tilde{\mathfrak{J}}\) of \(\mathfrak{J}\) and \(\tilde{\mathfrak{J}}.\)

**Lemma 2.16.** If \(u_j, \tilde{u}_k\) are orthonormal bases for \(\mathfrak{J}, \tilde{\mathfrak{J}},\) respectively, then \(u_j \otimes \tilde{u}_k\) is an orthonormal basis for \(\mathfrak{J} \otimes \tilde{\mathfrak{J}}.\)

**Proof.** That \(u_j \otimes \tilde{u}_k\) is an orthonormal set is immediate from (2.35). Moreover, since \(\text{span}\{u_j\}, \text{span}\{\tilde{u}_k\}\) are dense in \(\mathfrak{J}, \tilde{\mathfrak{J}},\) respectively, it is easy to see that \(u_j \otimes \tilde{u}_k\) is dense in \(\mathcal{F}(\mathfrak{J}, \tilde{\mathfrak{J}})/\mathcal{N}(\mathfrak{J}, \tilde{\mathfrak{J}}).\) But the latter is dense in \(\mathfrak{J} \otimes \tilde{\mathfrak{J}}.\) \(\square\)
2.5. Applications to Fourier series

Note that this in particular implies $\dim(\mathcal{H} \otimes \tilde{\mathcal{H}}) = \dim(\mathcal{H}) \dim(\tilde{\mathcal{H}})$.

**Example.** We have $\mathcal{H} \otimes \mathbb{C}^n = \mathcal{H}^n$.

**Example.** We have $\ell^2(\mathbb{N}) \otimes \ell^2(\mathbb{N}) \cong \ell^2(\mathbb{N} \times \mathbb{N})$ by virtue of the identification $(a_{jk}) \mapsto \sum_{jk} a_{jk} \delta^j \otimes \delta^k$ where $\delta^j$ is the standard basis for $\ell^2(\mathbb{N})$. In fact, this follows from the previous lemma as in the proof of Theorem 2.6.

It is straightforward to extend the tensor product to any finite number of Hilbert spaces. We even note

$$\bigoplus_{j=1}^{\infty} \mathcal{H}_j \otimes \mathcal{H} = \bigoplus_{j=1}^{\infty} (\mathcal{H}_j \otimes \mathcal{H}),$$

where equality has to be understood in the sense that both spaces are unitarily equivalent by virtue of the identification

$$\left(\sum_{j=1}^{\infty} f_j\right) \otimes f = \sum_{j=1}^{\infty} f_j \otimes f.$$  \tag{2.41}

**Problem 2.14.** Show that $f \otimes \tilde{f} = 0$ if and only if $f = 0$ or $\tilde{f} = 0$.

**Problem 2.15.** We have $f \otimes \tilde{f} = g \otimes \tilde{g} \neq 0$ if and only if there is some $\alpha \in \mathbb{C} \setminus \{0\}$ such that $f = \alpha g$ and $\tilde{f} = \alpha^{-1} \tilde{g}$.

**Problem 2.16.** Show (2.40).

2.5. Applications to Fourier series

We have already encountered the Fourier sine series during our treatment of the heat equation in Section 1.1. Given an integrable function $f$ we can define its **Fourier series**

$$S(f)(x) := \frac{a_0}{2} + \sum_{k \in \mathbb{N}} a_k \cos(kx) + b_k \sin(kx),$$

where the corresponding Fourier coefficients are given by

$$a_k := \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(kx) f(x) dx, \quad b_k := \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(kx) f(x) dx.$$ \tag{2.43}

At this point (2.42) is just a formal expression and it was (and to some extend still is) a fundamental question in mathematics to understand in what sense the above series converges. For example, does it converge at a given point (e.g. at every point of continuity) or when does it converge uniformly? We will give some first answers in the present section and then come back later to this when we have further tools at our disposal.
For our purpose the complex form
\[ S(f)(x) = \sum_{k \in \mathbb{Z}} \hat{f}_k e^{ikx}, \quad \hat{f}_k := \frac{1}{2\pi} \int_{-\pi}^{\pi} e^{-iy} f(y) dy \] (2.44)
will be more convenient. The connection is given via \( \hat{f}_{\pm k} = \frac{a_k \mp ib_k}{2} \). In this case the \( n \)'th partial sum can be written as
\[ S_n(f)(x) := \sum_{k=-n}^{n} \hat{f}_k e^{ikx} = \frac{1}{2\pi} \int_{-\pi}^{\pi} D_n(x-y) f(y) dy, \] (2.45)
where
\[ D_n(x) = \sum_{k=-n}^{n} e^{ikx} = \frac{\sin((n+1/2)x)}{\sin(x/2)} \] (2.46)
is known as the Dirichlet kernel (to obtain the second form observe that the left-hand side is a geometric series). Note that \( D_n(-x) = -D_n(x) \) and that \( |D_n(x)| \) has a global maximum \( D_n(0) = 2n+1 \) at \( x = 0 \). Moreover, by \( S_n(1) = 1 \) we see that \( \int_{-\pi}^{\pi} D_n(x) dx = 1 \).

Since
\[ \int_{-\pi}^{\pi} e^{-ikx} e^{ilx} dx = 2\pi \delta_{k,l} \] (2.47)
the functions \( e_k(x) = (2\pi)^{-1/2} e^{ikx} \) are orthonormal in \( L^2(-\pi, \pi) \) and hence the Fourier series is just the expansion with respect to this orthogonal set. Hence we obtain
**Theorem 2.17.** For every square integrable function \( f \in L^2(-\pi, \pi) \), the Fourier coefficients \( \hat{f}_k \) are square summable

\[
\sum_{k \in \mathbb{Z}} |\hat{f}_k|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |f(x)|^2 \quad (2.48)
\]

and the Fourier series converges to \( f \) in the sense of \( L^2 \). Moreover, this is a continuous bijection between \( L^2(-\pi, \pi) \) and \( \ell^2(\mathbb{Z}) \).

**Proof.** To show this theorem it suffices to show that the functions \( e_k \) form a basis. This will follow from Theorem 2.19 below (see the discussion after this theorem). It will also follow as a special case of Theorem 3.11 below (see the examples after this theorem) as well as from the Stone–Weierstraß theorem — Problem 2.19.

This gives a satisfactory answer in the Hilbert space \( L^2(-\pi, \pi) \) but does not answer the question about pointwise or uniform convergence. The latter will be the case if the Fourier coefficients are summable. First of all we note that for integrable functions the Fourier coefficients will at least tend to zero.

**Lemma 2.18** (Riemann–Lebesgue lemma). Suppose \( f \) is integrable, then the Fourier coefficients converge to zero.

**Proof.** By our previous theorem this holds for continuous functions. But the map \( f \rightarrow \hat{f} \) is bounded from \( C[-\pi, \pi] \subset L^1(-\pi, \pi) \) to \( c_0(\mathbb{Z}) \) (the sequences vanishing as \(|k| \to \infty\)) since \(|\hat{f}_k| \leq (2\pi)^{-1} \|f\|_1 \) and there is a unique extension to all of \( L^1(-\pi, \pi) \).

It turns out that this result is best possible in general and we cannot say more without additional assumptions on \( f \). For example, if \( f \) is periodic and differentiable, then integration by parts shows

\[
\hat{f}_k = \frac{1}{2\pi i k} \int_{-\pi}^{\pi} e^{-ikx} f'(x)dx. \tag{2.49}
\]

Then, since both \( k^{-1} \) and the Fourier coefficients of \( f' \) are square summable, we conclude that \( \hat{f}_k \) are summable and hence the Fourier series converges uniformly. So we have a simple sufficient criterion for summability of the Fourier coefficients, but can it be improved? Of course continuity of \( f \) is a necessary condition but this alone will not even be enough for pointwise convergence as we will see in the example on page 127. Moreover, continuity will not tell us more about the decay of the Fourier coefficients than what we already know in the integrable case from the Riemann–Lebesgue lemma (see the example on page 128).
A few improvements are easy: First of all, piecewise continuously differentiable would be sufficient for this argument. Or, slightly more general, an absolutely continuous function whose derivative is square integrable would also do (cf. Lemma 11.50). However, even for an absolutely continuous function the Fourier coefficients might not be summable: For an absolutely continuous function \( f \) we have a derivative which is integrable (Theorem 11.49) and hence the above formula combined with the Riemann–Lebesgue lemma implies \( \hat{f}_k = o(1/k) \). But on the other hand we can choose a summable sequence \( c_k \) which does not obey this asymptotic requirement, say \( c_k = 1/k \) for \( k = l^2 \) and \( c_k = 0 \) else. Then

\[
f(x) = \sum_{k \in \mathbb{Z}} c_ke^{ikx} = \sum_{l \in \mathbb{N}} \frac{1}{l^2}e^{il^2x}
\]

is a function with summable Fourier coefficients \( \hat{f}_k = c_k \) (by uniform convergence we can interchange summation and integration) but which is not absolutely continuous. There are further criteria for summability of the Fourier coefficients but no simple necessary and sufficient one.

Note however, that the situation looks much brighter if one looks at mean values

\[
\bar{S}_n(f)(x) = \frac{1}{n} \sum_{k=0}^{n-1} S_n(f)(x) = \frac{1}{2\pi} \int_{-\pi}^{\pi} F_n(x-y)f(y)dy,
\]

where

\[
F_n(x) = \frac{1}{n} \sum_{k=0}^{n-1} D_k(x) = \frac{1}{n} \left( \frac{\sin(nx/2)}{\sin(x/2)} \right)^2
\]

is the Fejér kernel. To see the second form we use the closed form for the Dirichlet kernel to obtain

\[
nF_n(x) = \sum_{k=0}^{n-1} \frac{\sin((k+1/2)x)}{\sin(x/2)} = \frac{1}{\sin(x/2)} \text{Im} \sum_{k=0}^{n-1} e^{i(k+1/2)x})
\]

\[
= \frac{1}{\sin(x/2)} \text{Im} \left( e^{ix/2} \frac{e^{inx} - 1}{e^{ix} - 1} \right) = \frac{1 - \cos(nx)}{2 \sin(x/2)^2} = \frac{\sin(nx/2)^2}{\sin(x/2)^2}.
\]

The main difference to the Dirichlet kernel is positivity: \( F_n(x) \geq 0 \). Of course the property \( \int_{-\pi}^{\pi} F_n(x)dx = 1 \) is inherited from the Dirichlet kernel.

**Theorem 2.19 (Fejér).** Suppose \( f \) is continuous and periodic with period \( 2\pi \). Then \( \bar{S}_n(f) \rightarrow f \) uniformly.

**Proof.** Let us set \( F_n = 0 \) outside \([−\pi,\pi] \). Then \( F_n(x) \leq \frac{1}{n\sin(x/2)^2} \) for \( \delta \leq |x| \leq \pi \) implies that a straightforward adaption of Lemma 1.2 to the periodic case is applicable. \( \square \)
In particular, this shows that the functions \( \{e_k\}_{k \in \mathbb{Z}} \) are total in \( C_{\text{per}}[-\pi, \pi] \) (continuous periodic functions) and hence also in \( L^p(-\pi, \pi) \) for \( 1 \leq p < \infty \) (Problem 2.18).

Note that this result shows that if \( S(f)(x) \) converges for a continuous function, then it must converge to \( f(x) \). We also remark that one can extend this result (see Lemma 10.18) to show that for \( f \in L^p(-\pi, \pi) \), \( 1 \leq p < \infty \), one has \( S_n(f) \to f \) in the sense of \( L^p \). As a consequence note that the Fourier coefficients uniquely determine \( f \) for integrable \( f \) (for square integrable \( f \) this follows from Theorem 2.17).

Finally, we look at pointwise convergence.

**Theorem 2.20.** Suppose
\[
\frac{f(x) - f(x_0)}{x - x_0}
\]
is integrable (e.g. \( f \) is Hölder continuous), then
\[
\lim_{m,n \to \infty} \sum_{k=-m}^{n} \hat{f}(k)e^{ikx_0} = f(x_0).
\]

**Proof.** Without loss of generality we can assume \( x_0 = 0 \) (by shifting \( x \to x - x_0 \) modulo \( 2\pi \) implying \( \hat{f}_k \to e^{-ikx_0} \hat{f}_k \) and \( f(x_0) = 0 \) (by linearity since
the claim is trivial for constant functions). Then by assumption
\[ g(x) = \frac{f(x)}{e^{ix} - 1} \]
is integrable and \( f(x) = (e^{ix} - 1)g(x) \) implies
\[ \hat{f}_k = \hat{g}_{k-1} - \hat{g}_k \]
and hence
\[ \sum_{k=m}^{n} \hat{f}_k = \hat{g}_{m-1} - \hat{g}_n. \]
Now the claim follows from the Riemann–Lebesgue lemma. \( \square \)

If one looks at symmetric partial sums \( S_n(f) \) we can do even better.

**Corollary 2.21** (Dirichlet–Dini criterion). Suppose there is some \( \alpha \) such that
\[ f(x_0 + x) + f(x_0 - x) - 2\alpha \]
is integrable. Then \( S_n(f)(x_0) \to \alpha \).

**Proof.** Without loss of generality we can assume \( x_0 = 0 \). Now observe (since \( D_n(-x) = D_n(x) \))
\[ S_n(f)(0) = \alpha + S_n(g)(0), \quad g(x) = \frac{f(x) + f(-x) - 2\alpha}{2} \]
and apply the previous result. \( \square \)

**Problem 2.17.** Compute the Fourier series of \( D_n \) and \( F_n \).

**Problem 2.18.** Show that \( C_{per}[-\pi, \pi] \) is dense in \( L^p(-\pi, \pi) \) for \( 1 \leq p < \infty \).

**Problem 2.19.** Show that the functions \( \varphi_n(x) = \frac{1}{\sqrt{2\pi}} e^{inx}, n \in \mathbb{Z} \), form an orthonormal basis for \( H = L^2(-\pi, \pi) \). (Hint: Start with \( K = [-\pi, \pi] \) where \( -\pi \) and \( \pi \) are identified and use the Stone–Weierstraß theorem.)
Typically, linear operators are much more difficult to analyze than matrices and many new phenomena appear which are not present in the finite dimensional case. So we have to be modest and slowly work our way up. A class of operators which still preserves some of the nice properties of matrices is the class of compact operators to be discussed in this chapter.

3.1. Compact operators

A linear operator $A : X \to Y$ defined between normed spaces $X$, $Y$ is called **compact** if every sequence $Af_n$ has a convergent subsequence whenever $f_n$ is bounded. Equivalently (cf. Corollary 0.20), $A$ is compact if it maps bounded sets to relatively compact ones. The set of all compact operators is denoted by $\mathcal{C}(X,Y)$. If $X = Y$ we will just write $\mathcal{C}(X) := \mathcal{C}(X,X)$ as usual.

**Example.** Every linear map between finite dimensional spaces is compact by the Bolzano–Weierstraß theorem. Slightly more general, an operator is compact if its range is finite dimensional.

The following elementary properties of compact operators are left as an exercise (Problem 3.1):

**Theorem 3.1.** Let $X$, $Y$, and $Z$ be normed spaces. Every compact linear operator is bounded, $\mathcal{C}(X,Y) \subseteq \mathcal{L}(X,Y)$. Linear combinations of compact operators are compact, that is, $\mathcal{C}(X,Y)$ is a subspace of $\mathcal{L}(X,Y)$. Moreover, the product of a bounded and a compact operator is again compact, that is, $A \in \mathcal{L}(X,Y)$, $B \in \mathcal{C}(Y,Z)$ or $A \in \mathcal{C}(X,Y)$, $B \in \mathcal{L}(Y,Z)$ implies $BA \in \mathcal{C}(X,Z)$. 

In particular, the set of compact operators $\mathcal{C}(X)$ is an ideal of the set of bounded operators. Moreover, if $X$ is a Banach space this ideal is even closed:

**Theorem 3.2.** Suppose $X$ is a normed and $Y$ a Banach space. Let $A_n \in \mathcal{C}(X,Y)$ be a convergent sequence of compact operators. Then the limit $A$ is again compact.

**Proof.** Let $f_j^{(0)}$ be a bounded sequence. Choose a subsequence $f_j^{(1)}$ such that $A_1f_j^{(1)}$ converges. From $f_j^{(1)}$ choose another subsequence $f_j^{(2)}$ such that $A_2f_j^{(2)}$ converges and so on. Since $f_j^{(n)}$ might disappear as $n \to \infty$, we consider the diagonal sequence $f_j := f_j^{(j)}$. By construction, $f_j$ is a subsequence of $f_j^{(n)}$ for $j \geq n$ and hence $A_nf_j$ is Cauchy for every fixed $n$. Now

$$\|Af_j - Af_k\| = \|(A - A_n)(f_j - f_k) + A_n(f_j - f_k)\|$$

$$\leq \|A - A_n\|\|f_j - f_k\| + \|A_nf_j - A_nf_k\|$$

shows that $Af_j$ is Cauchy since the first term can be made arbitrary small by choosing $n$ large and the second by the Cauchy property of $A_nf_j$.

\[ \Box \]

**Example.** Let $X := \ell^p(\mathbb{N})$ and consider the operator

$$(Qa)_j := q_j a_j$$

for some sequence $q = (q_j)_{j=1}^{\infty} \in c_0(\mathbb{N})$ converging to zero. Let $Q_n$ be associated with $q^n_j = q_j$ for $j \leq n$ and $q^n_j = 0$ for $j > n$. Then the range of $Q^n$ is finite dimensional and hence $Q_n$ is compact. Moreover, by $\|Q_n - Q\| = \sup_{j>n} |q_j|$ we see $Q_n \to Q$ and thus $Q$ is also compact by the previous theorem.

\[ \diamond \]

**Example.** Let $X = C^1[0,1], Y = C[0,1]$ (cf. Problem 1.28) then the embedding $X \hookrightarrow Y$ is compact. Indeed, a bounded sequence in $X$ has both the functions and the derivatives uniformly bounded. Hence by the mean value theorem the functions are equicontinuous and hence there is a uniformly convergent subsequence by the Arzelà–Ascoli theorem (Theorem 0.37). Of course the same conclusion holds if we take $X = C^{0,\gamma}[0,1]$ to be Hölder continuous functions or if we replace $[0,1]$ by a compact metric space.

\[ \diamond \]

If $A : X \to Y$ is a bounded operator there is a unique extension $\overline{A} : \overline{X} \to \overline{Y}$ to the completion by Theorem 1.12. Moreover, if $A \in \mathcal{C}(X,Y)$, then $A \in \mathcal{C}(\overline{X},\overline{Y})$ is immediate. That we also have $\overline{A} \in \mathcal{C}(\overline{X},\overline{Y})$ will follow from the next lemma. In particular, it suffices to verify compactness on a dense set.
3.1. Compact operators

Lemma 3.3. Let \( X, Y \) be normed spaces and \( A \in \mathcal{C}(X,Y) \). Let \( \overline{X}, \overline{Y} \) be the completion of \( X, Y \), respectively. Then \( \overline{A} \in \mathcal{C}(\overline{X}, \overline{Y}) \), where \( \overline{A} \) is the unique extension of \( A \).

Proof. Let \( f_n \in X \) be a given bounded sequence. We need to show that \( \overline{A}f_n \) has a convergent subsequence. Pick \( f_j^n \in X \) such that \( \|f_j^n - f_n\| \leq \frac{1}{j} \) and by compactness of \( A \) we can assume that \( Af_j^n \to g \). But then \( \|\overline{A}f_n - g\| \leq \|A\|\|f_n - f_j^n\| + \|Af_j^n - g\| \) shows that \( \overline{A}f_n \to g \). \( \square \)

One of the most important examples of compact operators are integral operators. The proof will be based on the Arzelà–Ascoli theorem (Theorem 0.37).

Lemma 3.4. Let \( X = C([a,b]) \) or \( X = L^2_{\text{cont}}(a,b) \). The integral operator \( K : X \to X \) defined by

\[
(Kf)(x) := \int_a^b K(x,y)f(y)dy, \quad (3.1)
\]

where \( K(x,y) \in C([a,b] \times [a,b]) \), is compact.

Proof. First of all note that \( K(\ldots) \) is continuous on \([a,b] \times [a,b]\) and hence uniformly continuous. In particular, for every \( \varepsilon > 0 \) we can find a \( \delta > 0 \) such that \( |K(y,t) - K(x,t)| \leq \varepsilon \) whenever \( |y-x| \leq \delta \). Moreover, \( \|K\|_{\infty} = \sup_{x,y\in[a,b]}|K(x,y)| < \infty \).

We begin with the case \( X = L^2_{\text{cont}}(a,b) \). Let \( g(x) = Kf(x) \). Then

\[
|g(x)| \leq \int_a^b |K(x,t)||f(t)|dt \leq \|K\|_{\infty} \int_a^b |f(t)|dt \leq \|K\|_{\infty} \|1\| \|f\|,
\]

where we have used Cauchy–Schwarz in the last step (note that \( \|1\| = \sqrt{b-a} \)). Similarly,

\[
|g(x) - g(y)| \leq \int_a^b |K(y,t) - K(x,t)||f(t)|dt \\
\leq \varepsilon \int_a^b |f(t)|dt \leq \varepsilon \|1\| \|f\|,
\]

whenever \( |y-x| \leq \delta \). Hence, if \( f_n(x) \) is a bounded sequence in \( L^2_{\text{cont}}(a,b) \), then \( g_n(x) = Kf_n(x) \) is bounded and equicontinuous and hence has a uniformly convergent subsequence by the Arzelà–Ascoli theorem (Theorem 0.37). But a uniformly convergent sequence is also convergent in the norm induced by the scalar product. Therefore \( K \) is compact.

The case \( X = C([a,b]) \) follows by the same argument upon observing

\[
\int_a^b |f(t)|dt \leq (b-a)\|f\|_{\infty}.
\]

\( \square \)
Compact operators are very similar to (finite) matrices as we will see in the next section.

**Problem 3.1.** Show Theorem 3.1.

**Problem 3.2.** Show that adjoint of the integral operator $K$ from Lemma 3.4 is the integral operator with kernel $K(y, x)^*$:

$$(K^* f)(x) = \int_a^b K(y, x)^* f(y) dy.$$  

(Hint: Fubini.)

**Problem 3.3.** Show that the mapping $\frac{d}{dx}: C^2[a, b] \rightarrow C[a, b]$ is compact.  

(Hint: Arzelà–Ascoli.)

### 3.2. The spectral theorem for compact symmetric operators

Let $\mathcal{H}$ be an inner product space. A linear operator $A$ is called symmetric if its domain is dense and if

$$\langle g, Af \rangle = \langle Ag, f \rangle \quad f, g \in D(A). \quad (3.2)$$

If $A$ is bounded (with $D(A) = \mathcal{H}$), then $A$ is symmetric precisely if $A = A^*$, that is, if $A$ is self-adjoint. However, for unbounded operators there is a subtle but important difference between symmetry and self-adjointness.

A number $z \in \mathbb{C}$ is called eigenvalue of $A$ if there is a nonzero vector $u \in D(A)$ such that

$$Au = zu. \quad (3.3)$$

The vector $u$ is called a corresponding eigenvector in this case. The set of all eigenvectors corresponding to $z$ is called the eigenspace

$$\text{Ker}(A - z) \quad (3.4)$$

corresponding to $z$. Here we have used the shorthand notation $A - z$ for $A - zI$. An eigenvalue is called simple if there is only one linearly independent eigenvector.

**Example.** Let $\mathcal{H} := \ell^2(\mathbb{N})$ and consider the shift operators $(S^\pm a)_j := a_{j \pm 1}$ (with $a_0 := 0$). Suppose $z \in \mathbb{C}$ is an eigenvalue, then the corresponding eigenvector $u$ must satisfy $u_{j \pm 1} = zu_j$. For $S^-$ the special case $j = 0$ gives $0 = u_0 = zu_1$. So either $z = 0$ and $u = u_1 \delta^1$ or $z \neq 0$ and $u = 0$. Hence the only eigenvalue is $z = 0$. For $S^+$ we get $u_j = z^j u_1$ and this will give an element in $\ell^2(\mathbb{N})$ if and only of $|z| < 1$. Hence $z$ with $|z| < 1$ is an eigenvalue. In both cases all eigenvalues are simple.

**Example.** Let $\mathcal{H} := \ell^2(\mathbb{N})$ and consider the multiplication operator $(Qa)_j := q_j a_j$ with a bounded sequence $q \in \ell^\infty(\mathbb{N})$. Suppose $z \in \mathbb{C}$ is an eigenvalue, then the corresponding eigenvector $u$ must satisfy $(q_j - z)u_j = 0$. Hence
every value $q_j$ is an eigenvalue with corresponding eigenvector $u = \delta^j$. If there is only one $j$ with $z = q_j$ the eigenvalue is simple (otherwise the numbers of independent eigenvectors equals the number of times $z$ appears in the sequence $q$). If $z$ is different from all entries of the sequence then $u = 0$ and $z$ is no eigenvalue.

Note that in the last example $Q$ will be self-adjoint if and only if $q$ is real-valued and hence if and only if all eigenvalues are real-valued. Moreover, the corresponding eigenfunctions are orthogonal. This has nothing to do with the simple structure of our operator and is in fact always true.

**Theorem 3.5.** Let $A$ be symmetric. Then all eigenvalues are real and eigenvectors corresponding to different eigenvalues are orthogonal.

**Proof.** Suppose $\lambda$ is an eigenvalue with corresponding normalized eigenvector $u$. Then $\lambda = \langle u, Au \rangle = \langle Au, u \rangle = \lambda^*$, which shows that $\lambda$ is real. Furthermore, if $Au_j = \lambda_j u_j$, $j = 1, 2$, we have

$$(\lambda_1 - \lambda_2)\langle u_1, u_2 \rangle = \langle Au_1, u_2 \rangle - \langle u_1, Au_2 \rangle = 0$$

finishing the proof. \hfill \□

Note that while eigenvectors corresponding to the same eigenvalue $\lambda$ will in general not automatically be orthogonal, we can of course replace each set of eigenvectors corresponding to $\lambda$ by an set of orthonormal eigenvectors having the same linear span (e.g. using Gram–Schmidt orthogonalization).

**Example.** Let $\mathcal{S} = \ell^2(\mathbb{N})$ and consider the Jacobi operator $J = \frac{1}{2}(S^++S^-)$ associated with the sequences $a_j = \frac{1}{2}, b_j = 0$:

$$(Jc)_j := \frac{1}{2}(c_{j+1} + c_{j-1})$$

with the convention $c_0 = 0$. Recall that $J^* = J$. If we look for an eigenvalue $Ju = zu$, we need to solve the corresponding recursion $u_{j+1} = 2zu_j - u_{j-1}$ starting from $u_0 = 0$ (our convention) and $u_1 = 1$ (normalization). Like an ordinary differential equation, a linear recursion relations with constant coefficients can be solved by an exponential ansatz $k^j$ which leads to the characteristic polynomial $k^2 = 2zk - 1$. This gives two linearly independent solutions and our requirements lead us to

$$u_j(z) = \frac{k^j - k^{-j}}{k - k^{-1}}, \quad k = z - \sqrt{z^2 - 1}.$$ 

Note that $k^{-1} = z + \sqrt{z^2 - 1}$ and in the case $k = z = \pm 1$ the above expression has to be understood as its limit $u_j(\pm 1) = (\pm 1)^{j+1}j$. In fact, $T_j(z) = u_{j-1}(z)$ are polynomials of degree $j$ known as *Chebyshev polynomials*. 

Now for \( z \in \mathbb{R} \setminus [-1, 1] \) we have \(|k| < 1\) and \( u_j \) explodes exponentially. For \( z \in [-1, 1] \) we have \(|k| = 1\) and hence we can write \( k = e^{i\kappa} \) with \( \kappa \in \mathbb{R} \). Thus \( u_j = \frac{\sin(kj)}{\sin(k)} \) is oscillating. So for no value of \( z \in \mathbb{R} \) our potential eigenvector \( u \) is square summable and thus \( J \) has no eigenvalues.

The previous example shows that in the infinite dimensional case symmetry is not enough to guarantee existence of even a single eigenvalue. In order to always get this, we will need an extra condition. In fact, we will see that compactness provides a suitable extra condition to obtain an orthonormal basis of eigenfunctions. The crucial step is to prove existence of one eigenvalue, the rest then follows as in the finite dimensional case.

**Theorem 3.6.** Let \( \mathcal{H} \) be an inner product space. A symmetric compact operator \( A \) has an eigenvalue \( \alpha_1 \) which satisfies \( |\alpha_1| = \|A\| \).

**Proof.** We set \( \alpha = \|A\| \) and assume \( \alpha \neq 0 \) (i.e., \( A \neq 0 \)) without loss of generality. Since

\[
\|A\|^2 = \sup_{f: \|f\|=1} \|Af\|^2 = \sup_{f: \|f\|=1} \langle Af, Af \rangle = \sup_{f: \|f\|=1} \langle f, A^2 f \rangle
\]

there exists a normalized sequence \( u_n \) such that

\[
\lim_{n \to \infty} \langle u_n, A^2 u_n \rangle = \alpha^2.
\]

Since \( A \) is compact, it is no restriction to assume that \( A^2 u_n \) converges, say \( \lim_{n \to \infty} A^2 u_n = \alpha^2 u \). Now

\[
\|(A^2 - \alpha^2)u_n\|^2 = \|A^2 u_n\|^2 - 2\alpha^2 \langle u_n, A^2 u_n \rangle + \alpha^4
\]

\[
\leq 2\alpha^2 (\alpha^2 - \langle u_n, A^2 u_n \rangle)
\]

(where we have used \( \|A^2 u_n\| \leq \|A\| \|Au_n\| \leq \|A\|^2 \|u_n\| = \alpha^2 \) implies \( \lim_{n \to \infty} (A^2 u_n - \alpha^2 u_n) = 0 \) and hence \( \lim_{n \to \infty} u_n = u \). In addition, \( u \) is a normalized eigenvector of \( A^2 \) since \( (A^2 - \alpha^2)u = 0 \). Factorizing this last equation according to \((A - \alpha)u = v \) and \((A + \alpha)v = 0\) shows that either \( v \neq 0 \) is an eigenvector corresponding to \(-\alpha \) or \( v = 0 \) and hence \( u \neq 0 \) is an eigenvector corresponding to \( \alpha \).

Note that for a bounded operator \( A \), there cannot be an eigenvalue with absolute value larger than \( \|A\| \), that is, the set of eigenvalues is bounded by \( \|A\| \) (Problem 3.4).

Now consider a symmetric compact operator \( A \) with eigenvalue \( \alpha_1 \) (as above) and corresponding normalized eigenvector \( u_1 \). Setting

\[
\mathcal{H}_1 := \{ u_1 \} = \{ f \in \mathcal{H} | \langle u_1, f \rangle = 0 \}
\]

we can restrict \( A \) to \( \mathcal{H}_1 \) since \( f \in \mathcal{H}_1 \) implies

\[
\langle u_1, Af \rangle = \langle Au_1, f \rangle = \alpha_1 \langle u_1, f \rangle = 0
\]
and hence $Af \in \mathcal{H}_1$. Denoting this restriction by $A_1$, it is not hard to see that $A_1$ is again a symmetric compact operator. Hence we can apply Theorem 3.6 iteratively to obtain a sequence of eigenvalues $\alpha_j$ with corresponding normalized eigenvectors $u_j$. Moreover, by construction, $u_j$ is orthogonal to all $u_k$ with $k < j$ and hence the eigenvectors $\{u_j\}$ form an orthonormal set. By construction we also have $|\alpha_j| = \|A_j\| \leq \|A_{j-1}\| = |\alpha_{j-1}|$. This procedure will not stop unless $\mathcal{H}$ is finite dimensional. However, note that $\alpha_j = 0$ for $j\geq n$ might happen if $A_n = 0$.

**Theorem 3.7** (Hilbert). Suppose $\mathcal{H}$ is an infinite dimensional Hilbert space and $A : \mathcal{H} \to \mathcal{H}$ is a compact symmetric operator. Then there exists a sequence of real eigenvalues $\alpha_j$ converging to $0$. The corresponding normalized eigenvectors $u_j$ form an orthonormal set and every $f \in \mathcal{H}$ can be written as

$$f = \sum_{j=1}^{\infty} \langle u_j, f \rangle u_j + h,$$

(3.7)

where $h$ is in the kernel of $A$, that is, $Ah = 0$.

In particular, if $0$ is not an eigenvalue, then the eigenvectors form an orthonormal basis (in addition, $\mathcal{H}$ need not be complete in this case).

**Proof.** Existence of the eigenvalues $\alpha_j$ and the corresponding eigenvectors $u_j$ has already been established. Since the sequence $|\alpha_j|$ is decreasing it has a limit $\varepsilon \geq 0$ and we have $|\alpha_j| \geq \varepsilon$. If this limit is nonzero, then $v_j = \alpha_j^{-1}u_j$ is a bounded sequence ($\|v_j\| \leq \frac{1}{\varepsilon}$) for which $Av_j$ has no convergent subsequence since $\|Av_j - Av_k\|^2 = \|u_j - u_k\|^2 = 2$, a contradiction.

Next, setting

$$f_n := \sum_{j=1}^{n} \langle u_j, f \rangle u_j,$$

we have

$$\|A(f - f_n)\| \leq |\alpha_n| \|f - f_n\| \leq \|\alpha_n\| \|f\|$$

since $f - f_n \in \mathcal{H}_n$ and $\|A_n\| = |\alpha_n|$. Letting $n \to \infty$ shows $A(f_\infty - f) = 0$ proving (3.7). Finally, note that without completeness $f_\infty$ might not be well-defined unless $h = 0$. □

By applying $A$ to (3.7) we obtain the following canonical form of compact symmetric operators.

**Corollary 3.8.** Every compact symmetric operator $A$ can be written as

$$Af = \sum_{j=1}^{N} \alpha_j \langle u_j, f \rangle u_j,$$

(3.8)
where \( \alpha_j \) are the nonzero eigenvalues with corresponding eigenvectors \( u_j \) from the previous theorem.

Remark: There are two cases where our procedure might fail to construct an orthonormal basis of eigenvectors. One case is where there is an infinite number of nonzero eigenvalues. In this case \( \alpha_n \) never reaches 0 and all eigenvectors corresponding to 0 are missed. In the other case, 0 is reached, but there might not be a countable basis and hence again some of the eigenvectors corresponding to 0 are missed. In any case, by adding vectors from the kernel (which are automatically eigenvectors), one can always extend the eigenvectors \( u_j \) to an orthonormal basis of eigenvectors.

**Corollary 3.9.** Every compact symmetric operator \( A \) has an associated orthonormal basis of eigenvectors \( \{u_j\}_{j \in J} \). The corresponding unitary map \( U : \mathcal{H} \to \ell^2(J), \ f \mapsto \{\langle u_j, f \rangle\}_{j \in J} \) diagonalizes \( A \) in the sense that \( UAU^{-1} \) is the operator which multiplies each basis vector \( \delta^j = Uu_j \) by the corresponding eigenvalue \( \alpha_j \).

**Example.** Let \( a, b \in c_0(\mathbb{N}) \) be real-valued sequences and consider the operator

\[
(Jc)_j := a_j c_{j+1} + b_j c_j + a_{j-1} c_{j-1}.
\]

If \( A, B \) denote the multiplication operators by the sequences \( a, b \), respectively, then we already know that \( A \) and \( B \) are compact. Moreover, using the shift operators \( S^\pm \) we can write

\[
J = AS^+ + B + S^- A,
\]

which shows that \( J \) is self-adjoint since \( A^* = A, B^* = B \), and \((S^\pm)^* = S^{\mp}\). Hence we can conclude that \( J \) has a countable number of eigenvalues converging to zero and a corresponding orthonormal basis of eigenvectors.

In particular, in the new picture it is easy to define functions of our operator (thus extending the functional calculus from Problem 1.32). To this end set \( \Sigma := \overline{\{\alpha_j\}_{j \in J}} \) and denote by \( B(K) \) the Banach algebra of bounded functions \( F : K \to \mathbb{C} \) together with the sup norm.

**Corollary 3.10** (Functional calculus). Let \( A \) be a compact symmetric operator with associated orthonormal basis of eigenvectors \( \{u_j\}_{j \in J} \) and corresponding eigenvalues \( \{\alpha_j\}_{j \in J} \). Suppose \( F \in B(\Sigma) \), then

\[
F(A)f = \sum_{j \in J} F(\alpha_j)(u_j, f)u_j
\]

defines a continuous algebra homomorphism from the the Banach algebra \( B(\Sigma) \) to the algebra \( \mathcal{L}(\mathcal{H}) \) with \( 1(A) = 1 \) and \( \|A\| = A \). Moreover \( F(A)^* = F^*(A) \), where \( F^* \) is the function which takes complex conjugate values.
3.2. The spectral theorem for compact symmetric operators

Proof. This is straightforward to check for multiplication operators in $\ell^2(J)$ and hence the result follows by the previous corollary. $\square$

In many applications $F$ will be given by a function on $\mathbb{R}$ (or at least on $[-\|A\|, \|A\|]$) and since only the values $F(\alpha_j)$ are used two functions which agree on all eigenvalues will give the same result.

As a brief application we will say a few words about general spectral theory for bounded operators $A \in \mathcal{L}(X)$ in a Banach space $X$. In the finite dimensional case, the spectrum is precisely the set of eigenvalues. In the infinite dimensional case one defines the spectrum as

$$\sigma(A) := \{ z \in \mathbb{C} | \exists (A - z)^{-1} \in \mathcal{L}(X) \}. \quad (3.10)$$

It is important to emphasize that the inverse is required to exist as a bounded operator. Hence there are several ways in which this can fail: First of all, $A - z$ could not be injective. In this case $z$ is an eigenvalue and thus all eigenvalues belong to the spectrum. Secondly, it could not be surjective. And finally, even if it is bijective it could be unbounded. However, it will follow from the open mapping theorem that this last case cannot happen for a bounded operator. The inverse of $A - z$ for $z \notin \sigma(A)$ is known as the resolvent of $A$ and plays a crucial role in spectral theory. Using Problem 1.31 one can show that the complement of the spectrum is open, and hence the spectrum is closed. Since we will discuss this in detail in Chapter 6 we will not pursue this here but only look at our special case of symmetric compact operators.

To compute the inverse of $A - z$ we will use the functional calculus: To this end consider $F(\alpha) = \frac{1}{\alpha - z}$. Of course this function is unbounded on $\mathbb{R}$ but if $z$ is neither an eigenvalue nor zero it is bounded on $\Sigma$ and hence satisfies our requirements. Then

$$R_A(z)f := \sum_{j \in J} \frac{1}{\alpha_j - z} \langle u_j, f \rangle u_j \quad (3.11)$$

satisfies $(A - z)R_A(z) = R_A(z)(A - z) = I$, that is, $R_A(z) = (A - z)^{-1} \in \mathcal{L}(\ell^2)$. Of course, if $z$ is an eigenvalue, then the above formula breaks down. However, in the infinite dimensional case it also breaks down if $z = 0$ even if 0 is not an eigenvalue! In this case the above definition will still give an operator which is the inverse of $A - z$, however, since the sequence $\alpha_j^{-1}$ is unbounded, so will be the corresponding multiplication operator in $\ell^2(J)$ and the sum in (3.11) will only converge if $\{\alpha_j^{-1} \langle u_j, f \rangle\}_{j \in J} \in \ell^2(J)$. So in the infinite dimensional case 0 is in the spectrum even if it is not an eigenvalue. In particular,

$$\sigma(A) = \overline{\{ \alpha_j \}_{j \in J}}. \quad (3.12)$$
Moreover, if we use \( \frac{1}{\alpha_j - z} = \frac{\alpha_j}{z(\alpha_j - z)} - \frac{1}{z} \) we can rewrite this as

\[
R_A(z)f = \frac{1}{z} \left( \sum_{j=1}^{N} \frac{\alpha_j}{\alpha_j - z} \langle u_j, f \rangle u_j - f \right)
\]

where it suffices to take the sum over all nonzero eigenvalues.

This is all we need and it remains to apply these results to Sturm–Liouville operators.

**Problem 3.4.** Show that if \( A \) is bounded, then every eigenvalue \( \alpha \) satisfies \( |\alpha| \leq \|A\| \).

**Problem 3.5.** Find the eigenvalues and eigenfunctions of the integral operator

\[
(Kf)(x) := \int_0^1 u(x)v(y)f(y)dy
\]

in \( L^2_{\text{cont}}(0,1) \), where \( u(x) \) and \( v(x) \) are some given continuous functions.

**Problem 3.6.** Find the eigenvalues and eigenfunctions of the integral operator

\[
(Kf)(x) := 2 \int_0^1 (2xy - x - y + 1)f(y)dy
\]

in \( L^2_{\text{cont}}(0,1) \).

### 3.3. Applications to Sturm–Liouville operators

Now, after all this hard work, we can show that our Sturm–Liouville operator

\[
L := -\frac{d^2}{dx^2} + q(x),
\]

where \( q \) is continuous and real, defined on

\[
\mathcal{D}(L) := \{ f \in C^2[0,1] | f(0) = f(1) = 0 \} \subset L^2_{\text{cont}}(0,1),
\]

has an orthonormal basis of eigenfunctions.

The corresponding eigenvalue equation \( Lu = zu \) explicitly reads

\[
- u''(x) + q(x)u(x) = zu(x).
\]

It is a second order homogeneous linear ordinary differential equations and hence has two linearly independent solutions. In particular, specifying two initial conditions, e.g. \( u(0) = 0, u'(0) = 1 \) determines the solution uniquely. Hence, if we require \( u(0) = 0 \), the solution is determined up to a multiple and consequently the additional requirement \( u(1) = 0 \) cannot be satisfied by a nontrivial solution in general. However, there might be some \( z \in \mathbb{C} \) for which the solution corresponding to the initial conditions \( u(0) = 0, u'(0) = 1 \)
happens to satisfy $u(1) = 0$ and these are precisely the eigenvalues we are looking for.

Note that the fact that $L^2_{cont}(0,1)$ is not complete causes no problems since we can always replace it by its completion $H = L^2(0,1)$. A thorough investigation of this completion will be given later, at this point this is not essential.

We first verify that $L$ is symmetric:

$$\langle f, Lg \rangle = \int_0^1 f'(x)g'(x)dx + \int_0^1 f(x)g(x)dx = \int_0^1 -f''(x)g(x)dx + \int_0^1 f(x)g(x)dx$$

(3.16)

$$= \langle Lf, g \rangle.$$  

Here we have used integration by parts twice (the boundary terms vanish due to our boundary conditions $f(0) = f(1) = 0$ and $g(0) = g(1) = 0$).

Of course we want to apply Theorem 3.7 and for this we would need to show that $L$ is compact. But this task is bound to fail, since $L$ is not even bounded (see the example on page 59)!

So here comes the trick: If $L$ is unbounded its inverse $L^{-1}$ might still be bounded. Moreover, $L^{-1}$ might even be compact and this is the case here! Since $L$ might not be injective (0 might be an eigenvalue), we consider $R_L(z) := (L - z)^{-1}$, $z \in \mathbb{C}$, which is also known as the resolvent of $L$.

In order to compute the resolvent, we need to solve the inhomogeneous equation $(L - z)f = g$. This can be done using the variation of constants formula from ordinary differential equations which determines the solution up to an arbitrary solution of the homogeneous equation. This homogeneous equation has to be chosen such that $f \in \mathcal{D}(L)$, that is, such that $f(0) = f(1) = 0$.

Define

$$f(x) := \frac{u_+(z, x)}{W(z)} \left( \int_0^x u_-(z, t)g(t)dt \right)$$

$$+ \frac{u_-(z, x)}{W(z)} \left( \int_x^1 u_+(z, t)g(t)dt \right),$$

(3.17)

where $u_\pm(z, x)$ are the solutions of the homogeneous differential equation $-u''_\pm(z, x) + (q(x) - z)u_\pm(z, x) = 0$ satisfying the initial conditions $u_-(z, 0) = 0$, $u'_-(z, 0) = 1$ respectively $u_+(z, 1) = 0$, $u'_+(z, 1) = 1$ and

$$W(z) := W(u_+(z), u_-(z)) = u'_-(z, x)u_+(z, x) - u_-(z, x)u'_+(z, x)$$

(3.18)
is the Wronski determinant, which is independent of $x$ (check this!).

Then clearly $f(0) = 0$ since $u_-(z,0) = 0$ and similarly $f(1) = 0$ since $u_+(z,1) = 0$. Furthermore, $f$ is differentiable and a straightforward computation verifies

$$f'(x) = \frac{u_+(z,x)'}{W(z)} \left( \int_0^x u_-(z,t) g(t) dt \right) + \frac{u_-(z,x)'}{W(z)} \left( \int_x^1 u_+(z,t) g(t) dt \right).$$

(3.19)

Thus we can differentiate once more giving

$$f''(x) = \frac{u_+(z,x)''}{W(z)} \left( \int_0^x u_-(z,t) g(t) dt \right) + \frac{u_-(z,x)''}{W(z)} \left( \int_x^1 u_+(z,t) g(t) dt \right) - g(x)$$

$$= (q(x) - z) f(x) - g(x).$$

(3.20)

In summary, $f$ is in the domain of $L$ and satisfies $(L - z) f = g$.

Note that $z$ is an eigenvalue if and only if $W(z) = 0$. In fact, in this case $u_+(z,x)$ and $u_-(z,x)$ are linearly dependent and hence $u_+(z,x) = c u_-(z,x)$ with $c = u_+(z,0)$. Evaluating this last identity at $x = 0$ shows $u_+(z,0) = c u_-(z,0) = 0$ that $u_-(z,x)$ satisfies both boundary conditions and is thus an eigenfunction.

Introducing the Green function

$$G(z,x,t) := \frac{1}{W(u_+(z),u_-(z))} \begin{cases} 
  u_+(z,x) u_-(z,t), & x \geq t, \\
  u_+(z,t) u_-(z,x), & x \leq t,
\end{cases}$$

we see that $(L - z)^{-1}$ is given by

$$(L - z)^{-1} g(x) = \int_0^1 G(z,x,t) g(t) dt.$$  

(3.22)

Moreover, from $G(z,x,t) = G(z,t,x)$ it follows that $(L - z)^{-1}$ is symmetric for $z \in \mathbb{R}$ (Problem 3.7) and from Lemma 3.4 it follows that it is compact. Hence Theorem 3.7 applies to $(L - z)^{-1}$ once we show that we can find a real $z$ which is not an eigenvalue.

**Theorem 3.11.** The Sturm–Liouville operator $L$ has a countable number of discrete and simple eigenvalues $E_n$ which accumulate only at $\infty$. They are bounded from below and can hence be ordered as follows:

$$\min_{x \in [a,b]} q(x) \leq E_0 < E_1 < \cdots.$$ 

(3.23)
The corresponding normalized eigenfunctions \( u_n \) form an orthonormal basis for \( L^2_{\text{cont}}(0,1) \), that is, every \( f \in L^2_{\text{cont}}(0,1) \) can be written as

\[
f(x) = \sum_{n=0}^{\infty} \langle u_n, f \rangle u_n(x).
\]  

Moreover, for \( f \in \mathcal{D}(L) \) this series is uniformly convergent.

**Proof.** If \( E_j \) is an eigenvalue with corresponding normalized eigenfunction \( u_j \) we have

\[
E_j = \langle u_j, Lu_j \rangle = \int_0^1 (|u_j'(x)|^2 + q(x)|u_j(x)|^2) \, dx \geq \min_{x \in [0,1]} q(x)
\]  

where we have used integration by parts as in (3.16). Hence the eigenvalues are bounded from below.

Now pick a value \( \lambda \in \mathbb{R} \) such that \( R_L(\lambda) \) exists (\( \lambda < \min_{x \in [0,1]} q(x) \) say). By Lemma 3.4 \( R_L(\lambda) \) is compact and by Lemma 3.3 this remains true if we replace \( L^2_{\text{cont}}(0,1) \) by its completion. By Theorem 3.7 there are eigenvalues \( \alpha_n \) of \( R_L(\lambda) \) with corresponding eigenfunctions \( u_n \). Moreover, \( R_L(\lambda)u_n = \alpha_n u_n \) is equivalent to \( Lu_n = (\lambda + \frac{1}{\alpha_n})u_n \), which shows that \( E_n = \lambda + \frac{1}{\alpha_n} \) are eigenvalues of \( L \) with corresponding eigenfunctions \( u_n \). Now everything follows from Theorem 3.7 except that the eigenvalues are simple.

To show this, observe that if \( u_n \) and \( v_n \) are two different eigenfunctions corresponding to \( E_n \), then \( u_n(0) = v_n(0) = 0 \) implies \( W(u_n, v_n) = 0 \) and hence \( u_n \) and \( v_n \) are linearly dependent.

To show that (3.24) converges uniformly if \( f \in \mathcal{D}(L) \) we begin by writing \( f = R_L(\lambda)g \), \( g \in L^2_{\text{cont}}(0,1) \), implying

\[
\sum_{n=0}^{\infty} \langle u_n, f \rangle u_n(x) = \sum_{n=0}^{\infty} \langle R_L(\lambda)u_n, g \rangle u_n(x) = \sum_{n=0}^{\infty} \alpha_n \langle u_n, g \rangle u_n(x).
\]

Moreover, the Cauchy–Schwarz inequality shows

\[
\left| \sum_{j=m}^{n} \alpha_j \langle u_j, g \rangle u_j(x) \right|^2 \leq \sum_{j=m}^{n} |\langle u_j, g \rangle|^2 \sum_{j=m}^{n} |\alpha_j u_j(x)|^2.
\]

Now, by (2.18), \( \sum_{j=0}^{\infty} |\langle u_j, g \rangle|^2 = \|g\|^2 \) and hence the first term is part of a convergent series. Similarly, the second term can be estimated independent of \( x \) since

\[
\alpha_n u_n(x) = R_L(\lambda)u_n(x) = \int_0^1 G(\lambda, x, t)u_n(t) \, dt = \langle u_n, G(\lambda, x, .) \rangle
\]
implies
\[ \sum_{j=m}^{n} |\alpha_j u_j(x)|^2 \leq \sum_{j=0}^{\infty} |\langle u_j, G(\lambda, x, \cdot) \rangle|^2 = \int_{0}^{1} |G(\lambda, x, t)|^2 dt \leq M(\lambda)^2, \]
where \( M(\lambda) := \max_{x, t \in [0,1]} |G(\lambda, x, t)|, \) again by (2.18). □

Moreover, it is even possible to weaken our assumptions for uniform convergence. To this end we consider the sequilinear form associated with \( L \):
\[ s_L(f, g) := \langle f, Lg \rangle = \int_{0}^{1} \left( f'(x)^* g'(x) + q(x) f(x)^* g(x) \right) dx \quad (3.26) \]
for \( f, g \in \mathcal{D}(L) \), where we have used integration by parts as in (3.16). In fact, the above formula continues to hold for \( f \) in a slightly larger class of functions,

\[ \Omega(L) := \{ f \in C^1_p[0,1] | f(0) = f(1) = 0 \} \supseteq \mathcal{D}(L), \quad (3.27) \]

which we call the form domain of \( L \). Here \( C^1_p[a, b] \) denotes the set of piecewise continuously differentiable functions \( f \) in the sense that \( f \) is continuously differentiable except for a finite number of points at which it is continuous and the derivative has limits from the left and right. In fact, any class of functions for which the partial integration needed to obtain (3.26) can be justified would be good enough (e.g. the set of absolutely continuous functions to be discussed in Section 11.8).

**Lemma 3.12.** For a regular Sturm–Liouville problem (3.24) converges uniformly provided \( f \in \Omega(L) \).

**Proof.** By replacing \( L \to L - q_0 \) for \( q_0 > \min_{x \in [0,1]} q(x) \) we can assume \( q(x) > 0 \) without loss of generality. (This will shift the eigenvalues \( E_n \to E_n - q_0 \) and leave the eigenvectors unchanged.) In particular, we have \( q_L(f) := s_L(f, f) > 0 \) after this change. By (3.26) we also have \( E_j = \langle u_j, Lu_j \rangle = q_L(u_j) > 0 \).

Now let \( f \in \Omega(L) \) and consider (3.24). Then, observing that \( s_L(f, g) \) is a symmetric sesquilinear form (after our shift it is even a scalar product) as
well as \( s_L(f, u_j) = E_j \langle f, u_j \rangle \) one obtains

\[
0 \leq q_L \left( f - \sum_{j=m}^{n} \langle u_j, f \rangle u_j \right) = q_L(f) - \sum_{j=m}^{n} \langle u_j, f \rangle s_L(f, u_j) \\
- \sum_{j=m}^{n} \langle u_j, f \rangle^* s_L(u_j, f) + \sum_{j,k=m}^{n} \langle u_j, f \rangle^* \langle u_k, f \rangle s_L(u_j, u_k) \\
= q_L(f) - \sum_{j=m}^{n} E_j |\langle u_j, f \rangle|^2
\]

which implies

\[
\sum_{j=m}^{n} E_j |\langle u_j, f \rangle|^2 \leq q_L(f).
\]

In particular, note that this estimate applies to \( f(y) = G(\lambda, x, y) \). Now we can proceed as in the proof of the previous theorem (with \( \lambda = 0 \) and \( \alpha_j = E_j^{-1} \))

\[
\sum_{j=m}^{n} |\langle u_j, f \rangle u_j(x)| = \sum_{j=m}^{n} E_j |\langle u_j, f \rangle \langle u_j, G(0, x, .) \rangle| \\
\leq \left( \sum_{j=m}^{n} E_j |\langle u_j, f \rangle|^2 \sum_{j=m}^{n} E_j |\langle u_j, G(0, x, .) \rangle|^2 \right)^{1/2} \\
< q_L(f)^{1/2} q_L(G(0, x, .))^{1/2},
\]

where we have used the Cauchy–Schwarz inequality for the weighted scalar product \( (f_j, g_j) \rightarrow \sum_j f_j^* g_j E_j \). Finally note that \( q_L(G(0, x, .)) \) is continuous with respect to \( x \) and hence can be estimated by its maximum over \([0, 1]\). \( \square \)

Another consequence of the computations in the previous proof is also worthwhile noting:

**Corollary 3.13.** We have

\[
G(z, x, y) = \sum_{j=0}^{\infty} \frac{1}{E_j - z} u_j(x) u_j(y),
\]

where the sum is uniformly convergent. Moreover, we have the following trace formula

\[
\int_0^1 G(z, x, x) dx = \sum_{j=0}^{\infty} \frac{1}{E_j - z}.
\]

(3.28)
Proof. Using the conventions from the proof of the previous lemma we compute
\[ \langle u_j, G(0, x, \cdot) \rangle = \int_0^1 G(0, x, y)u_j(y)dy = R_L(0)u_j(x) = E_j^{-1}u_j(x) \]
which already proves (3.28) if \( x \) is kept fixed and the convergence of the sum is regarded in \( L^2 \) with respect to \( y \). However, the calculations from our previous lemma show
\[ \sum_{j=0}^{\infty} \frac{1}{E_j} |u_j(x)|^2 = E_j \sum_{j=0}^{\infty} |\langle u_j, G(0, x, \cdot) \rangle|^2 \leq q_L(G(0, x, \cdot)) \]
which proves uniform convergence of our sum
\[ \sum_{j=0}^{\infty} \frac{1}{|E_j - z|} |u_j(x)u_j(y)| \leq \sup_j \frac{E_j}{|E_j - z|} q_L(G(0, x, \cdot))^{1/2}, \]
where we have used the Cauchy–Schwarz inequality for the weighted scalar product \((f_j, g_j) \mapsto \sum_j f_j^* g_j E_j^{-1}\).

Finally, the last claim follows upon computing the integral using (3.28) and observing \( \|u_j\| = 1 \). \( \square \)

Example. Let us look at the Sturm–Liouville problem with \( q = 0 \). Then the underlying differential equation is
\[ -u''(x) = zu(x) \]
whose solution is given by \( u(x) = c_1 \sin(\sqrt{z}x) + c_2 \cos(\sqrt{z}x) \). The solution satisfying the boundary condition at the left endpoint is \( u_-(z, x) = \sin(\sqrt{z}x) \) and it will be an eigenfunction if and only if \( u_-(z, 1) = \sin(\sqrt{z}) = 0 \). Hence the corresponding eigenvalues and normalized eigenfunctions are
\[ E_n = \pi^2 n^2, \quad u_n(x) = \sqrt{2} \sin(n\pi x), \quad n \in \mathbb{N}. \]
Moreover, every function \( f \in H_0 \) can be expanded into a Fourier sine series
\[ f(x) = \sum_{n=1}^{\infty} f_n u_n(x) \quad f_n := \int_0^1 u_n(x)f(x)dx, \]
which is convergent with respect to our scalar product. If \( f \in C^1_p[0, 1] \) with \( f(0) = f(1) = 0 \) the series will converge uniformly. For an application of the trace formula see Problem 3.10. \( \diamond \)

Example. We could also look at the same equation as in the previous problem but with different boundary conditions
\[ u'(0) = u'(1) = 0. \]
Then
\[ E_n = \pi^2 n^2, \quad u_n(x) = \begin{cases} 1, & n = 0, \\ \sqrt{2} \cos(n\pi x), & n \in \mathbb{N}. \end{cases} \]

Moreover, every function \( f \in \mathcal{S}_0 \) can be expanded into a Fourier cosine series
\[ f(x) = \sum_{n=1}^{\infty} f_n u_n(x), \quad f_n := \int_0^1 u_n(x)f(x)dx, \]
which is convergent with respect to our scalar product.

**Example.** Combining the last two examples we see that every symmetric function on \([-1,1]\) can be expanded into a Fourier cosine series and every anti-symmetric function into a Fourier sine series. Moreover, since every function \( f(x) \) can be written as the sum of a symmetric function \( f(x) + f(-x) \) and an anti-symmetric function \( \frac{f(x) - f(-x)}{2} \), it can be expanded into a Fourier series. Hence we recover Theorem 2.17.

Problem 3.7. Show that for our Sturm–Liouville operator \( u_{\pm}(z, x)^* = u_{\pm}(z^*, x) \). Conclude \( R_L(z)^* = R_L(z^*) \). (Hint: Problem 3.2.)

Problem 3.8. Show that the resolvent \( R_A(z) = (A - z)^{-1} \) (provided it exists and is densely defined) of a symmetric operator \( A \) is again symmetric for \( z \in \mathbb{R} \). (Hint: \( g \in \mathcal{D}(R_A(z)) \) if and only if \( g = (A - z)f \) for some \( f \in \mathcal{D}(A) \).)

Problem 3.9. Suppose \( E_0 > 0 \) and equip \( \Omega(L) \) with the scalar product \( s_L \). Show that
\[ f(x) = s_L(G(0, x, .), f). \]
In other words, point evaluations are continuous functionals associated with the vectors \( G(0, x, .) \in \Omega(L) \). In this context, \( G(0, x, y) \) is called a reproducing kernel.

Problem 3.10. Show that
\[ \sum_{n=1}^{\infty} \frac{1}{n^2 - z} = \frac{1 - \pi \sqrt{2} \cot(\pi \sqrt{z})}{2z}, \quad z \in \mathbb{C} \setminus \mathbb{N}. \]

In particular, for \( z = 0 \) this gives Euler’s solution of the Basel problem:
\[ \sum_{n=1}^{\infty} \frac{1}{n^2} = \frac{\pi^2}{6}. \]

In fact, comparing the power series of both sides at \( z = 0 \) gives
\[ \sum_{n=1}^{\infty} \frac{1}{n^{2k}} = \frac{(-1)^{k+1}(2\pi)^{2k}B_{2k}}{2(2k)!}, \quad k \in \mathbb{N}, \]
where $B_k$ are the Bernoulli numbers defined via $\frac{x}{e^x - 1} = \sum_{k=0}^{\infty} B_k \frac{x^k}{k!}$.

(Hint: Use the trace formula (3.29).)

**Problem 3.11.** Consider the Sturm–Liouville problem on a compact interval $[a, b]$ with domain

$$\mathcal{D}(L) = \{ f \in C^2[a, b] \mid f'(a) - \alpha f(a) = f'(b) - \beta f(b) = 0 \}$$

for some real constants $\alpha, \beta \in \mathbb{R}$. Show that Theorem 3.11 continues to hold except for the lower bound on the eigenvalues.

### 3.4. Estimating eigenvalues

In general, there is no way of computing eigenvalues and their corresponding eigenfunctions explicitly. Hence it is important to be able to determine the eigenvalues at least approximately.

Let $A$ be a self-adjoint operator which has a lowest eigenvalue $\alpha_1$ (e.g., $A$ is a Sturm–Liouville operator). Suppose we have a vector $f$ which is an approximation for the eigenvector $u_1$ of this lowest eigenvalue $\alpha_1$. Moreover, suppose we can write

$$A := \sum_{j=1}^{\infty} \alpha_j \langle u_j, \cdot \rangle u_j, \quad \mathcal{D}(A) := \{ f \in \mathcal{H} \mid \sum_{j=1}^{\infty} |\alpha_j \langle u_j, f \rangle|^2 < \infty \},$$

where $\{u_j\}_{j \in \mathbb{N}}$ is an orthonormal basis of eigenvectors. Since $\alpha_1$ is supposed to be the lowest eigenvalue we have $\alpha_j \geq \alpha_1$ for all $j \in \mathbb{N}$.

Writing $f = \sum_j \gamma_j u_j$, $\gamma_j = \langle u_j, f \rangle$, one computes

$$\langle f, Af \rangle = \langle f, \sum_{j=1}^{\infty} \alpha_j \gamma_j u_j \rangle = \sum_{j=1}^{\infty} \alpha_j |\gamma_j|^2, \quad f \in \mathcal{D}(A),$$

and we clearly have

$$\alpha_1 \leq \frac{\langle f, Af \rangle}{\|f\|^2}, \quad f \in \mathcal{D}(A),$$

with equality for $f = u_1$. In particular, any $f$ will provide an upper bound and if we add some free parameters to $f$, one can optimize them and obtain quite good upper bounds for the first eigenvalue. For example we could take some orthogonal basis, take a finite number of coefficients and optimize them. This is known as the Rayleigh–Ritz method.

**Example.** Consider the Sturm–Liouville operator $L$ with potential $q(x) = x$ and Dirichlet boundary conditions $f(0) = f(1) = 0$ on the interval $[0, 1]$. Our starting point is the quadratic form

$$q_L(f) := \langle f, Lf \rangle = \int_0^1 \left( |f'(x)|^2 + q(x)|f(x)|^2 \right) dx$$
which gives us the lower bound
\[ \langle f, Lf \rangle \geq \min_{0 \leq x \leq 1} q(x) = 0. \]

While the corresponding differential equation can in principle be solved in terms of Airy functions, there is no closed form for the eigenvalues.

First of all we can improve the above bound upon observing
\[ 0 \leq q(x) \leq 1 \]
which implies
\[ \langle f, L_0 f \rangle \leq \langle f, Lf \rangle \leq \langle f, (L_0 + 1)f \rangle, \quad f \in \mathcal{D}(L) = \mathcal{D}(L_0), \]
where \( L_0 \) is the Sturm–Liouville operator corresponding to \( q(x) = 0 \).

Since the lowest eigenvalue of \( L_0 \) is \( \pi^2 \), we obtain
\[ \pi^2 \leq E_1 \leq \pi^2 + 1 \]
for the lowest eigenvalue \( E_1 \) of \( L \).

Moreover, using the lowest eigenfunction \( f_1(x) = \sqrt{2} \sin(\pi x) \) of \( L_0 \) one obtains the improved upper bound
\[ E_1 \leq \langle f_1, Lf_1 \rangle = \pi^2 + \frac{1}{2} \approx 10.3696. \]

Taking the second eigenfunction \( f_2(x) = \sqrt{2} \sin(2\pi x) \) of \( L_0 \) we can make the ansatz \( f(x) = (1 + \gamma^2)^{-1/2} (f_1(x) + \gamma f_2(x)) \) which gives
\[ \langle f, Lf \rangle = \pi^2 + \frac{1}{2} + \frac{\gamma}{1 + \gamma^2} (3\pi^2 \gamma - \frac{32}{9\pi^2}). \]
The right-hand side has a unique minimum at \( \gamma = \frac{32}{27\pi^4 + \sqrt{1024 + 729\pi^8}} \) giving the bound
\[ E_1 \leq \frac{5}{2} \pi^2 + \frac{1}{2} - \frac{\sqrt{1024 + 729\pi^8}}{18\pi^2} \approx 10.3685 \]
which coincides with the exact eigenvalue up to five digits.

But is there also something one can say about the next eigenvalues? Suppose we know the first eigenfunction \( u_1 \). Then we can restrict \( A \) to the orthogonal complement of \( u_1 \) and proceed as before: \( E_2 \) will be the minimum of \( \langle f, Af \rangle \) over all \( f \) restricted to this subspace. If we restrict to the orthogonal complement of an approximating eigenfunction \( f_1 \), there will still be a component in the direction of \( u_1 \) left and hence the infimum of the expectations will be lower than \( E_2 \). Thus the optimal choice \( f_1 = u_1 \) will give the maximal value \( E_2 \).

**Theorem 3.14 (Max-min).** Let \( A \) be a symmetric operator and let \( \alpha_1 \leq \alpha_2 \leq \cdots \leq \alpha_N \) be eigenvalues of \( A \) with corresponding orthonormal eigenvectors
$u_1, u_2, \ldots, u_N$. Suppose

$$A = \sum_{j=1}^{N} \alpha_j \langle u_j, \cdot \rangle u_j + \tilde{A} \quad (3.33)$$

with $\tilde{A} \geq \alpha_N$. Then

$$\alpha_j = \sup_{f_1, \ldots, f_{j-1}} \inf_{f \in U(f_1, \ldots, f_{j-1})} \langle f, Af \rangle, \quad 1 \leq j \leq N, \quad (3.34)$$

where

$$U(f_1, \ldots, f_j) := \{ f \in \mathfrak{D}(A) \| f \| = 1, f \in \text{span}\{f_1, \ldots, f_j\} \}. \quad (3.35)$$

Proof. We have

$$\inf_{f \in U(f_1, \ldots, f_{j-1})} \langle f, Af \rangle \leq \alpha_j.$$

In fact, set $f = \sum_{k=1}^{j} \gamma_k u_k$ and choose $\gamma_k$ such that $f \in U(f_1, \ldots, f_{j-1})$. Then

$$\langle f, Af \rangle = \sum_{k=1}^{j} |\gamma_k|^2 \alpha_k \leq \alpha_j$$

and the claim follows.

Conversely, let $\gamma_k = \langle u_k, f \rangle$ and write $f = \sum_{k=1}^{j} \gamma_k u_k + \tilde{f}$. Then

$$\inf_{f \in U(u_1, \ldots, u_{j-1})} \langle f, Af \rangle = \inf_{f \in U(u_1, \ldots, u_{j-1})} \left( \sum_{k=j}^{N} |\gamma_k|^2 \alpha_k + \langle \tilde{f}, \tilde{A} \tilde{f} \rangle \right) = \alpha_j. \quad \Box$$

Of course if we are interested in the largest eigenvalues all we have to do is consider $-A$.

Note that this immediately gives an estimate for eigenvalues if we have a corresponding estimate for the operators. To this end we will write

$$A \leq B \quad \Leftrightarrow \quad \langle f, Af \rangle \leq \langle f, Bf \rangle, \quad f \in \mathfrak{D}(A) \cap \mathfrak{D}(B). \quad (3.36)$$

Corollary 3.15. Suppose $A$ and $B$ are symmetric operators with corresponding eigenvalues $\alpha_j$ and $\beta_j$ as in the previous theorem. If $A \leq B$ and $\mathfrak{D}(B) \subseteq \mathfrak{D}(A)$ then $\alpha_j \leq \beta_j$.

Proof. By assumption we have $\langle f, Af \rangle \leq \langle f, Bf \rangle$ for $f \in \mathfrak{D}(B)$ implying

$$\inf_{f \in U_A} \langle f, Af \rangle \leq \inf_{f \in U_B} \langle f, Af \rangle \leq \inf_{f \in U_B} \langle f, Bf \rangle,$$

where we have indicated the dependence of $U$ on the operator via a subscript. Taking the sup on both sides the claim follows. \quad \Box
Example. Let $L$ be again our Sturm–Liouville operator and $L_0$ the corresponding operator with $q(x) = 0$. Set $q_− = \min_{0 \leq x \leq 1} q(x)$ and $q_+ = \max_{0 \leq x \leq 1} q(x)$. Then $L_0 + q_− \leq L \leq L_0 + q_+$ implies

$$\pi^2 n^2 + q_− \leq E_n \leq \pi^2 n^2 + q_+.$$ 

In particular, we have proven the famous **Weyl asymptotic**

$$E_n = \pi^2 n^2 + O(1)$$

for the eigenvalues. ◦

There is also an alternative version which can be proven similar (Problem 3.12):

**Theorem 3.16** (Min-max). Let $A$ be as in the previous theorem. Then

$$\alpha_j = \inf_{V_j \subset D(A), \dim(V_j) = j} \sup_{f \in V_j, \|f\| = 1} \langle f, Af \rangle,$$  

(3.37)

where the inf is taken over subspaces with the indicated properties.

**Problem 3.12.** Prove Theorem 3.16.

**Problem 3.13.** Suppose $A, A_n$ are self-adjoint, bounded and $A_n \to A$. Then $\alpha_k(A_n) \to \alpha_k(A)$. (Hint: For $B$ self-adjoint $\|B\| \leq \varepsilon$ is equivalent to $-\varepsilon \leq B \leq \varepsilon$.)

### 3.5. Canonical form of compact operators

Our first aim is to find a generalization of Corollary 3.8 for general compact operators between Hilbert spaces. The key observation is that if $K \in \mathcal{C}(\mathcal{H}_1, \mathcal{H}_2)$ is compact, then $K^*K \in \mathcal{C}(\mathcal{H}_1)$ is compact and symmetric and thus, by Corollary 3.8, there is a countable orthonormal set $\{u_j\} \subset \mathcal{H}_1$ and nonzero real numbers $s_j^2 \neq 0$ such that

$$K^*Kf = \sum_j s_j^2 \langle u_j, f \rangle u_j.$$  

(3.38)

Moreover, $\|Ku_j\|^2 = \langle u_j, K^*Ku_j \rangle = \langle u_j, s_j^2 u_j \rangle = s_j^2$ shows that we can set

$$s_j := \|Ku_j\| > 0.$$  

(3.39)

The numbers $s_j = s_j(K)$ are called **singular values** of $K$. There are either finitely many singular values or they converge to zero.

**Theorem 3.17** (Canonical form of compact operators). Let $K \in \mathcal{C}(\mathcal{H}_1, \mathcal{H}_2)$ be compact and let $s_j$ be the singular values of $K$ and $\{u_j\} \subset \mathcal{H}_1$ corresponding orthonormal eigenvectors of $K^*K$. Then

$$K = \sum_j s_j \langle u_j, \cdot \rangle v_j,$$  

(3.40)
where \( v_j = s_j^{-1}Ku_j \). The norm of \( K \) is given by the largest singular value
\[
\|K\| = \max_j s_j(K). 
\] (3.41)

Moreover, the vectors \( \{v_j\} \subset \mathcal{H}_2 \) are again orthonormal and satisfy \( K^*v_j = s_ju_j \). In particular, \( v_j \) are eigenvectors of \( KK^* \) corresponding to the eigenvalues \( s_j^2 \).

**Proof.** For any \( f \in \mathcal{H}_1 \) we can write
\[
f = \sum_j \langle u_j, f \rangle u_j + f_{\perp}
\]
with \( f_{\perp} \in \text{Ker}(K^*K) = \text{Ker}(K) \) (Problem 3.14). Then
\[
Kf = \sum_j \langle u_j, f \rangle Ku_j = \sum_j s_j \langle u_j, f \rangle v_j
\]
as required. Furthermore,
\[
\langle v_j, v_k \rangle = (s_j s_k)^{-1} \langle Ku_j, Ku_k \rangle = (s_j s_k)^{-1} \langle K^*Ku_j, u_k \rangle = s_j s_k^{-1} \langle u_j, u_k \rangle
\]
shows that \( \{v_j\} \) are orthonormal. By definition \( K^*v_j = s_j^{-1}K^*Ku_j = s_ju_j \) which also shows \( KK^*v_j = s_jKu_j = s_j^2v_j \).

Finally, (3.41) follows using Bessel’s inequality
\[
\|Kf\|^2 = \| \sum_j s_j \langle u_j, f \rangle v_j \|^2 = \sum_j s_j^2 |\langle u_j, f \rangle|^2 \leq \left( \max_j s_j(K)^2 \right) \|f\|^2,
\]
where equality holds for \( f = u_{j_0} \) if \( s_{j_0} = \max_j s_j(K) \). \( \square \)

If \( K \in \mathcal{C}(\mathcal{H}) \) is self-adjoint, then \( u_j = \sigma_j v_j, \sigma_j^2 = 1 \), are the eigenvectors of \( K \) and \( \sigma_j s_j \) are the corresponding eigenvalues. In particular, for a self-adjoint operators the singular values are the absolute values of the nonzero eigenvalues.

The above theorem also gives rise to the **polar decomposition**
\[
K = U|K| = |K^*|U,
\] (3.42)
where
\[
|K| := \sqrt{K^*K} = \sum_j s_j \langle u_j, . \rangle u_j, \quad |K^*| = \sqrt{KK^*} = \sum_j s_j \langle v_j, . \rangle v_j
\] (3.43)
are self-adjoint (in fact nonnegative) and
\[
U := \sum_j \langle u_j, . \rangle v_j
\] (3.44)
is an isometry from \( \overline{\text{Ran}(K^*)} = \text{span}\{u_j\} \) onto \( \overline{\text{Ran}(K)} = \text{span}\{v_j\} \).

From the max-min theorem (Theorem 3.14) we obtain:
Lemma 3.18. Let $K \in \mathcal{C}(\mathcal{H}_1, \mathcal{H}_2)$ be compact; then
\begin{equation}
 s_j(K) = \min_{f_1, \ldots, f_{j-1}} \sup_{f \in U(f_1, \ldots, f_{j-1})} \|Kf\|,
\end{equation}
where $U(f_1, \ldots, f_j) := \{f \in \mathcal{H}_1 \mid \|f\| = 1, f \in \text{span}\{f_1, \ldots, f_j\}^\perp\}$.

In particular, note
\begin{align*}
 s_j(AK) &\leq \|A\| s_j(K), \\
 s_j(KA) &\leq \|A\| s_j(K)
\end{align*}
whenever $K$ is compact and $A$ is bounded (the second estimate follows from the first by taking adjoints).

An operator $K \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ is called a finite rank operator if its range is finite dimensional. The dimension
\[ \text{rank}(K) = \dim \text{Ran}(K) \]
is called the rank of $K$. Since for a compact operator
\begin{equation}
 \overline{\text{Ran}(K)} = \overline{\text{span}\{v_j\}}
\end{equation}
we see that a compact operator is finite rank if and only if the sum in (3.40) is finite. Note that the finite rank operators form an ideal in $\mathcal{L}(\mathcal{H})$ just as the compact operators do. Moreover, every finite rank operator is compact by the Heine–Borel theorem (Theorem 0.22).

Now truncating the sum in the canonical form gives us a simple way to approximate compact operators by finite rank ones. Moreover, this is in fact the best approximation within the class of finite rank operators:

Lemma 3.19. Let $K \in \mathcal{C}(\mathcal{H}_1, \mathcal{H}_2)$ be compact and let its singular values be ordered. Then
\begin{equation}
 s_j(K) = \min_{\text{rank}(F) < j} \|K - F\|,
\end{equation}
with equality for
\begin{equation}
 F_{j-1} := \sum_{k=1}^{j-1} s_k \langle u_k, \cdot, v_k \rangle.
\end{equation}
In particular, the closure of the ideal of finite rank operators in $\mathcal{L}(\mathcal{H})$ is the ideal of compact operators.

Proof. That there is equality for $F = F_{j-1}$ follows from (3.41). In general, the restriction of $F$ to $\text{span}\{u_1, \ldots, u_j\}$ will have a nontrivial kernel. Let $f = \sum_{k=1}^j \alpha_j u_j$ be a normalized element of this kernel, then $\|(K - F)f\|^2 = \|Kf\|^2 = \sum_{k=1}^j |\alpha_k s_k|^2 \geq s_j^2$.

In particular, every compact operator can be approximated by finite rank ones and since the limit of compact operators is compact, we cannot get more than the compact operators. \qed
Two more consequences are worthwhile noting.

**Corollary 3.20.** An operator $K \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ is compact if and only if $K^*K$ is.

**Proof.** Just observe that $K^*K$ compact is all that was used to show Theorem 3.17. □

**Corollary 3.21.** An operator $K \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ is compact (finite rank) if and only $K^* \in \mathcal{L}(\mathcal{H}_2, \mathcal{H}_1)$ is. In fact, $s_j(K) = s_j(K^*)$ and

$$K^* = \sum_j s_j\langle v_j, \cdot \rangle u_j.$$  

(3.50)

**Proof.** First of all note that (3.50) follows from (3.40) since taking adjoints is continuous and $(\langle u_j, \cdot \rangle v_j)^* = \langle v_j, \cdot \rangle u_j$ (cf. Problem 2.7). The rest is straightforward. □

From this last lemma one easily gets a number of useful inequalities for the singular values:

**Corollary 3.22.** Let $K_1$ and $K_2$ be compact and let $s_j(K_1)$ and $s_j(K_2)$ be ordered. Then

(i) $s_{j+k-1}(K_1 + K_2) \leq s_j(K_1) + s_k(K_2)$,

(ii) $s_{j+k-1}(K_1K_2) \leq s_j(K_1)s_k(K_2)$,

(iii) $|s_j(K_1) - s_j(K_2)| \leq \|K_1 - K_2\|$.

**Proof.** Let $F_1$ be of rank $j - 1$ and $F_2$ of rank $k - 1$ such that $\|K_1 - F_1\| = s_j(K_1)$ and $\|K_2 - F_2\| = s_k(K_2)$. Then $s_{j+k-1}(K_1 + K_2) \leq \|(K_1 + K_2) - (F_1 + F_2)\| = \|K_1 - F_1\| + \|K_2 - F_2\| = s_j(K_1) + s_k(K_2)$ since $F_1 + F_2$ is of rank at most $j + k - 2$.

Similarly $F = F_1(K_2 - F_2) + K_1 F_2$ is of rank at most $j + k - 2$ and hence

$s_{j+k-1}(K_1K_2) \leq \|K_1K_2 - F\| = \|(K_1 - F_1)(K_2 - F_2)\| \leq \|K_1 - F_1\|\|K_2 - F_2\| = s_j(K_1)s_k(K_2)$.

Next, choosing $k = 1$ and replacing $K_2 \rightarrow K_2 - K_1$ in (i) gives $s_j(K_2) \leq s_j(K_1) + \|K_2 - K_1\|$. Reversing the roles gives $s_j(K_1) \leq s_j(K_2) + \|K_1 - K_2\|$ and proves (iii). □

**Example.** One might hope that item (i) from the previous corollary can be improved to $s_j(K_1 + K_2) \leq s_j(K_1) + s_j(K_2)$. However, this is not the case as the following example shows:

$$K_1 := \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad K_2 := \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}.$$

Then $1 = s_2(K_1 + K_2) \not\leq s_2(K_1) + s_2(K_2) = 0$. □
Problem 3.14. Show that \( \text{Ker}(A^*A) = \text{Ker}(A) \) for any \( A \in \mathcal{L}(H_1, H_2) \).

Problem 3.15. Let \( K \) be multiplication by a sequence \( k \in c_0(\mathbb{N}) \) in the Hilbert space \( \ell^2(\mathbb{N}) \). What are the singular values of \( K \)?

Problem 3.16. Let \( K \) be multiplication by a sequence \( k \in c_0(\mathbb{N}) \) in the Hilbert space \( \ell^2(\mathbb{N}) \) and consider \( L = KS^* \). What are the singular values of \( L \)? Does \( L \) have any eigenvalues?

Problem 3.17. Let \( K \in \mathcal{C}(H_1, H_2) \) be compact and let its singular values be ordered. Let \( M \subseteq H_1, N \subseteq H_1 \) be subspaces with corresponding orthogonal projections \( P_M, P_N \), respectively. Then

\[
\sigma_j(K) = \min_{\dim(M)<j} \|K - KP_M\| = \min_{\dim(N)<j} \|K - P_NK\|,
\]

where the minimum is taken over all subspaces with the indicated dimension. Moreover, we have equality for

\[
M = \text{span}\{u_k\}_{k=1}^{j-1}, \quad N = \text{span}\{v_k\}_{k=1}^{j-1}.
\]

3.6. Hilbert–Schmidt and trace class operators

We can further subdivide the class of compact operators \( \mathcal{C}(\mathcal{H}) \) according to the decay of their singular values. We define

\[
\|K\|_p := \left( \sum_j \sigma_j(K)^p \right)^{1/p} \tag{3.51}
\]

plus corresponding spaces

\[
\mathcal{J}_p(\mathcal{H}) = \{ K \in \mathcal{C}(\mathcal{H}) \|K\|_p < \infty \}, \tag{3.52}
\]

which are known as Schatten \( p \)-classes. Even though our notation hints at the fact that \( \| \cdot \|_p \) is a norm, we will only prove this here for \( p = 1, 2 \) (the only nontrivial part is the triangle inequality). Note that by (3.41)

\[
\|K\| \leq \|K\|_p \tag{3.53}
\]

and that by \( \sigma_j(K) = \sigma_j(K^*) \) we have

\[
\|K\|_p = \|K^*\|_p. \tag{3.54}
\]

The two most important cases are \( p = 1 \) and \( p = 2 \): \( \mathcal{J}_2(\mathcal{H}) \) is the space of Hilbert–Schmidt operators and \( \mathcal{J}_1(\mathcal{H}) \) is the space of trace class operators.

Example. Any multiplication operator by a sequence from \( \ell^p(\mathbb{N}) \) is in the Schatten \( p \)-class of \( \mathcal{H} = \ell^2(\mathbb{N}) \). \( \diamond \)
Example. By virtue of the Weyl asymptotics (see the example on 113) the resolvent of our Sturm–Liouville operator is trace class.

Example. Let $k$ be a periodic function which is square integrable over $[-\pi, \pi]$. Then the integral operator

$$(Kf)(x) = \frac{1}{2\pi} \int_{-\pi}^{\pi} k(y-x)f(y)dy$$

has the eigenfunctions $u_j(x) = (2\pi)^{-1/2}e^{-ijx}$ with corresponding eigenvalues $\hat{k}_j$, $j \in \mathbb{Z}$, where $\hat{k}_j$ are the Fourier coefficients of $k$. Since $\{u_j\}_{j \in \mathbb{Z}}$ is an ONB we have found all eigenvalues. In particular, the Fourier transform maps $K$ to the multiplication operator with the sequence of its eigenvalues $\hat{k}_j$. Hence the singular values are the absolute values of the nonzero eigenvalues and (3.40) reads

$$K = \sum_{j \in \mathbb{Z}} \hat{k}_j \langle u_j, \cdot \rangle u_j.$$

Moreover, since the eigenvalues are in $l^2(\mathbb{Z})$ we see that $K$ is a Hilbert–Schmidt operator. If $k$ is continuous with summable Fourier coefficients (e.g. $k \in C^2_{\text{per}}[-\pi, \pi]$), then $K$ is trace class.

We first prove an alternate definition for the Hilbert–Schmidt norm.

Lemma 3.23. A bounded operator $K$ is Hilbert–Schmidt if and only if

$$\sum_{j \in J} \|Kw_j\|^2 < \infty$$  \hspace{1cm} (3.55)

for some orthonormal basis and

$$\|K\|_2 = \left(\sum_{j \in J} \|Kw_j\|^2\right)^{1/2},$$  \hspace{1cm} (3.56)

for every orthonormal basis in this case.

Proof. First of all note that (3.55) implies that $K$ is compact. To see this, let $P_n$ be the projection onto the space spanned by the first $n$ elements of the orthonormal basis $\{w_j\}$. Then $K_n = KP_n$ is finite rank and converges to $K$ since

$$\|(K - K_n)f\| = \|\sum_{j>n} c_j Kw_j\| \leq \sum_{j>n} |c_j|\|Kw_j\| \leq \left(\sum_{j>n} \|Kw_j\|^2\right)^{1/2}\|f\|,$$

where $f = \sum_j c_j w_j$. 

The rest follows from (3.40) and

\[ \sum_j \|Kw_j\|^2 = \sum_{k,j} |\langle v_k, Kw_j \rangle|^2 = \sum_k \|K^* v_k\|^2 = \sum_k s_k(K)^2 = \|K\|^2. \]

Here we have used \( \text{span}\{v_k\} = \text{Ker}(K^*)^\perp = \text{Ran}(K) \) in the first step. \( \square \)

**Corollary 3.24.** The Hilbert–Schmidt norm satisfies the triangle inequality and hence is indeed a norm.

**Proof.** This follows from (3.56) upon using the triangle inequality for \( H \) and for \( \ell^2(J) \). \( \square \)

Now we can show

**Lemma 3.25.** The set of Hilbert–Schmidt operators forms an ideal in \( L(H) \) and

\[ \|KA\|_2 \leq \|A\|\|K\|_2, \quad \|AK\|_2 \leq \|A\|\|K\|_2. \]  \( (3.57) \)

**Proof.** If \( K_1 \) and \( K_2 \) are Hilbert–Schmidt operators, then so is their sum since

\[ \|K_1 + K_2\|_2 = \left( \sum_{j \in J} \| (K_1 + K_2) w_j \|^2 \right)^{1/2} \leq \left( \sum_{j \in J} (\|K_1 w_j\| + \|K_2 w_j\|)^2 \right)^{1/2} \]

\[ \leq \|K_1\|_2 + \|K_2\|_2, \]

where we have used the triangle inequality for \( \ell^2(J) \).

Let \( K \) be Hilbert–Schmidt and \( A \) bounded. Then \( AK \) is compact and

\[ \|AK\|^2_2 = \sum_j \|AKw_j\|^2 \leq \|A\|^2 \sum_j \|Kw_j\|^2 = \|A\|^2 \|K\|^2_2. \]

For \( KA \) just consider adjoints. \( \square \)

**Example.** Consider \( \ell^2(\mathbb{N}) \) and let \( K \) be some compact operator. Let \( K_{jk} = \langle \delta^j, K \delta^k \rangle = (K \delta^j)_{k} \) be its matrix elements such that

\[ (Ka)_j = \sum_{k=1}^\infty K_{jk} a_k. \]

Then, choosing \( w_j = \delta^j \) in (3.56) we get

\[ \|K\|_2 = \left( \sum_{j=1}^\infty \|K \delta^j\|^2 \right)^{1/2} = \left( \sum_{j=1}^\infty \sum_{k=1}^\infty |K_{jk}|^2 \right)^{1/2}. \]

Hence \( K \) is Hilbert–Schmidt if and only if its matrix elements are in \( \ell^2(\mathbb{N} \times \mathbb{N}) \) and the Hilbert–Schmidt norm coincides with the \( \ell^2(\mathbb{N} \times \mathbb{N}) \) norm of
the matrix elements. Especially in the finite dimensional case the Hilbert–Schmidt norm is also known as **Frobenius norm**.

Of course the same calculation shows that a bounded operator is Hilbert–Schmidt if and only if its matrix elements \( \langle w_j, K w_k \rangle \) with respect to some orthonormal basis \( \{ w_j \}_{j \in J} \) are in \( \ell^2(J \times J) \) and the Hilbert–Schmidt norm coincides with the \( \ell^2(J \times J) \) norm of the matrix elements.

**Example.** Let \( I = [a, b] \) be a compact interval. Suppose \( K : L^2(I) \rightarrow C(I) \) is continuous, then \( K : L^2(I) \rightarrow L^2(I) \) is Hilbert–Schmidt with Hilbert–Schmidt norm \( \| K \|_2 \leq \sqrt{b - a} M \), where \( M := \| K \|_{L^2(I) \rightarrow C(I)} \).

To see this start by observing that point evaluations are continuous functionals on \( C(I) \) and hence \( f \mapsto \langle K f, x \rangle \) is a continuous linear functional on \( L^2(I) \) satisfying \( | \langle K f, x \rangle | \leq M \| f \| \). By the Riesz lemma there is some \( K_x \in L^2(I) \) with \( \| K_x \| \leq M \) such that

\[
\langle K f, x \rangle = \langle K_x, f \rangle
\]

and hence for any orthonormal basis \( \{ w_j \}_{j \in \mathbb{N}} \) we have

\[
\sum_{j \in \mathbb{N}} | \langle K w_j, x \rangle |^2 = \sum_{j \in \mathbb{N}} | \langle K_x, w_j \rangle |^2 = \| K_x \|^2 \leq M^2.
\]

But then

\[
\sum_{j \in \mathbb{N}} \| K w_j \|^2 = \sum_{j \in \mathbb{N}} \int_a^b | \langle K w_j, x \rangle |^2 dx = \int_a^b \left( \sum_{j \in \mathbb{N}} | \langle K w_j, x \rangle |^2 \right) dx
\]

\[
\leq (b - a) M^2
\]

as claimed.

Since Hilbert–Schmidt operators turn out easy to identify (cf. also Section 10.5), it is important to relate \( J_1(\mathcal{F}) \) with \( J_2(\mathcal{F}) \):

**Lemma 3.26.** An operator is trace class if and only if it can be written as the product of two Hilbert–Schmidt operators, \( K = K_1 K_2 \), and in this case we have

\[
\| K \|_1 \leq \| K_1 \|_2 \| K_2 \|_2.
\]  

(3.58)

**Proof.** Using (3.40) (where we can extend \( u_n \) and \( v_n \) to orthonormal bases if necessary) and Cauchy–Schwarz we have

\[
\| K \|_1 = \sum_{n} \langle v_n, K u_n \rangle = \sum_{n} | \langle K_1^* v_n, K_2 u_n \rangle |
\]

\[
\leq \left( \sum_{n} \| K_1^* v_n \|^2 \sum_{n} \| K_2 u_n \|^2 \right)^{1/2} = \| K_1 \|_2 \| K_2 \|_2
\]
and hence $K = K_1 K_2$ is trace class if both $K_1$ and $K_2$ are Hilbert–Schmidt operators. To see the converse, let $K$ be given by (3.40) and choose $K_1 = \sum_j \sqrt{s_j(\bar{K})} \langle u_j, \cdot \rangle v_j$, respectively, $K_2 = \sum_j \sqrt{s_j(\bar{K})} \langle u_j, \cdot \rangle u_j$. \hfill \Box

Now we can also explain the name trace class:

**Lemma 3.27.** If $K$ is trace class, then for every orthonormal basis $\{w_n\}$ the trace

$$\text{tr}(K) = \sum_n \langle w_n, K w_n \rangle$$

(3.59)

is finite,

$$|\text{tr}(K)| \leq \|K\|_1,$$

(3.60)

and independent of the orthonormal basis.

**Proof.** If we write $K = K_1 K_2$ with $K_1, K_2$ Hilbert–Schmidt, then the Cauchy–Schwarz inequality implies $|\text{tr}(K)| \leq \|K^*\|_2 \|K_2\|_2 \leq \|K\|_1$. Moreover, if $\{\tilde{w}_n\}$ is another orthonormal basis, we have

$$\sum_n \langle w_n, K_1 K_2 w_n \rangle = \sum_n \langle K_1^* w_n, K_2 w_n \rangle = \sum_{n,m} \langle K_1^* w_n, \tilde{w}_m \rangle \langle \tilde{w}_m, K_2 w_n \rangle$$

$$= \sum_{m,n} \langle K_2^* v_m, w_n \rangle \langle w_n, K_1 v_m \rangle = \sum_m \langle K_2^* \tilde{w}_m, K_1 \tilde{w}_m \rangle$$

$$= \sum_m \langle \tilde{w}_m, K_2 K_1 \tilde{w}_m \rangle.$$

In the special case $w = \tilde{w}$ we see $\text{tr}(K_1 K_2) = \text{tr}(K_2 K_1)$ and the general case now shows that the trace is independent of the orthonormal basis. \hfill \Box

Clearly for self-adjoint trace class operators, the trace is the sum over all eigenvalues (counted with their multiplicity). To see this, one just has to choose the orthonormal basis to consist of eigenfunctions. This is even true for all trace class operators and is known as Lidskij trace theorem (see [28] for an easy to read introduction).

**Example.** We already mentioned that the resolvent of our Sturm–Liouville operator is trace class. Choosing a basis of eigenfunctions we see that the trace of the resolvent is the sum over its eigenvalues and combining this with our trace formula (3.29) gives

$$\text{tr}(R_L(z)) = \sum_{j=0}^{\infty} \frac{1}{E_j - z} = \int_0^1 G(z, x, x) dx$$

for $z \in \mathbb{C}$ no eigenvalue. \hfill \diamond
Example. For our integral operator $K$ from the example on page 118 we have in the trace class case

$$\text{tr}(K) = \sum_{j \in \mathbb{Z}} \hat{k}_j = k(0).$$

Note that this can again be interpreted as the integral over the diagonal $(2\pi)^{-1}k(x-x) = (2\pi)^{-1}k(0)$ of the kernel. 

We also note the following elementary properties of the trace:

**Lemma 3.28.** Suppose $K$, $K_1$, $K_2$ are trace class and $A$ is bounded.

(i) The trace is linear.

(ii) $\text{tr}(K^*) = \text{tr}(K)^*$.

(iii) If $K_1 \leq K_2$, then $\text{tr}(K_1) \leq \text{tr}(K_2)$.

(iv) $\text{tr}(AK) = \text{tr}(KA)$.

**Proof.** (i) and (ii) are straightforward. (iii) follows from $K_1 \leq K_2$ if and only if $\langle f, K_1 f \rangle \leq \langle f, K_2 f \rangle$ for every $f \in \mathcal{H}$. (iv) By Problem 2.12 and (i), it is no restriction to assume that $A$ is unitary. Let $\{w_n\}$ be some ONB and note that $\{\tilde{w}_n = Aw_n\}$ is also an ONB. Then

$$\text{tr}(AK) = \sum_n \langle \tilde{w}_n, AK \tilde{w}_n \rangle = \sum_n \langle Aw_n, AKAw_n \rangle = \sum_n \langle w_n, KAw_n \rangle = \text{tr}(KA)$$

and the claim follows. \hfill \Box

We also mention a useful criterion for $K$ to be trace class.

**Lemma 3.29.** An operator $K$ is trace class if and only if it can be written as

$$K = \sum_j \langle f_j, . \rangle g_j \quad (3.61)$$

for some sequences $f_j$, $g_j$ satisfying

$$\sum_j \|f_j\| \|g_j\| < \infty. \quad (3.62)$$

Moreover, in this case

$$\|K\|_1 = \min \sum_j \|f_j\| \|g_j\|, \quad (3.63)$$

where the minimum is taken over all representations as in (3.61).
Proof. To see that a trace class operator (3.40) can be written in such a way choose \( f_j = u_j, \) \( g_j = s_j v_j. \) This also shows that the minimum in (3.63) is attained. Conversely note that the sum converges in the operator norm and hence \( K \) is compact. Moreover, for every finite \( N \) we have

\[
\sum_{k=1}^{N} s_k = \sum_{k=1}^{N} \langle v_k, Ku_k \rangle = \sum_{j=1}^{N} \sum_{k=1}^{N} \langle v_k, g_j \rangle \langle f_j, u_k \rangle = \sum_{j} \sum_{k=1}^{N} \langle v_k, g_j \rangle \langle f_j, u_k \rangle
\]

\[
\leq \sum_{j} \left( \sum_{k=1}^{N} |\langle v_k, g_j \rangle|^2 \right)^{1/2} \left( \sum_{k=1}^{N} |\langle f_j, u_k \rangle|^2 \right)^{1/2} \leq \sum_{j} \|f_j\| \|g_j\|.
\]

This also shows that the right-hand side in (3.63) cannot exceed \( \|K\|_1. \) To see the last claim we choose an ONB \( \{w_k\} \) to compute the trace

\[
\text{tr}(K) = \sum_{k} \langle w_k, Kw_k \rangle = \sum_{k} \sum_{j} \langle w_k, f_j \rangle \langle w_k, g_j \rangle = \sum_{j} \sum_{k} \langle w_k, f_j \rangle \langle w_k, g_j \rangle
\]

\[
= \sum_{j} \langle f_j, g_j \rangle. \quad \square
\]

An immediate consequence of (3.63) is:

**Corollary 3.30.** The trace norm satisfies the triangle inequality and hence is indeed a norm.

Finally, note that

\[
\|K\|_2 = \left( \text{tr}(K^* K) \right)^{1/2}
\]

which shows that \( J_2(\mathcal{H}) \) is in fact a Hilbert space with scalar product given by

\[
\langle K_1, K_2 \rangle = \text{tr}(K_1^* K_2).
\]

**Problem 3.18.** Let \( \mathcal{H} := \ell^2(\mathbb{N}) \) and let \( A \) be multiplication by a sequence \( a = (a_j)_{j=1}^\infty. \) Show that \( A \) is Hilbert–Schmidt if and only if \( a \in \ell^2(\mathbb{N}). \)

Furthermore, show that \( \|A\|_2 = \|a\| \) in this case.

**Problem 3.19.** An operator of the form \( K : \ell^2(\mathbb{N}) \to \ell^2(\mathbb{N}), \) \( f_n \mapsto \sum_{j \in \mathbb{N}} k_{n+j} f_j \) is called Hankel operator.

- Show that \( K \) is Hilbert–Schmidt if and only if \( \sum_{j \in \mathbb{N}} |k_{j+1}|^2 < \infty \) and this number equals \( \|K\|_2. \)
- Show that \( K \) is Hilbert–Schmidt with \( \|K\|_2 \leq \|c\|_1 \) if \( |k_j| \leq c_j, \) where \( c_j \) is decreasing and summable.

(Hint: For the first item use summation by parts.)
Chapter 4

The main theorems about Banach spaces

4.1. The Baire theorem and its consequences

Recall that the interior of a set is the largest open subset (that is, the union of all open subsets). A set is called nowhere dense if its closure has empty interior. The key to several important theorems about Banach spaces is the observation that a Banach space cannot be the countable union of nowhere dense sets.

Theorem 4.1 (Baire category theorem). Let $X$ be a (nonempty) complete metric space. Then $X$ cannot be the countable union of nowhere dense sets.

Proof. Suppose $X = \bigcup_{n=1}^{\infty} X_n$. We can assume that the sets $X_n$ are closed and none of them contains a ball; that is, $X \setminus X_n$ is open and nonempty for every $n$. We will construct a Cauchy sequence $x_n$ which stays away from all $X_n$.

Since $X \setminus X_1$ is open and nonempty, there is a ball $B_{r_1}(x_1) \subseteq X \setminus X_1$. Reducing $r_1$ a little, we can even assume $\overline{B}_{r_1}(x_1) \subseteq X \setminus X_1$. Moreover, since $X_2$ cannot contain $B_{r_1}(x_1)$, there is some $x_2 \in B_{r_1}(x_1)$ that is not in $X_2$. Since $B_{r_1}(x_1) \cap (X \setminus X_2)$ is open, there is a closed ball $\overline{B}_{r_2}(x_2) \subseteq B_{r_1}(x_1) \cap (X \setminus X_2)$. Proceeding recursively, we obtain a sequence (here we use the axiom of choice) of balls such that

$$\overline{B}_{r_n}(x_n) \subseteq B_{r_{n-1}}(x_{n-1}) \cap (X \setminus X_n).$$

Now observe that in every step we can choose $r_n$ as small as we please; hence without loss of generality $r_n \to 0$. Since by construction $x_n \in B_{r_N}(x_N)$ for
4. The main theorems about Banach spaces

\( n \geq N \), we conclude that \( x_n \) is Cauchy and converges to some point \( x \in X \). But \( x \in B_{r_n}(x_n) \subseteq X \setminus X_n \) for every \( n \), contradicting our assumption that the \( X_n \) cover \( X \).

In other words, if \( X_n \subseteq X \) is a sequence of closed subsets which cover \( X \), at least one \( X_n \) contains a ball of radius \( \varepsilon > 0 \).

**Example.** The set of rational numbers \( \mathbb{Q} \) can be written as a countable union of its elements. This shows that the completeness assumption is crucial.

Remark: Sets which can be written as the countable union of nowhere dense sets are said to be of **first category** or **meager**. All other sets are **second category** or **fat**. Hence explaining the name category theorem.

Since a closed set is nowhere dense if and only if its complement is open and dense (cf. Problem 0.6), there is a reformulation which is also worthwhile noting:

**Corollary 4.2.** Let \( X \) be a complete metric space. Then any countable intersection of open dense sets is again dense.

**Proof.** Let \( \{O_n\} \) be a family of open dense sets whose intersection is not dense. Then this intersection must be missing some closed ball \( B_{\varepsilon} \). This ball will lie in \( \bigcup_n X_n \), where \( X_n := X \setminus O_n \) are closed and nowhere dense. Now note that \( \tilde{X}_n := X_n \cap B_{\varepsilon} \) are closed nowhere dense sets in \( B_{\varepsilon} \). But \( B_{\varepsilon} \) is a complete metric space, a contradiction. \( \Box \)

Countable intersections of open sets are in some sense the next general sets after open sets (cf. also Section 8.6) and are called \( G_\delta \) sets (here \( G \) and \( \delta \) stand for the German words Gebiet and Durchschnitt, respectively). The complement of a \( G_\delta \) set is a countable union of closed sets also known as an \( F_\sigma \) set (here \( F \) and \( \sigma \) stand for the French words fermé and somme, respectively). The complement of a dense \( G_\delta \) set will be a countable intersection of nowhere dense sets and hence by definition meager. Consequently properties which hold on a dense \( G_\delta \) are considered **generic** in this context.

**Example.** The irrational numbers are a dense \( G_\delta \) set in \( \mathbb{R} \). To see this, let \( x_n \) be an enumeration of the rational numbers and consider the intersection of the open sets \( O_n := \mathbb{R} \setminus \{x_n\} \). The rational numbers are hence an \( F_\sigma \) set.

Now we are ready for the first important consequence:

**Theorem 4.3** (Banach–Steinhaus). Let \( X \) be a Banach space and \( Y \) some normed vector space. Let \( \{A_\alpha\} \subseteq \mathcal{L}(X,Y) \) be a family of bounded operators. Then
4.1. The Baire theorem and its consequences

- either \( \{A_\alpha\} \) is uniformly bounded, \( \|A_\alpha\| \leq C \),
- or the set \( \{ x \in X \mid \sup_\alpha \|A_\alpha x\| = \infty \} \) is a dense \( G_\delta \).

**Proof.** Consider the sets
\[
O_n := \{ x \mid \|A_\alpha x\| > n \text{ for all } \alpha \} = \bigcup_\alpha \{ x \mid \|A_\alpha x\| > n \}, \quad n \in \mathbb{N}.
\]
By continuity of \( A_\alpha \) and the norm, each \( O_n \) is a union of open sets and hence open. Now either all of these sets are dense and hence their intersection
\[
\bigcap_{n \in \mathbb{N}} O_n = \{ x \mid \sup_\alpha \|A_\alpha x\| = \infty \}
\]
is a dense \( G_\delta \) by Corollary 4.2. Otherwise, \( X \setminus \overline{O_n} \) is nonempty and open for one \( n \) and we can find a ball of positive radius \( B_\varepsilon(x_0) \subset X \setminus O_n \). Now observe
\[
\|A_\alpha y\| = \|A_\alpha(y + x_0 - x_0)\| \leq \|A_\alpha(y + x_0)\| + \|A_\alpha x_0\| \leq 2n
\]
for \( \|y\| \leq \varepsilon \). Setting \( y = \varepsilon \frac{x}{\|x\|} \), we obtain
\[
\|A_\alpha x\| \leq \frac{2n}{\varepsilon} \|x\|
\]
for every \( x \). \( \square \)

Note that there is also a variant of the Banach–Steinhaus theorem for pointwise limits of bounded operators which will be discussed in Lemma 4.33.

Hence there are two ways to use this theorem by excluding one of the two possible options. Showing that the pointwise bound holds on a sufficiently large set (e.g. a ball), thereby ruling out the second option, implies a uniform bound and is known as the **uniform boundedness principle**.

**Corollary 4.4.** Let \( X \) be a Banach space and \( Y \) some normed vector space. Let \( \{A_\alpha\} \subseteq \mathcal{L}(X,Y) \) be a family of bounded operators. Suppose \( \|A_\alpha x\| \leq C(x) \) is bounded for every fixed \( x \in X \). Then \( \{A_\alpha\} \) is uniformly bounded, \( \|A_\alpha\| \leq C \).

Conversely, if there is no uniform bound, the pointwise bound must fail on a dense \( G_\delta \). This is illustrated in the following example.

**Example.** Consider the Fourier series (2.44) of a continuous periodic function \( f \in C_{\text{per}}[-\pi,\pi] = \{ f \in C[-\pi,\pi] \mid f(-\pi) = f(\pi) \} \). (Note that this is a closed subspace of \( C[-\pi,\pi] \) and hence a Banach space — it is the kernel of the linear functional \( \ell(f) = f(-\pi) - f(\pi) \).) We want to show that for every fixed \( x \in [-\pi,\pi] \) there is a dense \( G_\delta \) set of functions in \( C_{\text{per}}[-\pi,\pi] \) for which the Fourier series will diverge at \( x \) (it will even be unbounded).
Without loss of generality we fix \( x = 0 \) as our point. Then the \( n \)'th partial sum gives rise to the linear functional

\[
\ell_n(f) := \frac{1}{2\pi} \int_{-\pi}^{\pi} D_n(x)f(x)dx
\]

and it suffices to show that the family \( \{\ell_n\}_{n \in \mathbb{N}} \) is not uniformly bounded.

By the example on page 61 (adapted to our present periodic setting) we have

\[
\|\ell_n\| = \frac{1}{2\pi} \|D_n\|_1.
\]

Now we estimate

\[
\|D_n\|_1 = \frac{1}{\pi} \int_0^\pi |D_n(x)|dx \geq \frac{1}{\pi} \int_0^\pi \frac{\sin((n + 1/2)x)}{x/2} dx
\]

\[
= \frac{2}{\pi} \int_0^{(n+1/2)\pi} |\sin(y)| \frac{dy}{y} \geq \frac{2}{\pi} \sum_{k=1}^n \int_{(k-1)\pi}^{k\pi} |\sin(y)| \frac{dy}{k\pi} = \frac{4}{\pi^2} \sum_{k=1}^n \frac{1}{k}
\]

and note that the harmonic series diverges.

In fact, we can even do better. Let \( G(x) \subset C_{per}[-\pi, \pi] \) be the dense \( G_\delta \) of functions whose Fourier series diverges at \( x \). Then, given countably many points \( \{x_j\}_{j \in \mathbb{N}} \subset [-\pi, \pi] \), the set \( G = \bigcap_{j \in \mathbb{N}} G(x_j) \) is still a dense \( G_\delta \) by Corollary 4.2. Hence there is a dense \( G_\delta \) of functions whose Fourier series diverges on a given countable set of points. \( \diamond \)

**Example.** Recall that the Fourier coefficients of an absolutely continuous function satisfy the estimate

\[
|\hat{f}_k| \leq \begin{cases} \|f\|_\infty, & k = 0, \\ \frac{\|f\|_\infty}{|k|}, & k \neq 0. \end{cases}
\]

This raises the question if a similar estimate can be true for continuous functions. More precisely, can we find a sequence \( c_k > 0 \) such that

\[
|\hat{f}_k| \leq C_f c_k,
\]

where \( C_f \) is some constant depending on \( f \). If this were true, the linear functionals

\[
\ell_k(f) := \frac{\hat{f}_k}{c_k}, \quad k \in \mathbb{Z},
\]

satisfy the assumptions of the uniform boundedness principle implying \( \|\ell_k\| \leq C \). In other words, we must have an estimate of the type

\[
|\hat{f}_k| \leq C\|f\|_\infty c_k
\]

which implies \( 1 \leq C c_k \) upon choosing \( f(x) = e^{ikx} \). Hence our assumption cannot hold for any sequence \( c_k \) converging to zero and there is no universal
decay rate for the Fourier coefficients of continuous functions beyond the fact that they must converge to zero by the Riemann–Lebesgue lemma.

The next application is

**Theorem 4.5 (Open mapping).** Let $A \in \mathcal{L}(X, Y)$ be a continuous linear operator between Banach spaces. Then $A$ is open (i.e., maps open sets to open sets) if and only if it is onto.

**Proof.** Set $B^X_\varepsilon := B^X_\varepsilon(0)$ and similarly for $B^Y_\varepsilon(0)$. By translating balls (using linearity of $A$), it suffices to prove that for every $\varepsilon > 0$ there is a $\delta > 0$ such that $B^Y_\delta \subseteq A(B^X_\varepsilon)$.

So let $\varepsilon > 0$ be given. Since $A$ is surjective we have

$$Y = AX = A \bigcup_{n=1}^{\infty} nB^X_\varepsilon = \bigcup_{n=1}^{\infty} A(nB^X_\varepsilon) = \bigcup_{n=1}^{\infty} nA(B^X_\varepsilon)$$

and the Baire theorem implies that for some $n$, $nA(B^X_\varepsilon)$ contains a ball. Since multiplication by $n$ is a homeomorphism, the same must be true for $n = 1$, that is, $B^Y_\delta(y) \subset A(B^X_\varepsilon)$. Consequently

$$B^Y_\delta \subseteq -y + A(B^X_\varepsilon) \subset A(B^X_\varepsilon) + A(B^X_\varepsilon) \subseteq A(B^X_\varepsilon) + A(B^X_\varepsilon) \subseteq A(B^X_{2\varepsilon}).$$

So it remains to get rid of the closure. To this end choose $\varepsilon_n > 0$ such that $\sum_{n=1}^{\infty} \varepsilon_n < \varepsilon$ and corresponding $\delta_n \to 0$ such that $B^Y_{\delta_n} \subset A(B^X_{\varepsilon_n})$. Now for $y \in B^Y_{\delta_1} \subset A(B^X_{\varepsilon_1})$ we have $x_1 \in B^X_{\varepsilon_1}$ such that $Ax_1$ is arbitrarily close to $y$, say $y - Ax_1 \in B^Y_{\delta_2} \subset A(B^X_{\varepsilon_2})$. Hence we can find $x_2 \in A(B^X_{\varepsilon_2})$ such that $(y - Ax_1) - Ax_2 \in B^Y_{\delta_3} \subset A(B^X_{\varepsilon_3})$ and proceeding like this a a sequence $x_n \in A(B^X_{\varepsilon_{n+1}})$ such that

$$y - \sum_{k=1}^{n} Ax_k \in B^Y_{\delta_{n+1}}.$$ 

By construction the limit $x := \sum_{k=1}^{\infty} Ax_k$ exists and satisfies $x \in B^X_\varepsilon$ as well as $y = Ax \in AB^X_\varepsilon$. That is, $B^Y_{\delta_1} \subseteq AB^X_\varepsilon$ as desired.

Conversely, if $A$ is open, then the image of the unit ball contains again some ball $B^Y_\delta \subseteq A(B^X_1)$. Hence by scaling $B^Y_\varepsilon \subseteq A(B^X_\varepsilon)$ and letting $r \to \infty$ we see that $A$ is onto: $Y = A(X)$.

As an immediate consequence we get the inverse mapping theorem:

**Theorem 4.6 (Inverse mapping).** Let $A \in \mathcal{L}(X, Y)$ be a continuous linear bijection between Banach spaces. Then $A^{-1}$ is continuous.
Example. Consider the operator \( (Aa)_{j=1}^n = (\frac{1}{j}a_j)_{j=1}^n \) in \( \ell^2(\mathbb{N}) \). Then its inverse \( (A^{-1}a)_{j=1}^n = (j a_j)_{j=1}^n \) is unbounded (show this!). This is in agreement with our theorem since its range is dense (why?) but not all of \( \ell^2(\mathbb{N}) \):

\[
\infty = \sum_{j=1}^{\infty} 1 = \sum_{j=1}^{\infty} |jb_j|^2 = \sum_{j=1}^{\infty} |a_j|^2 < \infty.
\]

This should also be compared with Corollary 4.9 below.

Example. Consider the Fourier series (2.44) of an integrable function. Using the inverse function theorem we can show that not every sequence tending to 0 (which is a necessary condition according to the Riemann–Lebesgue lemma) arises as the Fourier coefficients of an integrable function:

By the elementary estimate

\[
\|\hat{f}\|_\infty \leq \frac{1}{2\pi} \|f\|_1
\]

we see that the mapping \( F(f) := \hat{f} \) continuously maps \( F : L^1(-\pi, \pi) \to c_0(\mathbb{Z}) \) (the Banach space of sequences converging to 0). In fact, this estimate holds for continuous functions and hence there is a unique continuous extension of \( F \) to all of \( L^1(-\pi, \pi) \) by Theorem 1.12. Moreover, it can be shown that if \( F \) is injective (for \( f \in L^2 \) this follows from Theorem 2.17, the general case \( f \in L^1 \) will be established in the example on page 318). Now if \( F \) were onto, the inverse mapping theorem would show that the inverse is also continuous, that is, we would have an estimate \( \|\hat{f}\|_\infty \geq C\|f\|_1 \) for some \( C > 0 \). However, considering the Dirichlet kernel \( D_n \) we have \( \|D_n\|_1 = 1 \) but \( \|D_n\|_1 \to \infty \) as shown in the example on page 127.

Another important consequence is the closed graph theorem. The \textbf{graph} of an operator \( A \) is just

\[
\Gamma(A) := \{(x, Ax) | x \in \mathcal{D}(A)\}.
\]

If \( A \) is linear, the graph is a subspace of the Banach space \( X \oplus Y \) (provided \( X \) and \( Y \) are Banach spaces), which is just the Cartesian product together with the norm

\[
\|(x, y)\|_{X \oplus Y} := \|x\|_X + \|y\|_Y.
\]

Note that \((x_n, y_n) \to (x, y)\) if and only if \( x_n \to x \) and \( y_n \to y \). We say that \( A \) has a closed graph if \( \Gamma(A) \) is a closed set in \( X \oplus Y \).

**Theorem 4.7 (Closed graph).** Let \( A : X \to Y \) be a linear map from a Banach space \( X \) to another Banach space \( Y \). Then \( A \) is continuous if and only if its graph is closed.
4.1. The Baire theorem and its consequences

Proof. If \( \Gamma(A) \) is closed, then it is again a Banach space. Now the projection \( \pi_1(x, Ax) = x \) onto the first component is a continuous bijection onto \( X \). So by the inverse mapping theorem its inverse \( \pi_1^{-1} \) is again continuous. Moreover, the projection \( \pi_2(x, Ax) = Ax \) onto the second component is also continuous and consequently so is \( A = \pi_2 \circ \pi_1^{-1} \). The converse is easy. \( \square \)

Remark: The crucial fact here is that \( A \) is defined on \( all \) of \( X \)!

Operators whose graphs are closed are called **closed operators**. Being closed is the next option you have once an operator turns out to be unbounded. If \( A \) is closed, then \( x_n \to x \) does not guarantee you that \( Ax_n \) converges (like continuity would), but it at least guarantees that if \( Ax_n \) converges, it converges to the right thing, namely \( Ax \):

- \( A \) bounded (with \( D(A) = X \)): \( x_n \to x \) implies \( Ax_n \to Ax \).
- \( A \) closed (with \( D(A) \subseteq X \)): \( x_n \to x \), \( x_n \in D(A) \), and \( Ax_n \to y \) implies \( x \in D(A) \) and \( y = Ax \).

If an operator is not closed, you can try to take the closure of its graph, to obtain a closed operator. If \( A \) is bounded this always works (which is just the content of Theorem 1.12). However, in general, the closure of the graph might not be the graph of an operator as we might pick up points \((x, y_1) - (x, y_2) = (0, y_2 - y_1) \in \Gamma(A) \) in this case and thus \( \Gamma(A) \) is the graph of some operator if and only if

\[
\Gamma(A) \cap \{(0, y) | y \in Y\} = \{(0, 0)\}.
\]  \( (4.3) \)

If this is the case, \( A \) is called **closable** and the operator \( A \) associated with \( \Gamma(A) \) is called the **closure** of \( A \).

In particular, \( A \) is closable if and only if \( x_n \to 0 \) and \( Ax_n \to y \) implies \( y = 0 \). In this case

\[
D(\overline{A}) = \{ x \in X | \exists x_n \in D(A), y \in Y : x_n \to x \text{ and } Ax_n \to y \},
\]

\[
\overline{A}x = y.
\]  \( (4.4) \)

For yet another way of defining the closure see Problem 4.9.

**Example.** Consider the operator \( A \) in \( \ell^p(\mathbb{N}) \) defined by \( Aa_j := ja_j \) on \( D(A) = \{ a \in \ell^p(\mathbb{N}) | a_j \neq 0 \text{ for finitely many } j \} \).

(i). \( A \) is closable. In fact, if \( a^n \to 0 \) and \( Aa^n \to b \) then we have \( a^n_j \to 0 \) and thus \( ja^n_j \to 0 = b_j \) for any \( j \in \mathbb{N} \).

(ii). The closure of \( A \) is given by

\[
D(\overline{A}) = \left\{ \begin{array}{ll}
\{ a \in \ell^p(\mathbb{N}) | (ja_j)_{j=1}^\infty \in \ell^p(\mathbb{N}) \}, & 1 \leq p < \infty, \\
\{ a \in c_0(\mathbb{N}) | (ja_j)_{j=1}^\infty \in c_0(\mathbb{N}) \}, & p = \infty,
\end{array} \right.
\]
and \( \overline{\mathcal{A}} a_j = j a_j \). In fact, if \( a^n \to a \) and \( A a^n \to b \) then we have \( a^n_j \to a_j \) and \( j a^n_j \to b_j \) for any \( j \in \mathbb{N} \) and thus \( b_j = j a_j \) for any \( j \in \mathbb{N} \). In particular, \( (j a_j)_{j=1}^\infty = (b_j)_{j=1}^\infty \in \ell^p(\mathbb{N}) \) \((c_0(\mathbb{N}) \text{ if } p = \infty)\). Conversely, suppose \((j a_j)_{j=1}^\infty \in \ell^p(\mathbb{N}) \) \((c_0(\mathbb{N}) \text{ if } p = \infty)\) and consider

\[
a^n_j := \begin{cases} a_j, & j \leq n, \\ 0, & j > n.\end{cases}
\]

Then \( a^n \to a \) and \( A a^n \to (j a_j)_{j=1}^\infty \).

(iii). Note that the inverse of \( \overline{\mathcal{A}} \) is the bounded operator \( \overline{\mathcal{A}}^{-1} a_j = j^{-1} a_j \) defined on all of \( \ell^p(\mathbb{N}) \). Thus \( \overline{\mathcal{A}}^{-1} \) is closed. However, since its range \( \text{Ran}(\overline{\mathcal{A}}^{-1}) = \mathcal{D}(\overline{\mathcal{A}}) \) is dense but not all of \( \ell^p(\mathbb{N}) \), \( \overline{\mathcal{A}}^{-1} \) does not map closed sets to closed sets in general. In particular, the concept of a closed operator should not be confused with the concept of a closed map in topology!

(iv). Extending the basis vectors \( \{\delta^n\}_{n \in \mathbb{N}} \) to a Hamel basis (Problem 1.6) and setting \( A a = 0 \) for every other element from this Hamel basis we obtain a (still unbounded) operator which is everywhere defined. However, this extension cannot be closed! 

\( \diamond \)

**Example.** Here is a simple example of a nonclosable operator: Let \( X := \ell^2(\mathbb{N}) \) and consider \( Ba := (\sum_{j=1}^\infty a_j) \delta^1 \) defined on \( \ell^1(\mathbb{N}) \subset \ell^2(\mathbb{N}) \). Let \( a^n_j := \frac{1}{n} \) for \( 1 \leq j \leq n \) and \( a^n_j := 0 \) for \( j > n \). Then \( \|a^n\|_2 = \frac{1}{\sqrt{n}} \) implying \( a^n \to 0 \) but \( Ba^n = \delta^1 \not\to 0 \). 

\( \diamond \)

**Example.** Another example are point evaluations in \( L^2(0, 1) \): Let \( x_0 \in [0, 1] \) and consider \( \ell_{x_0} : \mathcal{D}(\ell_{x_0}) \to \mathbb{C}, f \mapsto f(x_0) \) defined on \( \mathcal{D}(\ell_{x_0}) := C[0, 1] \subseteq L^2(0, 1) \). Then \( f_n(x) := \max(0, 1 - n|x - x_0|) \) satisfies \( f_n \to 0 \) but \( \ell_{x_0}(f_n) = 1 \). 

\( \diamond \)

**Lemma 4.8.** Suppose \( A \) is closable and \( \overline{\mathcal{A}} \) is injective. Then \( \overline{A}^{-1} = \overline{\mathcal{A}}^{-1} \).

**Proof.** If we set

\[
\Gamma^{-1} = \{(y, x) | (x, y) \in \Gamma\}
\]

then \( \Gamma(A^{-1}) = \Gamma^{-1}(A) \) and

\[
\overline{\Gamma(A^{-1})} = \overline{\Gamma^{-1}(A)} = \overline{\Gamma(A)^{-1}} = \Gamma(\overline{A}^{-1}) = \Gamma(\overline{A}^{-1}). \quad \square
\]

Note that \( A \) injective does not imply \( \overline{A} \) injective in general.

**Example.** Let \( P_M \) be the projection in \( \ell^2(\mathbb{N}) \) on \( M := \{b\}_j^{1} \), where \( b := (2^{-j/2})_{j=1}^{\infty} \). Explicitly we have \( P_M a = a - \langle b, a \rangle b \). Then \( P_M \) restricted to the space of sequences with finitely many nonzero terms is injective, but its closure is not. 

\( \diamond \)
Corollary 4.9. Suppose $A : \mathcal{D}(A) \subseteq X \to Y$ is closed and injective. Then $A^{-1}$ defined on $\mathcal{D}(A^{-1}) = \text{Ran}(A)$ is closed. Moreover, in this case $\text{Ran}(A)$ is closed if and only if $A^{-1}$ is bounded.

The question when $\text{Ran}(A)$ is closed plays an important role when investigating solvability of the equation $Ax = y$ and the last part gives us a convenient criterion. Moreover, note that $A^{-1}$ is bounded if and only if there is some $c > 0$ such that
\[
\|Ax\| \geq c\|x\|, \quad x \in \mathcal{D}(A).
\] (4.5)
Indeed, this follows upon setting $x = A^{-1}y$ in the above inequality which also shows that $c = \|A^{-1}\|^{-1}$ is the best possible constant. Factoring out the kernel we even get a criterion for the general case:

Corollary 4.10. Suppose $A : \mathcal{D}(A) \subseteq X \to Y$ is closed. Then $\text{Ran}(A)$ is closed if and only if
\[
\|Ax\| \geq c \text{dist}(x, \text{Ker}(A)), \quad x \in \mathcal{D}(A),
\] (4.6)
for some $c > 0$.

\textbf{Proof.} Consider the quotient space $\tilde{X} := X/\text{Ker}(A)$ and the induced operator $\tilde{A} : \mathcal{D}(\tilde{A}) \to Y$ where $\mathcal{D}(\tilde{A}) = \mathcal{D}(A)/\text{Ker}(A) \subseteq \tilde{X}$. By construction $\tilde{A}[x] = 0$ iff $x \in \text{Ker}(A)$ and hence $\tilde{A}$ is injective. To see that $\tilde{A}$ is closed we use $\tilde{\pi} : X \times Y \to \tilde{X} \times Y$, $(x, y) \mapsto ([x], y)$ which is bounded, surjective and hence open. Moreover, $\tilde{\pi}(\Gamma(A)) = \Gamma(\tilde{A})$. In fact, we even have $(x, y) \in \Gamma(A)$ iff $([x], y) \in \Gamma(\tilde{A})$ and thus $\tilde{\pi}(X \times Y \setminus \Gamma(A)) = \tilde{X} \times Y \setminus \Gamma(\tilde{A})$ implying that $Y \setminus \Gamma(\tilde{A})$ is open. Finally, observing $\text{Ran}(A) = \text{Ran}(\tilde{A})$ we have reduced it to the previous corollary. \hfill $\Box$

There is also another criterion which does not involve the distance to the kernel.

Corollary 4.11. Suppose $A : \mathcal{D}(A) \subseteq X \to Y$ is closed. Then $\text{Ran}(A)$ is closed if for some given $\varepsilon > 0$ and $0 \leq \delta < 1$ we can find for every $y \in \text{Ran}(A)$ a corresponding $x \in \mathcal{D}(X)$ such that
\[
\varepsilon \|x\| + \|y - Ax\| \leq \delta \|y\|.
\] (4.7)
Conversely, if $\text{Ran}(A)$ is closed this can be done whenever $\varepsilon < c\delta$ with $c$ from the previous corollary.

\textbf{Proof.} If $\text{Ran}(A)$ is closed and $\varepsilon < c\delta$ there is some $x \in \mathcal{D}(A)$ with $y = Ax$ and $\|Ax\| \geq \frac{\varepsilon}{\delta} \|x\|$ after maybe adding an element from the kernel to $x$. This $x$ satisfies $\varepsilon \|x\| + \|y - Ax\| = \varepsilon \|x\| \leq \delta \|y\|$ as required.
Conversely, fix \( y \in \text{Ran}(A) \) and recursively choose a sequence \( x_n \) such that
\[
\varepsilon \|x_n\| + \|(y - A\tilde{x}_{n-1}) - Ax_n\| \leq \delta \|y - A\tilde{x}_{n-1}\|, \quad \tilde{x}_n := \sum_{m \leq n} x_m.
\]
In particular, \( \|y - A\tilde{x}_n\| \leq \delta^n \|y\| \) as well as \( \varepsilon \|x_n\| \leq \delta^n \|y\| \), which shows \( \tilde{x}_n \to x \) and \( A\tilde{x}_n \to y \). Hence \( x \in \mathcal{D}(A) \) and \( y = Tx \in \text{Ran}(A) \). \( \square \)

The closed graph theorem tells us that closed linear operators can be defined on all of \( X \) if and only if they are bounded. So if we have an unbounded operator we cannot have both! That is, if we want our operator to be at least closed, we have to live with domains. This is the reason why in quantum mechanics most operators are defined on domains. In fact, there is another important property which does not allow unbounded operators to be defined on the entire space:

**Theorem 4.12** (Hellinger–Toeplitz). Let \( A : \mathfrak{H} \to \mathfrak{H} \) be a linear operator on some Hilbert space \( \mathfrak{H} \). If \( A \) is symmetric, that is \( \langle g, Af \rangle = \langle Ag, f \rangle \), \( f, g \in \mathfrak{H} \), then \( A \) is bounded.

**Proof.** It suffices to prove that \( A \) is closed. In fact, \( f_n \to f \) and \( Af_n \to g \) implies
\[
\langle h, g \rangle = \lim_{n \to \infty} \langle h, Af_n \rangle = \lim_{n \to \infty} \langle Ah, f_n \rangle = \langle Ah, f \rangle = \langle h, Af \rangle
\]
for every \( h \in \mathfrak{H} \). Hence \( Af = g \). \( \square \)

**Problem 4.1.** An infinite dimensional Banach space cannot have a countable Hamel basis (see Problem 1.6). (Hint: Apply Baire’s theorem to \( X_n := \text{span}\{u_j\}_{j=1}^n \).)

**Problem 4.2.** Let \( X := C[0,1] \). Show that the set of functions which are nowhere differentiable contains a dense \( G_\delta \). (Hint: Consider \( F_k := \{ f \in X \mid \exists x \in [0,1] : |f(x) - f(y)| \leq k|x - y|, \forall y \in [0,1] \} \). Show that this set is closed and nowhere dense. For the first property Bolzano–Weierstraß might be useful, for the latter property show that the set of piecewise linear functions whose slopes are bounded below by some fixed number in absolute value are dense.)

**Problem 4.3.** Let \( X \) be the space of sequences with finitely many nonzero terms together with the sup norm. Consider the family of operators \( \{A_n\}_{n \in \mathbb{N}} \) given by \( (A_n a)_j := j a_j, j \leq n \) and \( (A_n a)_j := 0, j > n \). Then this family is pointwise bounded but not uniformly bounded. Does this contradict the Banach–Steinhaus theorem?

**Problem 4.4.** Let \( X \) be a complete metric space without isolated points. Show that a dense \( G_\delta \) set cannot be countable. (Hint: A single point is nowhere dense.)
4.2. The Hahn–Banach theorem and its consequences

Problem 4.5. Consider a Schauder basis as in (1.34). Show that the coordinate functionals $\alpha_n$ are continuous. (Hint: Denote the set of all possible sequences of Schauder coefficients by $A$ and equip it with the norm $\|\alpha\| := \sup_n \| \sum_{k=1}^{n} \alpha_k u_k \|$; note that $A$ is precisely the set of sequences for which this norm is finite. By construction the operator $A : A \to X$, $\alpha \mapsto \sum_k \alpha_k u_k$ has norm one. Now show that $A$ is complete and apply the inverse mapping theorem.)

Problem 4.6. Show that a compact symmetric operator in an infinite-dimensional Hilbert space cannot be surjective.

Problem 4.7. Show that if $A$ is closed and $B$ bounded, then $A + B$ is closed. Show that this in general fails if $B$ is not bounded. (Here $A + B$ is defined on $D(A + B) = D(A) \cap D(B)$.)

Problem 4.8. Show that the differential operator $A = \frac{d}{dx}$ defined on $D(A) = C^1[0,1] \subset C[0,1]$ (sup norm) is a closed operator. (Compare the example in Section 1.5.)

Problem 4.9. Consider a linear operator $A : D(A) \subseteq X \to Y$, where $X$ and $Y$ are Banach spaces. Define the graph norm associated with $A$ by
\[
\|x\|_A := \|x\|_X + \|Ax\|_Y, \quad x \in D(A).
\] (4.8)
Show that $A : D(A) \to Y$ is bounded if we equip $D(A)$ with the graph norm. Show that the completion $X_A$ of $(D(A), \|\cdot\|_A)$ can be regarded as a subset of $X$ if and only if $A$ is closable. Show that in this case the completion can be identified with $D(\overline{A})$ and that the closure of $A$ in $X$ coincides with the extension from Theorem 1.12 of $A$ in $X_A$.

Problem 4.10. Let $X := \ell^2(\mathbb{N})$ and $(Aa)_j := j a_j$ with $D(A) := \{ a \in \ell^2(\mathbb{N}) | (ja_j)_{j \in \mathbb{N}} \in \ell^2(\mathbb{N}) \}$ and $Ba := (\sum_{j \in \mathbb{N}} a_j) \delta_1$. Then we have seen that $A$ is closed while $B$ is not closable. Show that $A + B$, $D(A + B) = D(A) \cap D(B) = D(A)$ is closed.

4.2. The Hahn–Banach theorem and its consequences

Let $X$ be a Banach space. Recall that we have called the set of all bounded linear functionals the dual space $X^*$ (which is again a Banach space by Theorem 1.13).

Example. Consider the Banach space $\ell^p(\mathbb{N})$, $1 \leq p < \infty$. Taking the Kronecker deltas $\delta^n$ as a Schauder basis the $n$'th term $x_n$ of a sequence $x \in \ell^p(\mathbb{N})$ can also be considered as the $n$'th coordinate of $x$ with respect to this basis. Moreover, the map $l_n(x) = x_n$ is a bounded linear functional, that is, $l_n \in \ell^p(\mathbb{N})^*$, since $|l_n(x)| = |x_n| \leq \|x\|_p$. It is a special case of the following more general example (in fact, we have $l_n = l_{\delta^n}$). Since the coordinates of
4. The main theorems about Banach spaces

a vector carry all the information this explains why understanding linear functionals is of key importance.

Example. Consider the Banach space $\ell^p(\mathbb{N})$, $1 \leq p < \infty$. We have already seen that by Hölder’s inequality (1.28) every $y \in \ell^q(\mathbb{N})$ gives rise to a bounded linear functional

$$l_y(x) := \sum_{n \in \mathbb{N}} y_n x_n$$

whose norm is $\|l_y\| = \|y\|_q$ (Problem 4.15). But can every element of $\ell^p(\mathbb{N})^*$ be written in this form?

Suppose $p := 1$ and choose $l \in \ell^1(\mathbb{N})^*$. Now define $y_n := l(\delta^n)$.

Then

$$|y_n| = |l(\delta^n)| \leq \|l\| \|\delta^n\|_1 = \|l\|$$

shows $\|y\|_\infty \leq \|l\|$, that is, $y \in \ell^\infty(\mathbb{N})$. By construction $l(x) = l_y(x)$ for every $x \in \text{span}\{\delta^n\}$. By continuity of $l$ it even holds for $x \in \text{span}\{\delta^n\} = \ell^1(\mathbb{N})$. Hence the map $y \mapsto l_y$ is an isomorphism, that is, $\ell^1(\mathbb{N})^* \cong \ell^\infty(\mathbb{N})$. A similar argument shows $\ell^p(\mathbb{N})^* \cong \ell^q(\mathbb{N})$, $1 \leq p < \infty$ (Problem 4.16). One usually identifies $\ell^p(\mathbb{N})^*$ with $\ell^q(\mathbb{N})$ using this canonical isomorphism and simply writes $\ell^p(\mathbb{N})^* = \ell^q(\mathbb{N})$. In the case $p = \infty$ this is not true, as we will see soon.

It turns out that many questions are easier to handle after applying a linear functional $l \in X^*$. For example, suppose $x(t)$ is a function $\mathbb{R} \to X$ (or $\mathbb{C} \to X$), then $l(x(t))$ is a function $\mathbb{R} \to \mathbb{C}$ (respectively $\mathbb{C} \to \mathbb{C}$) for any $l \in X^*$. So to investigate $l(x(t))$ we have all tools from real/complex analysis at our disposal. But how do we translate this information back to $x(t)$? Suppose we have $l(x(t)) = l(y(t))$ for all $l \in X^*$. Can we conclude $x(t) = y(t)$? The answer is yes and will follow from the Hahn–Banach theorem.

We first prove the real version from which the complex one then follows easily.

**Theorem 4.13** (Hahn–Banach, real version). Let $X$ be a real vector space and $\varphi : X \to \mathbb{R}$ a convex function (i.e., $\varphi(\lambda x + (1-\lambda)y) \leq \lambda \varphi(x) + (1-\lambda)\varphi(y)$ for $\lambda \in (0,1)$).

If $l$ is a linear functional defined on some subspace $Y \subset X$ which satisfies $l(y) \leq \varphi(y)$, $y \in Y$, then there is an extension $\tilde{l}$ to all of $X$ satisfying $\tilde{l}(x) \leq \varphi(x)$, $x \in X$. 


Proof. Let us first try to extend $\ell$ in just one direction: Take $x \notin Y$ and set $\bar{Y} = \text{span}\{x, Y\}$. If there is an extension $\tilde{\ell}$ to $\bar{Y}$ it must clearly satisfy

$$\tilde{\ell}(y + \alpha x) = \ell(y) + \alpha \tilde{\ell}(x).$$

So all we need to do is to choose $\tilde{\ell}(x)$ such that $\tilde{\ell}(y + \alpha x) \leq \varphi(y + \alpha x)$. But this is equivalent to

$$\sup_{\alpha > 0, y \in Y} \frac{\varphi(y - \alpha x) - \ell(y)}{-\alpha} \leq \tilde{\ell}(x) \leq \inf_{\alpha > 0, y \in Y} \frac{\varphi(y + \alpha x) - \ell(y)}{\alpha}$$

and is hence only possible if

$$\frac{\varphi(y_1 - \alpha_1 x) - \ell(y_1)}{-\alpha_1} \leq \frac{\varphi(y_2 + \alpha_2 x) - \ell(y_2)}{\alpha_2}$$

for every $\alpha_1, \alpha_2 > 0$ and $y_1, y_2 \in Y$. Rearranging this last equations we see that we need to show

$$\alpha_2 \ell(y_1) + \alpha_1 \ell(y_2) \leq \alpha_2 \varphi(y_1 - \alpha_1 x) + \alpha_1 \varphi(y_2 + \alpha_2 x).$$

Starting with the left-hand side we have

$$\alpha_2 \ell(y_1) + \alpha_1 \ell(y_2) = (\alpha_1 + \alpha_2) \ell(\lambda y_1 + (1 - \lambda) y_2)$$

$$\leq (\alpha_1 + \alpha_2) \varphi(\lambda y_1 + (1 - \lambda) y_2)$$

$$= (\alpha_1 + \alpha_2) \varphi(\lambda(y_1 - \alpha_1 x) + (1 - \lambda)(y_2 + \alpha_2 x))$$

$$\leq \alpha_2 \varphi(y_1 - \alpha_1 x) + \alpha_1 \varphi(y_2 + \alpha_2 x),$$

where $\lambda = \frac{\alpha_2}{\alpha_1 + \alpha_2}$. Hence one dimension works.

To finish the proof we appeal to Zorn’s lemma (see Appendix A): Let $E$ be the collection of all extensions $\tilde{\ell}$ satisfying $\tilde{\ell}(x) \leq \varphi(x)$. Then $E$ can be partially ordered by inclusion (with respect to the domain) and every linear chain has an upper bound (defined on the union of all domains). Hence there is a maximal element $\tilde{\ell}$ by Zorn’s lemma. This element is defined on $X$, since if it were not, we could extend it as before contradicting maximality. \(\square\)

Note that linearity gives us a corresponding lower bound $-\varphi(-x) \leq \tilde{\ell}(x)$, $x \in X$, for free. In particular, if $\varphi(x) = \varphi(-x)$ then $|\tilde{\ell}(x)| \leq \varphi(x).

Theorem 4.14 (Hahn–Banach, complex version). Let $X$ be a complex vector space and $\varphi : X \to \mathbb{R}$ a convex function satisfying $\varphi(\alpha x) \leq \varphi(x)$ if $|\alpha| = 1$.

If $\ell$ is a linear functional defined on some subspace $Y \subset X$ which satisfies $|\ell(y)| \leq \varphi(y)$, $y \in Y$, then there is an extension $\tilde{\ell}$ to all of $X$ satisfying $|\tilde{\ell}(x)| \leq \varphi(x)$, $x \in X$.

Proof. Set $\ell_r = \text{Re}(\ell)$ and observe

$$\ell(x) = \ell_r(x) - i\ell_r(ix).$$
4. The main theorems about Banach spaces

By our previous theorem, there is a real linear extension \( \ell_r \) satisfying \( \ell_r(x) \leq \varphi(x) \). Now set \( \ell(x) = \ell_r(x) - i\ell_r(ix) \). Then \( \ell(x) \) is real linear and by \( \ell(ix) = \ell_r(ix) + i\ell_r(x) = i\ell(x) \) also complex linear. To show \( |\ell(x)| \leq \varphi(x) \) we abbreviate \( \alpha = \ell(x)^* \) and use

\[
|\ell(x)| = \alpha \ell(x) = \ell(\alpha x) = \ell_r(\alpha x) \leq \varphi(\alpha x) \leq \varphi(x),
\]

which finishes the proof. \( \square \)

Note that \( \varphi(\alpha x) \leq \varphi(x) \), \( |\alpha| = 1 \) is in fact equivalent to \( \varphi(\alpha x) = \varphi(x) \), \( |\alpha| = 1 \).

If \( \ell \) is a bounded linear functional defined on some subspace, the choice \( \varphi(x) = \|\ell\||x| \) implies:

**Corollary 4.15.** Let \( X \) be a normed space and let \( \ell \) be a bounded linear functional defined on some subspace \( Y \subseteq X \). Then there is an extension \( \ell \in X^* \) preserving the norm.

Moreover, we can now easily prove our anticipated result

**Corollary 4.16.** Let \( X \) be a normed space and \( x \in X \) fixed. Suppose \( \ell(x) = 0 \) for all \( \ell \) in some total subset \( Y \subseteq X^* \). Then \( x = 0 \).

**Proof.** Clearly, if \( \ell(x) = 0 \) holds for all \( \ell \) in some total subset, this holds for all \( \ell \in X^* \). If \( x \neq 0 \) we can construct a bounded linear functional on \( \operatorname{span}\{x\} \) by setting \( \ell(\alpha x) = \alpha \) and extending it to \( X^* \) using the previous corollary. But this contradicts our assumption. \( \square \)

**Example.** Let us return to our example \( \ell^\infty(\mathbb{N}) \). Let \( c(\mathbb{N}) \subset \ell^\infty(\mathbb{N}) \) be the subspace of convergent sequences. Set

\[
l(x) = \lim_{n \to \infty} x_n, \quad x \in c(\mathbb{N}), \tag{4.10}
\]

then \( l \) is bounded since

\[
|l(x)| = \lim_{n \to \infty} |x_n| \leq \|x\|_\infty. \tag{4.11}
\]

Hence we can extend it to \( \ell^\infty(\mathbb{N}) \) by Corollary 4.15. Then \( l(x) \) cannot be written as \( l(x) = l_y(x) \) for some \( y \in \ell^1(\mathbb{N}) \) (as in (4.9)) since \( y_n = l(\delta^n) = 0 \) shows \( y = 0 \) and hence \( \ell_y = 0 \). The problem is that \( \operatorname{span}\{\delta^n\} = c_0(\mathbb{N}) \neq \ell^\infty(\mathbb{N}) \), where \( c_0(\mathbb{N}) \) is the subspace of sequences converging to 0.

Moreover, there is also no other way to identify \( \ell^\infty(\mathbb{N})^* \) with \( \ell^1(\mathbb{N}) \), since \( \ell^1(\mathbb{N}) \) is separable whereas \( \ell^\infty(\mathbb{N}) \) is not. This will follow from Lemma 4.21 (iii) below.

Another useful consequence is
Corollary 4.17. Let $Y \subseteq X$ be a subspace of a normed vector space and let $x_0 \in X \setminus Y$. Then there exists an $\ell \in X^*$ such that (i) $\ell(y) = 0$, $y \in Y$, (ii) $\ell(x_0) = \text{dist}(x_0, Y)$, and (iii) $\|\ell\| = 1$.

Proof. Replacing $Y$ by $\overline{Y}$ we see that it is no restriction to assume that $Y$ is closed. (Note that $x_0 \in X \setminus Y$ if and only if $\text{dist}(x_0, Y) > 0$.) Let $\tilde{Y} = \text{span}\{x_0, Y\}$. Since every element of $\tilde{Y}$ can be uniquely written as $y + \alpha x_0$ we can define

$$\ell(y + \alpha x_0) = \alpha \text{dist}(x_0, Y).$$

By construction $\ell$ is linear on $\tilde{Y}$ and satisfies (i) and (ii). Moreover, by $\text{dist}(x_0, Y) \leq \|x_0 - y\|$ for every $y \in Y$ we have

$$|\ell(y + \alpha x_0)| = |\alpha| \text{dist}(x_0, Y) \leq \|y + \alpha x_0\|, \quad y \in Y.$$

Hence $\|\ell\| \leq 1$ and there is an extension to $X^*$ by Corollary 4.15. To see that the norm is in fact equal to one, take a sequence $y_n \in Y$ such that $\text{dist}(x_0, Y) \geq (1 - \frac{1}{n})\|x_0 + y_n\|$. Then

$$|\ell(y_n + x_0)| = \text{dist}(x_0, Y) \geq (1 - \frac{1}{n})\|y_n + x_0\|$$

establishing (iii). 

Two more straightforward consequences of the last corollary are also worthwhile noting:

Corollary 4.18. Let $Y \subseteq X$ be a subspace of a normed vector space. Then $x \in Y$ if and only if $\ell(x) = 0$ for every $\ell \in X^*$ which vanishes on $Y$.

Corollary 4.19. Let $Y$ be a closed subspace and $\{x_j\}_{j=1}^n$ be a linearly independent subset of $X$. If $Y \cap \text{span}\{x_j\}_{j=1}^n = \{0\}$, then there exists a biorthogonal system $\{\ell_j\}_{j=1}^n \subset X^*$ such that $\ell_j(x_k) = 0$ for $j \neq k$, $\ell_j(x_j) = 1$ and $\ell(y) = 0$ for $y \in Y$.

Proof. Fix $j_0$. Since $Y_{j_0} = Y + \text{span}\{x_j\}_{1 \leq j \leq n, j \neq j_0}$ is closed (Corollary 1.15), $x_{j_0} \notin Y_{j_0}$ implies $\text{dist}(x_{j_0}, Y_{j_0}) > 0$ and existence of $\ell_{j_0}$ follows from Corollary 4.17. 

If we take the bidual (or double dual) $X^{**}$ of a normed space $X$, then the Hahn–Banach theorem tells us, that $X$ can be identified with a subspace of $X^{**}$. In fact, consider the linear map $J : X \to X^{**}$ defined by $J(x)(\ell) = \ell(x)$ (i.e., $J(x)$ is evaluation at $x$). Then

Theorem 4.20. Let $X$ be a normed space. Then $J : X \to X^{**}$ is isometric (norm preserving).
Proof. Fix \( x_0 \in X \). By \( |J(x_0)(\ell)| = \|\ell(x_0)\| \leq \|\ell\| \|x_0\| \) we have at least \( \|J(x_0)\|_{**} \leq \|x_0\| \). Next, by Hahn–Banach there is a linear functional \( \ell_0 \) with norm \( \|\ell_0\|_* = 1 \) such that \( \ell_0(x_0) = \|x_0\| \). Hence \( |J(x_0)(\ell_0)| = \|\ell_0(x_0)\| = \|x_0\| \) shows \( \|J(x_0)\|_{**} = \|x_0\| \). \( \square \)

Example. This gives another quick way of showing that a normed space has a completion: Take \( X = J(X) \subseteq X^{**} \) and recall that a dual space is always complete (Theorem 1.13). \( \diamond \)

Thus \( J : X \to X^{**} \) is an isometric embedding. In many cases we even have \( J(X) = X^{**} \) and \( X \) is called reflexive in this case.

Example. The Banach spaces \( \ell^p(\mathbb{N}) \) with \( 1 < p < \infty \) are reflexive: Identify \( \ell^p(\mathbb{N})^* \) with \( \ell^q(\mathbb{N}) \) (cf. Problem 4.16) and choose \( z \in \ell^p(\mathbb{N})^{**} \). Then there is some \( x \in \ell^p(\mathbb{N}) \) such that
\[
(\ell^q(\mathbb{N}))^* \ni z(y) = \sum_{j \in \mathbb{N}} y_j x_j, \quad y \in \ell^q(\mathbb{N}) \cong \ell^p(\mathbb{N})^*.
\]
But this implies \( z(y) = y(x) \), that is, \( z = J(x) \), and thus \( J \) is surjective. (Warning: It does not suffice to just argue \( \ell^p(\mathbb{N})^{**} \cong \ell^q(\mathbb{N})^* \cong \ell^p(\mathbb{N}) \).)

However, \( \ell^1 \) is not reflexive since \( \ell^1(\mathbb{N})^* \cong \ell^\infty(\mathbb{N}) \) but \( \ell^\infty(\mathbb{N})^* \not\cong \ell^1(\mathbb{N}) \) as noted earlier. Things get even a bit more explicit if we look at \( c_0(\mathbb{N}) \), where we can identify (cf. Problem 4.17) \( c_0(\mathbb{N})^* \) with \( \ell^1(\mathbb{N}) \) and \( c_0(\mathbb{N})^{**} \) with \( \ell^\infty(\mathbb{N}) \). Under this identification \( J(c_0(\mathbb{N})) = c_0(\mathbb{N}) \subseteq \ell^\infty(\mathbb{N}) \). \( \diamond \)

Example. By the same argument, every Hilbert space is reflexive. In fact, by the Riesz lemma we can identify \( H^* \) with \( H \) via the (conjugate linear) map \( x \mapsto (\cdot, x) \). Taking \( z \in H^{**} \) we have, again by the Riesz lemma, that \( z(y) = \langle (\cdot, y), (\cdot, z) \rangle_H^* = \langle x, y \rangle^* = \langle y, x \rangle = J(x)(y) \). \( \diamond \)

Lemma 4.21. Let \( X \) be a Banach space.

(i) If \( X \) is reflexive, so is every closed subspace.

(ii) \( X \) is reflexive if and only if \( X^* \) is.

(iii) If \( X^* \) is separable, so is \( X \).

Proof. (i) Let \( Y \) be a closed subspace. Denote by \( j : Y \hookrightarrow X \) the natural inclusion and define \( j^{**} : Y^{**} \to X^{**} \) via \( j^{**}(y'')(\ell) = y''(\ell|_Y) \) for \( y'' \in Y^{**} \) and \( \ell \in X^* \). Note that \( j^{**} \) is isometric by Corollary 4.15. Then
\[
\begin{array}{ccc}
X & \xrightarrow{J_X} & X^{**} \\
\uparrow j & & \uparrow j^{**} \\
Y & \xrightarrow{J_Y} & Y^{**}
\end{array}
\]
commutes. In fact, we have \( j_{**}(J_Y(y))(\ell) = J_Y(y) (\ell|_Y) = \ell(y) = J_X(y)(\ell) \). Moreover, since \( J_X \) is surjective, for every \( y'' \in Y^{**} \) there is an \( x \in X \) such that \( j_{**}(y'')(x) = J_X(y)(\ell) \). Since \( j_{**}(y'')(\ell) = y''(\ell|_Y) \) vanishes on all \( \ell \in X^* \) which vanish on \( Y \), so does \( \ell(x) = J_X(x)(\ell) = j_{**}(y'')(\ell) \) and thus \( x \in Y \) by Corollary 4.18. That is, \( j_{**}(Y^{**}) = J_X(Y) \) and \( J_Y = j \circ J_X \circ j_{**}^{-1} \) is surjective.

(ii) Suppose \( X \) is reflexive. Then the two maps

\[
(J_X)_*: X^* \to X^{***} \quad (J_X)^*: X^{**} \to X^*
\]

are inverse of each other. Moreover, fix \( x'' \in X^{**} \) and let \( x = J_X^{-1}(x'') \). Then \( J_X \cdot (x')(x'') = x''(x') = J(x)(x') = x'(x) = x'(J_X^{-1}(x'')) \), that is \( J_X = (J_X)_* \), respectively \( (J_X)_*^{-1} = (J_X)^* \), which shows \( X^* \) reflexive if \( X \) reflexive.

To see the converse, observe that \( X^* \) reflexive implies \( X^{**} \) reflexive and hence \( J_X(X) \cong X \) is reflexive by (i).

(iii) Let \( \{x_n\}_{n=1}^\infty \) be a dense set in \( X^* \). Then we can choose \( x_n \in X \) such that \( |x_n| = 1 \) and \( \ell_n(x_n) \geq \|\ell_n\|/2 \). We will show that \( \{x_n\}_{n=1}^\infty \) is total in \( X \). If it were not, we could find some \( x \in X \setminus \text{span}\{x_n\}_{n=1}^\infty \) and hence there is a functional \( \ell \in X^* \) as in Corollary 4.17. Choose a subsequence \( \ell_{n_k} \to \ell \).

Then

\[
\|\ell - \ell_{n_k}\| \geq |(\ell - \ell_{n_k})(x_{n_k})| = |\ell_{n_k}(x_{n_k})| \geq \|\ell_{n_k}\|/2,
\]

which implies \( \ell_{n_k} \to \ell \) and contradicts \( \|\ell\| = 1 \).

If \( X \) is reflexive, then the converse of (iii) is also true (since \( X \cong X^{**} \) separable implies \( X^* \) separable), but in general this fails as the example \( \ell^1(\mathbb{N})^* \cong \ell^\infty(\mathbb{N}) \) shows. In fact, this can be used to show that a separable space is not reflexive, by showing that its dual is not separable.

**Example.** The space \( C(I) \) is not reflexive. To see this observe that the dual space contains point evaluations \( \ell_{x_0}(f) := f(x_0) \), \( x_0 \in I \). Moreover, for \( x_0 \neq x_1 \) we have \( \|\ell_{x_0} - \ell_{x_1}\| = 2 \) and hence \( C(I)^* \) is not separable. You should appreciate the fact that it was not necessary to know the full dual space which is quite intricate (see Theorem 12.5).

Note that the product of two reflexive spaces is also reflexive. In fact, this even holds for countable products — Problem 4.19.

**Problem 4.11.** Let \( X = \mathbb{R}^3 \) equipped with the norm \( |(x,y,z)|_1 = |x| + |y| + |z| \) and \( Y = \{(x,y,z) | x+y = 0, z = 0 \} \). Find at least two extensions of \( \ell(x,y,z) = x \) from \( Y \) to \( X \) which preserve the norm. What if we take the usual Euclidean norm \( |(x,y,z)|_2 = (|x|^2 + |y|^2 + |z|^2)^{1/2} ? \)

**Problem 4.12.** Let \( X \) be some normed space. Show that

\[
\|x\| = \sup_{\ell \in V, \|\ell\|=1} |\ell(x)|,
\]

(4.12)
where $V \subset X^*$ is some dense subspace. Show that equality is attained if $V = X^*$.

**Problem 4.13.** Let $X$ be some normed space. By definition we have

$$
\|\ell\| = \sup_{x \in X, \|x\|=1} |\ell(x)|
$$

for every $\ell \in X^*$. One calls $\ell \in X^*$ norm-attaining, if the supremum is attained, that is, there is some $x \in X$ such that $\|\ell\| = |\ell(x)|$.

Show that in a reflexive Banach space every linear functional is norm-attaining. Give an example of a linear functional which is not norm-attaining. For uniqueness see Problem 5.27. (Hint: For the first part apply the previous problem to $X^*$. For the second part consider Problem 4.17 below.)

**Problem 4.14.** Let $X,Y$ be some normed spaces and $A : \mathcal{D}(A) \subseteq X \to Y$. Show

$$
\|A\| = \sup_{x \in X, \|x\|=1; \ell \in V, \|\ell\|=1} |\ell(Ax)|,
$$

where $V \subset Y^*$ is a dense subspace.

**Problem 4.15.** Show that $\|l_y\| = \|y\|_q$, where $l_y \in \ell^p(\mathbb{N})^*$ as defined in (4.9). (Hint: Choose $x \in \ell^p$ such that $x_n y_n = |y_n|^q$.)

**Problem 4.16.** Show that every $\ell \in \ell^p(\mathbb{N})^*$, $1 \leq p < \infty$, can be written as

$$
\ell(x) = \sum_{n \in \mathbb{N}} y_n x_n
$$

with some $y \in \ell^q(\mathbb{N})$. (Hint: To see $y \in \ell^q(\mathbb{N})$ consider $x^N$ defined such that $x_n^N = |y_n|^q/y_n$ for $n \leq N$ with $y_n \neq 0$ and $x_n^N = 0$ else. Now look at $|\ell(x^N)| \leq \|\ell\| \|x^N\|_p$.)

**Problem 4.17.** Let $c_0(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ be the subspace of sequences which converge to 0, and $c(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ the subspace of convergent sequences.

(i) Show that $c_0(\mathbb{N})$, $c(\mathbb{N})$ are both Banach spaces and that $c(\mathbb{N}) = \text{span}\{c_0(\mathbb{N}), e\}$, where $e = (1,1,1,\ldots) \in c(\mathbb{N})$.

(ii) Show that every $\ell \in c_0(\mathbb{N})^*$ can be written as

$$
\ell(a) = \sum_{n \in \mathbb{N}} b_n a_n
$$

with some $b \in \ell^1(\mathbb{N})$ which satisfies $\|b\|_1 = \|\ell\|$.

(iii) Show that every $\ell \in c(\mathbb{N})^*$ can be written as

$$
\ell(a) = \sum_{n \in \mathbb{N}} b_n a_n + b_0 \lim_{n \to \infty} a_n
$$

with some $b \in \ell^1(\mathbb{N})$ which satisfies $|b_0| + \|b\|_1 = \|\ell\|$. 
4.3. The adjoint operator

Problem 4.18. Let \( u_n \in X \) be a Schauder basis and suppose the complex numbers \( c_n \) satisfy \( |c_n| \leq c \|u_n\| \). Is there a bounded linear functional \( \ell \in X^* \) with \( \ell(u_n) = c_n \)? (Hint: Consider e.g. \( X = \ell^2(\mathbb{Z}) \).)

Problem 4.19. Let \( X = \bigtimes_{p,j \in \mathbb{N}} X_j \) be defined as in Problem 1.35 and let \( \frac{1}{p} + \frac{1}{q} = 1 \). Show that for \( 1 \leq p < \infty \) we have \( X^* \cong \bigtimes_{q,j \in \mathbb{N}} X_j^* \), where the identification is given by

\[
y(x) = \sum_{j \in \mathbb{N}} y_j(x_j), \quad x = (x_j)_{j \in \mathbb{N}} \in \bigtimes_{p,j \in \mathbb{N}} X_j, \quad y = (y_j)_{j \in \mathbb{N}} \in \bigtimes_{q,j \in \mathbb{N}} X_j^*.
\]

Moreover, if all \( X_j \) are reflexive, so is \( X \).

Problem 4.20 (Banach limit). Let \( c(\mathbb{N}) \subset \ell^\infty(\mathbb{N}) \) be the subspace of all bounded sequences for which the limit of the Cesàro means

\[
L(x) = \lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^{n} x_k
\]

exists. Note that \( c(\mathbb{N}) \subseteq c(\mathbb{N}) \) and \( L(x) = \lim_{n \to \infty} x_n \) for \( x \in c(\mathbb{N}) \).

Show that \( L \) can be extended to all of \( \ell^\infty(\mathbb{N}) \) such that

(i) \( L \) is linear,
(ii) \( |L(x)| \leq \|x\|_{\infty} \),
(iii) \( L(Sx) = L(x) \) where \( (Sx)_n = x_{n+1} \) is the shift operator,
(iv) \( L(x) \geq 0 \) when \( x_n \geq 0 \) for all \( n \),
(v) \( \lim \inf_n x_n \leq L(x) \leq \lim \sup_n x_n \) for all real-valued sequences.

(Hint: Of course existence follows from Hahn–Banach and (i), (ii) will come for free. Also (iii) will be inherited from the construction. For (iv) note that the extension can assumed to be real-valued and investigate \( L(e - x) \) for \( x \geq 0 \) with \( \|x\|_{\infty} = 1 \) where \( e = (1, 1, 1, \ldots) \). (v) then follows from (iv).)

Problem 4.21. Show that a finite dimensional subspace \( M \) of a Banach space \( X \) can be complemented. (Hint: Start with a basis \( \{x_j\} \) for \( M \) and choose a corresponding dual basis \( \{\ell_k\} \) with \( \ell_k(x_j) = \delta_{j,k} \) which can be extended to \( X^* \).)

4.3. The adjoint operator

Given two normed spaces \( X \) and \( Y \) and a bounded operator \( A \in \mathcal{L}(X, Y) \) we can define its adjoint \( A' : Y^* \to X^* \) via \( A'y = y' \circ A, \ y' \in Y^* \). It is
immediate that $A'$ is linear and boundedness follows from
\[
\|A'\| = \sup_{y' \in Y^* : \|y'\| = 1} \|A'y'\| = \sup_{y' \in Y^* : \|y'\| = 1} \left( \sup_{x \in X : \|x\| = 1} |(A'y')(x)| \right)
\]
\[
= \sup_{y' \in Y^* : \|y'\| = 1} \left( \sup_{x \in X : \|x\| = 1} |y'(Ax)| \right) = \sup_{x \in X : \|x\| = 1} \|Ax\| = \|A\|,
\]
where we have used Problem 4.12 to obtain the fourth equality. In summary,

**Theorem 4.22.** Let $A \in \mathcal{L}(X,Y)$, then $A' \in \mathcal{L}(Y^*,X^*)$ with $\|A\| = \|A'\|$.

Note that for $A,B \in \mathcal{L}(X,Y)$ and $\alpha,\beta \in \mathbb{C}$ we have
\[
(\alpha A + \beta B)' = \alpha A' + \beta B'
\]
and for $A \in \mathcal{L}(X,Y)$ and $B \in \mathcal{L}(Y,Z)$ we have
\[
(BA)' = A'B'
\]
which is immediate from the definition. Moreover, note that $(I_X)' = I_{X^*}$ which shows that if $A$ is invertible then so is $A'$ with
\[
(A^{-1})' = (A')^{-1}.
\]
That $A$ is invertible if $A'$ is will follow from Theorem 4.26 below.

**Example.** Given a Hilbert space $\mathcal{H}$ we have the conjugate linear isometry $C : \mathcal{H} \rightarrow \mathcal{H}^*$, $f \mapsto \langle f, \cdot \rangle$. Hence for given $A \in \mathcal{L}(\mathcal{H}_1, \mathcal{H}_2)$ we have $A'C_2f = \langle f, A \cdot \rangle = (A^*f, \cdot)$ which shows $A' = C_1A^*C_2^{-1}$.

**Example.** Let $X = Y = \ell^p(\mathbb{N})$, $1 \leq p < \infty$, such that $X^* = \ell^q(\mathbb{N})$, $\frac{1}{p} + \frac{1}{q} = 1$. Consider the right shift $R \in \mathcal{L}(\ell^p(\mathbb{N}))$ given by
\[
Rx = (0, x_1, x_2, \ldots).
\]
Then for $y' \in \ell^q(\mathbb{N})$
\[
y'(Rx) = \sum_{j=1}^{\infty} y'_j(Rx)_j = \sum_{j=2}^{\infty} y'_j x_{j-1} = \sum_{j=1}^{\infty} y'_{j+1} x_j
\]
which shows $(R'y')_k = y_{k+1}$ upon choosing $x = \delta_k$. Hence $R' = L$ is the left shift: $Ly = (y_2, y_3, \ldots)$.

Of course we can also consider the doubly adjoint operator $A''$. Then a simple computation
\[
A''(J_X(x))(y') = J_X(x)(A'y') = (A'y')(x) = y'(Ax) = J_Y(Ax)(y')
\]
shows that the following diagram commutes
\[
\begin{array}{ccc}
X & \xrightarrow{A} & Y \\
J_X \downarrow & & \downarrow J_Y \\
X^{**} & \xrightarrow{A''} & Y^{**}
\end{array}
\]

Consequently
\[
A'' \mid_{\text{Ran}(J_X)} = J_Y AJ_X^{-1}, \quad A = J_Y^{-1}A''J_X. \tag{4.18}
\]

Hence, regarding \(X\) as a subspace \(J_X(X) \subseteq X^{**}\) and \(Y\) as a subspace \(J_Y(Y) \subseteq Y^{**}\), then \(A''\) is an extension of \(A\) to \(X^{**}\) but with values in \(Y^{**}\).

In particular, note that \(B \in \mathcal{L}(Y^*, X^*)\) is the adjoint of some other operator \(B = A\) if and only if \(B'(J_X(X)) = A''(J_X(X)) \subseteq J_Y(Y)\) (for the converse note that \(A := J_Y^{-1}B'J_X\) will do the trick). This can be used to show that not every operator is an adjoint (Problem 4.22).

**Theorem 4.23** (Schauder). Suppose \(X, Y\) are Banach spaces and \(A \in \mathcal{L}(X,Y)\). Then \(A\) is compact if and only if \(A'\) is.

**Proof.** If \(A\) is compact, then \(A(B_1^X(0))\) is relatively compact and hence \(K = A(B_1^X(0))\) is a compact metric space. Let \(y''_n \in Y^*\) be a bounded sequence and consider the family of functions \(f_n = y''_n|_K \in C(K)\). Then this family is bounded and equicontinuous since
\[
|f_n(y_1) - f_n(y_2)| \leq \|y''_n\|\|y_1 - y_2\| \leq C\|y_1 - y_2\|.
\]

Hence the Arzelà–Ascoli theorem (Theorem 0.37) implies existence of a uniformly converging subsequence \(f_{n_j}\). For this subsequence we have
\[
\|A'y_{n_j} - A'y_{n_k}\| \leq \sup_{x \in B_1^X(0)} |y_{n_j}(Ax) - y_{n_k}(Ax)| = \|f_{n_j} - f_{n_k}\|_{\infty}
\]
since \(A(B_1^X(0)) \subseteq K\) is dense. Thus \(y'_{n_j}\) is the required subsequence and \(A'\) is compact.

To see the converse note that if \(A'\) is compact then so is \(A''\) by the first part and hence also \(A = J_Y^{-1}A''J_X\). \(\square\)

Finally we discuss the relation between solvability of \(Ax = y\) and the corresponding adjoint equation \(A'y = x\). To this end we need the analog of the orthogonal complement of a set. Given subsets \(M \subseteq X\) and \(N \subseteq X^*\) we define their **annihilator** as
\[
M^\perp := \{\ell \in X^*| \ell(x) = 0 \ \forall x \in M\} = \{\ell \in X^*| M \subseteq \ker(\ell)\} = \bigcap_{x \in M} \{x\}^\perp,
\]
\[
N^\perp := \{x \in X| \ell(x) = 0 \ \forall \ell \in N\} = \bigcap_{\ell \in N} \ker(\ell) = \bigcap_{\ell \in N} \{\ell\}^\perp. \tag{4.19}
\]
In particular, \( \{ \ell \}^\perp = \text{Ker}(\ell) \) while \( \{ x \}^\perp = \text{Ker}(J(x)) \) (with \( J : X \hookrightarrow X^{**} \) the canonical embedding).

**Example.** In a Hilbert space the annihilator is simply the orthogonal complement. \( \Diamond \)

The following properties are immediate from the definition (by linearity and continuity)
- \( M^\perp \) is a closed subspace of \( X^* \) and \( M^\perp = (\text{span}(M))^\perp \).
- \( N^\perp \) is a closed subspace of \( X \) and \( N^\perp = (\text{span}(N))^\perp \).

Note also that \( M^\perp = \{ 0 \} \) if and only if \( \text{span}(M) = X \) (cf. Corollary 4.17) and \( N^\perp = \{ 0 \} \) if \( \text{span}(N) = X^* \) (cf. Corollary 4.16). The converse of the last statement is wrong in general.

**Example.** Consider \( X := \ell^1(\mathbb{N}) \) and \( N := \{ \delta^n \}_{n \in \mathbb{N}} \subset \ell^\infty(\mathbb{N}) \simeq X^* \). Then \( \text{span}(N) = c_0(\mathbb{N}) \) but \( N^\perp = \{ 0 \} \). \( \Diamond \)

**Lemma 4.24.** We have \( (M^\perp)^\perp = \overline{\text{span}(M)} \) and \( (N^\perp)^\perp \supseteq \overline{\text{span}(N)} \).

**Proof.** By the preceding remarks we can assume \( M, N \) to be closed subspaces. The first part
\[
(M^\perp)^\perp = \{ x \in X | \ell(x) = 0 \ \forall \ell \in X^* \text{ with } M \subseteq \text{Ker}(\ell) \} = \overline{\text{span}(M)}
\]
is Corollary 4.18 and for the second part one just has to spell out the definition:
\[
(N^\perp)^\perp = \{ \ell \in X^* | \bigcap_{\hat{\ell} \in \hat{N}} \text{Ker}(\hat{\ell}) \subseteq \text{Ker}(\ell) \} \supseteq \overline{\text{span}(N)}. \quad \Box
\]

Note that we have equality in the preceding lemma if \( N \) is finite dimensional (Problem 4.27). Moreover, with a little more machinery one can show equality if \( X \) is reflexive (Problem 5.10). For non-reflexive spaces the inclusion can be strict as the previous example shows.

Warning: Some authors call a set \( N \subseteq X^* \) total if \( \{ N \}^\perp = \{ 0 \} \). By the preceding discussion this is equivalent to our definition if \( X \) is reflexive, but otherwise might differ.

Furthermore, we have the following analog of (2.28).

**Lemma 4.25.** Suppose \( X, Y \) are normed spaces and \( A \in \mathcal{L}(X, Y) \). Then \( \text{Ran}(A')^\perp = \text{Ker}(A) \) and \( \text{Ran}(A)^\perp = \text{Ker}(A') \).

**Proof.** For the first claim observe: \( x \in \text{Ker}(A) \iff Ax = 0 \iff \ell(Ax) = 0, \forall \ell \in X^* \iff (A'\ell)(x) = 0, \forall \ell \in X^* \iff x \in \text{Ran}(A')^\perp \).

For the second claim observe: \( \ell \in \text{Ker}(A') \iff A'\ell = 0 \iff (A'\ell)(x) = 0, \forall x \in X \iff \ell(Ax) = 0, \forall x \in X \iff \ell \in \text{Ran}(A)^\perp. \quad \Box \)
Taking annihilators in these formulas we obtain
\[ \ker(A') \perp = (\operatorname{ran}(A^\perp)) \perp = \overline{\operatorname{ran}(A)} \] (4.20)
and
\[ \ker(A) = (\operatorname{ran}(A^\prime)^\perp) \perp \supseteq \overline{\operatorname{ran}(A^\prime)} \] (4.21)
which raises the question of equality in the latter.

**Theorem 4.26** (Closed range). Suppose \( X, Y \) are Banach spaces and \( A \in \mathcal{L}(X,Y) \). Then the following items are equivalent:

(i) \( \operatorname{ran}(A) \) is closed.
(ii) \( \ker(A)^\perp = \operatorname{ran}(A^\prime) \).
(iii) \( \operatorname{ran}(A^\prime) \) is closed.
(iv) \( \ker(A^\prime)^\perp = \operatorname{ran}(A) \).

**Proof.** (i) \( \Leftrightarrow \) (vi): Immediate from (4.20).

(i) \( \Rightarrow \) (ii): Note that if \( \ell \in \operatorname{ran}(A^\prime) \) then \( \ell = A^\prime(\tilde{\ell}) = \tilde{\ell} \circ A \) vanishes on \( \ker(A) \) and hence \( \ell \in \ker(A)^\perp \). Conversely, if \( \ell \in \ker(A)^\perp \) we can set \( \tilde{\ell}(y) = \ell(A^{-1}y) \) for \( y \in \operatorname{ran}(A) \) and extend it to all of \( Y \) using Corollary 4.15. Here \( \tilde{A} : X/\ker(A) \to \operatorname{ran}(A) \) is the induced map (cf. Problem 1.38) which has a bounded inverse by Theorem 4.6. By construction \( \ell = A^\prime(\tilde{\ell}) \in \operatorname{ran}(A^\prime) \).

(ii) \( \Rightarrow \) (iii): Clear since annihilators are closed.

(iii) \( \Rightarrow \) (i): Let \( Z = \overline{\operatorname{ran}(A)} \) and let \( \tilde{A} : X \to Z \) be the range restriction of \( A \). Then \( \tilde{A}^\prime \) is injective (since \( \ker(\tilde{A}^\prime) = \overline{\operatorname{ran}(\tilde{A})^\perp} = \{0\} \)) and has the same range \( \overline{\operatorname{ran}(\tilde{A}^\prime)} = \operatorname{ran}(A^\prime) \) (since every linear functional in \( Z^* \) can be extended to one in \( Y^* \) by Corollary 4.15). Hence we can assume \( Z = Y \) and hence \( A^\prime \) injective without loss of generality.

Suppose \( \operatorname{ran}(A) \) were not closed. Then, given \( \varepsilon > 0 \) and \( 0 \leq \delta < 1 \), by Corollary 4.11 there is some \( y \in Y \) such that \( \varepsilon \|x\| + \|y - Ax\| > \delta \|y\| \) for all \( x \in X \). Hence there is a linear functional \( \ell \in Y^* \) such that \( \delta \leq \|\ell\| \leq 1 \) and \( \|A^\prime \ell\| \leq \varepsilon \). Indeed consider \( X \oplus Y \) and use Corollary 4.17 to choose \( \ell \in (X \oplus Y)^* \) such that \( \ell \) vanishes on the closed set \( V := \{(\varepsilon x, Ax) : x \in X\} \), \( \|\ell\| = 1 \), and \( \ell(0, y) = \operatorname{dist}((0, y), V) \) (note that \( (0, y) \notin V \) since \( y \neq 0 \)). Then \( \ell(x) = \ell(0,.) \) is the functional we are looking for since \( \operatorname{dist}((0, y), V) \geq \delta \|y\| \) and \( \|A^\prime \ell\| = \ell(0, Ax) = \ell(-\varepsilon x, 0) = -\varepsilon \ell(x, 0) \). Now this allows us to choose \( \ell_n \) with \( \|\ell_n\| \to 1 \) and \( \|A^\prime \ell_n\| \to 0 \) such that Corollary 4.10 implies that \( \operatorname{ran}(A^\prime) \) is not closed.

With the help of annihilators we can also describe the dual spaces of subspaces.
Theorem 4.27. Let $M$ be a closed subspace of a normed space $X$. Then there are canonical isometries

$$(X/M)^* \cong M^\perp, \quad M^* \cong X^*/M^\perp. \quad (4.22)$$

Proof. In the first case the isometry is given by $\ell \mapsto \ell \circ j$, where $j : X \to X/M$ is the quotient map. In the second case $x' + M^\perp \mapsto x'|_M$. The details are easy to check. \qed

Problem 4.22. Let $X = Y = c_0(\mathbb{N})$ and recall that $X^* = \ell^1(\mathbb{N})$ and $X^{**} = \ell^\infty(\mathbb{N})$. Consider the operator $A \in \mathcal{L}(\ell^1(\mathbb{N}))$ given by

$$Ax = (\sum_{n \in \mathbb{N}} x_n, 0, \ldots).$$

Show that

$$A'x' = (x'_1, x'_1, \ldots).$$

Conclude that $A$ is not the adjoint of an operator from $\mathcal{L}(c_0(\mathbb{N}))$.

Problem 4.23. Show

$$\text{Ker}(A') \cong \text{Coker}(A)^*, \quad \text{Coker}(A') \cong \text{Ker}(A)^*$$

for $A \in \mathcal{L}(X, Y)$ with $\text{Ran}(A)$ closed.

Problem 4.24. Let $X_j$ be Banach spaces. A sequence of operators $A_j \in \mathcal{L}(X_j, X_{j+1})$

$$X_1 \xrightarrow{A_1} X_2 \xrightarrow{A_2} X_3 \cdots X_n \xrightarrow{A_n} X_{n+1}$$

is said to be exact if $\text{Ran}(A_j) = \text{Ker}(A_{j+1})$ for $1 \leq j \leq n$. Show that a sequence is exact if and only if the corresponding dual sequence

$$X_1^* \xleftarrow{A_1'} X_2^* \xleftarrow{A_2'} X_3^* \cdots X_n^* \xleftarrow{A_n'} X_{n+1}^*$$

is exact.

Problem 4.25. Suppose $X$ is separable. Show that there exists a countable set $N \subset X^*$ with $N \perp = \{0\}$.

Problem 4.26. Let $X$ be a normed vector space and $Y \subset X$ some subspace. Show that if $Y \neq X$, then for every $\varepsilon \in (0, 1)$ there exists an $x_\varepsilon$ with $\|x_\varepsilon\| = 1$ and

$$\inf_{y \in Y} \|x_\varepsilon - y\| \geq 1 - \varepsilon. \quad (4.23)$$

Note: In a Hilbert space the claim holds with $\varepsilon = 0$ for any normalized $x$ in the orthogonal complement of $Y$ and hence $x_\varepsilon$ can be thought of a replacement of an orthogonal vector. (Hint: Choose a $y_\varepsilon \in Y$ which is close to $x$ and look at $x - y_\varepsilon$.)
4.4. Weak convergence

Problem 4.27. Suppose $X$ is a vector space and $\ell, \ell_1, \ldots, \ell_n$ are linear functionals such that $\bigcap_{j=1}^n \ker(\ell_j) \subseteq \ker(\ell)$. Then $\ell = \sum_{j=0}^n \alpha_j \ell_j$ for some constants $\alpha_j \in \mathbb{C}$. (Hint: Find a dual basis $x_k \in X$ such that $\ell_j(x_k) = \delta_{j,k}$ and look at $x - \sum_{j=1}^n \ell_j(x)x_j$.)

Problem 4.28. Let us write $\ell_n \rightharpoonup \ell$ provided the sequence converges pointwise, that is, $\ell_n(x) \to \ell(x)$ for all $x \in X$. Let $N \subseteq X^*$ and suppose $\ell_n \rightharpoonup \ell$ with $\ell_n \in N$. Show that $\ell \in (N_\perp)\perp$.

4.4. Weak convergence

In Section 4.2 we have seen that $\ell(x) = 0$ for all $\ell \in X^*$ implies $x = 0$. Now what about convergence? Does $\ell(x_n) \to \ell(x)$ for every $\ell \in X^*$ imply $x_n \to x$? In fact, in a finite dimensional space component-wise convergence is equivalent to convergence. Unfortunately in the infinite dimensional this is no longer true in general:

Example. Let $u_n$ be an infinite orthonormal set in some Hilbert space. Then $\langle g, u_n \rangle \to 0$ for every $g$ since these are just the expansion coefficients of $g$ which are in $\ell^2(\mathbb{N})$ by Bessel’s inequality. Since by the Riesz lemma (Theorem 2.10), every bounded linear functional is of this form, we have $\ell(u_n) \to 0$ for every bounded linear functional. (Clearly $u_n$ does not converge to 0, since $\|u_n\| = 1$.)

If $\ell(x_n) \to \ell(x)$ for every $\ell \in X^*$ we say that $x_n$ converges weakly to $x$ and write

$$\text{w-lim}_{n \to \infty} x_n = x \quad \text{or} \quad x_n \rightharpoonup x.$$  \hfill (4.24)

Clearly, $x_n \to x$ implies $x_n \rightharpoonup x$ and hence this notion of convergence is indeed weaker. Moreover, the weak limit is unique, since $\ell(x_n) \to \ell(x)$ and $\ell(x_n) \to \ell(\tilde{x})$ imply $\ell(x - \tilde{x}) = 0$. A sequence $x_n$ is called a weak Cauchy sequence if $\ell(x_n)$ is Cauchy (i.e. converges) for every $\ell \in X^*$.

Lemma 4.28. Let $X$ be a Banach space.

(i) $x_n \to x$, $y_n \to y$ and $\alpha_n \to \alpha$ implies $x_n + y_n \to x + y$ and $\alpha_n x_n \to \alpha x$.

(ii) $x_n \to x$ implies $\|x\| \leq \liminf \|x_n\|$.

(iii) Every weak Cauchy sequence $x_n$ is bounded: $\|x_n\| \leq C$.

(iv) If $X$ is reflexive, then every weak Cauchy sequence converges weakly.

(v) A sequence $x_n$ is Cauchy if and only if $\ell(x_n)$ is Cauchy, uniformly for $\ell \in X^*$ with $\|\ell\| = 1$. 
Proof. (i) Follows from $\ell(\alpha_n x_n + y_n) = \alpha_n \ell(x_n) + \ell(y_n) \to \alpha \ell(x) + \ell(y)$. (ii) Choose $\ell \in X^*$ such that $\ell(x) = \|x\|$ (for the limit $x$) and $\|\ell\| = 1$. Then $\|x\| = \ell(x) = \lim \inf \ell(x_n) \leq \lim \inf \|x_n\|$. (iii) For every $\ell$ we have that $|J(x_n)(\ell)| = \|x_n\| \leq C(\ell)$ is bounded. Hence by the uniform boundedness principle we have $\|x_n\| = \|J(x_n)\| \leq C$. (iv) If $x_n$ is a weak Cauchy sequence, then $\ell(x_n)$ converges and we can define $j(\ell) = \lim \ell(x_n)$. By construction $j$ is a linear functional on $X^*$. Moreover, by (ii) we have $|j(\ell)| \leq \sup \|\ell(x_n)\| \leq \|\ell\| \sup \|x_n\| \leq C\|\ell\|$ which shows $j \in X^{**}$. Since $X$ is reflexive, $j = J(x)$ for some $x \in X$ and by construction $\ell(x_n) \to J(x)(\ell) = \ell(x)$, that is, $x_n \rightharpoonup x$.

(v) This follows from $\|x_n - x_m\| = \sup_{\|\ell\|=1} |\ell(x_n - x_m)|$ (cf. Problem 4.12). □

Item (ii) says that the norm is sequentially weakly lower semicontinuous (cf. Problem 8.19) while the previous example shows that it is not sequentially weakly continuous (this will in fact be true for any convex function as we will see later). However, bounded linear operators turn out to be sequentially weakly continuous (Problem 4.30).

Example. Consider $L^2(0,1)$ and recall (see the example on page 108) that $u_n(x) = \sqrt{2} \sin(n\pi x), \quad n \in \mathbb{N},$

form an ONB and hence $u_n \rightharpoonup 0$. However, $v_n = u_n^2 \to 1$. In fact, one easily computes

$\langle u_m, v_n \rangle = \frac{\sqrt{2}((-1)^m - 1)}{m\pi} \frac{4k^2}{m^2 + 4k^2} \to \frac{\sqrt{2}((-1)^m - 1)}{m\pi} = \langle u_m, 1 \rangle$

and the claim follows from Problem 4.32 since $\|v_n\| = \sqrt{\frac{3}{2}}$. ○

Remark: One can equip $X$ with the weakest topology for which all $\ell \in X^*$ remain continuous. This topology is called the weak topology and it is given by taking all finite intersections of inverse images of open sets as a base. By construction, a sequence will converge in the weak topology if and only if it converges weakly. By Corollary 4.17 the weak topology is Hausdorff, but it will not be metrizable in general. In particular, sequences do not suffice to describe this topology. Nevertheless we will stick with sequences for now and come back to this more general point of view in Section 5.3.

In a Hilbert space there is also a simple criterion for a weakly convergent sequence to converge in norm (see Theorem 5.19 for a generalization).
4.4. Weak convergence

Lemma 4.29. Let $\mathcal{H}$ be a Hilbert space and let $f_n \rightharpoonup f$. Then $f_n \to f$ if and only if $\lim \sup \|f_n\| \leq \|f\|$. 

Proof. By (ii) of the previous lemma we have $\lim \|f_n\| = \|f\|$ and hence 
$$
\|f - f_n\|^2 = \|f\|^2 - 2\text{Re}(\langle f, f_n \rangle) + \|f_n\|^2 \to 0.
$$
The converse is straightforward.  

Now we come to the main reason why weakly convergent sequences are of interest: A typical approach for solving a given equation in a Banach space is as follows:

(i) Construct a (bounded) sequence $x_n$ of approximating solutions (e.g. by solving the equation restricted to a finite dimensional subspace and increasing this subspace).

(ii) Use a compactness argument to extract a convergent subsequence.

(iii) Show that the limit solves the equation.

Our aim here is to provide some results for the step (ii). In a finite dimensional vector space the most important compactness criterion is boundedness (Heine–Borel theorem, Theorem 0.22). In infinite dimensions this breaks down:

Theorem 4.30. The closed unit ball in $X$ is compact if and only if $X$ is finite dimensional. 

Proof. If $X$ is finite dimensional, then $X$ is isomorphic to $\mathbb{C}^n$ and the closed unit ball is compact by the Heine–Borel theorem (Theorem 0.22).

Conversely, suppose $S = \{x \in X|\|x\| = 1\}$ is compact. Then the open cover $\{X \setminus \text{Ker}(\ell)\}_{\ell \in X^*}$ has a finite subcover, $S \subset \bigcup_{j=1}^n X \setminus \text{Ker}(\ell_j) = X \setminus \bigcap_{j=1}^n \text{Ker}(\ell_j)$. Hence $\bigcap_{j=1}^n \text{Ker}(\ell_j) = \{0\}$ and the map $X \to \mathbb{C}^n$, $x \mapsto (\ell_1(x), \cdots, \ell_n(x))$ is injective, that is, $\dim(X) \leq n$.  

If we are willing to treat convergence for weak convergence, the situation looks much brighter!

Theorem 4.31. Let $X$ be a reflexive Banach space. Then every bounded sequence has a weakly convergent subsequence.

Proof. Let $x_n$ be some bounded sequence and consider $Y = \overline{\text{span}\{x_n\}}$. Then $Y$ is reflexive by Lemma 4.21 (i). Moreover, by construction $Y$ is separable and so is $Y^*$ by the remark after Lemma 4.21.

Let $\ell_k$ be a dense set in $Y^*$. Then by the usual diagonal sequence argument we can find a subsequence $x_{n_m}$ such that $\ell_k(x_{n_m})$ converges for
every $k$. Denote this subsequence again by $x_n$ for notational simplicity. Then,

$$
\|\ell(x_n) - \ell(x_m)\| \leq \|\ell(x_n) - \ell_k(x_n)\| + \|\ell_k(x_n) - \ell_k(x_m)\| \\
+ \|\ell_k(x_m) - \ell(x_m)\| \\
\leq 2C\|\ell - \ell_k\| + \|\ell_k(x_n) - \ell_k(x_m)\|
$$

shows that $\ell(x_n)$ converges for every $\ell \in \text{span}\{\ell_k\} = Y^\ast$. Thus there is a limit by Lemma 4.28 (iv). \qed

Note that this theorem breaks down if $X$ is not reflexive.

**Example.** Consider the sequence of vectors $\delta_n$ (with $\delta_n = 1$ and $\delta_{nm} = 0$, $n \neq m$) in $\ell^p(\mathbb{N})$, $1 \leq p < \infty$. Then $\delta_n \rightharpoonup 0$ for $1 < p < \infty$. In fact, since every $l \in \ell^p(\mathbb{N})^\ast$ is of the form $l = l_y$ for some $y \in \ell^q(\mathbb{N})$ we have $l_y(\delta_n) = y_n \to 0$.

If we consider the same sequence in $\ell^1(\mathbb{N})$ there is no weakly convergent subsequence. In fact, since $l_y(\delta_n) \to 0$ for every sequence $y \in \ell^\infty(\mathbb{N})$ with finitely many nonzero entries, the only possible weak limit is zero. On the other hand choosing the constant sequence $y = (1)_{j=1}^\infty$ we see $l_y(\delta_n) = 1 \not\to 0$, a contradiction. \diamond

**Example.** Let $X = L^1[-1,1]$. Every bounded integrable $\varphi$ gives rise to a linear functional

$$
\ell_\varphi(f) = \int f(x)\varphi(x) \, dx
$$

in $L^1[-1,1]^\ast$. Take some nonnegative $u_1$ with compact support, $\|u_1\|_1 = 1$, and set $u_k(x) = ku_1(kx)$ (implying $\|u_k\|_1 = 1$). Then we have

$$
\int u_k(x)\varphi(x) \, dx \to \varphi(0)
$$

(see Problem 10.23) for every continuous $\varphi$. Furthermore, if $u_{k_j} \rightharpoonup u$ we conclude

$$
\int u(x)\varphi(x) \, dx = \varphi(0).
$$

In particular, choosing $\varphi_k(x) = \max(0,1 - k|x|)$ we infer from the dominated convergence theorem

$$
1 = \int u(x)\varphi_k(x) \, dx \to \int u(x)\chi_{\{0\}}(x) \, dx = 0,
$$

a contradiction.

In fact, $u_k$ converges to the Dirac measure centered at 0, which is not in $L^1[-1,1]$. \diamond
4.4. Weak convergence

Note that the above theorem also shows that in an infinite dimensional reflexive Banach space weak convergence is always weaker than strong convergence since otherwise every bounded sequence had a weakly, and thus by assumption also norm, convergent subsequence contradicting Theorem 4.30. In a non-reflexive space this situation can however occur.

**Example.** In \( \ell^1(\mathbb{N}) \) every weakly convergent sequence is in fact (norm) convergent (such Banach spaces are said to have the **Schur property**). First of all recall that \( \ell^1(\mathbb{N}) \cong \ell^\infty(\mathbb{N}) \) and \( a^n \rightarrow 0 \) implies

\[
 l_b(a^n) = \sum_{k=1}^{\infty} b_k a^n_k \rightarrow 0, \quad \forall b \in \ell^\infty(\mathbb{N}).
\]

Now suppose we could find a sequence \( a^n \rightarrow 0 \) for which \( \liminf_n \|a^n\|_1 \geq \varepsilon > 0 \). After passing to a subsequence we can assume \( \|a^n\|_1 \geq \varepsilon/2 \) and after rescaling the norm even \( \|a^n\|_1 = 1 \). Now weak convergence \( a^n \rightarrow 0 \) implies \( a^n_j = l_{b^j}(a^n) \rightarrow 0 \) for every fixed \( j \in \mathbb{N} \). Hence the main contribution to the norm of \( a^n \) must move towards \( \infty \) and we can find a subsequence \( n_j \) and a corresponding increasing sequence of integers \( k_j \) such that

\[
 \sum_{k_j \leq k < k_{j+1}} |a^n_{k_j}| \geq \frac{2}{3}.
\]

Now set

\[
 b_k = \text{sign}(a^n_{k_j}), \quad k_j \leq k < k_{j+1}.
\]

Then

\[
 |l_b(a^n_j)| = \sum_{k_j \leq k < k_{j+1}} |a^n_{k_j}| + \sum_{1 \leq k < k_j; k_{j+1} \leq k} b_k a^n_{k_j} \geq \frac{2}{3} - \frac{1}{3} = \frac{1}{3},
\]

contradicting \( a^n_j \rightarrow 0 \). \( \Box \)

It is also useful to observe that compact operators will turn weakly convergent into (norm) convergent sequences.

**Theorem 4.32.** Let \( A \in \mathcal{C}(X,Y) \) be compact. Then \( x_n \rightarrow x \) implies \( Ax_n \rightarrow Ax \). If \( X \) is reflexive the converse is also true.

**Proof.** If \( x_n \rightarrow x \) we have \( \sup_n \|x_n\| \leq C \) by Lemma 4.28 (ii). Consequently \( Ax_n \) is bounded and we can pass to a subsequence such that \( Ax_{n_k} \rightarrow y \). Moreover, by Problem 4.30 we even have \( y = Ax \) and Lemma 0.5 shows \( Ax_n \rightarrow Ax \).

Conversely, if \( X \) is reflexive, then by Theorem 4.31 every bounded sequence \( x_n \) has a subsequence \( x_{n_k} \rightarrow x \) and by assumption \( Ax_{n_k} \rightarrow x \). Hence \( A \) is compact. \( \Box \)

Operators which map weakly convergent sequences to convergent sequences are also called **completely continuous**. However, be warned that some authors use completely continuous for compact operators. By the
above theorem every compact operator is completely continuous and the converse also holds in reflexive spaces. However, the last example shows that the identity map in $\ell^1(\mathbb{N})$ is completely continuous but it is clearly not compact by Theorem 4.30.

Let me remark that similar concepts can be introduced for operators. This is of particular importance for the case of unbounded operators, where convergence in the operator norm makes no sense at all.

A sequence of operators $A_n$ is said to converge strongly to $A$,
\[
s\lim_{n\to\infty} A_n = A \quad \iff \quad A_n x \to A x \quad \forall x \in \mathcal{D}(A) \subseteq \mathcal{D}(A_n). \tag{4.25}
\]
It is said to converge weakly to $A$,
\[
w\lim_{n\to\infty} A_n = A \quad \iff \quad A_n x \rightharpoonup A x \quad \forall x \in \mathcal{D}(A) \subseteq \mathcal{D}(A_n). \tag{4.26}
\]
Clearly norm convergence implies strong convergence and strong convergence implies weak convergence. If $Y$ is finite dimensional strong and weak convergence will be the same and this is in particular the case for $Y = \mathbb{C}$.

**Example.** Consider the operator $S_n \in \mathcal{L}(\ell^2(\mathbb{N}))$ which shifts a sequence $n$ places to the left, that is,
\[
S_n (x_1, x_2, \ldots) = (x_{n+1}, x_{n+2}, \ldots) \tag{4.27}
\]
and the operator $S_n^* \in \mathcal{L}(\ell^2(\mathbb{N}))$ which shifts a sequence $n$ places to the right and fills up the first $n$ places with zeros, that is,
\[
S_n^* (x_1, x_2, \ldots) = (0, \ldots, 0, x_1, x_2, \ldots). \tag{4.28}
\]
Then $S_n$ converges to zero strongly but not in norm (since $\|S_n\| = 1$) and $S_n^*$ converges weakly to zero (since $\langle x, S_n^* y \rangle = \langle S_n x, y \rangle$) but not strongly (since $\|S_n^* x\| = \|x\|$).

**Lemma 4.33.** Suppose $A_n, B_n \in \mathcal{L}(X,Y)$ are sequences of bounded operators.

(i) $s\lim_{n\to\infty} A_n = A$, $s\lim_{n\to\infty} B_n = B$, and $\alpha_n \to \alpha$ implies $s\lim_{n\to\infty}(A_n + B_n) = A + B$ and $s\lim_{n\to\infty} \alpha_n A_n = \alpha A$.

(ii) $s\lim_{n\to\infty} A_n = A$ implies $\|A\| \leq \liminf_{n\to\infty} \|A_n\|$.

(iii) If $A_n x$ converges for all $x \in X$ then $\|A_n\| \leq C$ and there is an operator $A \in \mathcal{L}(X,Y)$ such that $s\lim_{n\to\infty} A_n = A$.

(iv) If $A_n y$ converges for $y$ in a total set and $\|A_n\| \leq C$, then there is an operator $A \in \mathcal{L}(X,Y)$ such that $s\lim_{n\to\infty} A_n = A$.

The same result holds if strong convergence is replaced by weak convergence.
4.4. Weak convergence

Proof. (i) \( \lim_{n \to \infty} (\alpha_n A_n + B_n)x = \lim_{n \to \infty} (\alpha_n A_n x + B_n x) = \alpha Ax + Bx. \)
(ii) follows from
\[ \|Ax\| = \lim_{n \to \infty} \|A_n x\| \leq \liminf_{n \to \infty} \|A_n\| \]
for every \( x \in \mathcal{D}(A) \) with \( \|x\| = 1. \)
(iii) by linearity of the limit, \( Ax := \lim_{n \to \infty} A_n x \) is a linear operator. Moreover, since convergent sequences are bounded, \( \|A_n x\| \leq C(x) \), the uniform boundedness principle implies \( \|A_n\| \leq C \). Hence \( \|Ax\| = \lim_{n \to \infty} \|A_n x\| \leq C \|x\| \).
(iv) By taking linear combinations we can replace the total set by a dense one. Moreover, we can define a linear operator \( A \) on this dense set via \( Ay := \lim_{n \to \infty} A_n y \). By \( \|A_n\| \leq C \) we see \( \|A\| \leq C \) and there is a unique extension to all of \( X \). Now just use
\[ \|A_n x - Ax\| \leq \|A_n x - A_n y\| + \|A_n y - Ay\| + \|Ay - Ax\| \]
and choose \( y \) in the dense subspace such that \( \|x - y\| \leq \frac{\varepsilon}{4C} \) and \( n \) large such that \( \|A_n y - Ay\| \leq \frac{\varepsilon}{2} \).

The case of weak convergence is left as an exercise (Problem 4.12 might be useful).

Item (iii) of this lemma is sometimes also known as Banach–Steinhaus theorem. For an application of this lemma see Lemma 10.18.

Example. Let \( X \) be a Banach space of functions \( f : [-\pi, \pi] \to \mathbb{C} \) such the functions \( \{e_k(x) := e^{ikx}\}_{k \in \mathbb{Z}} \) are total. E.g. \( X = C_{\text{per}}[-\pi, \pi, 1] \) or \( X = L^p[-\pi, \pi] \) for \( 1 \leq p < \infty \). Then the Fourier series (2.44) converges on a total set and hence it will converge on all of \( X \) if and only if \( \|S_n\| \leq C \). For example, if \( X = C_{\text{per}}[-\pi, \pi] \) then
\[ \|S_n\| = \sup_{\|f\|_1 = 1} \|S_n(f)\| = \sup_{\|f\|_1 = 1} |S_n(f)(0)| = \frac{1}{2\pi} \|D_n\|_1 \]
which is unbounded as we have seen in the example on page 127. In fact, in this example we have even shown failure of pointwise convergence and hence this is nothing new. However, if we consider \( X = L^1[-\pi, \pi] \) we have (recall the Fejér kernel which satisfies \( \|F_n\|_1 = 1 \) and use (2.52) together with \( S_n(D_m) = D_{\text{min}(m,n)} \))
\[ \|S_n\| = \sup_{\|f\|_1 = 1} \|S_n(f)\| \geq \lim_{m \to \infty} \|S_n(F_m)\|_1 = \|D_n\|_1 \]
and we get that the Fourier series does not converge for every \( L^1 \) function.

Lemma 4.34. Suppose \( A_n \in \mathcal{L}(Y,Z) \), \( B_n \in \mathcal{L}(X,Y) \) are two sequences of bounded operators.
4. The main theorems about Banach spaces

(i) \( \text{s-lim}_{n \to \infty} A_n = A \) and \( \text{s-lim}_{n \to \infty} B_n = B \) implies \( \text{s-lim}_{n \to \infty} A_n B_n = AB \).

(ii) \( \text{w-lim}_{n \to \infty} A_n = A \) and \( \text{s-lim}_{n \to \infty} B_n = B \) implies \( \text{w-lim}_{n \to \infty} A_n B_n = AB \).

(iii) \( \lim_{n \to \infty} A_n = A \) and \( \text{w-lim}_{n \to \infty} B_n = B \) implies \( \text{w-lim}_{n \to \infty} A_n B_n = AB \).

**Proof.** For the first case just observe

\[
\| (A_n B_n - AB)x \| \leq \| (A_n - A)Bx \| + \| A_n \| \| (B_n - B)x \| \to 0.
\]

The remaining cases are similar and again left as an exercise. \( \square \)

**Example.** Consider again the last example. Then

\[
S_n^* S_n (x_1, x_2, \ldots) = (0, \ldots, 0, x_{n+1}, x_{n+2}, \ldots)
\]

converges to 0 weakly (in fact even strongly) but

\[
S_n^* S_n (x_1, x_2, \ldots) = (x_1, x_2, \ldots)
\]

does not! Hence the order in the second claim is important. \( \diamond \)

For a sequence of linear functionals \( \ell_n \), strong convergence is also called \textbf{weak-* convergence.} That is, the weak-* limit of \( \ell_n \) is \( \ell \) if \( \ell_n(x) \to \ell(x) \) for all \( x \in X \) and we will write

\[
\text{w*-lim}_{n \to \infty} x_n = x \quad \text{or} \quad x_n \xrightarrow{*} x \quad (4.29)
\]

in this case. Note that this is not the same as weak convergence on \( X^* \) unless \( X \) is reflexive: \( \ell \) is the weak limit of \( \ell_n \) if

\[
j(\ell_n) \to j(\ell) \quad \forall j \in X^{**}, \quad (4.30)
\]

whereas for the weak-* limit this is only required for \( j \in J(X) \subseteq X^{**} \) (recall \( J(x)(\ell) = \ell(x) \)).

**Example.** In a Hilbert space weak-* convergence of the linear functionals \( \langle x_n, \cdot \rangle \) is the same as weak convergence of the vectors \( x_n \). \( \diamond \)

**Example.** Consider \( X = c_0(\mathbb{N}) \), \( X^* \simeq l^1(\mathbb{N}) \), and \( X^{**} \simeq l^\infty(\mathbb{N}) \) with \( J \) corresponding to the inclusion \( c_0(\mathbb{N}) \hookrightarrow l^\infty(\mathbb{N}) \). Then weak convergence on \( X^* \) implies

\[
l_b(a^n - a) = \sum_{k=1}^{\infty} b_k (a_k^n - a_k) \to 0
\]

for all \( b \in l^\infty(\mathbb{N}) \) and weak-* convergence implies that this holds for all \( b \in c_0(\mathbb{N}) \). Whereas we already have seen that weak convergence is equivalent to norm convergence, it is not hard to see that weak-* convergence is equivalent to the fact that the sequence is bounded and each component converges (cf. Problem 4.33). \( \diamond \)
With this notation it is also possible to slightly generalize Theorem 4.31 (Problem 4.34):

**Lemma 4.35** (Helly). *Suppose $X$ is a separable Banach space. Then every bounded sequence $\ell_n \in X^*$ has a weak-$*$ convergent subsequence.*

**Example.** Let us return to the example after Theorem 4.31. Consider the Banach space of continuous functions $X = C[-1,1]$. Using $\ell_f(\varphi) = \int \varphi f \, dx$ we can regard $L^1[-1,1]$ as a subspace of $X^*$. Then the Dirac measure centered at 0 is also in $X^*$ and it is the weak-$*$ limit of the sequence $u_k$.

**Problem 4.29.** Suppose $\ell_n \to \ell$ in $X^*$ and $x_n \to x$ in $X$. Then $\ell_n(x_n) \to \ell(x)$. Similarly, suppose $s$-$\lim \ell_n \to \ell$ and $x_n \to x$. Then $\ell_n(x_n) \to \ell(x)$. Does this still hold if $s$-$\lim \ell_n \to \ell$ and $x_n \to x$?

**Problem 4.30.** Show that $x_n \to x$ implies $Ax_n \to Ax$ for $A \in \mathcal{L}(X,Y)$. Conversely, show that if $x_n \to 0$ implies $Ax_n \to 0$ then $A \in \mathcal{L}(X,Y)$.

**Problem 4.31.** Suppose $A_n, A \in \mathcal{L}(X,Y)$. Show that $s$-$\lim A_n = A$ and $\lim x_n = x$ implies $\lim A_n x_n = Ax$.

**Problem 4.32.** Show that if $\{\ell_j\} \subseteq X^*$ is some total set, then $x_n \to x$ if and only if $x_n$ is bounded and $\ell_j(x_n) \to \ell_j(x)$ for all $j$. Show that this is wrong without the boundedness assumption (Hint: Take e.g. $X = l^2(\mathbb{N})$).

**Problem 4.33.** Show that if $\{x_j\} \subseteq X$ is some total set, then $\ell_n \rightharpoonup \ell$ if and only if $\ell_n \in X^*$ is bounded and $\ell_n(x_j) \to \ell(x_j)$ for all $j$.

**Problem 4.34.** Prove Lemma 4.35.

### 4.5. Applications to minimizing nonlinear functionals

Finally, let me discuss a simple application of the above ideas to the **calculus of variations.** Many problems lead to finding the minimum of a given function. For example, many physical problems can be described by an energy functional and one seeks a solution which minimizes this energy. So we have a Banach space $X$ (typically some function space) and a functional $F : M \subseteq X \to \mathbb{R}$ (of course this functional will in general be nonlinear). If $M$ is compact and $F$ is continuous, then we can proceed as in the finite-dimensional case to show that there is a minimizer: Start with a sequence $x_n$ such that $F(x_n) \to \inf_M F$. By compactness we can assume that $x_n \to x_0$ after passing to a subsequence and by continuity $F(x_n) \to F(x_0) = \inf_M F$. Now in the infinite dimensional case we will use weak convergence to get compactness and hence we will also need weak (sequential) continuity of $F$. However, since there are more weakly than strongly convergent subsequences, weak (sequential) continuity is in fact a stronger property than just continuity!
Example. By Lemma 4.28 (ii) the norm is weakly sequentially lower semi-
continuous but it is in general not weakly sequentially continuous as any
infinite orthonormal set in a Hilbert space converges weakly to 0. However,
note that this problem does not occur for linear maps. This is an immediate
consequence of the very definition of weak convergence (Problem 4.30).

Hence weak continuity might be to much to hope for in concrete appli-
cations. In this respect note that, for our argument to work lower,
semicontinuity (cf. Problem 8.19) will already be sufficient:

**Theorem 4.36 (Variational principle).** Let $X$ be a reflexive Banach space
and let $F : M \subseteq X \to (-\infty, \infty]$. Suppose $M$ is nonempty, weakly sequen-
tially closed and that either $F$ is **weakly coercive**, that is $F(x) \to \infty$ whenever $\|x\| \to \infty$, or that $M$ is bounded. If in addition, $F$ is weakly sequentially
lower semicontinuous, then there exists some $x_0 \in M$ with $F(x_0) = \inf_M F$.

**Proof.** Without loss of generality we can assume $F(x) < \infty$ for some $x \in
M$. As above we start with a sequence $x_n \in M$ such that $F(x_n) \to \inf_M F$.
If $M = X$ then the fact that $F$ is coercive implies that $x_n$ is bounded.
Otherwise, it is bounded since we assumed $M$ to be bounded. Hence we can
pass to a subsequence such that $x_n \rightharpoonup x_0$ with $x_0 \in M$ since $M$ is assumed
sequentially closed. Now since $F$ is weakly sequentially lower semicontinuous
we finally get $\inf_M F = \lim_{n \to \infty} F(x_n) = \lim \inf_{n \to \infty} F(x_n) \geq F(x_0)$. □

Of course in a metric space the definition of closedness in terms of se-
quences agrees with the corresponding topological definition. In the present
situation sequentially weakly closed implies (sequentially) closed and the
converse holds at least for convex sets.

**Lemma 4.37.** Suppose $M \subseteq X$ is convex. Then $M$ is closed if and only if
it is sequentially weakly closed.

**Proof.** Suppose $x$ is in the weak sequential closure of $M$, that is, there is
a sequence $x_n \rightharpoonup x$. If $x \notin M$, then by Corollary 5.4 we can find a linear
functional $\ell$ which separates $\{x\}$ and $M$. But this contradicts $\ell(x) = d <
c < \ell(x_n) \to \ell(x)$. □

Similarly, the same is true with lower semicontinuity. In fact, a slightly
weaker assumption suffices. Let $X$ be a vector space and $M \subseteq X$ a convex
subset. A function $F : M \to \mathbb{R}$ is called **quasiconvex** if

$$F(\lambda x + (1 - \lambda)y) \leq \max\{F(x), F(y)\}, \quad \lambda \in (0, 1), \quad x, y \in M. \quad (4.31)$$

It is called **strictly quasiconvex** if the inequality is strict for $x \neq y$. By
$\lambda F(x) + (1 - \lambda)F(y) \leq \max\{F(x), F(y)\}$ every (strictly) convex function
is (strictly) quasiconvex. The converse is not true as the following example shows.

**Example.** Every (strictly) monotone function on $\mathbb{R}$ is (strictly) quasiconvex. Moreover, the same is true for symmetric functions which are (strictly) monotone on $[0, \infty)$. Hence the function $F(x) = \sqrt{|x|}$ is strictly quasiconvex. But it is clearly not convex on $M = \mathbb{R}$. $\diamond$

Now we are ready for the next

**Lemma 4.38.** Suppose $M \subseteq X$ is a closed convex set and suppose $F : M \to \mathbb{R}$ is quasiconvex. Then $F$ is weakly sequentially lower semicontinuous if and only if it is (sequentially) lower semicontinuous.

**Proof.** Suppose $F$ is lower semicontinuous. If it were not sequentially lower semicontinuous we could find a sequence $x_n \rightharpoonup x_0$ with $F(x_n) \to a < F(x_0)$. But then $x_n \in F^{-1}((\infty, a])$ for $n$ sufficiently large implying $x_0 \in F^{-1}((\infty, a])$ as this set is convex (Problem 4.36) and closed. But this gives the contradiction $a < F(x_0) \leq a$. $\square$

**Corollary 4.39.** Let $X$ be a reflexive Banach space and let $M$ be a nonempty closed convex subset. If $F : M \subseteq X \to \mathbb{R}$ is quasiconvex, lower semicontinuous, and, if $M$ is unbounded, weakly coercive, then there exists some $x_0 \in M$ with $F(x_0) = \inf_M F$. If $F$ is strictly quasiconvex then $x_0$ is unique.

**Proof.** It remains to show uniqueness. Let $x_0$ and $x_1$ be two different minima. Then $F(\lambda x_0 + (1 - \lambda) x_1) < \max\{F(x_0), F(x_1)\} = \inf_M F$, a contradiction. $\square$

**Example.** Let $X$ be a reflexive Banach space. Suppose $M \subseteq X$ is a nonempty closed convex set. Then for every $x \in X$ there is a point $x_0 \in M$ with minimal distance, $\|x - x_0\| = \text{dist}(x, M)$. Indeed, $F(z) = \text{dist}(x, z)$ is convex, continuous and, if $M$ is unbounded weakly coercive. Hence the claim follows from Corollary 4.39. Note that the assumption that $X$ is reflexive is crucial (Problem 4.35). Moreover, we also get that $x_0$ is unique if $X$ is strictly convex (see Problem 1.12). $\diamond$

**Example.** Let $\mathfrak{H}$ be a Hilbert space and $\ell \in \mathfrak{H}^*$ a linear functional. We will give a variational proof of the Riesz lemma (Theorem 2.10). To this end consider

$$F(x) = \frac{1}{2}\|x\|^2 - \text{Re}(\ell(x)), \quad x \in \mathfrak{H}.$$  

Then $F$ is convex, continuous, and weakly coercive. Hence there is some $x_0 \in \mathfrak{H}$ with $F(x_0) = \inf_{x \in \mathfrak{H}} F(x)$. Moreover, for fixed $x \in \mathfrak{H}$, 

$$\mathbb{R} \to \mathbb{R}, \quad \varepsilon \mapsto F(x_0 + \varepsilon x) = F(x_0) + \varepsilon \text{Re}(\langle x_0, x \rangle - \ell(x)) + \frac{\varepsilon^2}{2}\|x\|^2$$
is a smooth map which has a minimum at \( \varepsilon = 0 \). Hence its derivative at \( \varepsilon = 0 \) must vanish: \( \Re(\langle x_0, x \rangle - \ell(x)) = 0 \) for all \( x \in \mathcal{H} \). Replacing \( x \to -ix \) we also get \( \Im(\langle x_0, x \rangle - \ell(x)) = 0 \) and hence \( \ell(x) = \langle x_0, x \rangle \).

**Example.** Let \( \mathcal{H} \) be a Hilbert space and let us consider the problem of finding the lowest eigenvalue of a positive operator \( A \geq 0 \). Of course this is bound to fail since the eigenvalues could accumulate at 0 without 0 being an eigenvalue (e.g. the multiplication operator with the sequence \( \frac{1}{n} \) in \( \ell^2(N) \)). Nevertheless it is instructive to see how things can go wrong (and it underlines the importance of our various assumptions).

To this end consider its quadratic form \( q_A(f) = \langle f, Af \rangle \). Then, since \( \frac{1}{2} q_A \) is a seminorm (Problem 1.21) and taking squares is convex, \( q_A \) is convex. If we consider it on \( M = \bar{B}_1(0) \) we get existence of a minimum from Theorem 4.36. However this minimum is just \( q_A(0) = 0 \) which is not very interesting. In order to obtain a minimal eigenvalue we would need to take \( M = S_1 = \{ f \| f \| = 1 \} \), however, this set is not weakly closed (its weak closure is \( \bar{B}_1(0) \) as we will see in the next section). In fact, as pointed out before, the minimum is in general not attained on \( M \) in this case.

Note that our problem with the trivial minimum at 0 would also disappear if we would search for a maximum instead. However, our lemma above only guarantees us weak sequential lower semicontinuity but not weak sequential upper semicontinuity. In fact, note that not even the norm (the quadratic form of the identity) is weakly sequentially upper continuous (cf. Lemma 4.28 (ii) versus Lemma 4.29). If we make the additional assumption that \( A \) is compact, then \( q_A \) is weakly sequentially continuous as can be seen from Theorem 4.32. Hence for compact operators the maximum is attained at some vector \( f_0 \). Of course we will have \( \| f_0 \| = 1 \) but is it an eigenvalue?

To see this we resort to a small ruse: Consider the real function

\[
\phi(t) = \frac{q_A(f_0 + tf)}{\|f_0 + tf\|^2} = \frac{\alpha_0 + 2t\Re\langle f, Af_0 \rangle + t^2 q_A(f)}{1 + 2t\Re\langle f, f_0 \rangle + t^2 \|f\|^2}, \quad \alpha_0 = q_A(f_0),
\]

which has a maximum at \( t = 0 \) for any \( f \in \mathcal{H} \). Hence we must have \( \phi'(0) = 2\Re\langle f, (A - \alpha_0)f_0 \rangle = 0 \) for all \( f \in \mathcal{H} \). Replacing \( f \to if \) we get

\[
2\Im\langle f, (A - \alpha_0)f_0 \rangle = 0 \text{ and hence } \langle f, (A - \alpha_0)f_0 \rangle = 0 \text{ for all } f, \text{ that is } Af_0 = \alpha_0 f. \]

So we have recovered Theorem 3.6.

\( \diamond \)

**Problem 4.35.** Consider \( X = C[0,1] \) and \( M = \{ f \| f_0 f(x)dx = 1, f(0) = 0 \} \). Show that \( M \) is closed and convex. Show that \( d(0, M) = 1 \) but there is no minimizer. If we replace the boundary condition by \( f(0) = 1 \) there is a unique minimizer and for \( f(0) = 2 \) there are infinitely many minimizers.

**Problem 4.36.** Show that \( F : M \to \mathbb{R} \) is quasiconvex if and only if the sublevel sets \( F^{-1}((\infty, a]) \) are convex for every \( a \in \mathbb{R} \).
Further topics on
Banach spaces

5.1. The geometric Hahn–Banach theorem

Finally we turn to a geometric version of the Hahn–Banach theorem. Let $X$ be a vector space. For every subset $U \subset X$ we define its Minkowski functional (or gauge)

$$p_U(x) = \inf\{t > 0 | x \in tU\}. \quad (5.1)$$

Here $tU = \{tx | x \in U\}$. Note that $0 \in U$ implies $p_U(0) = 0$ and $p_U(x)$ will be finite for all $x$ when $U$ is absorbing, that is, for every $x \in X$ there is some $r$ such that $x \in \alpha U$ for every $|\alpha| \geq r$. Note that every absorbing set contains 0 and every neighborhood of 0 in a Banach space is absorbing.

**Example.** Let $X$ be a Banach space and $U = B_1(0)$, then $p_U(x) = \|x\|$. If $X = \mathbb{R}^2$ and $U = (-1,1) \times \mathbb{R}$ then $p_U(x) = |x_1|$. If $X = \mathbb{R}^2$ and $U = (-1,1) \times \{0\}$ then $p_U(x) = |x_1|$ if $x_2 = 0$ and $p_U(x) = \infty$ else.

We will only need minimal requirements and it will suffice if $X$ is a topological vector space, that is, a vector space which carries a topology such that both vector addition $X \times X \to X$ and scalar multiplication $\mathbb{C} \times X \to X$ are continuous mappings. Of course every normed vector space is a topological vector space with the usual topology generated by open balls. As in the case of normed linear spaces, $X^*$ will denote the vector space of all continuous linear functionals on $X$. 

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Lemma 5.1. Let $X$ be a vector space and $U$ a convex subset containing 0. Then

$$p_U(x + y) \leq p_U(x) + p_U(y), \quad p_U(\lambda x) = \lambda p_U(x), \quad \lambda \geq 0. \quad (5.2)$$

Moreover, $\{x|p_U(x) < 1\} \subseteq U \subseteq \{x|p_U(x) \leq 1\}$. If, in addition, $X$ is a topological vector space and $U$ is open, then $U = \{x|p_U(x) < 1\}$.

**Proof.** The homogeneity condition $p(\lambda x) = \lambda p(x)$ for $\lambda > 0$ is straightforward. To see the sublinearity, let $t, s > 0$ with $x \in tU$ and $y \in sU$, then

$$\frac{t}{t+s} x + \frac{s}{t+s} y = \frac{x+y}{t+s}$$

is in $U$ by convexity. Moreover, $p_U(x + y) \leq s + t$ and taking the infimum over all $t$ and $s$ we find $p_U(x + y) \leq p_U(x) + p_U(y)$.

Suppose $p_U(x) < 1$, then $t^{-1}x \in U$ for some $t < 1$ and thus $x \in U$ by convexity. Similarly, if $x \in U$ then $t^{-1}x \in U$ for $t \geq 1$ by convexity and thus $p_U(x) \leq 1$. Finally, let $U$ be open and $x \in U$, then $(1+\varepsilon)x \in U$ for some $\varepsilon > 0$ and thus $p(x) \leq (1+\varepsilon)^{-1}$. \hfill \Box

Note that (5.2) implies convexity

$$p_U(\lambda x + (1-\lambda)y) \leq \lambda p_U(x) + (1-\lambda)p_U(y), \quad \lambda \in [0,1]. \quad (5.3)$$

**Theorem 5.2** (geometric Hahn–Banach, real version). Let $U, V$ be disjoint nonempty convex subsets of a real topological vector space $X$ and let $U$ be open. Then there is a linear functional $\ell \in X^*$ and some $c \in \mathbb{R}$ such that

$$\ell(x) < c \leq \ell(y), \quad x \in U, y \in V. \quad (5.4)$$

If $V$ is also open, then the second inequality is also strict.

**Proof.** Choose $x_0 \in U$ and $y_0 \in V$, then

$$W = (U - x_0) + (V - y_0) = \{(x-x_0) - (y-y_0)|x \in U, y \in V\}$$

is open (since $U$ is), convex (since $U$ and $V$ are) and contains 0. Moreover, since $U$ and $V$ are disjoint we have $z_0 = y_0 - x_0 \notin W$. By the previous lemma,
the associated Minkowski functional $p_W$ is convex and by the Hahn–Banach theorem there is a linear functional satisfying

$$\ell(tz_0) = t, \quad |\ell(x)| \leq p_W(x).$$

Note that since $z_0 \notin W$ we have $p_W(z_0) \geq 1$. Moreover, $W = \{x | p_U(x) < 1\} \subseteq \{x | |\ell(x)| < 1\}$ which shows that $\ell$ is continuous at 0 by scaling and by translations $\ell$ is continuous everywhere.

Finally we again use $p_W(z) < 1$ for $z \in W$ implying

$$\ell(x) - \ell(y) + 1 = \ell(x - y + z_0) \leq p_W(x - y + z_0) < 1$$

and hence $\ell(x) < \ell(y)$ for $x \in U$ and $y \in V$. Therefore $\ell(U)$ and $\ell(V)$ are disjoint convex subsets of $\mathbb{R}$. Finally, let us suppose that there is some $x_1$ for which $\ell(x_1) = \sup \ell(U)$. Then, by continuity of the map $t \mapsto x_1 + tz_0$ there is some $\varepsilon > 0$ such that $x_1 + \varepsilon z_0 \in U$. But this gives a contradiction $\ell(x_1) + \varepsilon = \ell(x_1 + \varepsilon z_0) \leq \ell(x_1)$. Thus the claim holds with $c = \sup \ell(U)$. If $V$ is also open an analogous argument shows $\inf \ell(V) < \ell(y)$ for all $y \in V$. \hfill \Box

Of course there is also a complex version.

**Theorem 5.3** (geometric Hahn–Banach, complex version). Let $U$, $V$ be disjoint nonempty convex subsets of a topological vector space $X$ and let $U$ be open. Then there is a linear functional $\ell \in X^*$ and some $c \in \mathbb{R}$ such that

$$\text{Re}(\ell(x)) < c \leq \text{Re}(\ell(y)), \quad x \in U, \ y \in V. \quad (5.5)$$

If $V$ is also open, then the second inequality is also strict.

**Proof.** Consider $X$ as a real Banach space. Then there is a continuous real-linear functional $\ell_r : X \to \mathbb{R}$ by the real version of the geometric Hahn–Banach theorem. Then $\ell(x) = \ell_r(x) - i\ell_r(ix)$ is the functional we are looking for (check this). \hfill \Box

**Example.** The assumption that one set is open is crucial as the following example shows. Let $X = c_0(\mathbb{N})$, $U = \{a \in c_0(\mathbb{N}) | \exists N : a_N > 0 \text{ and } a_n = 0, n > N\}$ and $V = \{0\}$. Note that $U$ is convex but not open and that $U \cap V = \emptyset$. Suppose we could find a linear functional $\ell$ as in the geometric Hahn–Banach theorem (of course we can choose $\alpha = \ell(0) = 0$ in this case). Then by Problem 4.17 there is some $b_j \in \ell^\infty(\mathbb{N})$ such that $\ell(a) = \sum_{j=1}^\infty b_j a_j$. Moreover, we must have $b_j = \ell(\delta_j) < 0$. But then $a = (b_2, -b_1, 0, \ldots) \in U$ and $\ell(a) = 0 \not< 0$. \hfill \diamond

Note that two disjoint closed convex sets can be separated strictly if one of them is compact. However, this will require that every point has a neighborhood base of convex open sets. Such topological vector spaces are called **locally convex spaces** and they will be discussed further in
Section 5.4. For now we just remark that every normed vector space is locally convex since balls are convex.

Corollary 5.4. Let $U, V$ be disjoint nonempty closed convex subsets of a locally convex space $X$ and let $U$ be compact. Then there is a linear functional $\ell \in X^*$ and some $c, d \in \mathbb{R}$ such that
\[
\text{Re}(\ell(x)) \leq d < c \leq \text{Re}(\ell(y)), \quad x \in U, y \in V.
\] (5.6)

Proof. Since $V$ is closed, for every $x \in U$ there is a convex open neighborhood $N_x$ of 0 such that $x + N_x$ does not intersect $V$. By compactness of $U$ there are $x_1, \ldots, x_n$ such that the corresponding neighborhoods $x_j + \frac{1}{2}N_{x_j}$ cover $U$. Set $N = \bigcap_{j=1}^n N_{x_j}$ which is a convex open neighborhood of 0. Then
\[
\tilde{U} = U + \frac{1}{2}N \subseteq \bigcup_{j=1}^n (x_j + \frac{1}{2}N_{x_j}) + \frac{1}{2}N \subseteq \bigcup_{j=1}^n (x_j + \frac{1}{2}N_{x_j} + \frac{1}{2}N_{x_j}) = \bigcup_{j=1}^n (x_j + N_{x_j})
\]
is a convex open set which is disjoint from $V$. Hence by the previous theorem we can find some $\ell$ such that $\text{Re}(\ell(x)) < c \leq \text{Re}(\ell(y))$ for all $x \in \tilde{U}$ and $y \in V$. Moreover, since $\ell(U)$ is a compact interval $[c, d]$ the claim follows. □

Note that if $U$ and $V$ are absolutely convex (i.e., $\alpha U + \beta U \subseteq U$ for $|\alpha| + |\beta| \leq 1$), then we can write the previous condition equivalently as
\[
|\ell(x)| \leq d < c \leq |\ell(y)|, \quad x \in U, y \in V,
\] (5.7)
since $x \in U$ implies $\theta x \in U$ for $\theta = \text{sign}(\ell(x))$ and thus $|\ell(x)| = \theta \ell(x) = \ell(\theta x) = \text{Re}(\ell(\theta x))$.

From the last corollary we can also obtain versions of Corollaries 4.17 and 4.15 for locally convex vector spaces.

Corollary 5.5. Let $Y \subseteq X$ be a subspace of a locally convex space and let $x_0 \in X \setminus Y$. Then there exists an $\ell \in X^*$ such that (i) $\ell(y) = 0$, $y \in Y$ and (ii) $\ell(x_0) = 1$.

Proof. Consider $\ell$ from Corollary 5.4 applied to $U = \{x_0\}$ and $V = Y$. Now observe that $\ell(Y)$ must be a subspace of $\mathbb{C}$ and hence $\ell(Y) = \{0\}$ implying $\text{Re}(\ell(x_0)) < 0$. Finally $\ell(x_0)^{-1}\ell$ is the required functional. □

Corollary 5.6. Let $Y \subseteq X$ be a subspace of a locally convex space and let $\ell : Y \to \mathbb{C}$ be a continuous linear functional. Then there exists a continuous extension $\tilde{\ell} \in X^*$.

Proof. Without loss of generality we can assume that $\ell$ is nonzero such that we can find $x_0 \in y$ with $\ell(x_0) = 1$. Since $Y$ has the subset topology...
5.2. Convex sets and the Krein–Milman theorem

Let $X$ be a locally convex vector space. The intersection of arbitrary convex sets is again convex. Hence we can define the convex hull of a set $U$ as the smallest convex set containing $U$, that is, the intersection of all convex sets containing $U$. It is straightforward to show (Problem 5.5) that the convex hull is given by

$$\text{hull}(U) := \left\{ \sum_{j=1}^{n} \lambda_j x_j \big| n \in \mathbb{N}, x_j \in U, \sum_{j=1}^{n} \lambda_j = 1, \lambda_j \geq 0 \right\}. \quad (5.8)$$

A line segment is convex and can be generated as the convex hull of its endpoints. Similarly, a full triangle is convex and can be generated as the convex hull of its vertices. However, if we look at a ball, then we need its entire boundary to recover it as the convex hull. So how can we characterize those points which determine a convex sets via the convex hull?

$x_0 \not\in Y_0 := \overline{\text{Ker}(\ell)}$, where the closure is taken in $X$. Now Corollary 5.5 gives a functional $\bar{\ell}$ with $\bar{\ell}(x_0) = 1$ and $Y_0 \subseteq \text{Ker}(\bar{\ell})$. Moreover,

$$\bar{\ell}(x) - \ell(x) = \bar{\ell}(x) - \ell(x)x_0 = \bar{\ell}(x - \ell(x)x_0) = 0, \quad x \in Y,$$

since $x - \ell(x)x_0 \in \text{Ker}(\ell)$.

**Problem 5.1.** Let $X$ be a topological vector space. Show that $U + V$ is open if one of the sets is open.

**Problem 5.2.** Show that Corollary 5.4 fails even in $\mathbb{R}^2$ unless one set is compact.

**Problem 5.3.** Let $X$ be a topological vector space and $M \subseteq X$, $N \subseteq X^*$. Then the corresponding polar, prepolar sets are

$$M^o = \{ \ell \in X^*| |\ell(x)| \leq 1 \forall x \in M \}, \quad N_o = \{ x \in X| |\ell(x)| \leq 1 \forall \ell \in N \},$$

respectively. Show

(i) $M^o$ is closed and absolutely convex.

(ii) $M_1 \subseteq M_2$ implies $M_2^o \subseteq M_1^o$.

(iii) For $\alpha \neq 0$ we have $(\alpha M)^o = |\alpha|^{-1} M^o$.

(iv) If $M$ is a subspace we have $M^o = M^\perp$.

The same claims hold for prepolar sets.

**Problem 5.4** (Bipolar theorem). Let $X$ be a locally convex space and suppose $M \subseteq X$ is absolutely convex, that is, for $|\alpha| + |\beta| \leq 1$ we have $\alpha M + \beta M \subseteq M$. Show $(M^o)_o = \overline{M}$. (Hint: Use Corollary 5.4 to show that for every $y \not\in \overline{M}$ there is some $\ell \in X^*$ with $\text{Re}(\ell(x)) \leq 1 < \ell(y)$, $x \in \overline{M}$.)

5.2. Convex sets and the Krein–Milman theorem

Let $X$ be a locally convex vector space. The intersection of arbitrary convex sets is again convex. Hence we can define the convex hull of a set $U$ as the smallest convex set containing $U$, that is, the intersection of all convex sets containing $U$. It is straightforward to show (Problem 5.5) that the convex hull is given by

$$\text{hull}(U) := \left\{ \sum_{j=1}^{n} \lambda_j x_j \big| n \in \mathbb{N}, x_j \in U, \sum_{j=1}^{n} \lambda_j = 1, \lambda_j \geq 0 \right\}. \quad (5.8)$$

A line segment is convex and can be generated as the convex hull of its endpoints. Similarly, a full triangle is convex and can be generated as the convex hull of its vertices. However, if we look at a ball, then we need its entire boundary to recover it as the convex hull. So how can we characterize those points which determine a convex sets via the convex hull?
Let $K$ be a set and $M \subseteq K$ a nonempty subset. Then $M$ is called an extremal subset of $K$ if no point of $M$ can be written as a convex combination of two points unless both are in $M$: For given $x, y \in K$ and $\lambda \in (0, 1)$ we have that
\[
\lambda x + (1 - \lambda)y \in M \implies x, y \in M.
\]  
(5.9)

If $M = \{x\}$ is extremal, then $x$ is called an extremal point of $K$. Hence an extremal point cannot be written as a convex combination of two other points from $K$.

Note that we did not require $K$ to be convex. If $K$ is convex, then $M$ is extremal if and only if $K \setminus M$ is convex. Moreover, if $L \subseteq M$ is extremal and $M \subseteq K$ is extremal, then $L \subseteq K$ is extremal as well (Problem 5.6).

Example. Consider $\mathbb{R}^2$ with the norms $\| \cdot \|_p$. Then the extremal points of the closed unit ball (cf. Figure 1) are the boundary points for $1 < p < \infty$ and the vertices for $p = 1, \infty$. In any case the boundary is an extremal set. Slightly more general, in a strictly convex space, (ii) of Problem 1.12 says that the extremal points of the unit ball are precisely its boundary points. ◦

Example. Consider $\mathbb{R}^3$ and let $C = \{(x_1, x_2, 0) \in \mathbb{R}^3 | x_1^2 + x_2^2 = 1\}$. Take two more points $x_{\pm} = (0, 0, \pm 1)$ and consider the convex hull $K$ of $M = C \cup \{x_+, x_-\}$. Then $M$ is extremal in $K$ and, moreover, every point from $M$ is an extremal point. However, if we change the two extra points to be $x_{\pm} = (1, 0, \pm 1)$, then the point $(1, 0, 0)$ is no longer extremal. Hence the extremal points are now $M \setminus \{(1, 0, 0)\}$. Note in particular that the set of extremal points is not closed in this case. ◦

Extremal sets arise naturally when minimizing linear functionals.

**Lemma 5.7.** Suppose $K \subseteq X$ and $\ell \in X^*$. If
\[
K_\ell := \{x \in K | \ell(x) = \inf_{y \in K} \text{Re}(\ell(y))\}
\]
is nonempty (e.g. if $K$ is compact), then it is extremal in $K$. If $K$ is closed and convex, then $K_\ell$ is closed and convex.

**Proof.** Set $m = \inf_{y \in K} \text{Re}(\ell(y))$. Let $x, y \in K$, $\lambda \in (0, 1)$ and suppose $\lambda x + (1 - \lambda)y \in K_\ell$. Then
\[
m = \text{Re}(\ell(\lambda x + (1 - \lambda)y)) = \lambda \text{Re}(\ell(x)) + (1 - \lambda)\text{Re}(\ell(y)) \geq \lambda m + (1 - \lambda)m = m
\]
with strict inequality if $\text{Re}(\ell(x)) > m$ or $\text{Re}(\ell(y)) > m$. Hence we must have $x, y \in K_\ell$. Finally by linearity $K_\ell$ is convex and by continuity it is closed. ◯
5.2. Convex sets and the Krein–Milman theorem

If $K$ is a closed convex set, then nonempty subsets of the type $K_\ell$ are called **faces** of $K$ and $H_\ell := \{ x \in X | \ell(x) = \inf_{y \in K} \text{Re}(\ell(y)) \}$ is called a **support hyperplane** of $K$.

Conversely, if $K$ is convex with nonempty interior, then every point $x$ on the boundary has a supporting hyperplane (observe that the interior is convex and apply the geometric Hahn–Banach theorem with $U = K^\circ$ and $V = \{x\}$).

Next we want to look into existence of extremal points.

**Example.** Note that an interior point can never be extremal as it can be written as convex combination of some neighboring points. In particular, an open convex set will not have any extremal points (e.g. $X$, which is also closed, has no extremal points). Conversely, if $K$ is closed and convex, then the boundary is extremal since $K \setminus \partial K = K^\circ$ is convex (Problem 5.7).

**Example.** Suppose $X$ is a strictly convex Banach space. Then every nonempty compact subset $K$ has an extremal point. Indeed, let $x \in K$ be such that $\|x\| = \sup_{y \in K} \|y\|$, then $x$ is extremal: If $x = \lambda y + (1 - \lambda)z$ then $\|x\| \leq \lambda \|y\| + (1 - \lambda) \|z\| \leq \|x\|$ shows that we have equality in the triangle inequality and hence $x = y = z$ by Problem 1.12 (i).

**Example.** In a not strictly convex space the situation is quite different. For example, consider the closed unit ball in $\ell^\infty(\mathbb{N})$. Let $a \in \ell^\infty(\mathbb{N})$. If there is some index $j$ such that $\lambda := |a_j| < 1$ then $a = \frac{1}{2} b + \frac{1}{2} c$ where $b = a + \varepsilon \delta^j$ and $c = a - \varepsilon \delta^j$ with $\varepsilon \leq 1 - |a_j|$. Hence the only possible extremal points are those with $|a_j| = 1$ for all $j \in \mathbb{N}$. If we have such an $a$, then if $a = \lambda b + (1 - \lambda)c$ we must have $1 = |\lambda b_n + (1 - \lambda)c_n| \leq \lambda |b_n| + (1 - \lambda) |c_n| \leq 1$ and hence $a_n = b_n = c_n$ by strict convexity of the absolute value. Hence all such sequences are extremal.

However, if we consider $c_0(\mathbb{N})$ the same argument shows that the closed unit ball contains no extremal points. In particular, the following lemma implies that there is no locally convex topology for which the closed unit ball in $c_0(\mathbb{N})$ is compact. Together with the Banach–Alaoglu theorem (Theorem 5.10) this will show that $c_0(\mathbb{N})$ is not the dual of any Banach space.

**Lemma 5.8** (Krein–Milman). Let $X$ be a locally convex space. Suppose $K \subseteq X$ is compact and nonempty. Then it contains at least one extremal point.

**Proof.** We want to apply Zorn’s lemma. To this end consider the family

$$\mathcal{M} = \{ M \subseteq K | \text{compact and extremal in } K \}$$

with the partial order given by reversed inclusion. Since $K \in \mathcal{M}$ this family is nonempty. Moreover, given a linear chain $C \subseteq \mathcal{M}$ we consider $M := \bigcap C$. 
Then $M \subseteq K$ is nonempty by the finite intersection property and since it is closed also compact. Moreover, as the nonempty intersection of extremal sets it is also extremal. Hence $M \in \mathcal{M}$ and thus $\mathcal{M}$ has a maximal element. Denote this maximal element by $M$.

We will show that $M$ contains precisely one point (which is then extremal by construction). Indeed, suppose $x, y \in M$. If $x \neq y$ we can, by Corollary 5.4, choose a linear functional $\ell \in X^*$ with $\text{Re}(\ell(x)) \neq \text{Re}(\ell(y))$. Then by Lemma 5.7 $M_\ell \subset M$ is extremal in $M$ and hence also in $K$. But by $\text{Re}(\ell(x)) \neq \text{Re}(\ell(y))$ it cannot contain both $x$ and $y$ contradicting maximality of $M$. □

Finally, we want to recover a convex set as the convex hull of its extremal points. In our infinite dimensional setting an additional closure will be necessary in general.

Since the intersection of arbitrary closed convex sets is again closed and convex we can define the closed convex hull of a set $U$ as the smallest closed convex set containing $U$, that is, the intersection of all closed convex sets containing $U$. Since the closure of a convex set is again convex (Problem 5.7) the closed convex hull is simply the closure of the convex hull.

**Theorem 5.9** (Krein–Milman). Let $X$ be a locally convex space. Suppose $K \subseteq X$ is convex and compact. Then it is the closed convex hull of its extremal points.

**Proof.** Let $E$ be the extremal points and $M := \text{hull}(E) \subseteq K$ be its closed convex hull. Suppose $x \in K \setminus M$ and use Corollary 5.4 to choose a linear functional $\ell \in X^*$ with

$$\min_{y \in M} \text{Re}(\ell(y)) > \text{Re}(\ell(x)) \geq \min_{y \in K} \text{Re}(\ell(y)).$$

Now consider $K_\ell$ from Lemma 5.7 which is nonempty and hence contains an extremal point $y \in E$. But $y \notin M$, a contradiction. □

While in the finite dimensional case the closure is not necessary (Problem 5.8), it is important in general as the following example shows.

**Example.** Consider the closed unit ball in $\ell^1(\mathbb{N})$. Then the extremal points are $\{e^{i\theta}\delta^n | n \in \mathbb{N}, \theta \in \mathbb{R}\}$. Indeed, suppose $\|a\|_1 = 1$ with $\lambda := |a_j| \in (0, 1)$ for some $j \in \mathbb{N}$. Then $a = \lambda b + (1 - \lambda)c$ where $b := \lambda^{-1}a_j\delta^j$ and $c := (1 - \lambda)^{-1}(a - a_j\delta^j)$. Hence the only possible extremal points are of the form $e^{i\theta}\delta^n$. Moreover, if $e^{i\theta}\delta^n = \lambda b + (1 - \lambda)c$ we must have $1 = |\lambda b_n + (1 - \lambda)c_n| \leq \lambda|b_n| + (1 - \lambda)|c_n| \leq 1$ and hence $a_n = b_n = c_n$ by strict convexity of the absolute value. Thus the convex hull of the extremal points are the sequences from the unit ball which have finitely many terms nonzero. While the closed unit ball is not compact in the norm topology it will be in the
5.2. Convex sets and the Krein–Milman theorem

weak-* topology by the the Banach–Alaoglu theorem (Theorem 5.10). To this end note that \( \ell^1(\mathbb{N}) \cong c_0(\mathbb{N})^* \).

Also note that in the infinite dimensional case the extremal points can be dense.

**Example.** Let \( X = C([0,1], \mathbb{R}) \) and consider the convex set \( K = \{ f \in C^1([0,1], \mathbb{R}) | f(0) = 0, \| f' \|_\infty \leq 1 \} \). Note that the functions \( f( \pm x) = \pm x \) are extremal. For example, assume

\[
x = \lambda f(x) + (1 - \lambda)g(x)
\]

then

\[
1 = \lambda f'(x) + (1 - \lambda)g'(x)
\]

which implies \( f'(x) = g'(x) = 1 \) and hence \( f(x) = g(x) = x \).

To see that there are no other extremal functions, suppose \( |f'(x)| \leq 1 - \varepsilon \) on some interval \( I \). Choose a nontrivial continuous function \( g \) which is 0 outside \( I \) and has integral 0 over \( I \) and \( \|g\|_\infty \leq \varepsilon \). Let \( G = \int_0^x g(t)dt \). Then \( f = \frac{1}{2}(f + G) + \frac{1}{2}(f - G) \) and hence \( f \) is not extremal. Thus \( f_\pm \) are the only extremal points and their (closed) convex is given by \( f_\lambda(x) = \lambda x \) for \( \lambda \in [-1,1] \).

**Problem 5.5.** Show that the convex hull is given by (5.8).

**Problem 5.6.** Show that if \( L \subseteq M \) is extremal and \( M \subseteq K \) is extremal, then \( L \subseteq K \) is extremal as well.

**Problem 5.7.** Let \( X \) be a topological vector space. Show that the closure and the interior of a convex set is convex. (**Hint:** One way of showing the first claim is to consider the the continuous map \( f : X \times X \to X \) given by \( (x,y) \mapsto \lambda x + (1 - \lambda)y \) and use Problem 0.12.)

**Problem 5.8** (Carathéodory). Show that for a compact convex set \( K \subseteq \mathbb{R}^n \) every point can be written as convex combination of \( n + 1 \) extremal points.
(Hint: Induction on $n$. Without loss assume that 0 is an extremal point. If $K$ is contained in an $n-1$ dimensional subspace we are done. Otherwise $K$ has an open interior. Now for a given point the line through this point and 0 intersects the boundary where we have a corresponding face.)

5.3. Weak topologies

In Section 4.4 we have defined weak convergence for sequences and this raises the question about a natural topology associated with this convergence. To this end we define the **weak topology** on $X$ as the weakest topology for which all $\ell \in X^*$ remain continuous. Recall that a base for this topology is given by sets of the form

$$\ell + \bigcap_{j=1}^{n} |\ell_j|^{-1}([-\epsilon, \epsilon]) = \{ \tilde{x} \in X | |\ell_j(x) - \ell_j(\tilde{x})| < \epsilon, \ 1 \leq j \leq n \},$$

$$x \in X, \ \ell_j \in X^*, \ \epsilon_j > 0. \quad (5.10)$$

In particular, it is straightforward to check that a sequence converges with respect to this topology if and only if it converges weakly. Since the linear functionals separate points (cf. Corollary 4.16) the weak topology is Hausdorff.

Similarly, we define the **weak-$\ast$ topology** on $X^*$ as the weakest topology for which all $j \in J(X) \subseteq X^{**}$ remain continuous. In particular, the weak-$\ast$ topology is weaker than the weak topology on $X^*$ and both are equal if $X$ is reflexive. Since different linear functionals must differ at least at one point the weak-$\ast$ topology is also Hausdorff. A base for the weak-$\ast$ topology is given by sets of the form

$$\ell + \bigcap_{j=1}^{n} |J(x_j)|^{-1}([-\epsilon, \epsilon]) = \{ \tilde{\ell} \in X^* | |\ell(x_j) - \tilde{\ell}(x_j)| < \epsilon, \ 1 \leq j \leq n \},$$

$$\ell \in X^*, \ x_j \in X, \ \epsilon_j > 0. \quad (5.11)$$

Note that given a total set $\{x_n\}_{n \in \mathbb{N}} \subset X$ of (w.l.o.g.) normalized vectors

$$d(\ell, \tilde{\ell}) = \sum_{n=1}^{\infty} \frac{1}{2^n} |\ell(x_n) - \tilde{\ell}(x_n)| \quad (5.12)$$

defines a metric on the unit ball $\bar{B}_1^*(0) \subset X^*$ which can be shown to generate the weak-$\ast$ topology (cf. (iv) of Lemma 4.33). Hence Lemma 4.35 could also be stated as $\bar{B}_1^*(0) \subset X^*$ being weak-$\ast$ compact. This is in fact true without assuming $X$ to be separable and is known as Banach–Alaoglu theorem.

**Theorem 5.10 (Banach–Alaoglu).** Let $X$ be a Banach space. Then $\bar{B}_1^*(0) \subset X^*$ is compact in the weak-$\ast$ topology.
5.3. Weak topologies

Proof. Abbreviate $B = \overline{B}_1^X(0)$, $B^* = \overline{B}_{X^*}^X(0)$, and $B_x = \overline{B}_x^X(0)$. Consider the (injective) map $\Phi : X^* \to C^X$ given by $\Phi(\ell)(x) = \ell(x)$ and identify $X^*$ with $\Phi(X^*)$. Then the weak-$*$ topology on $X^*$ coincides with the relative topology on $\Phi(X^*) \subseteq C^X$ (recall that the product topology on $C^X$ is the weakest topology which makes all point evaluations continuous). Moreover, $\Phi(\ell) \leq \|\ell\|\|x\|$ implies $\Phi(B^*) \subset \bigtimes_{x \in X} B_x$ where the last product is compact by Tychonoff's theorem. Hence it suffices to show that $\Phi(B^*)$ is closed. To this end let $l \in \overline{\Phi}(B^*)$. We need to show that $l$ is linear and bounded. Fix $x_1, x_2 \in X$, $\alpha \in \mathbb{C}$, and consider the open neighborhood

$$U(l) = \left\{ h \in \bigtimes_{x \in B} B_x \mid \left| h(x_1 + x_2) - l(x_1 + \alpha x_2) \right| < \varepsilon, \left| h(x_1) - l(x_1) \right| < \varepsilon, \left| \alpha |h(x_2) - l(x_2)| \right| < \varepsilon \right\}$$

of $l$. Since $U(l) \cap \Phi(X^*)$ is nonempty we can choose an element $h$ from this intersection to show $|l(x_1 + \alpha x_2) - l(x_1) - \alpha l(x_2)| < 3\varepsilon$. Since $\varepsilon > 0$ is arbitrary we conclude $l(x_1 + \alpha x_2) = l(x_1) - \alpha l(x_2)$. Moreover, $|l(x_1)| \leq |h(x_1)| + \varepsilon \leq \|x_1\| + \varepsilon$ shows $\|l\| \leq 1$ and thus $l \in \Phi(B^*)$.

Since the weak topology is weaker than the norm topology every weakly closed set is also (norm) closed. Moreover, the weak closure of a set will in general be larger than the norm closure. However, for convex sets both will coincide. In fact, we have the following characterization in terms of closed (affine) half-spaces, that is, sets of the form $\{x \in X | \Re(\ell(x)) \leq \alpha\}$ for some $\ell \in X^*$ and some $\alpha \in \mathbb{R}$.

**Theorem 5.11** (Mazur). The weak as well as the norm closure of a convex set $K$ is the intersection of all half-spaces containing $K$. In particular, a convex set $K \subseteq X$ is weakly closed if and only if it is closed.

**Proof.** Since the intersection of closed-half spaces is (weakly) closed, it suffices to show that for every $x$ not in the (weak) closure there is a closed half-plane not containing $x$. Moreover, if $x$ is not in the weak closure it is also not in the norm closure (the norm closure is contained in the weak closure) and by Theorem 5.3 with $U = B_{\text{dist}(x,K)}(x)$ and $V = K$ there is a functional $\ell \in X^*$ such that $K \subseteq \Re(\ell)^{-1}([c, \infty))$ and $x \not\in \Re(\ell)^{-1}([c, \infty))$. □

**Example.** Suppose $X$ is infinite dimensional. The weak closure $\overline{S}^w$ of $S = \{ x \in X | \|x\| = 1 \}$ is the closed unit ball $\overline{B}_1(0)$. Indeed, since $\overline{B}_1(0)$ is convex the previous lemma shows $\overline{S}^w \subseteq \overline{B}_1(0)$. Conversely, if $x \in \overline{B}_1(0)$ is not in the weak closure, then there must be an open neighborhood $x + \bigcup_{j=1}^n |\ell_j|^{-1}(0, \varepsilon)$ not contained in the weak closure. Since $X$ is infinite dimensional we can find a nonzero element $x_0 \in \bigcap_{j=1}^n \ker(\ell_j)$ such that the affine line $x + tx_0$ is in this neighborhood and hence also avoids $\overline{S}^w$. But this
is impossible since by the intermediate value theorem there is some \( t_0 > 0 \)
such that \( \|x + t_0x_0\| = 1 \). Hence \( B_1(0) \subseteq S^w \).

Note that this example also shows that in an infinite dimensional space
the weak and norm topologies are always different! In a finite dimensional
space both topologies of course agree.

**Corollary 5.12** (Mazur lemma). Suppose \( x_k \to x \), then there are convex
combinations \( y_k = \sum_{j=1}^{n_k} \lambda_{k,j}x_j \) (with \( \sum_{j=1}^{n_k} \lambda_{k,j} = 1 \) and \( \lambda_{k,j} \geq 0 \)) such that
\( y_k \to x \).

**Proof.** Let \( K = \{\sum_{j=1}^{n_k} \lambda_{j}x_j \mid n \in \mathbb{N}, \sum_{j=1}^{n_k} \lambda_{j} = 1, \lambda_{j} \geq 0\} \) be the convex
hull of the points \( \{x_n\} \). Then by the previous result \( x \in \overline{K} \).

**Example.** Let \( \mathcal{H} \) be a Hilbert space and \( \{\varphi_j\} \) some infinite ONS. Then we
already know \( \varphi_j \rightharpoonup 0 \). Moreover, the convex combination
\( \psi_j = \frac{1}{j} \sum_{k=1}^{j} \varphi_k \to 0 \) since \( \|\psi_j\| = j^{-1/2} \).

Finally, we note two more important results. For the first note that
since \( X^* \) is the dual of \( X^* \) it has a corresponding weak-* topology and
by Banach–Alaoglu theorem \( \overline{B_1}^*(0) \) is weak-* compact and hence weak-*
closed.

**Theorem 5.13** (Goldstine). The image of the closed unit ball \( \overline{B_1}(0) \) under
the canonical embedding \( J \) into the closed unit ball \( \overline{B_1}^*(0) \) is weak-*
dense.

**Proof.** Let \( j \in \overline{B_1}^*(0) \) be given. Since sets of the form \( j + \bigcap_{k=1}^{n} [\ell_k^{-1}(0,\varepsilon)] \)
provide a neighborhood base (where we can assume the \( \ell_k \in X^* \) to be
linearly independent without loss of generality) it suffices to find some \( x \in \overline{B_1+\varepsilon}(0) \) with \( \ell_k(x) = j(\ell_k) \) for \( 1 \leq k \leq n \) since then \( (1 + \varepsilon)^{-1}J(x) \) will
be in the above neighborhood. Without the requirement \( \|x\| \leq 1 + \varepsilon \) this
follows from surjectivity of the map \( F : X \to \mathbb{C}^n, x \mapsto (\ell_1(x), \ldots, \ell_n(x)) \).
Moreover, given one such \( x \) the same is true for every element from \( x + Y \),
where \( Y = \bigcap_k \text{Ker}(\ell_k) \). So if \( (x + Y) \cap \overline{B_1+\varepsilon}(0) \) were empty, we would have
\( \text{dist}(x,Y) \geq 1 + \varepsilon \) and by Corollary 4.17 we could find some normalized
\( \ell \in X^* \) which vanishes on \( Y \) and satisfies \( \ell(x) \geq 1 + \varepsilon \). But by Problem 4.27
we have \( \ell \in \text{span}(\ell_1, \ldots, \ell_n) \) implying
\[ 1 + \varepsilon \leq \ell(x) = j(\ell) \leq \|j\|\|\ell\| \leq 1 \]
a contradiction. \( \square \)

**Example.** Consider \( X = c_0(\mathbb{N}), X^* \simeq \ell^1(\mathbb{N}), \) and \( X^{**} \simeq \ell^\infty(\mathbb{N}) \) with \( J \)
corresponding to the inclusion \( c_0(\mathbb{N}) \hookrightarrow \ell^\infty(\mathbb{N}) \). Then we can consider the
linear functionals \( \ell_j(x) = x_j \) which are total in \( X^* \) and a sequence in \( X^{**} \)
will be weak-* convergent if and only if it is bounded and converges when
composed with any of the $\ell_j$ (in other words, when the sequence converges componentwise — cf. Problem 4.33). So for example, cutting off a sequence in $B_1^*(0)$ after $n$ terms (setting the remaining terms equal to 0) we get a sequence from $B_1(0) \hookrightarrow B_1^*(0)$ which is weak-* convergent (but of course not norm convergent).

\[\diamond\]

**Theorem 5.14.** A Banach space $X$ is reflexive if and only if the closed unit ball $B_1(0)$ is weakly compact.

**Proof.** If $X$ is reflexive that this result follows from the Banach–Alaoglu theorem since in this case $J(B_1(0)) = B_1^*(0)$ and the weak-$*$ topology agrees with the weak topology on $X^*$. Conversely, suppose $B_1(0)$ is weakly compact. Since the weak topology on $J(X)$ is the relative topology of the weak-$*$ topology on $X^*$ we conclude that $J(B_1(0))$ is compact (and thus closed) in the weak-$*$ topology on $X^*$. But now Glaustine’s theorem implies $J(B_1(0)) = B_1^*(0)$ and hence $X$ is reflexive. \[\square\]

**Problem 5.9.** Show that a weakly sequentially compact set is bounded.

**Problem 5.10.** Show that the annihilator $M^\perp$ of a set $M \subseteq X$ is weak-* closed. Moreover show that $(N^\perp)^{\perp} = \overline{\text{span}(N)}$. In particular $(N^\perp)^{\perp} = \overline{\text{span}(N)}$ if $X$ is reflexive. (Hint: The first part and hence one inclusion of the second part are straightforward. For the other inclusion use Corollary 4.19.)

**Problem 5.11.** Suppose $K \subseteq X$ is convex and $x$ is a boundary point of $K$. Then there is a supporting hyperplane at $x$. That is, there is some $\ell \in X^*$ such that $\ell(x) = 0$ and $K$ is contained in the closed half-plane $\{y | \text{Re}(\ell(y - x)) \leq 0\}$.

### 5.4. Beyond Banach spaces: Locally convex spaces

We have already seen that it is often important to weaken the notion of convergence (i.e., to weaken the underlying topology) to get a larger class of converging sequences. It turns out that all cases considered so far fit within a general framework which we want to discuss in this section. We start with an alternate definition of a locally convex vector space which we already briefly encountered in Corollary 5.4 (equivalence of both definitions will be established below).

A vector space $X$ together with a topology is called a **locally convex vector space** if there exists a family of seminorms $\{q_{\alpha}\}_{\alpha \in A}$ which generates the topology in the sense that the topology is the weakest topology for which the family of functions $\{q_{\alpha}(\cdot - x)\}_{\alpha \in A, x \in X}$ is continuous. Hence the topology
is generated by sets of the form \( x + q_{\alpha}^{-1}(I) \), where \( I \subseteq [0, \infty) \) is open (in the relative topology). Moreover, sets of the form

\[
x + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j])
\]

are a neighborhood base at \( x \) and hence it is straightforward to check that a locally convex vector space is a topological vector space, that is, both vector addition and scalar multiplication are continuous. For example, if \( z = x + y \) then the preimage of the open neighborhood \( z + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j]) \) contains the open neighborhood \( (x + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j/2]), y + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j/2])) \) by virtue of the triangle inequality. Similarly, if \( z = \gamma x \) then the preimage of the open neighborhood \( z + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j]) \) contains the open neighborhood \( (B_{\varepsilon}(\gamma x) + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j/2])) \) with \( \varepsilon < \frac{\varepsilon_j}{2q_{\alpha_j}(x)} \).

Moreover, note that a sequence \( x_n \) will converge to \( x \) in this topology if and only if \( q_{\alpha}(x_n - x) \to 0 \) for all \( \alpha \).

**Example.** Of course every Banach space equipped with the norm topology is a locally convex vector space if we choose the single seminorm \( q(x) = \|x\| \).

**Example.** A Banach space \( X \) equipped with the weak topology is a locally convex vector space. In this case we have used the continuous linear functionals \( \ell \in X^* \) to generate the topology. However, note that the corresponding seminorms \( q_{\ell}(x) := |\ell(x)| \) generate the same topology since \( x + q_{\ell}^{-1}([0, \varepsilon]) = \ell^{-1}(B_{\varepsilon}(x)) \) in this case. The same is true for \( X^* \) equipped with the weak or the weak-\( * \) topology.

**Example.** The bounded linear operators \( \mathcal{L}(X, Y) \) together with the seminorms \( q_x(A) := \|Ax\| \) for all \( x \in X \) (strong convergence) or the seminorms \( q_{\ell,x}(A) := |\ell(Ax)| \) for all \( x \in X, \ell \in Y^* \) (weak convergence) are locally convex vector spaces.

**Example.** The continuous functions \( C(I) \) together with the pointwise topology generated by the seminorms \( q_x(f) := |f(x)| \) for all \( x \in I \) is a locally convex vector space.

In all these examples we have one additional property which is often required as part of the definition: The seminorms are called **separated** if for every \( x \in X \) there is a seminorm with \( q_{\alpha}(x) \neq 0 \). In this case the corresponding locally convex space is Hausdorff, since for \( x \neq y \) the neighborhoods \( U(x) = x + q_{\alpha}^{-1}([0, \varepsilon)) \) and \( U(y) = y + q_{\alpha}^{-1}([0, \varepsilon)) \) will be disjoint for \( \varepsilon = \frac{1}{2q_{\alpha}(x-y)} > 0 \) (the converse is also true; Problem 5.18).

It turns out crucial to understand when a seminorm is continuous.
Lemma 5.15. Let $X$ be a locally convex vector space with corresponding family of seminorms $\{q_\alpha\}_{\alpha \in A}$. Then a seminorm $q$ is continuous if and only if there are seminorms $q_{\alpha_j}$ and constants $c_j > 0$, $1 \leq j \leq n$, such that $q(x) \leq \sum_{j=1}^{n} c_j q_{\alpha_j}(x)$.

Proof. If $q$ is continuous, then $q^{-1}(B_1(0))$ contains an open neighborhood of 0 of the form $\bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j])$ and choosing $c_j = \max_{1 \leq j \leq n} \varepsilon_j^{-1}$ we obtain that $\sum_{j=1}^{n} c_j q_{\alpha_j}(x) < 1$ implies $q(x) < 1$ and the claim follows from Problem 5.13. Conversely note that if $q(x) = r$ then $q^{-1}(B_r(0))$ contains the set $U(x) = x + \bigcap_{j=1}^{n} q_{\alpha_j}^{-1}([0, \varepsilon_j])$ provided $\sum_{j=1}^{n} c_j \varepsilon_j \leq \varepsilon$ since $|q(y) - q(x)| \leq q(y - x) \leq \sum_{j=1}^{n} c_j q_{\alpha_j}(x - y) < \varepsilon$ for $y \in U(x)$.

Example. The weak topology on an infinite dimensional space cannot be generated by a norm. Indeed, let $q$ be a continuous seminorm and $q_{\alpha_j} = |\ell_{\alpha_j}|$ as in the lemma. Then $\bigcap_{j=1}^{n} \ker(\ell_{\alpha_j})$ has codimension at most $n$ and hence contains some $x \neq 0$ implying that $q(x) \leq \sum_{j=1}^{n} c_j q_{\alpha_j}(x) = 0$. Thus $q$ is no norm. Similarly, the other examples cannot be generated by a norm except in finite dimensional cases.

Moreover, note that the topology is translation invariant in the sense that $U(x)$ is a neighborhood of $x$ if and only if $U(x) - x = \{y - x | y \in U(x)\}$ is a neighborhood of 0. Hence we can restrict our attention to neighborhoods of 0 (this is of course true for any topological vector space). Hence if $X$ and $Y$ are topological vector spaces, then a linear map $A : X \rightarrow Y$ will be continuous if and only if it is continuous at 0. Moreover, if $Y$ is a locally convex space with respect to some seminorms $p_\beta$, then $A$ will be continuous if and only if $p_\beta \circ A$ is continuous for every $\beta$ (Lemma 0.11). Finally, since $p_\beta \circ A$ is a seminorm, the previous lemma implies:

Corollary 5.16. Let $(X, \{q_\alpha\})$ and $(Y, \{p_\beta\})$ be locally convex vector spaces. Then a linear map $A : X \rightarrow Y$ is continuous if and only if for every $\beta$ there are some seminorms $q_{\alpha_j}$ and constants $c_j > 0$, $1 \leq j \leq n$, such that $p_\beta(Ax) \leq \sum_{j=1}^{n} c_j q_{\alpha_j}(x)$.

It will shorten notation when sums of the type $\sum_{j=1}^{n} c_j q_{\alpha_j}(x)$, which appeared in the last two results, can be replaced by a single expression $cq_\alpha$. This can be done if the family of seminorms $\{q_\alpha\}_{\alpha \in A}$ is directed, that is, for given $\alpha, \beta \in A$ there is a $\gamma \in A$ such that $q_\alpha(x) + q_\beta(x) \leq C q_\gamma(x)$ for some $C > 0$. Moreover, if $F(A)$ is the set of all finite subsets of $A$, then $\{\tilde{q}_F = \sum_{\alpha \in F} q_\alpha\}_{F \in F(\mathcal{A})}$ is a directed family which generates the same topology (since every $\tilde{q}_F$ is continuous with respect to the original family we do not get any new open sets).
While the family of seminorms is in most cases more convenient to work with, it is important to observe that different families can give rise to the same topology and it is only the topology which matters for us. In fact, it is possible to characterize locally convex vector spaces as topological vector spaces which have a neighborhood basis at 0 of absolutely convex sets. Here a set \( U \) is called **absolutely convex**, if for \( |\alpha| + |\beta| \leq 1 \) we have \( \alpha U + \beta U \subseteq U \). Since the sets \( q_\alpha^{-1}([0,\varepsilon]) \) are absolutely convex we always have such a basis in our case. To see the converse note that such a neighborhood \( U \) of 0 is also absorbing (Problem 5.12) and hence the corresponding Minkowski functional (5.1) is a seminorm (Problem 5.17). By construction, these seminorms generate the topology since if \( U_0 = \bigcap_{j=1}^n q_\alpha^{-1}([0,\varepsilon_j]) \subseteq U \) we have for the corresponding Minkowski functionals \( p_U(x) \leq p_{U_0}(x) \leq \varepsilon^{-1} \sum_{j=1}^n q_{\alpha_j}(x) \), where \( \varepsilon = \min \varepsilon_j \). With a little more work (Problem 5.16), one can even show that it suffices to assume to have a neighborhood basis at 0 of convex open sets.

Given a topological vector space \( X \) we can define its dual space \( X^* \) as the set of all continuous linear functionals. However, while it can happen in general that the dual space is empty, \( X^* \) will always be nontrivial for a locally convex space since the Hahn–Banach theorem can be used to construct linear functionals (using a continuous seminorm for \( \varphi \) in Theorem 4.14) and also the geometric Hahn–Banach theorem (Theorem 5.3) holds (see also its corollaries). In this respect note that for every continuous linear functional \( \ell \) in a topological vector space \( |\ell|^{-1}(0,\varepsilon) \) is an absolutely convex open neighborhoods of 0 and hence existence of such sets is necessary for the existence of nontrivial continuous functionals. As a natural topology on \( X^* \) we could use the weak-* topology defined to be the weakest topology generated by the family of all point evaluations \( q_{\ell}(\ell) = |\ell(x)| \) for all \( x \in X \). Since different linear functionals must differ at least at one point the weak-* topology is Hausdorff. Given a continuous linear operator \( A : X \to Y \) between locally convex spaces we can define its adjoint \( A' : Y^* \to X^* \) as before,

\[
(A'y^*)(x) := y^*(Ax).
\]  

A brief calculation

\[
q_\ell(A'y^*) = |(A'y^*)(x)| = |y^*(Ax)| = q_{A\ell}(y^*)
\]  

verifies that \( A' \) is continuous in the weak-* topology by virtue of Corollary 5.16.

The remaining theorems we have established for Banach spaces were consequences of the Baire theorem (which requires a complete metric space) and this leads us to the question when a locally convex space is a metric space. From our above analysis we see that a locally convex vector space
will be first countable if and only if countably many seminorms suffice to
determine the topology. In this case \( X \) turns out to be metrizable.

**Theorem 5.17.** A locally convex Hausdorff space is metrizable if and only
if it is first countable. In this case there is a countable family of separated
seminorms \( \{ q_n \}_{n \in \mathbb{N}} \) generating the topology and a metric is given by
\[
d(x, y) := \max_{n \in \mathbb{N}} \frac{1}{2^n} \frac{q_n(x - y)}{1 + q_n(x - y)}. \tag{5.16}
\]

**Proof.** If \( X \) is first countable there is a countable neighborhood base at
0 and hence also a countable neighborhood base of absolutely convex sets.
The Minkowski functionals corresponding to the latter base are seminorms
of the required type.

Now in this case it is straightforward to check that (5.16) defines a metric
(see also Problem 0.3). Moreover, the balls \( B^m_{\frac{r}{2^{\alpha} - n}}(x) \) are clearly open and convex (note that the intersection is
finite). Conversely, for every set of the form (5.13) we can choose
\( \varepsilon = \min \{ 2^{-\alpha_j \frac{s_j}{1 + s_j}} | 1 \leq j \leq n \} \) such that
\( B_{\varepsilon}(x) \) will be contained in this set. Hence both topologies are equivalent (cf. Lemma 0.2).

In general, a locally convex vector space \( X \) which has a separated count-
able family of seminorms is called a **Fréchet space** if it is complete with
respect to the metric (5.16). Note that the metric (5.16) is **translation
invariant**
\[
d(f, g) = d(f - h, g - h). \tag{5.17}
\]

**Example.** The continuous functions \( C(\mathbb{R}) \) together with local uniform con-
vergence are a Fréchet space. A countable family of seminorms is for example
\[
\| f \|_j = \sup_{|x| \leq j} | f(x) |, \quad j \in \mathbb{N}. \tag{5.18}
\]
Then \( f_k \to f \) if and only if \( \| f_k - f \|_j \to 0 \) for all \( j \in \mathbb{N} \) and it follows that
\( C(\mathbb{R}) \) is complete.

**Example.** The space \( C^\infty(\mathbb{R}^m) \) together with the seminorms
\[
\| f \|_{j,k} = \sum_{|\alpha| \leq k} \sup_{|x| \leq j} | \partial^\alpha f(x) |, \quad j \in \mathbb{N}, \ k \in \mathbb{N}_0, \tag{5.19}
\]
is a Fréchet space.

Note that \( \partial^\alpha : C^\infty(\mathbb{R}^m) \to C^\infty(\mathbb{R}^m) \) is continuous. Indeed by Corol-
lary 5.16 it suffices to observe that \( \| \partial^\alpha f \|_{j,k} \leq \| f \|_{j,k+|\alpha|} \).

**Example.** The **Schwartz space**
\[
S(\mathbb{R}^m) = \{ f \in C^\infty(\mathbb{R}^m) | \sup_x | x^\alpha (\partial^\beta f)(x) | < \infty, \ \forall \alpha, \beta \in \mathbb{N}^m_0 \} \tag{5.20}
\]
5. Further topics on Banach spaces

Together with the seminorms
\[ q_{\alpha,\beta}(f) = \| x^\alpha (\partial_\beta f)(x) \|_\infty, \quad \alpha, \beta \in \mathbb{N}_0^m. \] (5.21)

To see completeness note that a Cauchy sequence \( f_n \) is in particular a Cauchy sequence in \( C^\infty(\mathbb{R}^m) \). Hence there is a limit \( f \in C^\infty(\mathbb{R}^m) \) such that all derivatives converge uniformly. Moreover, since Cauchy sequences are bounded \( \| x^\alpha (\partial_\beta f_n)(x) \|_\infty \leq C_{\alpha,\beta} \) we obtain \( \| x^\alpha (\partial_\beta f)(x) \|_\infty \leq C_{\alpha,\beta} \) and thus \( f \in \mathcal{S}(\mathbb{R}^m) \).

Again \( \partial_\gamma : \mathcal{S}(\mathbb{R}^m) \to \mathcal{S}(\mathbb{R}^m) \) is continuous since \( q_{\alpha,\beta}(\partial_\gamma f) \leq q_{\alpha,\beta+\gamma}(f) \).

The dual space \( \mathcal{S}^*(\mathbb{R}^m) \) is known as the space of tempered distributions.

**Example.** The space of all entire functions \( f(z) \) (i.e. functions which are holomorphic on all of \( \mathbb{C} \)) together with the seminorms \( \| f \|_j = \sup_{|z| \leq j} |f(z)| \), \( j \in \mathbb{N} \), is a Fréchet space. Completeness follows from the Weierstraß convergence theorem which states that a limit of holomorphic functions which is uniform on every compact subset is again holomorphic.

**Example.** In all of the previous examples the topology cannot be generated by a norm. For example, if \( q \) is a norm for \( C(\mathbb{R}) \), then Lemma 5.15 that there is some index \( j \) such that \( q(f) \leq C\|f\|_j \). Now choose a nonzero function which vanishes on \(-j, j \) to get a contradiction.

There is another useful criterion when the topology can be described by a single norm. To this end we call a set \( B \subseteq X \) bounded if \( \sup_{x \in B} q_\alpha(x) < \infty \) for every \( \alpha \). By Corollary 5.16 this will then be true for any continuous seminorm on \( X \).

**Theorem 5.18 (Kolmogorov).** A locally convex vector space can be generated from a single seminorm if and only if it contains a bounded open set.

**Proof.** In a Banach space every open ball is bounded and hence only the converse direction is nontrivial. So let \( U \) be a bounded open set. By shifting and decreasing \( U \) if necessary we can assume \( U \) to be an absolutely convex open neighborhood of 0 and consider the associated Minkowski functional \( q = pu \). Then since \( U = \{ x | q(x) < 1 \} \) and \( \sup_{x \in U} q_\alpha(x) = C_\alpha < \infty \) we infer \( q_\alpha(x) \leq C_\alpha q(x) \) (Problem 5.13) and thus the single seminorm \( q \) generates the topology.

Finally, we mention that, since the Baire category theorem holds for arbitrary complete metric spaces, the open mapping theorem (Theorem 4.5), the inverse mapping theorem (Theorem 4.6) and the closed graph (Theorem 4.7) hold for Fréchet spaces without modifications. In fact, they are
formulated such that it suffices to replace Banach by Fréchet in these theorems as well as their proofs (concerning the proof of Theorem 4.5 take into account Problems 5.12 and 5.19).

**Problem 5.12.** In a topological vector space every neighborhood $U$ of 0 is absorbing.

**Problem 5.13.** Let $p, q$ be two seminorms. Then $p(x) \leq Cq(x)$ if and only if $q(x) < 1$ implies $p(x) < C$.

**Problem 5.14.** Let $X$ be a vector space. We call a set $U$ balanced if $\alpha U \subseteq U$ for every $|\alpha| \leq 1$. Show that a set is balanced and convex if and only if it is absolutely convex.

**Problem 5.15.** The intersection of arbitrary absolutely convex/balanced sets is again absolutely convex/balanced convex. Hence we can define the absolutely convex/balanced hull of a set $U$ as the smallest absolutely convex/balanced set containing $U$, that is, the intersection of all absolutely convex/balanced sets containing $U$. Show that the absolutely convex hull is given by

$$ahull(U) := \left\{ \sum_{j=1}^{n} \lambda_j x_j | n \in \mathbb{N}, x_j \in U, \sum_{j=1}^{n} |\lambda_j| \leq 1 \right\}$$

and the balanced hull by

$$bhull(U) := \{ \alpha x | x \in U, |\alpha| \leq 1 \}.$$ 

Show that $ahull(U) = hull(bhull(U))$.

**Problem 5.16.** In a topological vector space every convex open neighborhood $U$ of zero contains an absolutely convex open neighborhood of zero. (Hint: By continuity of the scalar multiplication $U$ contains a set of the form $B^C_{\varepsilon}(0) \cdot V$, where $V$ is an open neighborhood of zero.)

**Problem 5.17.** Let $X$ be a vector space. Show that the Minkowski functional of a balanced, convex, absorbing set is a seminorm.

**Problem 5.18.** If a locally convex space is Hausdorff then any corresponding family of seminorms is separated.

**Problem 5.19.** Suppose $X$ is a complete vector space with a translation invariant metric $d$. Show that \(\sum_{j=1}^{\infty} d(0,x_j) < \infty\) implies that

$$\sum_{j=1}^{\infty} x_j = \lim_{n \to \infty} \sum_{j=1}^{n} x_j$$

exists and

$$d(0,\sum_{j=1}^{\infty} x_j) \leq \sum_{j=1}^{\infty} d(0,x_j)$$
in this case (compare also Problem 1.4).

**Problem 5.20.** Instead of (5.16) one frequently uses

\[
\tilde{d}(x, y) := \sum_{n \in \mathbb{N}} \frac{q_n(x - y)}{1 + q_n(x - y)},
\]

Show that this metric generates the same topology.

Consider the Fréchet space \( C(\mathbb{R}) \) with \( q_n(f) = \sup_{[-n,n]} |f| \). Show that the metric balls with respect to \( \tilde{d} \) are not convex.

**Problem 5.21.** Suppose \( X \) is a metric vector space. Then balls are convex if and only if the metric is quasiconvex:

\[
d(\lambda x + (1 - \lambda)y, z) \leq \max\{d(x, z), d(y, z)\}, \quad \lambda \in (0, 1).
\]

(See also Problem 4.36.)

**Problem 5.22.** Consider \( \ell^p(\mathbb{N}) \) for \( p \in (0, 1) \) — compare Problem 1.14. Show that \( \| \cdot \|_p \) is not convex. Show that every convex open set is unbounded. Conclude that it is not a locally convex vector space. (Hint: Consider \( B_R(0) \). Then for \( r < R \) all vectors which have one entry equal to \( r \) and all other entries zero are in this ball. By taking convex combinations all vectors which have \( n \) entries equal to \( r/n \) are in the convex hull. The quasinorm of such a vector is \( n^{1/p-1}r \).)

**Problem 5.23.** Show that \( C^\infty_c(\mathbb{R}^m) \) is dense in \( \mathcal{S}(\mathbb{R}^m) \).

**Problem 5.24.** Let \( X \) be a topological vector space and \( M \) a closed subspace. Show that the quotient space \( X/M \) is again a topological vector space and that \( \pi : X \to X/M \) is linear, continuous, and open. Show that points in \( X/M \) are closed.

### 5.5. Uniformly convex spaces

In a Banach space \( X \), the unit ball is convex by the triangle inequality. Moreover, \( X \) is called **strictly convex** if the unit ball is a strictly convex set, that is, if for any two points on the unit sphere their average is inside the unit ball. See Problem 1.12 for some equivalent definitions. This is illustrated in Figure 1 which shows that in \( \mathbb{R}^2 \) this is only true for \( 1 < p < \infty \).

**Example.** By Problem 1.12 \( \ell^p(\mathbb{N}) \) is strictly convex for \( 1 < p < \infty \) but not for \( p = 1, \infty \). \( \diamond \)

A more qualitative is to require that if two unit vectors \( x, y \) satisfy \( \|x - y\| \geq \varepsilon \) for some \( \varepsilon > 0 \), then there is some \( \delta > 0 \) such that \( \|\frac{x+y}{2}\| \leq 1 - \delta \). In this case one calls \( X \) **uniformly convex** and

\[
\delta(\varepsilon) := \inf \left\{ 1 - \frac{\|x+y\|}{\|x\|} \mid \|x\| = \|y\| = 1, \|x - y\| \geq \varepsilon \right\}, \quad 0 \leq \varepsilon \leq 2, \quad (5.22)
\]
is called the modulus of convexity. Of course every uniformly convex space is
strictly convex. In finite dimensions the converse is also true (Problem 5.26).

Note that \( \delta \) is nondecreasing and

\[
\| \frac{x+y}{2} \| = \| x - \frac{x-y}{2} \| \geq 1 - \frac{\varepsilon}{2}
\]

shows \( 0 \leq \delta(\varepsilon) \leq \frac{\varepsilon}{2} \). Moreover, \( \delta(2) = 1 \) implies \( X \) strictly convex. In fact in this case \( 1 = \delta(2) \leq 1 - \| x+y \| \leq 1 \) for \( 2 \leq \| x-y \| \leq 2 \). That is, \( x = -y \) whenever \( \| x-y \| = 2 = \| x \| + \| y \| \).

**Example.** Every Hilbert space is uniformly convex with modulus of con-
vexity \( \delta(\varepsilon) = 1 - \sqrt{1 - \frac{\varepsilon^2}{4}} \) (Problem 5.25).

**Example.** Consider \( C[0,1] \) with the norm

\[
\| x \| := \| x \|_\infty + \| x \|_2 = \max_{t \in [0,1]} |x(t)| + \left( \int_0^1 |x(t)|^2 dt \right)^{-1}.
\]

Note that by \( \| x \|_2 \leq \| x \|_\infty \) this norm is equivalent to the usual one: \( \| x \|_\infty \leq \| x \| \leq 2 \| x \|_\infty \). While with the usual norm \( \| . \|_\infty \) this space is not strictly
convex, it is with the new one. To see this we use (i) from Problem 1.12. Then if \( \| x+y \| = \| x \| + \| y \| \) we must have both \( \| x+y \|_\infty = \| x \|_\infty + \| y \|_\infty \) and \( \| x+y \|_2 = \| x \|_2 + \| y \|_2 \). Hence strict convexity of \( \| . \|_2 \) implies strict
convexity of \( \| . \| \).

Note however, that \( \| . \| \) is not uniformly convex. In fact, since by the
Milman–Pettis theorem below, every uniformly convex space is reflexive,
there cannot be an equivalent norm on \( C[0,1] \) which is uniformly convex
(cf. the example on page 141).

**Example.** It can be shown that \( \ell^p(N) \) is uniformly convex for \( 1 < p < \infty \)
(see Theorem 10.11).

Equivalently, uniform convexity implies that if the average of two unit
vectors is close to the boundary, then they must be close to each other.
Specifically, if \( \| x \| = \| y \| = 1 \) and \( \| \frac{x+y}{2} \| > 1 - \delta(\varepsilon) \) then \( \| x - y \| < \varepsilon \). The
following result (which generalizes Lemma 4.29) uses this observation:

**Theorem 5.19** (Radon–Riesz theorem). Let \( X \) be a uniformly convex Ban-
ach space and let \( x_n \to x \). Then \( x_n \to x \) if and only if \( \limsup \| x_n \| \leq \| x \| \).

**Proof.** If \( x = 0 \) there is nothing to prove. Hence we can assume \( x_n \neq 0 \) for
all \( n \) and consider \( y_n := \frac{x_n}{\| x_n \|} \). Then \( y_n \to y := \frac{x}{\| x \|} \) and it sufices to show
\( y_n \to y \). Next choose a linear functional \( \ell \in X^* \) with \( \| \ell \| = 1 \) and \( \ell(y) = 1 \). Then

\[
\ell \left( \frac{y_n + y}{2} \right) \leq \left\| \frac{y_n + y}{2} \right\| \leq 1
\]
and letting \( n \to \infty \) shows \( \|y_{n+1}\| \to 1 \). Finally uniform convexity shows \( y_n \to y \).

For the proof of the next result we need to following equivalent condition.

**Lemma 5.20.** Let \( X \) be a Banach space. Then

\[
\delta(\varepsilon) = \inf \left\{ 1 - \frac{x+y}{2} \left\| \frac{x+y}{2} \right\| \left| \frac{x+y}{2} \right\| \leq 1, \|y\| \leq 1, \|x - y\| \geq \varepsilon \right\}, \quad 0 \leq \varepsilon \leq 2.
\]

**Proof.** It suffices to show that for given \( x \) and \( y \) which are not both on the unit sphere there is a better pair in the real subspace spanned by these vectors. By scaling we could get a better pair if both were strictly inside the unit ball and hence we can assume at least one vector to have norm one, say \( \|x\| = 1 \). Moreover, consider

\[
u(t) := \frac{\cos(t) x + \sin(t) y}{\|\cos(t) x + \sin(t) y\|}, \quad v(t) := u(t) + (y - x).
\]

Then \( \|v(0)\| = \|y\| < 1 \). Moreover, let \( t_0 \in (\frac{\pi}{2}, \frac{3\pi}{2}) \) be the value such that the line from \( x \) to \( u(t_0) \) passes through \( y \). Then, by convexity we must have \( \|v(t_0)\| > 1 \) and by the intermediate value theorem there is some \( 0 < t_1 < t_0 \) with \( \|v(t_1)\| = 1 \). Let \( u := u(t_1), v := v(t_1) \). The line through \( u \) and \( x \) is not parallel to the line through \( 0 \) and \( x + y \) and hence there are \( \alpha, \lambda \geq 0 \) such that

\[
\frac{\alpha}{2} (x + y) = \lambda u + (1 - \lambda)x.
\]

Moreover, since the line from \( x \) to \( u \) is above the line from \( x \) to \( y \) (since \( t_1 < t_0 \)) we have \( \alpha \geq 1 \). Rearranging this equation we get

\[
\frac{\alpha}{2} (u + v) = (\alpha + \lambda)u + (1 - \alpha - \lambda)x.
\]

Now, by convexity of the norm, if \( \lambda \leq 1 \) we have \( \lambda + \alpha > 1 \) and thus \( \|\lambda u + (1 - \lambda)x\| \leq 1 < \|(\alpha + \lambda)u + (1 - \alpha - \lambda)x\| \). Similarly, if \( \lambda > 1 \) we have \( \|\lambda u + (1 - \lambda)x\| < \|(\alpha + \lambda)u + (1 - \alpha - \lambda)x\| \) again by convexity of the norm. Hence \( \frac{1}{2} (x + y) \leq \frac{1}{2} (u + v) \) and \( u, v \) is a better pair.

Now we can prove:

**Theorem 5.21** (Milman–Pettis). A uniformly convex Banach space is reflexive.

**Proof.** Pick some \( x'' \in X^{**} \) with \( \|x''\| = 1 \). It suffices to find some \( x \in B_1(0) \) with \( \|x'' - J(x)\| \leq \varepsilon \). So fix \( \varepsilon > 0 \) and \( \delta := \delta(\varepsilon) \), where \( \delta(\varepsilon) \) is the modulus of convexity. Then \( \|x''\| = 1 \) implies that we can find some \( \ell \in X^* \) with \( \|\ell\| = 1 \) and \( |x''(\ell)| > 1 - \frac{\delta}{2} \). Consider the weak-* neighborhood

\[
U := \{ y'' \in X^{**} \mid |(y'' - x'')(\ell)| < \frac{\delta}{2} \}
\]
of $x''$. By Goldstine’s theorem (Theorem 5.13) there is some $x \in \overline{B}_1(0)$ with $J(x) \in U$ and this is the $x$ we are looking for. In fact, suppose this were not the case. Then the set $V := X^{**} \setminus B_{**}^*(J(x))$ is another weak-* neighborhood of $x''$ (since $B_{**}^*(J(x))$ is weak-* compact by the Banach-Alaoglu theorem) and appealing again to Goldstine’s theorem there is some $y \in \overline{B}_1(0)$ with $J(y) \in U \cap V$. Since $x, y \in U$ we obtain

$$1 - \frac{\delta}{2} < |x''(\ell)| \leq |\ell(\frac{x+y}{2})| + \frac{\delta}{2} \quad \Rightarrow \quad 1 - \delta < |\ell(\frac{x+y}{2})| \leq \frac{x+y}{2},$$

a contradiction to uniform convexity since $\|x - y\| \geq \varepsilon$. □

**Problem 5.25.** Show that a Hilbert space is uniformly convex. (Hint: Use the parallelogram law.)

**Problem 5.26.** Show that a finite dimensional space is uniformly convex if and only if it is strictly convex.

**Problem 5.27.** Let $X$ be strictly convex. Show that every nonzero linear functional attains its norm for at most one unit vector (cf. Problem 4.13).
Bounded linear operators

We have started our study by looking at eigenvalue problems which, from a historic viewpoint, were one of the key problems driving the development of functional analysis. In Chapter 3 we have investigated compact operators in Hilbert space and we have seen that they allow a treatment similar to what is known from matrices. However, more sophisticated problems will lead to operators whose spectra consist of more than just eigenvalues. Hence we want to go one step further and look at spectral theory for bounded operators. Here one of the driving forces was the development of quantum mechanics (there even the boundedness assumption is too much — but first things first). A crucial role is played by the algebraic structure, namely recall from Section 1.5 that the bounded linear operators on $X$ form a Banach space which has a (non-commutative) multiplication given by composition. In order to emphasize that it is only this algebraic structure which matters, we will develop the theory from this abstract point of view. While the reader should always remember that bounded operators on a Hilbert space is what we have in mind as the prime application, examples will apply these ideas also to other cases thereby justifying the abstract approach.

To begin with, the operators could be on a Banach space (note that even if $X$ is a Hilbert space, $L^2(X)$ will only be a Banach space) but eventually again self-adjointness will be needed. Hence we will need the additional operation of taking adjoints.
6.1. Banach algebras

A Banach space \( X \) together with a multiplication satisfying
\[
(x + y)z = xz + yz, \quad x(y + z) = xy + xz, \quad x, y, z \in X,
\]
and
\[
(xy)z = x(yz), \quad \alpha(xy) = (\alpha x)y = x(\alpha y), \quad \alpha \in \mathbb{C},
\]
and
\[
\|xy\| \leq \|x\|\|y\|.
\]
is called a **Banach algebra**. In particular, note that (6.3) ensures that multiplication is continuous (Problem 6.1). An element \( e \in X \) satisfying
\[
ex = xe = x, \quad \forall x \in X
\]
is called **identity** (show that \( e \) is unique) and we will assume \( \|e\| = 1 \) in this case.

**Example.** The continuous functions \( C(I) \) over some compact interval form a commutative Banach algebra with identity 1.

**Example.** The bounded linear operators \( \mathcal{L}(X) \) form a Banach algebra with identity \( \mathbb{I} \).

**Example.** The bounded sequences \( \ell^\infty(\mathbb{N}) \) together with the component-wise product form a commutative Banach algebra with identity 1.

**Example.** The space of all periodic continuous functions which have an absolutely convergent Fourier series \( \mathcal{A} \) together with the norm
\[
\|f\|_\mathcal{A} := \sum_{k \in \mathbb{Z}} |\hat{f}_k|
\]
and the usual product is known as the **Wiener algebra**. Of course as a Banach space it is isomorphic to \( \ell^1(\mathbb{Z}) \) via the Fourier transform. To see that it is a Banach algebra note that
\[
f(x)g(x) = \sum_{k \in \mathbb{Z}} \hat{f}_k e^{ikx} \sum_{j \in \mathbb{Z}} \hat{g}_j e^{ijx} = \sum_{k,j \in \mathbb{Z}} \hat{f}_k \hat{g}_j e^{i(k+j)x}
\]
\[
= \sum_{k \in \mathbb{Z}} \left( \sum_{j \in \mathbb{Z}} \hat{f}_j \hat{g}_{k-j} \right) e^{ikx}.
\]
Moreover, interchanging the order of summation
\[
\|fg\|_\mathcal{A} = \sum_{k \in \mathbb{Z}} \left| \sum_{j \in \mathbb{Z}} \hat{f}_j \hat{g}_{k-j} \right| \leq \sum_{j \in \mathbb{Z}} \sum_{k \in \mathbb{Z}} |\hat{f}_j| |\hat{g}_{k-j}| = \|f\|_\mathcal{A} \|g\|_\mathcal{A}
\]
shows that \( \mathcal{A} \) is a Banach algebra. The identity is of course given by \( e(x) \equiv 1 \). Moreover, note that \( \mathcal{A} \subseteq C_{\text{per}}[-\pi, \pi] \) and \( \|f\|_\infty \leq \|f\|_\mathcal{A} \).
Example. The space $L^1(\mathbb{R}^n)$ together with the convolution
\[(g * f)(x) := \int_{\mathbb{R}^n} g(x - y)f(y)dy = \int_{\mathbb{R}^n} g(y)f(x - y)dy\] (6.5)
is a commutative Banach algebra (Problem 6.7) without identity. □

A Banach algebra with identity is also known as **unital** and we will assume $X$ to be a Banach algebra with identity $e$ throughout the rest of this section. Note that an identity can always be added if needed (Problem 6.2).

An element $x \in X$ is called **invertible** if there is some $y \in X$ such that
\[xy = yx = e.\] (6.6)
In this case $y$ is called the inverse of $x$ and is denoted by $x^{-1}$. It is straightforward to show that the inverse is unique (if one exists at all) and that
\[(xy)^{-1} = y^{-1}x^{-1}, \quad (x^{-1})^{-1} = x.\] (6.7)
In particular, the set of invertible elements $\mathcal{G}(X)$ forms a group under multiplication.

Example. If $X = \mathcal{L}(\mathbb{C}^n)$ is the set of $n$ by $n$ matrices, then $\mathcal{G}(X) = \text{GL}(n)$ is the general linear group. □

Example. Let $X = \mathcal{L}(\ell^p(\mathbb{N}))$ and recall the shift operators $S^\pm$ defined via $(S^\pm a)_j = a_{j\pm1}$ with the convention that $a_0 = 0$. Then $S^+S^- = I$ but $S^-S^+ \neq I$. Moreover, note that $S^+S^-$ is invertible while $S^-S^+$ is not. So you really need to check both $xy = e$ and $yx = e$ in general. □

If $x$ is invertible then the same will be true for any nearby elements. This will be a consequence from the following straightforward generalization of the geometric series to our abstract setting.

**Lemma 6.1.** Let $X$ be a Banach algebra with identity $e$. Suppose $\|x\| < 1$. Then $e - x$ is invertible and
\[(e - x)^{-1} = \sum_{n=0}^{\infty} x^n.\] (6.8)

**Proof.** Since $\|x\| < 1$ the series converges and
\[(e - x) \sum_{n=0}^{\infty} x^n = \sum_{n=0}^{\infty} x^n - \sum_{n=1}^{\infty} x^n = e\]
respectively
\[\left(\sum_{n=0}^{\infty} x^n\right)(e - x) = \sum_{n=0}^{\infty} x^n - \sum_{n=1}^{\infty} x^n = e.\] □
Corollary 6.2. Suppose $x$ is invertible and $\|x^{-1}y\| < 1$ or $\|yx^{-1}\| < 1$. Then $(x - y)$ is invertible as well and

$$(x - y)^{-1} = \sum_{n=0}^{\infty} (x^{-1}y)^n x^{-1} \quad \text{or} \quad (x - y)^{-1} = \sum_{n=0}^{\infty} x^{-1}(yx^{-1})^n.$$  \hspace{1cm} (6.9)

In particular, both conditions are satisfied if $\|y\| < \|x^{-1}\|^{-1}$ and the set of invertible elements $\mathcal{G}(X)$ is open and taking the inverse is continuous:

$$\|(x - y)^{-1} - x^{-1}\| \leq \frac{\|y\|\|x^{-1}\|^2}{1 - \|x^{-1}y\|}.$$  \hspace{1cm} (6.10)

**Proof.** Just observe $x - y = x(e - x^{-1}y) = (e - yx^{-1})x$. \hfill \Box

The **resolvent set** is defined as

$$\rho(x) := \{\alpha \in \mathbb{C} | \exists (x - \alpha)^{-1}\} \subseteq \mathbb{C},$$  \hspace{1cm} (6.11)

where we have used the shorthand notation $x - \alpha = x - \alpha e$. Its complement is called the **spectrum**

$$\sigma(x) := \mathbb{C} \setminus \rho(x).$$  \hspace{1cm} (6.12)

It is important to observe that the inverse has to exist as an element of $X$. That is, if the elements of $X$ are bounded linear operators, it does not suffice that $x - \alpha$ is injective, as it might not be surjective. If it is bijective, boundedness of the inverse will come for free from the inverse mapping theorem.

**Example.** If $X := \mathcal{L}(\mathbb{C}^n)$ is the space of $n$ by $n$ matrices, then the spectrum is just the set of eigenvalues. More general, if $X$ are the bounded linear operators on an infinite-dimensional Hilbert or Banach space, then every eigenvalue will be in the spectrum but the converse is not true in general as an injective operator might not be surjective. In fact, this already can happen for compact operators where 0 could be in the spectrum without being an eigenvalue. \hfill \Diamond

**Example.** If $X := C(I)$, then the spectrum of a function $x \in C(I)$ is just its range, $\sigma(x) = x(I)$. Indeed, if $\alpha \not\in \text{Ran}(x)$ then $t \mapsto (x(t) - \alpha)^{-1}$ is the inverse of $x - \alpha$ (note that $\text{Ran}(x)$ is compact). Conversely, if $\alpha \in \text{Ran}(x)$ and $y$ were an inverse, then $y(t_0)(x(t_0) - \alpha) = 1$ gives a contradiction for any $t_0 \in I$ with $f(t_0) = \alpha$. \hfill \Diamond

**Example.** If $X = \mathcal{A}$ is the Wiener algebra, then, as in the previous example, every function which vanishes at some point cannot be inverted. If it does not vanish anywhere, it can be inverted and the inverse will be a continuous function. But will it again have a convergent Fourier series, that is, will it be in the Wiener Algebra? The affirmative answer of this question is a famous theorem of Wiener, which will be given later in Theorem 6.18. \hfill \Diamond
The map $\alpha \mapsto (x - \alpha)^{-1}$ is called the **resolvent** of $x \in X$. If $\alpha_0 \in \rho(x)$ we can choose $x \to x - \alpha_0$ and $y \to \alpha - \alpha_0$ in (6.9) which implies

$$(x - \alpha)^{-1} = \sum_{n=0}^{\infty} (\alpha - \alpha_0)^n (x - \alpha_0)^{-n - 1}, \quad |\alpha - \alpha_0| < \| (x - \alpha_0)^{-1} \|^{-1}. \quad (6.13)$$

In particular, since the radius of convergence cannot exceed the distance to the spectrum (since everything within the radius of convergent must belong to the resolvent set), we see that the norm of the resolvent must diverge

$$\| (x - \alpha)^{-1} \| \geq \frac{1}{\text{dist}(\alpha, \sigma(x))} \quad (6.14)$$

as $\alpha$ approaches the spectrum. Moreover, this shows that $(x - \alpha)^{-1}$ has a convergent power series with coefficients in $X$ around every point $\alpha_0 \in \rho(x)$. As in the case of coefficients in $\mathbb{C}$, such functions will be called **analytic**.

In particular, $\ell((x - \alpha)^{-1})$ is a complex-valued analytic function for every $\ell \in X^*$ and we can apply well-known results from complex analysis:

**Theorem 6.3.** For every $x \in X$, the spectrum $\sigma(x)$ is compact, nonempty and satisfies

$$\sigma(x) \subseteq \{ \alpha \mid |\alpha| \leq \| x \| \}. \quad (6.15)$$

**Proof.** Equation (6.13) already shows that $\rho(x)$ is open. Hence $\sigma(x)$ is closed. Moreover, $x - \alpha = -\alpha (e - \frac{1}{\alpha} x)$ together with Lemma 6.1 shows

$$(x - \alpha)^{-1} = -\frac{1}{\alpha} \sum_{n=0}^{\infty} \left( \frac{x}{\alpha} \right)^n, \quad |\alpha| > \| x \|,$$

which implies $\sigma(x) \subseteq \{ \alpha \mid |\alpha| \leq \| x \| \}$ is bounded and thus compact. Moreover, taking norms shows

$$\| (x - \alpha)^{-1} \| \leq \frac{1}{|\alpha|} \sum_{n=0}^{\infty} \| x \|^n |\alpha|^{-n} = \frac{1}{|\alpha| - \| x \|}, \quad |\alpha| > \| x \|,$$

which implies $(x - \alpha)^{-1} \to 0$ as $\alpha \to \infty$. In particular, if $\sigma(x)$ is empty, then $\ell((x - \alpha)^{-1})$ is an entire analytic function which vanishes at infinity. By Liouville’s theorem we must have $\ell((x - \alpha)^{-1}) = 0$ for all $\ell \in X^*$ in this case, and so $(x - \alpha)^{-1} = 0$, which is impossible. □

**Example.** The spectrum of the matrix

$$A := \begin{pmatrix}
0 & 1 & 0 & 1 \\
0 & 1 & \cdots & \cdots \\
-\epsilon_0 & -\epsilon_1 & \cdots & \cdots & -\epsilon_{n-1}
\end{pmatrix}$$
is given by the zeros of the polynomial (show this)
\[ \det(zI - A) = z^n + c_{n-1}z^{n-1} + \cdots + c_1z + c_0. \]
Hence the fact that \( \sigma(A) \) is nonempty implies the **fundamental theorem of algebra**, that every non-constant polynomial has at least one zero. \( \diamond \)

As another simple consequence we obtain:

**Theorem 6.4** (Gelfand–Mazur). Suppose \( X \) is a Banach algebra in which every element except 0 is invertible. Then \( X \) is isomorphic to \( \mathbb{C} \).

**Proof.** Pick \( x \in X \) and \( \alpha \in \sigma(x) \). Then \( x - \alpha \) is not invertible and hence \( x - \alpha = 0 \), that is \( x = \alpha \). Thus every element is a multiple of the identity. \( \square \)

Now we look at functions of \( x \). Given a polynomial \( p(\alpha) = \sum_{j=0}^{n} p_j \alpha^j \) we of course set
\[ p(x) := \sum_{j=0}^{n} p_j x^j. \] (6.16)
In fact, we could easily extend this definition to arbitrary convergent power series whose radius of convergence is larger than \( \|x\| \) (cf. Problem 1.32).

While this will give a nice functional calculus sufficient for many applications our aim is the spectral theorem which will allow us to handle arbitrary continuous functions. Since continuous functions can be approximated by polynomials by the Weierstrass theorem, polynomials will be sufficient for now. Moreover, the following result will be one of the two key ingredients for the proof of the spectral theorem.

**Theorem 6.5** (Spectral mapping). For every polynomial \( p \) and \( x \in X \) we have
\[ \sigma(p(x)) = p(\sigma(x)), \] (6.17)
where \( p(\sigma(x)) := \{p(\alpha)|\alpha \in \sigma(x)\} \).

**Proof.** Fix \( \alpha_0 \in \mathbb{C} \) and observe
\[ p(x) - p(\alpha_0) = (x - \alpha_0)q_0(x). \]
If \( p(\alpha_0) \notin \sigma(p(x)) \) we have
\[ (x - \alpha_0)^{-1} = q_0(x)((x - \alpha_0)q_0(x))^{-1} = ((x - \alpha_0)q_0(x))^{-1}q_0(x) \]
(check this — since \( q_0(x) \) commutes with \( (x - \alpha_0)q_0(x) \) it also commutes with its inverse). Hence \( \alpha_0 \notin \sigma(x) \).

Conversely, let \( \alpha_0 \in \sigma(p(x)) \). Then
\[ p(x) - \alpha_0 = a(x - \lambda_1) \cdots (x - \lambda_n) \]
and at least one \( \lambda_j \in \sigma(x) \) since otherwise the right-hand side would be invertible. But then \( p(\lambda_j) = \alpha_0 \), that is, \( \alpha_0 \in p(\sigma(x)) \). \( \square \)
The second key ingredient for the proof of the spectral theorem is the spectral radius

\[ r(x) := \sup_{\alpha \in \sigma(x)} |\alpha| \]  

(6.18)
of \( x \). Note that by (6.15) we have

\[ r(x) \leq \|x\|. \]  

(6.19)

As our next theorem shows, it is related to the radius of convergence of the Neumann series for the resolvent

\[ (x - \alpha)^{-1} = -\frac{1}{\alpha} \sum_{n=0}^{\infty} \left( \frac{x}{\alpha} \right)^n \]  

(6.20)

encountered in the proof of Theorem 6.3 (which is just the Laurent expansion around infinity).

**Theorem 6.6** (Beurling–Gelfand). The spectral radius satisfies

\[ r(x) = \inf_{n \in \mathbb{N}} \|x^n\|^{1/n} = \lim_{n \to \infty} \|x^n\|^{1/n}. \]  

(6.21)

**Proof.** By spectral mapping we have \( r(x)^n = r(x^n) \leq \|x^n\| \) and hence

\[ r(x) \leq \inf \|x^n\|^{1/n}. \]

Conversely, fix \( \ell \in X^* \), and consider

\[ \ell((x - \alpha)^{-1}) = -\frac{1}{\alpha} \sum_{n=0}^{\infty} \frac{1}{\alpha^n} \ell(x^n). \]  

(6.22)

Then \( \ell((x - \alpha)^{-1}) \) is analytic in \( |\alpha| > r(x) \) and hence (6.22) converges absolutely for \( |\alpha| > r(x) \) by Cauchy’s integral formula for derivatives. Hence for fixed \( \alpha \) with \( |\alpha| > r(x) \), \( \ell(x^n/\alpha^n) \) converges to zero for every \( \ell \in X^* \).

Since every weakly convergent sequence is bounded we have

\[ \frac{\|x^n\|}{|\alpha|^n} \leq C(\alpha) \]

and thus

\[ \limsup_{n \to \infty} \|x^n\|^{1/n} = \limsup_{n \to \infty} C(\alpha)^{1/n} |\alpha| = |\alpha|. \]

Since this holds for every \( |\alpha| > r(x) \) we have

\[ r(x) \leq \inf \|x^n\|^{1/n} \leq \liminf_{n \to \infty} \|x^n\|^{1/n} \leq \limsup_{n \to \infty} \|x^n\|^{1/n} \leq r(x), \]

which finishes the proof. \( \square \)
To end this section let us look at two examples illustrating these ideas.

**Example.** If $X := \mathcal{L}(\mathbb{C}^2)$ and $x := \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ such that $\|x\| = 1$ in our example. The same is true for any nilpotent matrix.

In general $x$ will be called **nilpotent** if $x^n = 0$ for some $n \in \mathbb{N}$ and any nilpotent element will satisfy $r(x) = 0$.

**Example.** Consider the linear Volterra integral operator

$$K(x)(t) := \int_0^t k(t, s)x(s)ds, \quad x \in C([0, 1]),$$

then, using induction, it is not hard to verify (Problem 6.6)

$$|K^n(x)(t)| \leq \frac{\|k\|\|x\|}{n!}t^n\|x\|.$$  

Consequently

$$\|K^n x\|_\infty \leq \frac{\|k\|\|x\|}{n!} \|x\|,$$

that is $\|K^n\| \leq \frac{\|k\|\|x\|}{n!}$, which shows

$$r(K) \leq \lim_{n \to \infty} \frac{\|k\|}{(n!)^{1/n}} = 0.$$

Hence $r(K) = 0$ and for every $\lambda \in \mathbb{C}$ and every $y \in C(I)$ the equation

$$x - \lambda K x = y$$

has a unique solution given by

$$x = (I - \lambda K)^{-1} y = \sum_{n=0}^{\infty} \lambda^n K^n y.$$  

Elements of a Banach algebra with $r(x) = 0$ are called **quasinilpotent**.

In the last two examples we have seen a strict inequality in (6.19). If we regard $r(x)$ as a spectral norm for $x$, then the spectral norm does not control the algebraic norm in such a situation. On the other hand, if we had equality for some $x$, and moreover, this were also true for any polynomial $p(x)$, then spectral mapping would imply that the spectral norm $\sup_{x \in \sigma(x)} |p(x)|$ equals the algebraic norm $\|p(x)\|$ and convergence on one side would imply convergence on the other side. So by taking limits we could get an isometric identification of elements of the form $f(x)$ with functions $f \in C(\sigma(x))$. But this is nothing but the content of the spectral theorem and self-adjointness will be the property which will make all this work.

**Problem 6.1.** Show that the multiplication in a Banach algebra $X$ is continuous: $x_n \to x$ and $y_n \to y$ imply $x_n y_n \to xy$. 

Problem 6.2 (Unitization). Show that if $X$ is a Banach algebra then $\mathbb{C} \oplus X$ is a unital Banach algebra, where we set $\|(\alpha, x)\| = |\alpha| + \|x\|$ and $(\alpha, x)(\beta, y) = (\alpha \beta, \alpha y + \beta x + xy)$.

Problem 6.3. Show $\sigma(x^{-1}) = \sigma(x)^{-1}$ if $x$ is invertible.

Problem 6.4. Suppose $x$ has both a right inverse $y$ (i.e., $xy = e$) and a left inverse $z$ (i.e., $zx = e$). Show that $y = z = x^{-1}$.

Problem 6.5. Suppose $xy$ and $yx$ are both invertible, then so are $x$ and $y$:

$$y^{-1} = (xy)^{-1}x = x(yx)^{-1}, \quad x^{-1} = (yx)^{-1}y = y(xy)^{-1}.$$ (Hint: Previous problem.)


Problem 6.7. Show that $L^1(\mathbb{R}^n)$ with convolution as multiplication is a commutative Banach algebra without identity (Hint: Lemma 10.17).

Problem 6.8. Show the first resolvent identity

$$(x - \alpha)^{-1} - (x - \beta)^{-1} = (\alpha - \beta)(x - \alpha)^{-1}(x - \beta)^{-1}$$

$$= (\alpha - \beta)(x - \beta)^{-1}(x - \alpha)^{-1}, \quad (6.27)$$

for $\alpha, \beta \in \rho(x)$.

Problem 6.9. Show $\sigma(xy) \setminus \{0\} = \sigma(yx) \setminus \{0\}$. (Hint: Find a relation between $(xy - \alpha)^{-1}$ and $(yx - \alpha)^{-1}$.)

6.2. The $C^*$ algebra of operators and the spectral theorem

We begin by recalling that if $\mathcal{H}$ is some Hilbert space, then for every $A \in \mathcal{L}(\mathcal{H})$ we can define its adjoint $A^* \in \mathcal{L}(\mathcal{H})$. Hence the Banach algebra $\mathcal{L}(\mathcal{H})$ has an additional operation in this case which will also give us self-adjointness, a property which has already turned out crucial for the spectral theorem in the case of compact operators. Even though this is not immediately evident, in some sense this additional structure adds the convenient geometric properties of Hilbert spaces to the picture.

A Banach algebra $X$ together with an involution satisfying

$$(x + y)^* = x^* + y^*, \quad (\alpha x)^* = \alpha^*x^*, \quad x^{**} = x, \quad (xy)^* = y^*x^*, \quad (6.28)$$

and

$$\|x\|^2 = \|x^*x\| \quad (6.29)$$

is called a $C^*$ algebra. Any subalgebra (we do not require a subalgebra to contain the identity) which is also closed under involution, is called a $*$-subalgebra.
The condition (6.29) might look a bit artificial at this point. Maybe
a requirement like \( \|x^*\| = \|x\| \) might seem more natural. In fact, at this
point the only justification is that it holds for our guiding example \( \mathcal{L}(\mathfrak{H}) \)
(cf. Lemma 2.13). Furthermore, it is important to emphasize that (6.29)
is a rather strong condition as it implies that the norm is already uniquely
determined by the algebraic structure. More precisely, Lemma 6.7 below
implies that the norm of \( x \) can be computed from the spectral radius of \( x^*x \)
via \( \|x\| = r(x^*x)^{1/2} \). So while there might be several norms which turn \( X \)
into a Banach algebra, there is at most one which will give a \( C^\ast \) algebra.

Note that (6.29) implies \( \|x^2\| = \|x^*x\| \leq \|x\|\|x^*\| \) and hence \( \|x\| \leq \|x^*\| \).
By \( x^{\ast\ast} = x \) this also implies \( \|x^\ast\| \leq \|x^{\ast\ast}\| = \|x\| \) and hence
\[
\|x\| = \|x^\ast\|, \quad \|x\|^2 = \|x^*x\| = \|xx^\ast\|.
\]
(6.30)

Example. The continuous functions \( C(I) \) together with complex conjuga-
tion form a commutative \( C^\ast \) algebra.

Example. The Banach algebra \( \mathcal{L}(\mathfrak{H}) \) is a \( C^\ast \) algebra by Lemma 2.13. The
compact operators \( \mathcal{C}(\mathfrak{H}) \) are a \( \ast \)-subalgebra.

Example. The bounded sequences \( \ell^\infty(\mathbb{N}) \) together with complex conjuga-
tion form a commutative \( C^\ast \) algebra. The set \( c_0(\mathbb{N}) \) of sequences converging
to 0 are a \( \ast \)-subalgebra.

If \( X \) has an identity \( e \), we clearly have \( e^\ast = e, \|e\| = 1, (x^{-1})^\ast = (x^\ast)^{-1} \)
(show this), and
\[
\sigma(x^\ast) = \sigma(x)^\ast.
\]
(6.31)

We will always assume that we have an identity and we note that it is always
possible to add an identity (Problem 6.10).

If \( X \) is a \( C^\ast \) algebra, then \( x \in X \) is called normal if \( x^\ast x = xx^\ast \), self-
adjoint if \( x^\ast = x \), and unitary if \( x^\ast = x^{-1} \). Moreover, \( x \) is called positive if \( x = y^2 \) for some \( y = y^\ast \in X \). Clearly both self-adjoint and unitary elements
are normal and positive elements are self-adjoint. If \( x \) is normal, then so is
any polynomial \( p(x) \) (it will be self-adjoint if \( x \) is and \( p \) is real-valued).

As already pointed out in the previous section, it is crucial to identify
elements for which the spectral radius equals the norm. The key ingredient
will be (6.29) which implies \( \|x^2\| = \|x\|^2 \) if \( x \) is self-adjoint. For unitary
elements we have \( \|x\| = \sqrt{\|x^*x\|} = \sqrt{\|e\|} = 1 \). Moreover, for normal
elements we get

Lemma 6.7. If \( x \in X \) is normal, then \( \|x^2\| = \|x\|^2 \) and \( r(x) = \|x\| \).

Proof. Using (6.29) three times we have
\[
\|x^2\| = \|(x^2)^\ast(x^2)\|^{1/2} = \|(xx^\ast)(xx^\ast)\|^{1/2} = \|x^\ast x\| = \|x\|^2
\]
6.2. The $C^*$ algebra of operators and the spectral theorem

and hence $r(x) = \lim_{k \to \infty} ||x^{2k}||^{1/2k} = ||x||$. \qed

The next result generalizes the fact that self-adjoint operators have only real eigenvalues.

**Lemma 6.8.** If $x$ is self-adjoint, then $\sigma(x) \subseteq \mathbb{R}$. If $x$ is positive, then $\sigma(x) \subseteq [0, \infty)$. 

**Proof.** Suppose $\alpha + i\beta \in \sigma(x)$, $\lambda \in \mathbb{R}$. Then $\alpha + i(\beta + \lambda) \in \sigma(x + i\lambda)$ and $\alpha^2 + (\beta + \lambda)^2 \leq ||x + i\lambda||^2 = ||(x + i\lambda)(x - i\lambda)|| = ||x^2 + \lambda^2|| \leq ||x||^2 + \lambda^2$. Hence $\alpha^2 + \beta^2 + 2\beta\lambda \leq ||x||^2$ which gives a contradiction if we let $|\lambda| \to \infty$ unless $\beta = 0$.

The second claim follows from the first using spectral mapping (Theorem 6.5). \qed

**Example.** If $X := \mathcal{L}(\mathbb{C}^2)$ and $x := \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ then $\sigma(x) = \{0\}$. Hence the converse of the above lemma is not true in general. \diamond

Given $x \in X$ we can consider the $C^*$ algebra $C^*(x)$ (with identity) generated by $x$ (i.e., the smallest closed $*$-subalgebra containing $e$ and $x$). If $x$ is normal we explicitly have

$$C^*(x) = \{p(x,x^*)|p : \mathbb{C}^2 \to \mathbb{C} \text{ polynomial}\}, \quad xx^* = x^*x, \tag{6.32}$$

and, in particular, $C^*(x)$ is commutative (Problem 6.11). In the self-adjoint case this simplifies to

$$C^*(x) := \{p(x)|p : \mathbb{C} \to \mathbb{C} \text{ polynomial}\}, \quad x = x^*. \tag{6.33}$$

Moreover, in this case $C^*(x)$ is isomorphic to $C(\sigma(x))$ (the continuous functions on the spectrum):

**Theorem 6.9** (Spectral theorem). If $X$ is a $C^*$ algebra and $x \in X$ is self-adjoint, then there is an isometric isomorphism $\Phi : C(\sigma(x)) \to C^*(x)$ such that $f(t) = t$ maps to $\Phi(t) = x$ and $f(t) = 1$ maps to $\Phi(1) = e$.

Moreover, for every $f \in C(\sigma(x))$ we have

$$\sigma(f(x)) = f(\sigma(x)), \tag{6.34}$$

where $f(x) = \Phi(f)$.

**Proof.** First of all, $\Phi$ is well defined for polynomials $p$ and given by $\Phi(p) = p(x)$. Moreover, since $p(x)$ is normal spectral mapping implies

$$||p(x)|| = r(p(x)) = \sup_{\alpha \in \sigma(p(x))} |\alpha| = \sup_{\alpha \in \sigma(x)} |p(\alpha)| = ||p||_{\infty}$$

for every polynomial $p$. Hence $\Phi$ is isometric. Next we use that the polynomials are dense in $C(\sigma(x))$. In fact, to see this one can either consider
a compact interval $I$ containing $\sigma(x)$ and use the Tietze extension theorem (Theorem 0.29) to extend $f$ to $I$ and then approximate the extension using polynomials (Theorem 1.3) or use the Stone–Weierstraß theorem (Theorem 0.39). Thus $\Phi$ uniquely extends to a map on all of $C(\sigma(x))$ by Theorem 1.12. By continuity of the norm this extension is again isometric. Similarly, we have $\Phi(fg) = \Phi(f)\Phi(g)$ and $\Phi(f^*) = \Phi(f^*)$ since both relations hold for polynomials.

To show $\sigma(f(x)) = f(\sigma(x))$ fix some $\alpha \in \mathbb{C}$. If $\alpha \notin f(\sigma(x))$, then $g(t) = \frac{1}{f(t) - \alpha} \in C(\sigma(x))$ and $\Phi(g) = (f(x) - \alpha)^{-1} \in X$ shows $\alpha \notin \sigma(f(x))$. Conversely, if $\alpha \notin \sigma(f(x))$ then $g = \Phi^{-1}((f(x) - \alpha)^{-1}) = \frac{1}{f - \alpha}$ is continuous, which shows $\alpha \notin f(\sigma(x))$. □

In particular, this last theorem tells us that we have a functional calculus for self-adjoint operators, that is, if $A \in \mathcal{L}(\mathfrak{H})$ is self-adjoint, then $f(A)$ is well defined for every $f \in C(\sigma(A))$. Specifically, we can compute $f(A)$ by choosing a sequence of polynomials $p_n$ which converge to $f$ uniformly on $\sigma(A)$, then we have $p_n(A) \to f(A)$ in the operator norm. In particular, if $f$ is given by a power series, then $f(A)$ defined via $\Phi$ coincides with $f(A)$ defined via its power series (cf. Problem 1.32).

**Problem 6.10** (Unitization). Show that if $X$ is a non-unital $C^*$ algebra then $\mathbb{C} \oplus X$ is a unital $C^*$ algebra, where we set $\| (\alpha, x) \| := \sup \{ \| \alpha y + x y \| : y \in X, \| y \| \leq 1 \}$, $(\alpha, x)(\beta, y) = (\alpha \beta, \alpha y + \beta x + xy)$ and $(\alpha, x)^* = (\alpha^*, x^*)$. (Hint: It might be helpful to identify $x \in X$ with the operator $L_x : X \to X$, $y \mapsto xy$ in $\mathcal{L}(X)$. Moreover, note $\| L_x \| = \| x \|$.)

**Problem 6.11.** Let $X$ be a $C^*$ algebra and $Y$ a $*$-subalgebra. Show that if $Y$ is commutative, then so is $\overline{Y}$.

**Problem 6.12.** Show that the map $\Phi$ from the spectral theorem is positivity preserving, that is, $f \geq 0$ if and only if $\Phi(f)$ is positive.

**Problem 6.13.** Let $x$ be self-adjoint. Show that the following are equivalent:

1. $\sigma(x) \subseteq [0, \infty)$.
2. $x$ is positive.
3. $\| \lambda - a \| \leq \lambda$ for all $\lambda \geq \| x \|$.
4. $\| \lambda - a \| \leq \lambda$ for one $\lambda \geq \| x \|$.

**Problem 6.14.** Let $A \in \mathcal{L}(\mathfrak{H})$. Show that $A$ is normal if and only if

$$\| Au \| = \| A^* u \|, \quad \forall u \in \mathfrak{H}. \quad (6.35)$$

(Hint: Problem 1.19.)
6.3. Spectral measures

Problem 6.15. Show that the Cayley transform of a self-adjoint element \(x\),
\[
y = (x - i)(x + i)^{-1}
\]
is unitary. Show that \(1 \notin \sigma(y)\) and
\[
x = i(1 + y)(1 - y)^{-1}.
\]

Problem 6.16. Show if \(x\) is unitary then \(\sigma(x) \subseteq \{\alpha \in \mathbb{C} \mid |\alpha| = 1\}\).

Problem 6.17. Suppose \(x\) is self-adjoint. Show that
\[
\| (x - \alpha)^{-1} \| = \frac{1}{\text{dist}(\alpha, \sigma(x))}.
\]

6.3. Spectral measures

The purpose of this section is to derive another formulation of the spectral theorem which is important in quantum mechanics. This reformulation requires familiarity with measure theory and can be skipped as the results will not be needed in the sequel.

Using the Riesz representation theorem we get another formulation in terms of spectral measures:

**Theorem 6.10.** Let \(\mathcal{H}\) be a Hilbert space, and let \(A \in \mathcal{L}(\mathcal{H})\) be self-adjoint. For every \(u, v \in \mathcal{H}\) there is a corresponding complex Borel measure \(\mu_{u,v}\) supported on \(\sigma(A)\) (the spectral measure) such that
\[
\langle u, f(A)v \rangle = \int_{\sigma(A)} f(t) d\mu_{u,v}(t), \quad f \in C(\sigma(A)). \quad (6.36)
\]
We have
\[
\mu_{u,v_1 + v_2} = \mu_{u,v_1} + \mu_{u,v_2}, \quad \mu_{u,\alpha v} = \alpha \mu_{u,v}, \quad \mu_{v,u} = \mu_{u,v}^* \quad (6.37)
\]
and \(|\mu_{u,v}(\sigma(A))| \leq ||u|| ||v||\). Furthermore, \(\mu_u = \mu_{u,u}\) is a positive Borel measure with \(\mu_u(\sigma(A)) = ||u||^2\).

**Proof.** Consider the continuous functions on \(I = [-\|A\|, \|A\|]\) and note that every \(f \in C(I)\) gives rise to some \(f \in C(\sigma(A))\) by restricting its domain. Clearly \(\ell_{u,v}(f) = \langle u, f(A)v \rangle\) is a bounded linear functional and the existence of a corresponding measure \(\mu_{u,v}\) with \(|\mu_{u,v}(I)| = \|\ell_{u,v}\| \leq ||u|| ||v||\) follows from Theorem 12.5. Since \(\ell_{u,v}(f)\) depends only on the value of \(f\) on \(\sigma(A) \subseteq I\), \(\mu_{u,v}\) is supported on \(\sigma(A)\).

Moreover, if \(f \geq 0\) we have \(\ell_u(f) = \langle u, f(A)u \rangle = \langle f(A)^{1/2}u, f(A)^{1/2}u \rangle = ||f(A)^{1/2}u||^2 \geq 0\) and hence \(\ell_u\) is positive and the corresponding measure \(\mu_u\) is positive. The rest follows from the properties of the scalar product. \(\square\)
It is often convenient to regard $\mu_{u,v}$ as a complex measure on $\mathbb{R}$ by using $\mu_{u,v}(\Omega) = \mu_{u,v}(\Omega \cap \sigma(A))$. If we do this, we can also consider $f$ as a function on $\mathbb{R}$. However, note that $f(A)$ depends only on the values of $f$ on $\sigma(A)$! Moreover, it suffices to consider $\mu_u$ since using the polarization identity (1.60) we have

$$\mu_{u,v}(\Omega) = \frac{1}{4}(\mu_{u+v}(\Omega) - \mu_{u-v}(\Omega) + i\mu_{u-iv}(\Omega) - i\mu_{u+iv}(\Omega)).$$  \hspace{1cm} (6.38)

Now the last theorem can be used to define $f(A)$ for every bounded measurable function $f \in B(\sigma(A))$ via Lemma 2.11 and extend the functional calculus from continuous to measurable functions:

**Theorem 6.11** (Spectral theorem). If $\mathcal{H}$ is a Hilbert space and $A \in \mathcal{L}(\mathcal{H})$ is self-adjoint, then there is an homomorphism $\Phi : B(\sigma(A)) \to \mathcal{L}(\mathcal{H})$ given by

$$\langle u, f(A)v \rangle = \int_{\sigma(A)} f(t) d\mu_{u,v}(t), \quad f \in B(\sigma(A)).$$  \hspace{1cm} (6.39)

Moreover, if $f_n(t) \to f(t)$ pointwise and $\sup_n \|f_n\|_\infty$ is bounded, then $f_n(A)u \to f(A)u$ for every $u \in \mathcal{H}$.

**Proof.** The map $\Phi$ is a well-defined linear operator by Lemma 2.11 since we have

$$\left| \int_{\sigma(A)} f(t) d\mu_{u,v}(t) \right| \leq \|f\|_\infty \|\mu_{u,v}\|(\sigma(A)) \leq \|f\|_\infty \|u\|\|v\|$$

and (6.37). Next, observe that $\Phi(f^*) = \Phi(f^*)$ and $\Phi(fg) = \Phi(f)\Phi(g)$ holds at least for continuous functions. To obtain it for arbitrary bounded functions, choose a (bounded) sequence $f_n$ converging to $f$ in $L^2(\sigma(A), d\mu_u)$ and observe

$$\|f_n(A) - f(A)\|_2^2 = \int |f_n(t) - f(t)|^2 d\mu_u(t)$$

(use $\|h(A)u\|^2 = \langle h(A)u, h(A)u \rangle = \langle u, h(A)^* h(A)u \rangle$). Thus $f_n(A)u \to f(A)u$ and for bounded $g$ we also have that $(gf_n)(A)u \to (gf)(A)u$ and $g(A)f_n(A)u \to g(A)f(A)u$. This establishes the case where $f$ is bounded and $g$ is continuous. Similarly, approximating $g$ removes the continuity requirement from $g$.

The last claim follows since $f_n \to f$ in $L^2$ by dominated convergence in this case. \hfill $\square$

Our final aim is to generalize Corollary 3.8 to bounded self-adjoint operators. Since the spectrum of an arbitrary self-adjoint might contain more than just eigenvalues we need to replace the sum by an integral. To this end
we recall the family of Borel sets $\mathfrak{B}(\mathbb{R})$ and begin by defining the spectral projections

$$P_A(\Omega) = \chi_{\Omega}(A), \quad \Omega \in \mathfrak{B}(\mathbb{R}),$$

such that

$$\mu_{u,v}(\Omega) = \langle u, P_A(\Omega)v \rangle.$$  

By $\chi_{\Omega}^2 = \chi_{\Omega}$ and $\chi_{\Omega}^* = \chi_{\Omega}$ they are orthogonal projections, that is $P^2 = P$ and $P^* = P$. Recall that any orthogonal projection $P$ decomposes $\mathcal{H}$ into an orthogonal sum

$$\mathcal{H} = \ker(P) \oplus \text{ran}(P),$$

where $\ker(P) = (I - P)\mathcal{H}$, $\text{ran}(P) = P\mathcal{H}$.

In addition, the spectral projections satisfy

$$P_A(\mathbb{R}) = I, \quad P_A(\bigcup_{n=1}^{\infty} \Omega_n)u = \sum_{n=1}^{\infty} P_A(\Omega_n)u, \quad \Omega_n \cap \Omega_m = \emptyset, m \neq n, \quad (6.43)$$

for every $u \in \mathcal{H}$. Here the dot inside the union just emphasizes that the sets are mutually disjoint. Such a family of projections is called a projection-valued measure. Indeed the first claim follows since $\chi_{\mathbb{R}} = 1$ and by $\chi_{\Omega_1 \cup \Omega_2} = \chi_{\Omega_1} + \chi_{\Omega_2}$ if $\Omega_1 \cap \Omega_2 = \emptyset$ the second claim follows at least for finite unions. The case of countable unions follows from the last part of the previous theorem since $\sum_{n=1}^{N} \chi_{\Omega_n} = \chi_{\bigcup_{n=1}^{N} \Omega_n} \to \chi_{\bigcup_{n=1}^{\infty} \Omega_n}$ pointwise (note that the limit will not be uniform unless the $\Omega_n$ are eventually empty and hence there is no chance that this series will converge in the operator norm). Moreover, since all spectral measures are supported on $\sigma(A)$ the same is true for $P_A$ in the sense that

$$P_A(\sigma(A)) = I.$$

I also remark that in this connection the corresponding distribution function

$$P_A(t) := P_A((-\infty, t])$$

is called a resolution of the identity.

Using our projection-valued measure we can define an operator-valued integral as follows: For every simple function $f = \sum_{j=1}^{n} \alpha_j \chi_{\Omega_j}$ (where $\Omega_j = f^{-1}(\alpha_j)$), we set

$$\int_{\mathbb{R}} f(t) dP_A(t) := \sum_{j=1}^{n} \alpha_j P_A(\Omega_j).$$

By (6.41) we conclude that this definition agrees with $f(A)$ from Theorem 6.11:

$$\int_{\mathbb{R}} f(t) dP_A(t) = f(A).$$

Extending this integral to functions from $B(\sigma(A))$ by approximating such functions with simple functions we get an alternative way of defining $f(A)$.
for such functions. This can in fact be done by just using the definition of a projection-valued measure and hence there is a one-to-one correspondence between projection-valued measures (with bounded support) and (bounded) self-adjoint operators such that

\[ A = \int t \, dP_A(t). \]  

(6.48)

If \( P_A(\{\alpha\}) \neq 0 \), then \( \alpha \) is an eigenvalue and \( \text{Ran}(P_A(\{\alpha\})) \) is the corresponding eigenspace (Problem 6.19). The fact that eigenspaces to different eigenvalues are orthogonal now generalizes to

**Lemma 6.12.** Suppose \( \Omega_1 \cap \Omega_2 = \emptyset \). Then

\[ \text{Ran}(P_A(\Omega_1)) \perp \text{Ran}(P_A(\Omega_2)). \]  

(6.49)

**Proof.** Clearly \( \chi_{\Omega_1} \chi_{\Omega_2} = \chi_{\Omega_1 \cap \Omega_2} \) and hence

\[ P_A(\Omega_1)P_A(\Omega_2) = P_A(\Omega_1 \cap \Omega_2). \]

Now if \( \Omega_1 \cap \Omega_2 = \emptyset \), then

\[ \langle P_A(\Omega_1)u, P_A(\Omega_2)v \rangle = \langle u, P_A(\Omega_1)P_A(\Omega_2)v \rangle = \langle u, P_A(\emptyset)v \rangle = 0, \]

which shows that the ranges are orthogonal to each other. \( \square \)

**Example.** Let \( A \in L(\mathbb{C}^n) \) be some symmetric matrix and let \( \alpha_1, \ldots, \alpha_m \) be its (distinct) eigenvalues. Then

\[ A = \sum_{j=1}^{m} \alpha_j P_A(\{\alpha_j\}), \]

where \( P_A(\{\alpha_j\}) \) is the projection onto the eigenspace \( \text{Ker}(A - \alpha_j) \) corresponding to the eigenvalue \( \alpha_j \) by Problem 6.19. In fact, using that \( P_A \) is supported on the spectrum, \( P_A(\sigma(A)) = I \), we see

\[ P(\Omega) = P_A(\sigma(A))P(\Omega) = P(\sigma(A) \cap \Omega) = \sum_{\alpha_j \in \Omega} P_A(\{\alpha_j\}). \]

Hence using that any \( f \in B(\sigma(A)) \) is given as a simple function \( f = \sum_{j=1}^{m} f(\alpha_j)\chi_{\{\alpha_j\}} \) we obtain

\[ f(A) = \int f(t) \, dP_A(t) = \sum_{j=1}^{m} f(\alpha_j)P_A(\{\alpha_j\}). \]

In particular, for \( f(t) = t \) we recover the above representation for \( A \). \( \diamond \)

**Example.** Let \( A \in L(\mathbb{C}^n) \) be self-adjoint and let \( \alpha \) be an eigenvalue. Let \( \tilde{P} = P_A(\{\alpha\}) \) be the projection onto the corresponding eigenspace and consider the restriction \( \tilde{A} = A|_{\tilde{H}} \) onto the orthogonal complement of this eigenspace \( \tilde{H} = (1 - \tilde{P})\tilde{H} \). Then by Lemma 6.12 we have \( \mu_{u,v}(\{\alpha\}) = 0 \) for
6.4. The Gelfand representation theorem

In this section we look at an alternative approach to the spectral theorem by trying to find a canonical representation for a Banach algebra. The idea is as follows: Given the Banach algebra $C[a,b]$ we have a one-to-one correspondence between points $x_0 \in [a,b]$ and point evaluations $m_{x_0}(f) = f(x_0)$. These point evaluations are linear functionals which at the same time preserve multiplication. In fact, we will see that these are the only (nontrivial) multiplicative functionals and hence we also have a one-to-one correspondence between points in $[a,b]$ and multiplicative functionals. Now $m_{x_0}(f) = f(x_0)$ says that the action of a multiplicative functional on a function is the same as the action of the function on a point. So for a general algebra $X$ we can try to go the other way: Consider the multiplicative functionals $m$ as points and the elements $x \in X$ as functions acting on these points (the value of this function being $m(x)$). This gives a map, the Gelfand representation, from $X$ into an algebra of functions.

A nonzero algebra homeomorphism $m : X \to \mathbb{C}$ will be called a multiplicative linear functional or character:

$$m(xy) = m(x)m(y), \quad m(e) = 1. \quad (6.50)$$

Note that the last equation comes for free from multiplicativity since $m$ is nontrivial. Moreover, there is no need to require that $m$ is continuous as this will also follow automatically (cf. Lemma 6.14 below).
As we will see, they are closely related to ideals, that is linear subspaces $I$ of $X$ for which $a \in I$, $x \in X$ implies $ax \in I$ and $xa \in I$. An ideal is called proper if it is not equal to $X$ and it is called maximal if it is not contained in any other proper ideal.

**Example.** Let $X := C([a,b])$ be the continuous functions over some compact interval. Then for fixed $x_0 \in [a,b]$, the linear functional $m_{x_0}(f) := f(x_0)$ is multiplicative. Moreover, its kernel $\text{Ker}(m_{x_0}) = \{ f \in C([a,b]) | f(x_0) = 0 \}$ is a maximal ideal (we will prove this in more generality below).

**Example.** Let $X$ be a Banach space. Then the compact operators are a closed ideal in $L(X)$ (cf. Theorem 3.1).

We first collect a few elementary properties of ideals.

**Lemma 6.13.** Let $X$ be a unital Banach algebra.

(i) A proper ideal can never contain an invertible element.

(ii) If $X$ is commutative every non-invertible element is contained in a proper ideal.

(iii) The closure of a (proper) ideal is again a (proper) ideal.

(iv) Maximal ideals are closed.

(v) Every proper ideal is contained in a maximal ideal.

**Proof.** (i). If $x \in I$ is invertible then $y = x(x^{-1}y) \in I$ shows $I = X$. (ii). Consider the ideal $xX = \{ xy | y \in X \}$. Then $xX = X$ if and only if there is some $y \in X$ with $xy = e$, that is, $y = x^{-1}$. (iii) and (iv). That the closure of an ideal is again an ideal follows from continuity of the product. Indeed, for $a \in \mathcal{T}$ choose a sequence $a_n \in I$ converging to $a$. Then $xa_n \in I \to xa \in \mathcal{T}$ as well as $a_nx \in I \to ax \in \mathcal{T}$. Moreover, note that by Lemma 6.1 all elements in the ball $B_1(e)$ are invertible and hence every proper ideal must be contained in the closed set $X \setminus B_1(e)$. So the closure of a proper ideal is proper and any maximal ideal must be closed. (v). To see that every ideal $I$ is contained in a maximal ideal consider the family of proper ideals containing $I$ ordered by inclusion. Then, since any union of a chain of proper ideals is again a proper ideal (that the union is again an ideal is straightforward, to see that it is proper note that it does not contain $B_1(e)$). Consequently Zorn’s lemma implies existence of a maximal element.

Note that if $I$ is a closed ideal, then the quotient space $X/I$ (cf. Lemma 1.14) is again a Banach algebra if we define

$$[x][y] = [xy].$$

Indeed $(x + I)(y + I) = xy + I$ and hence the multiplication is well-defined and inherits the distributive and associative laws from $X$. Also $[e]$ is an
identity. Finally,
\[
\left\| [xy] \right\| = \inf_{a \in I} \| xy + a \| = \inf_{b,c \in I} \left\| (x + b)(y + c) \right\| \leq \inf_{b \in I} \| x + b \| \inf_{c \in I} \| y + c \| = \left\| [x] \right\| \left\| [y] \right\|. 
\]
(6.52)

In particular, the projection map \( \pi : X \to X/I \) is a Banach algebra homomorphism.

**Example.** Consider the Banach algebra \( \mathcal{L}(X) \) together with the ideal of compact operators \( \mathcal{C}(X) \). Then the Banach algebra \( \mathcal{L}(X)/\mathcal{C}(X) \) is known as the **Calkin algebra**. Atkinson’s theorem (Theorem 6.28) says that the invertible elements in the Calkin algebra are precisely the images of the Fredholm operators. \( \diamond \)

**Lemma 6.14.** Let \( X \) be a unital Banach algebra and \( m \) a character. Then \( \ker(m) \) is a maximal ideal and \( m \) is continuous with \( \|m\| = m(e) = 1 \).

**Proof.** It is straightforward to check that \( \ker(m) \) is an ideal. Moreover, every \( x \) can be written as
\[
x = m(x)e + y, \quad y \in \ker(m).
\]
Let \( I \) be an ideal containing \( \ker(m) \). If there is some \( x \in I \setminus \ker(m) \) then \( m(x)^{-1}x = e + m(x)^{-1}y \in I \) and hence also \( e = (e + m(x)^{-1}y) - m(x)^{-1}y \in I \). Thus \( \ker(m) \) is maximal. Since maximal ideals are closed by the previous lemma, we conclude that \( m \) is continuous by Problem 1.33.

Clearly \( \|m\| \geq m(e) = 1 \). Conversely, suppose we can find some \( x \in X \) with \( \|x\| < 1 \) and \( m(x) = 1 \). Consequently \( \|x^n\| \leq \|x\|^n \to 0 \) contradicting \( m(x^n) = m(x)^n = 1 \).

In a commutative algebra the other direction is also true.

**Lemma 6.15.** In a commutative unital Banach algebra the characters and maximal ideals are in one-to-one correspondence.

**Proof.** We have already seen that for a character \( m \) there is corresponding maximal ideal \( \ker(m) \). Conversely, let \( I \) be a maximal ideal and consider the projection \( \pi : X \to X/I \). We first claim that every nontrivial element in \( X/I \) is invertible. To this end suppose \( [x_0] \neq [0] \) were not invertible. Then \( J = [x_0]X/I \) is a proper ideal (if it would contain the identity, \( X/I \) would contain an inverse of \( [x_0] \)). Moreover, \( I' = \{ y \in X | [y] \in J \} \) is a proper ideal of \( X \) (since \( e \in I' \) would imply \( [e] \in J \) which contains \( I \) since \( [x] = [0] \in J \) for \( x \in I \) but is strictly larger as \( x_0 \in I' \setminus I \). This contradicts maximality and hence by the Gelfand–Mazur theorem (Theorem 6.4), every element of \( X/I \) is of the form \( \alpha[e] \). If \( h : X/I \to \mathbb{C}, h(\alpha[e]) \mapsto \alpha \) is the corresponding algebra isomorphism, then \( m = h \circ \pi \) is a character with \( \ker(m) = I \). \( \square \)
Now we continue with the following observation: For fixed \( x \in X \) we get a map \( X^* \to \mathbb{C}, \ell \mapsto \ell(x) \). Moreover, if we equip \( X^* \) with the weak-* topology then this map will be continuous (by the very definition of the weak-* topology). So we have a map \( X \to C(X^*) \) and restricting this map to the set of all characters \( \mathcal{M} \subseteq X^* \) (equipped with the relative topology of the weak-* topology) it is known as the **Gelfand transform**:

\[
\Gamma : X \to C(\mathcal{M}), \quad x \mapsto \hat{x}(m) = m(x).
\] (6.53)

**Theorem 6.16** (Gelfand representation theorem). Let \( X \) be a unital Banach algebra. Then the set of all characters \( \mathcal{M} \subseteq X^* \) is a compact Hausdorff space with respect to the weak-* topology and the Gelfand transform is a continuous algebra homomorphism with \( \hat{e} = 1 \).

Moreover, \( \hat{x}(\mathcal{M}) \subseteq \sigma(x) \) and hence \( \|\hat{x}\|_\infty \leq r(x) \leq \|x\| \) where \( r(x) \) is the spectral radius of \( x \). If \( X \) is commutative then \( \hat{x}(\mathcal{M}) = \sigma(x) \) and hence \( \|\hat{x}\|_\infty = r(x) \).

**Proof.** As pointed out before, for fixed \( x, y \in X \) the map \( X^* \to \mathbb{C}^3, \ell \mapsto (\ell(x), \ell(y), \ell(xy)) \) is continuous and so is the map \( X^* \to \mathbb{C}, \ell \mapsto \ell(x)\ell(y) - \ell(xy) \) as a composition of continuous maps. Hence the kernel of this map \( M_{x,y} = \{\ell \in X^* | \ell(x)\ell(y) = \ell(xy)\} \) is weak-* closed and so is \( \mathcal{M} = M_0 \cap \bigcap_{x,y \in X} M_{x,y} \) where \( M_0 = \{\ell \in X^* | \ell(e) = 1\} \). So \( \mathcal{M} \) is a weak-* closed subset of the unit ball in \( X^* \) and the first claim follows form the Banach–Alaoglu theorem (Theorem 5.10).

Next \( (x+y)^\wedge(m) = m(x+y) = m(x)+m(y) = \hat{x}(m)+\hat{y}(m) \), \( (xy)^\wedge(m) = m(xy) = m(x)m(y) = \hat{x}(m)\hat{y}(m) \), and \( \hat{e}(m) = m(e) = 1 \) shows that the Gelfand transform is an algebra homomorphism.

Moreover, if \( m(x) = \alpha \) then \( x - \alpha \in \text{Ker}(m) \) implying that \( x - \alpha \) is not invertible (as maximal ideals cannot contain invertible elements), that is \( \alpha \in \sigma(x) \). Conversely, if \( X \) is commutative and \( \alpha \in \sigma(x) \), then \( x - \alpha \) is not invertible and hence contained in some maximal ideal, which in turn is the kernel of some character \( m \). Whence \( m(x - \alpha) = 0 \), that is \( m(x) = \alpha \) for some \( m \).

Of course this raises the question about injectivity or surjectivity of the Gelfand transform. Clearly

\[
x \in \text{Ker}(\Gamma) \iff x \in \bigcap_{m \in \mathcal{M}} \text{Ker}(m)
\] (6.54)

and it can only be injective if \( X \) is commutative. In this case

\[
x \in \text{Ker}(\Gamma) \iff x \in \text{Rad}(X) := \bigcap_{I \text{ maximal ideal}} I
\] (6.55)
where \( \text{Rad}(X) \) is known as the **Jacobson radical** of \( X \) and a Banach algebra is called **semi-simple** if the Jacobson radical is zero. So to put this result to use one needs to understand the set of characters, or equivalently, the set of maximal ideals. Two examples where this can be done are given below. The first one is not very surprising.

**Example.** If we start with a compact Hausdorff space \( K \) and consider \( C(K) \) we get nothing new. Indeed, first of all notice that the map \( K \to \mathcal{M}, \, x_0 \mapsto m_{x_0} \) which assigns each \( x_0 \) the corresponding point evaluation \( m_{x_0}(f) = f(x_0) \) is injective and continuous. Hence, by compactness of \( K \), it will be a homeomorphism once we establish surjectivity (Corollary 0.17). To this end we will show that all maximal ideals are of the form \( I = \ker(m_{x_0}) \) for some \( x_0 \in K \). So let \( I \) be an ideal and suppose there is no point where all functions vanish. Then for every \( x \in K \) there is a ball \( B_{r(x)}(x) \) and a function \( f_x \in C(K) \) such that \( |f_x(y)| \geq 1 \) for \( y \in B_{r(x)}(x) \). By compactness finitely many of these balls will cover \( K \). Now consider \( f = \sum_j f_{x_j}^* f_{x_j} \in I \). Then \( f \geq 1 \) and hence \( f \) is invertible, that is \( I = C(K) \). Thus maximal ideals are of the form \( I_{x_0} = \{ f \in C(K) | f(x_0) = 0 \} \) which are precisely the kernels of the characters \( m_{x_0}(f) = f(x_0) \). Thus \( \mathcal{M} \simeq K \) as well as \( \hat{f} \simeq f \).

**Example.** Consider the Wiener algebra \( A \) of all periodic continuous functions which have an absolutely convergent Fourier series. As in the previous example it suffices to show that all maximal ideals are of the form \( I_{x_0} = \{ f \in A | f(x_0) = 0 \} \). To see this set \( e_k(x) = e^{ikx} \) and note \( \|e_k\|_A = 1 \). Hence for every character \( m(e_k) = m(e_1)^k \) and \( |m(e_k)| \leq 1 \). Since the last claim holds for both positive and negative \( k \), we conclude \( |m(e_k)| = 1 \) and thus there is some \( x_0 \in [-\pi, \pi] \) with \( m(e_k) = e^{ikx_0} \). Consequently \( m(f) = f(x_0) \) and point evaluations are the only characters. Equivalently, every maximal ideal is of the form \( \ker(m_{x_0}) = I_{x_0} \).

So, as in the previous example, \( \mathcal{M} \simeq [-\pi, \pi] \) (with \( -\pi \) and \( \pi \) identified) as well hat \( \hat{f} \simeq f \). Moreover, the Gelfand transform is injective but not surjective since there are continuous functions whose Fourier series are not absolutely convergent. Incidentally this also shows that the Wiener algebra is no \( C^* \) algebra (despite the fact that we have a natural conjugation which satisfies \( \|f^*\|_A = \|f\|_A \) — this again underlines the special role of (6.29)) as the Gelfand–Naimark theorem below will show that the Gelfand transform is bijective for commutative \( C^* \) algebras.

Since \( 0 \notin \sigma(x) \) implies that \( x \) is invertible the Gelfand representation theorem also contains a useful criterion for invertibility.

**Corollary 6.17.** In a commutative unital Banach algebra an element \( x \) is invertible if and only if \( m(x) \neq 0 \) for all characters \( m \).
And applying this to the last example we get the following famous theorem of Wiener:

**Theorem 6.18 (Wiener).** Suppose \( f \in \mathcal{C}_{\text{per}}[-\pi,\pi] \) has an absolutely convergent Fourier series and does not vanish on \([-\pi,\pi]\). Then the function \( \frac{1}{f} \) also has an absolutely convergent Fourier series.

If we turn to a commutative \( \mathcal{C}^* \) algebra the situation further simplifies. First of all note that characters respect the additional operation automatically.

**Lemma 6.19.** If \( X \) is a unital \( \mathcal{C}^* \) algebra, then every character satisfies \( \hat{m}(x^*) = \hat{m}(x)^* \). In particular, the Gelfand transform is a continuous \(*\)-algebra homomorphism with \( \hat{\varepsilon} = 1 \) in this case.

**Proof.** If \( x \) is self-adjoint then \( \sigma(x) \subseteq \mathbb{R} \) (Lemma 6.8) and hence \( m(x) \in \mathbb{R} \) by the Gelfand representation theorem. Now for general \( x \) we can write \( x = a + ib \) with \( a = \frac{x + x^*}{2} \) and \( b = \frac{x - x^*}{2i} \) self-adjoint implying

\[
m(x^*) = m(a - ib) = m(a) - im(b) = (m(a) + im(b))^* = m(x)^*.
\]

Consequently the Gelfand transform preserves the involution: \( (x^*)^\wedge(m) = m(x^*) = m(x)^* = \hat{x}^*(m) \). \( \square \)

**Theorem 6.20 (Gelfand–Naimark).** Suppose \( X \) is a unital commutative \( \mathcal{C}^* \) algebra. Then the Gelfand transform is an isometric isomorphism between \( \mathcal{C}^* \) algebras.

**Proof.** Since in a commutative \( \mathcal{C}^* \) algebra every element is normal, Lemma 6.7 implies that the Gelfand transform is isometric. Moreover, by the previous lemma the image of \( X \) under the Gelfand transform is a closed \(*\)-subalgebra which contains \( \hat{\varepsilon} \equiv 1 \) and separates points (if \( \hat{x}(m_1) = \hat{x}(m_2) \) for all \( x \in X \) we have \( m_1 = m_2 \)). Hence it must be all of \( \mathcal{C}(\mathcal{M}) \) by the Stone–Weierstraß theorem (Theorem 0.39). \( \square \)

The first moral from this theorem is that from an abstract point of view there is only one commutative \( \mathcal{C}^* \) algebra, namely \( \mathcal{C}(K) \) with \( K \) some compact Hausdorff space. Moreover, it also very much reassembles the spectral theorem and in fact, we can derive the spectral theorem by applying it to \( \mathcal{C}^*(x) \), the \( \mathcal{C}^* \) algebra generated by \( x \) (cf. (6.32)). This will even give us the more general version for normal elements. As a preparation we show that it makes no difference whether we compute the spectrum in \( X \) or in \( \mathcal{C}^*(x) \).

**Lemma 6.21 (Spectral permanence).** Let \( X \) be a \( \mathcal{C}^* \) algebra and \( Y \subseteq X \) a closed \(*\)-subalgebra containing the identity. Then \( \sigma(y) = \sigma_Y(y) \) for every \( y \in Y \), where \( \sigma_Y(y) \) denotes the spectrum computed in \( Y \).
6.4. The Gelfand representation theorem

**Proof.** Clearly we have $\sigma(y) \subseteq \sigma_Y(y)$ and it remains to establish the reverse inclusion. If $(y-\alpha)$ has an inverse in $X$, then the same is true for $(y-\alpha)^*(y-\alpha)$ and $(y-\alpha)(y-\alpha)^*$. But the last two operators are self-adjoint and hence have real spectrum in $Y$. Thus $((y-\alpha)^*(y-\alpha) + \frac{1}{n})^{-1} \in Y$ and letting $n \to \infty$ shows $((y-\alpha)^*(y-\alpha))^{-1} \in Y$ since taking the inverse is continuous and $Y$ is closed. Similarly $((y-\alpha)(y-\alpha)^*)^{-1} \in Y$ and whence $(y-\alpha)^{-1} \in Y$ by Problem 6.5.

Now we can show

**Theorem 6.22** (Spectral theorem). *If $X$ is a $C^*$ algebra and $x$ is normal, then there is an isometric isomorphism $\Phi : C(\sigma(x)) \to C^*(x)$ such that $f(t) = t$ maps to $\Phi(t) = x$ and $f(t) = 1$ maps to $\Phi(1) = e$. Moreover, for every $f \in C(\sigma(x))$ we have

$$
\sigma(f(x)) = f(\sigma(x)),
$$

(6.56)

where $f(x) = \Phi(f)$.*

**Proof.** Given a normal element $x \in X$ we want to apply the Gelfand–Naimark theorem in $C^*(x)$. By our lemma we have $\sigma(x) = \sigma_{C^*(x)}(x)$. We first show that we can identify $\mathcal{M}$ with $\sigma(x)$. By the Gelfand representation theorem (applied in $C^*(x)$), $\hat{x} : \mathcal{M} \to \sigma(x)$ is continuous and surjective. Moreover, if for given $m_1, m_2 \in \mathcal{M}$ we have $\hat{x}(m_1) = m_1(x) = m_2(x) = \hat{x}(m_2)$ then

$$
m_1(p(x, x^*)) = p(m_1(x), m_1(x^*)) = p(m_2(x), m_2(x^*)) = m_2(p(x, x^*))
$$

for any polynomial $p : \mathbb{C}^2 \to \mathbb{C}$ and hence $m_1(y) = m_2(y)$ for every $y \in C^*(x)$ implying $m_1 = m_2$. Thus $\hat{x}$ is injective and hence a homeomorphism as $\mathcal{M}$ is compact. Thus we have an isometric isomorphism

$$
\Psi : C(\sigma(x)) \to C(\mathcal{M}), \quad f \mapsto f \circ \hat{x},
$$

and the isometric isomorphism we are looking for is $\Phi = \Gamma^{-1} \circ \Psi$. Finally, $\sigma(f(x)) = \sigma_{C^*(x)}(\Phi(f)) = \sigma_{C(\sigma(x))}(f) = f(\sigma(x))$.

**Example.** Let $X$ be a $C^*$ algebra and $x \in X$ normal. By the spectral theorem $C^*(x)$ is isomorphic to $C(\sigma(x))$. Hence every $y \in C^*(x)$ can be written as $y = f(x)$ for some $f \in C(\sigma(x))$ and every character is of the form $m(y) = m(f(x)) = f(\alpha)$ for some $\alpha \in \sigma(x)$.

**Problem 6.20.** Show that $C^1[0,1]$ is a Banach algebra. What are the characters? Is it semi-simple?

**Problem 6.21.** Consider the subalgebra of the Wiener algebra consisting of all functions whose negative Fourier coefficients vanish. What are the
characters? (Hint: Observe that these functions can be identified with holomorphic functions inside the unit disc with summable Taylor coefficients via 
\[ f(z) = \sum_{k=0}^{\infty} \hat{f}_k z^k \] known as the **Hardy space** \( H^1 \) of the disc.)

### 6.5. Fredholm operators

In this section we want to investigate solvability of the equation

\[ Ax = y \quad (6.57) \]

for \( A \in \mathcal{L}(X, Y) \) given \( y \in Y \). Clearly there exists a solution if \( y \in \text{Ran}(A) \) and this solution is unique if \( \text{Ker}(A) = \{0\} \). Hence these subspaces play a crucial role. Moreover, if the underlying Banach spaces are finite dimensional, the kernel has a complement \( X = \text{Ker}(A) \oplus X_0 \) and after factoring out the kernel this complement is isomorphic to the range of \( A \). As a consequence, the dimensions of these spaces are connected by the famous **rank-nullity theorem**

\[ \dim \text{Ker}(A) + \dim \text{Ran}(A) = \dim X \quad (6.58) \]

from linear algebra. In our infinite dimensional setting (apart from the technical difficulties that the kernel might not be complemented and the range might not be closed) this formula does not contain much information, but if we rewrite it in terms of the index,

\[ \text{ind}(A) := \dim \text{Ker}(A) - \dim \text{Coker}(A) = \dim(X) - \dim(Y), \quad (6.59) \]

at least the left-hand side will be finite if we assume both \( \text{Ker}(A) \) and \( \text{Coker}(A) \) to be finite dimensional. One of the most useful consequences of the rank-nullity theorem is that in the case \( X = Y \) the index will vanish and hence uniqueness of solutions for \( Ax = y \) will automatically give you existence for free (and vice versa). Indeed, for equations of the form \( x + Kx = y \) with \( K \) compact originally studied by Fredholm this is still true and this is the famous Fredholm alternative. It took a while until Noether found an example of singular integral equations which have a nonzero index and started investigating the general case.

We first note that in this case \( \text{Ran}(A) \) will be automatically closed.

**Lemma 6.23.** Suppose \( A \in \mathcal{L}(X, Y) \) with finite dimensional cokernel. Then \( \text{Ran}(A) \) is closed.

**Proof.** First of all note that the induced map \( \tilde{A} : X/\text{Ker}(A) \to Y \) is injective (Problem 1.38). Moreover, the assumption that the cokernel is finite says that there is a finite subspace \( Y_0 \subset Y \) such that \( Y = Y_0 \oplus \text{Ran}(A) \). Then

\[ \tilde{A} : X \oplus Y_0 \to Y, \quad \tilde{A}(x, y) = \tilde{Ax} + y \]
is bijective and hence a homeomorphism by Theorem 4.6. Since $X$ is a closed subspace of $X \oplus Y_0$ we see that $\text{Ran}(A) = \hat{A}(X)$ is closed in $Y$. □

Hence we call an operator $A \in \mathcal{L}(X, Y)$ a **Fredholm operator** (also Noether operator) if both its kernel and cokernel are finite dimensional. In this case we define its **index** as

$$\text{ind}(A) := \dim \text{Ker}(A) - \dim \text{Coker}(A).$$

(6.60)

The set of Fredholm operators will be denoted by $\mathcal{F}(X, Y)$. An immediate consequence of Theorem 4.26 is:

**Theorem 6.24** (Riesz). A bounded operator $A$ is Fredholm if and only if $A'$ is and in this case

$$\text{ind}(A') = -\text{ind}(A).$$

(6.61)

Note that by Problem 4.23 we have

$$\text{ind}(A) = \dim \text{Ker}(A) - \dim \text{Ker}(A') = \dim \text{Coker}(A') - \dim \text{Coker}(A)$$

(6.62)

since for a finite dimensional space the dual space has the same dimension.

**Example.** The right shift operator $S$ in $X = Y = \ell^p(\mathbb{N})$, $1 \leq p < 1$ is Fredholm. In fact, recall that $S'$ is the right shift and $\text{Ker}(S) = \{0\}$, $\text{Ker}(S') = \text{span}\{\delta^1\}$. In particular, $\text{ind}(S) = 1$ and $\text{ind}(S') = -1$. □

In the case of Hilbert spaces $\text{Ran}(A)$ closed implies $\mathcal{H} = \text{Ran}(A) \oplus \text{Ran}(A)^\perp$ and thus $\text{Coker}(A) \cong \text{Ran}(A)^\perp$. Hence an operator is Fredholm if $\text{Ran}(A)$ is closed and $\text{Ker}(A)$ and $\text{Ran}(A)^\perp$ are both finite dimensional. In this case

$$\text{ind}(A) = \dim \text{Ker}(A) - \dim \text{Ran}(A)^\perp$$

(6.63)

and $\text{ind}(A^*) = -\text{ind}(A)$ as is immediate from (2.28).

**Example.** Suppose $\mathcal{H}$ is a Hilbert space and $A = A^*$ is a self-adjoint Fredholm operator, then (2.28) shows that $\text{ind}(A) = 0$. In particular, a self-adjoint operator is Fredholm if $\dim \text{Ker}(A) < \infty$ and $\text{Ran}(A)$ is closed. For example, according to the example on page 200, $A - \lambda$ is Fredholm if $\lambda$ is an eigenvalue of finite multiplicity (in fact, inspecting this example shows that the converse is also true).

It is however important to notice that $\text{Ran}(A)^\perp$ finite dimensional does not imply $\text{Ran}(A)$ closed! For example consider $(Ax)_n = \frac{1}{n}x_n$ in $\ell^2(\mathbb{N})$ whose range is dense but not closed. □

Another useful formula concerns the product of two Fredholm operators. For its proof it will be convenient to use the notion of an exact sequence: Let $X_j$ be Banach spaces. A sequence of operators $A_j \in \mathcal{L}(X_j, X_{j+1})$

$$X_1 \xrightarrow{A_1} X_2 \xrightarrow{A_2} X_3 \cdots X_n \xrightarrow{A_n} X_{n+1}$$


is said to be **exact** if \( \text{Ran}(A_j) = \text{Ker}(A_{j+1}) \) for \( 0 \leq j < n \). We will also need the following two easily checked facts: If \( X_{j-1} \) and \( X_{j+1} \) are finite dimensional, so is \( X_j \) (Problem 6.22) and if the sequence of finite dimensional spaces starts with \( X_0 = \{0\} \) and ends with \( X_{n+1} = \{0\} \), then the alternating sum over the dimensions vanishes (Problem 6.23).

**Lemma 6.25.** Suppose \( A \in \mathcal{L}(X,Y) \), \( B \in \mathcal{L}(Y,Z) \). If two of the operators \( A, B, AB \) are Fredholm, so is the third and we have

\[
\text{ind}(AB) = \text{ind}(A) + \text{ind}(B). \tag{6.64}
\]

**Proof.** It is straightforward to check that the sequence

\[
0 \longrightarrow \text{Ker}(A) \longrightarrow \text{Ker}(BA) \xrightarrow{A} \text{Ker}(B) \longrightarrow \text{Coker}(A) \xrightarrow{B} \text{Coker}(BA) \longrightarrow \text{Coker}(B) \longrightarrow 0
\]

is exact. Here the maps which are not explicitly stated are canonical inclusions/projections. Hence by Problem 6.22, if two operators are Fredholm, so is the third. Moreover, the formula for the index follows from Problem 6.23. \( \square \)

Next we want to look a bit further into the structure of Fredholm operators. First of all, since \( \text{Ker}(A) \) is finite dimensional it is complemented (Problem 4.21), that is, there exists a closed subspace \( X_0 \subseteq X \) such that \( X = \text{Ker}(A) + X_0 \) and a corresponding projection \( P \in \mathcal{L}(X) \) with \( \text{Ran}(P) = \text{Ker}(A) \). Similarly, \( \text{Ran}(A) \) is complemented (Problem 1.39) and there exists a closed subspace \( Y_0 \subseteq Y \) such that \( Y = Y_0 + \text{Ran}(A) \) and a corresponding projection \( Q \in \mathcal{L}(Y) \) with \( \text{Ran}(Q) = Y_0 \). With respect to the decomposition \( \text{Ker}(A) \oplus X_0 \to Y_0 \oplus \text{Ran}(A) \) our Fredholm operator is given by

\[
A = \begin{pmatrix} 0 & 0 \\ 0 & A_0 \end{pmatrix}, \tag{6.65}
\]

where \( A_0 \) is the restriction of \( A \) to \( X_0 \to \text{Ran}(A) \). By construction \( A_0 \) is bijective and hence a homeomorphism (Theorem 4.6). Defining

\[
B := \begin{pmatrix} 0 & 0 \\ 0 & A_0^{-1} \end{pmatrix} \tag{6.66}
\]

we get

\[
AB = \mathbb{I} - Q, \quad BA = \mathbb{I} - P \tag{6.67}
\]

and hence \( A \) is invertible up to finite rank operators. Now we are ready for showing that the index is stable under small perturbations.

**Theorem 6.26** (Dieudonné). The set of Fredholm operators \( \mathcal{F}(X,Y) \) is open in \( \mathcal{L}(X,Y) \) and the index is locally constant.
Proof. Let \( C \in \mathcal{L}(X,Y) \) and write it as
\[
C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}
\]
with respect to the above splitting. Then if \( \|C_{22}\| < \|A_0^{-1}\|^{-1} \) we have that \( A_0 + C_{22} \) is still invertible (Problem 1.31). Now introduce
\[
D_1 = \begin{pmatrix} \mathbb{I} & -C_{12}(A_0 + C_{22})^{-1} \\ 0 & \mathbb{I} \end{pmatrix}, \quad D_2 = \begin{pmatrix} \mathbb{I} \\ -(A_0 + C_{22})^{-1}C_{21} & \mathbb{I} \end{pmatrix}
\]
and observe
\[
D_1(A + C)D_2 = \begin{pmatrix} C_{11} - C_{12}(A_0 + C_{22})^{-1}C_{21} & 0 \\ 0 & A_0 + C_{22} \end{pmatrix}.
\]
Since \( D_1, D_2 \) are homeomorphisms we see that \( A + C \) is Fredholm since the right-hand side obviously is. Moreover, \( \text{ind}(A + C) = \text{ind}(C_{11} - C_{12}(A_0 + C_{22})^{-1}C_{21}) = \dim(\text{Ker}(A)) - \dim(Y_0) = \text{ind}(A) \) since the second operator is between finite dimensional spaces and the index can be evaluated using (6.59).

Finally, the index is constant along \([0, 1] \to \mathcal{F}(X,Y), \alpha \mapsto \mathbb{I} - \alpha K\) and hence \( \text{ind}(\mathbb{I} - K) = \text{ind}(\mathbb{I}) = 0. \)

Next we show that an operator is Fredholm if and only if it has a left/right inverse up to compact operators.
Theorem 6.28 (Atkinson). An operator \( A \in \mathcal{L}(X, Y) \) is Fredholm if and only if there exist \( B_1, B_2 \in \mathcal{L}(Y, X) \) such that \( B_1 A - I \in \mathcal{C}(X) \) and \( AB_2 - I \in \mathcal{C}(Y) \).

Proof. If \( A \) is Fredholm we have already given an operator \( B \) in (6.66) such that \( BA - I \) and \( AB - I \) are finite rank. Conversely, according to Theorem 6.27 \( B_1 A \) and \( AB_2 \) are Fredholm. Since \( \text{Ker}(A) \subseteq \text{Ker}(B_2 A) \) and \( \text{Ran}(AB_2) \subseteq \text{Ran}(A) \) this shows that \( A \) is Fredholm. \( \square \)

Operators \( B_1 \) and \( B_2 \) as in the previous theorem are also known as a left and right parametrix, respectively. As a consequence we can now strengthen Theorem 6.27:

Corollary 6.29. For every \( A \in \mathcal{F}(X, Y) \) and \( K \in \mathcal{C}(X, Y) \) we have \( A + K \in \mathcal{F}(X, Y) \) with \( \text{ind}(A + K) = \text{ind}(A) \).

Proof. Using (6.66) we see that \( (A + K)B - I = -P + BK \in \mathcal{C}(Y) \) and \( (A + K)B - I = -Q + KB \in \mathcal{C}(X) \) and hence \( A + K \) is Fredholm. In fact, \( A + \alpha K \) is Fredholm for \( \alpha \in [0, 1] \) and hence \( \text{ind}(A + K) = \text{ind}(A) \) by continuity of the index. \( \square \)

Note that we have established the famous Fredholm alternative alluded to at the beginning.

Theorem 6.30 (Fredholm alternative). Suppose \( A \in \mathcal{F}(X, Y) \) has index \( \text{ind}(A) = 0 \). Then \( A \) is a homeomorphism if and only if it is injective. In particular, either the inhomogeneous equation

\[
Ax = y
\]

has a unique solution for every \( y \in Y \) or the corresponding homogeneous equation

\[
Ax = 0
\]

has a nontrivial solution.

Note that the more commonly found version of this theorem is for the special case \( A = I + K \) with \( K \) compact. In particular, it applies to the case where \( K \) is a compact integral operator (cf. Lemma 3.4).

Problem 6.22. Suppose \( X \xrightarrow{A} Y \xrightarrow{B} Z \) is exact. Show that if \( X \) and \( Z \) are finite dimensional, so is \( Y \).

Problem 6.23. Let \( X_j \) be finite dimensional vector spaces and suppose

\[
0 \rightarrow X_1 \xrightarrow{A_1} X_2 \xrightarrow{A_2} X_3 \cdots X_{n-1} \xrightarrow{A_{n-1}} X_n \rightarrow 0
\]
is exact. Show that
\[ \sum_{j=1}^{n} (-1)^j \dim(X_j) = 0. \]
(Hint: Rank-nullity theorem.)

**Problem 6.24.** Let \( A \in \mathcal{F}(X,Y) \) with a corresponding parametrix \( B_1, B_2 \in \mathcal{C}(Y,X) \). Set \( K_1 := I - B_1 A \in \mathcal{C}(X) \), \( K_2 := I - AB_2 \in \mathcal{C}(Y) \) and show
\[ B_1 - B = B_1 Q - K_1 B \in \mathcal{C}(Y,X), \quad B_2 - B = PB_2 - BK_2 \in \mathcal{C}(Y,X). \]
Hence a parametrix is unique up to compact operators. Moreover, \( B_1, B_2 \in \mathcal{F}(Y,X) \).

### 6.6. Spectral theory for bounded operators

So far, except for Section 6.3, we have developed spectral theory on an algebraic level based on the fact that bounded operators form a Banach algebra. In this section we want to take a more operator centered view and consider bounded linear operators \( \mathcal{L}(X) \), where \( X \) is some Banach space. Now we can make a finer subdivision of the spectrum based on why our operator fails to have a bounded inverse. Since in the bijective case boundedness of the inverse comes for free from the inverse mapping theorem (Theorem 4.6), there are basically two things which can go wrong: Either our map is not injective or it is not surjective. Moreover, in the latter case one can also ask how far it is away from being surjective, that is, if the range is dense or not. Accordingly one defines the **point spectrum**
\[ \sigma_p(A) := \{ \alpha \in \sigma(A) \mid \text{Ker}(A - \alpha) \neq \{0\} \} \]
(6.70)
as the set of all eigenvalues, the **continuous spectrum**
\[ \sigma_c(A) := \{ \alpha \in \sigma(A) \setminus \sigma_p(A) \mid \text{Ran}(A - \alpha) = X \} \]
(6.71)and finally the **residual spectrum**
\[ \sigma_r(A) := \{ \alpha \in \sigma(A) \setminus \sigma_p(A) \mid \text{Ran}(A - \alpha) \neq X \}. \]
(6.72)
Clearly we have
\[ \sigma(A) = \sigma_p(A) \cup \sigma_c(A) \cup \sigma_r(A). \]
(6.73)

**Example.** Suppose \( \mathcal{H} \) is a Hilbert space and \( A = A^* \) is self-adjoint. Then by (2.28), \( \sigma_r(A) = \emptyset \).

**Example.** Suppose \( X := \ell^p(\mathbb{N}) \) and \( L \) is the left shift. Then \( \sigma(L) = \mathcal{B}_1(0) \). Indeed, a simple calculation shows that Ker\((L - z) = \text{span}\{(z^j)_{j \in \mathbb{N}}\} \) for \( |z| < 1 \) if \( 1 \leq p < \infty \) and for \( |z| \leq 1 \) if \( p = \infty \). Hence \( \sigma_p(L) = \mathcal{B}_1(0) \) for \( 1 \leq p < \infty \) and \( \sigma_p(L) = \mathcal{B}_1(0) \) if \( p = \infty \). In particular, since the spectrum is closed and \( \|L\| = 1 \) we have \( \sigma(L) = \mathcal{B}_1(0) \). Moreover, for \( y \in \ell_c(\mathbb{N}) \) we set \( x_j := -\sum_{k=j}^{\infty} z^{j-k-1} y_k \) such that \((L - z)x = y\). In
particular, $\ell_2(\mathbb{N}) \subset \text{Ran}(L-z)$ and hence $\text{Ran}(S-z)$ is dense for $1 \leq p < \infty$. Thus $\sigma_c(L) = \partial B_1(0)$ for $1 \leq p < \infty$. Consequently, $\sigma_r(L) = \emptyset$. \hfill \Diamond

Since $A$ is invertible if and only if $A'$ is by Theorem 4.26 we obtain:

**Lemma 6.31.** Suppose $A \in \mathcal{L}(X)$. Then

$$\sigma(A) = \sigma(A').$$

Moreover, $\sigma_r(A) \subseteq \sigma_p(A') \subseteq \sigma_p(A) \cup \sigma_r(A)$ as well as $\sigma_c(A') \subseteq \sigma_c(A)$. If in addition $X$ is reflexive we have $\sigma_r(A') \subseteq \sigma_p(A) \subseteq \sigma_p(A') \cup \sigma_r(A')$ as well as $\sigma_c(A') = \sigma_c(A)$.

**Proof.** This follows from Theorem 4.26 and (4.20). In the reflexive case use $A \cong A''$. \hfill \Box

Note that in a Hilbert space $\sigma_x(A^*) = \sigma_x(A')^*$ for $x \in \{p, c, r\}$.

**Example.** Consider $L'$ from the previous example, which is just the right shift. Then $\sigma(L') = \sigma(L) = B_1(0)$. Moreover, it is easy to see that $\sigma_p(L') = \emptyset$ and thus $\sigma_c(L') = \sigma_p(L) = B_1(0)$ as well as $\sigma_r(L') = \sigma_p(L) = B_1(0)$ in the reflexive case $1 < p < \infty$. If $p = 1$ we have $\sigma_p(L') = \emptyset$ and the previous lemma tells us $\sigma_c(L') \subseteq \sigma_c(L) = \partial B_1(0)$. Hence $B_1(0) \subseteq \sigma_r(L')$. If we have $(L' - z)x = y$ with some $y \in \ell^\infty(\mathbb{N})$ we must have $x_j := \sum_{k=1}^{j} z^{k-j-1} y_k$. But it is nontrivial to see which $y$ lead to a bounded $x$. \hfill \Diamond

Next we want to have a closer look at eigenvalues. Note that eigenvectors corresponding to different eigenvalues are always linearly independent (Problem 6.26).

Finally, let us also discuss the connection with eigenvalues. We have seen that for a symmetric compact operator in a Hilbert space we can choose an orthonormal basis of eigenfunctions. Without the symmetry assumption we know that even in the finite dimensional case we can in general no longer find a basis of eigenfunctions and that the Jordan canonical form is the best one can do. There the generalized eigenspaces $\text{Ker}((A - z)^k)$ play an important role. In this respect one looks at the following ascending and descending chains of subspaces associated to $A \in \mathcal{L}(X)$ (where we have assumed $z = 0$ without loss of generality):

$$\{0\} \subseteq \text{Ker}(A) \subseteq \text{Ker}(A^2) \subseteq \text{Ker}(A^3) \subseteq \cdots \quad (6.74)$$

and

$$X \supseteq \text{Ran}(A) \supseteq \text{Ran}(A^2) \supseteq \text{Ran}(A^3) \supseteq \cdots \quad (6.75)$$

We will say that the kernel chain stabilizes at $n$ if $\text{Ker}(A^{n+1}) = \text{Ker}(A^n)$. Substituting $x = Ay$ in the equivalence $A^nx = 0 \iff A^{n+1}x = 0$ gives $A^{n+1}y = 0 \iff A^{n+2}y = 0$ and hence by induction we have $\text{Ker}(A^{n+k}) = \text{Ker}(A^n)$ for all $k \in \mathbb{N}_0$ in this case. Similarly, will say that the range
chain stabilizes at m if Ran(A^{m+1}) = Ran(A^m). Again, if x = A^{m+2}y ∈ Ran(A^{m+2}) we can write A^{m+1}y = A^mz for some z which shows x = A^{m+1}z ∈ Ran(A^{m+1}) and thus Ran(A^{m+k}) = Ran(A^m) for all k ∈ \mathbb{N}_0 in this case. While in a finite dimensional case both chains eventually have to stabilize, there is no reason why the same should happen in an infinite dimensional space.

**Example.** For the left shift operator L we have Ran(L^n) = \ell^p(\mathbb{N}) for all n ∈ \mathbb{N} while the kernel chain does not stabilize as Ker(L^n) = \{a ∈ \ell^p(\mathbb{N})|a_j = 0, j > n\}. Similarly, for the right shift operator R we have Ker(R^n) = \{0\} while the range chain does not stabilize as Ran(R^n) = \{a ∈ \ell^p(\mathbb{N})|a_j = 0, 1 ≤ j ≤ n\}.

**Lemma 6.32.** Suppose A : X → X is a linear operator.

(i) The kernel chain stabilizes at n if Ran(A^n) ∩ Ker(A) = \{0\}. Conversely, if the kernel chain stabilizes at n then Ran(A^n) ∩ Ker(A) = \{0\}.

(ii) The range chain stabilizes at m if Ker(A^m) + Ran(A) = X. Conversely, if the range chain stabilizes at m then Ker(A^m) + Ran(A^m) = X = \{0\}.

(iii) If both chains stabilize, then m = n and X = Ker(A^m) + Ran(A^m) = X.

**Proof.**

(i). If Ran(A^n) ∩ Ker(A) = \{0\} then A^{n+1}x = 0 implies A^n x ∈ Ran(A^n) ∩ Ker(A) = \{0\} and the kernel chain stabilizes at n. Conversely, let x ∈ Ran(A^n) ∩ Ker(A^n), then x = A^ny and A^n x = A^{2n}y = 0 implying y ∈ Ker(A^{2n}) = Ker(A^n), that is, x = A^ny = 0.

(ii). If Ker(A^m) + Ran(A) = X, then for any x = z + Ty we have T^m x = T^{m+1}y and hence Ran(T^m) = Ran(T^{m+1}). Conversely, if the range chain stabilizes at m, then T^m x = T^{2m}y and x = T^m y + (x - T^m y).

(iii). Suppose Ran(A^{m+1}) = Ran(A^m) but Ker(A^m) ⊊ Ker(A^{m+1}). Let x ∈ Ker(A^{m+1}) \ Ker(A^m) and observe that by 0 ≠ A^m x = A^{m+1}y there is an x ∈ Ker(A^{m+2}) \ Ker(A^{m+1}). Iterating this argument would shows that the kernel chain does not stabilize contradiction our assumption. Hence n ≤ m.

Conversely, suppose Ker(A^{n+1}) = Ker(A^n) and Ran(A^{m+1}) = Ran(A^m) for m ≥ n. Then

A^m x = A^{m+1} y \Rightarrow x - Ay ∈ Ker(A^m) = Ker(A^n) \Rightarrow A^n x = A^{n+1} y

shows Ran(A^{n+1}) = Ran(A^n), that is, m ≤ n.

The last claim now follows by combining (i) and (ii).
In particular this applies to compact operators:

Lemma 6.33. Suppose that \( K \in \mathcal{C}(X) \). Then for every \( z \in \mathbb{C} \setminus \{0\} \) either \( K - z \) has a bounded inverse or \( z \) is an eigenvalue and there is some \( n = n(z) \in \mathbb{N} \) such that \( \ker(K - z)^n = \ker(K - z)^{n+k} \) and \( \operatorname{ran}(K - z)^n = \operatorname{ran}(K - z)^{n+k} \) for every \( k \geq 0 \) and

\[
X = \ker(K - z)^n + \operatorname{ran}(K - z)^n. \tag{6.76}
\]

Proof. Considering \( I - z^{-1}K \) we can assume \( z = 1 \) without loss of generality. Moreover, since \( (I - K)^n - I \in \mathcal{C}(X) \) we see that \( \ker(K - z)^n \) is finite dimensional and \( \operatorname{ran}(K - z)^n \) is closed and finite codimensional for every \( n \in \mathbb{N} \). Next suppose the kernel chain does not stabilize. Abbreviate \( K_n := \ker(I - K)^n \). Then, by Problem 4.26, we can choose \( x_n \in K_{n+1} \setminus K_n \) such that \( \|x_n\| = 1 \) and \( \operatorname{dist}(x_n, K_n) \geq \frac{1}{2} \). But since \( (I - K)x_n \in K_n \) and \( Kx_n \in K_{n+1} \), we see that

\[
\|Kx_n - Kx_m\| = \|x_n - (I - K)x_n - Ax_m\| \geq \operatorname{dist}(x_n, K_n) \geq \frac{1}{2}
\]

for \( n > m \) and hence the bounded sequence \( Kx_n \) has no convergent subsequence, a contradiction. Consequently the kernel sequence for \( K \) also stabilizes, thus by Problem 4.23 \( \operatorname{coker}(I - K)^n \approx \ker(I - K^n) \) the sequence of cokernels stabilizes, which finally implies that the range sequence stabilizes. The rest follows from the previous lemma. \( \square \)

If \( \alpha \) is an eigenvalue and the kernel chain \( \ker((A - \alpha)^n) \) stabilizes at \( n \), the \( n \) is called the index of the eigenvalue. The order of a generalized eigenvector \( u \) corresponding to an eigenvalue \( \alpha \) is the smallest \( n \) such that \((A - \alpha)^n u \neq 0 \).

Problem 6.25. Discuss the spectrum of the right shift on \( \ell^1(\mathbb{N}) \).

Problem 6.26. Suppose \( A \in \mathcal{L}(X) \). Show that generalized eigenvectors corresponding to different eigenvalues or with different order are linearly independent.

Problem 6.27. Suppose \( A \in \mathcal{F}(X) \). If the kernel chain stabilizes then \( \operatorname{ind}(A) \leq 0 \). If the range chain stabilizes then \( \operatorname{ind}(A) \geq 0 \).
In this chapter we want to look at ordinary linear differential equations in Banach spaces. As a preparation we briefly discuss a few relevant facts about differentiation and integration for Banach space valued functions.

7.1. Analysis for Banach space valued functions

Let $X$ be a Banach space. Let $I \subseteq \mathbb{R}$ be some interval and denote by $C(I,X)$ the set of continuous functions from $I$ to $X$. Given $t \in I$ we call $f: I \to X$ differentiable at $t$ if the limit
\[
\dot{f}(t) := \lim_{\varepsilon \to 0} \frac{f(t + \varepsilon) - f(t)}{\varepsilon}
\] (7.1)
exists. The set of functions $f: I \to X$ which are differentiable at all $t \in I$ and for which $\dot{f} \in C(I,X)$ is denoted by $C^1(I,X)$. Clearly $C^1(I,X) \subset C(I,X)$. As usual we set $C^{k+1}(I,X) := \{ f \in C^1(I,x) | \dot{f} \in C^k(I,X) \}$.

Note that if $U \in \mathcal{L}(X,Y)$ and $f \in C^k(I,X)$, then $Uf \in C^k(I,Y)$ and $\frac{d}{dt}Uf = U \dot{f}$.

The following version of the mean value theorem will be crucial.

**Theorem 7.1 (Mean value theorem).** Suppose $f(t) \in C^1(I,X)$. Then
\[
\|f(t) - f(s)\| \leq M|t - s|, \quad M := \sup_{\tau \in [s,t]} \|\dot{f}(\tau)\|,
\] (7.2)
for $s \leq t \in I$.

**Proof.** Fix $\tilde{M} > M$ and consider $d(\tau) := \|f(\tau) - f(s)\| - \tilde{M}(\tau - s)$ for $\tau \in [s,t]$. Suppose $\tau_0$ is the largest $\tau$ for which the claim holds. So there is
a sequence $\varepsilon_n \downarrow 0$ such that for sufficiently large $n$

\[ 0 > d(\tau_0 + \varepsilon_n) = \|f(\tau_0 + \varepsilon_n) - f(\tau_0) + f(\tau_0) - f(s)\| - \tilde{M}(\tau_0 + \varepsilon_n - s) \leq \|f(\tau_0 + \varepsilon_n) - f(\tau_0)\| - \tilde{M}\varepsilon_n = \|\dot{f}(\tau)\varepsilon_n + o(\varepsilon_n)\| - \tilde{M}\varepsilon_n \leq (M - \tilde{M})\varepsilon_n + o(\varepsilon_n) < 0. \]

This contradicts our assumption. \(\square\)

In particular,

**Corollary 7.2.** For $f \in C^1(I, X)$ we have $\dot{f} = 0$ if and only if $f$ is constant.

Next we turn to integration. Let $I := [a, b]$ be compact. A function $f : I \to X$ is called a **step function** provided there are numbers $t_0 = a < t_1 < t_2 < \cdots < t_{n-1} < t_n = b$ (7.3) such that $f(t)$ is constant on each of the open intervals $(t_{i-1}, t_i)$. The set of all step functions $S(I, X)$ forms a linear space and can be equipped with the sup norm. The corresponding Banach space obtained after completion is called the set of **regulated functions** $R(I, X)$. In other words, a regulated function is the uniform limit of a step function.

Observe that $C(I, X) \subset R(I, X)$. In fact, consider the functions $f_n := \sum_{j=0}^{n-1} f(t_j)\chi(t_j, t_{j+1}) \in S(I, X)$, where $t_j := a + j\frac{b-a}{n}$ and $\chi$ is the characteristic function. Since $f \in C(I, X)$ is uniformly continuous, we infer that $f_n$ converges uniformly to $f$.

For a step function $f \in S(I, X)$ we can define a linear map $\int : S(I, X) \to X$ by

\[ \int_I f(t) dt := \sum_{i=1}^{n} x_i(t_i - t_{i-1}), \]  

where $x_i$ is the value of $f$ on $(t_{i-1}, t_i)$. This map satisfies

\[ \left\| \int_I f(t) dt \right\| \leq (b - a)\|f\|_{\infty} \]  

and hence it can be extended uniquely to a bounded linear map $\int : R(I, X) \to X$ with the same norm $(b - a)$ by Theorem 1.12. Of course if $X = \mathbb{C}$ this coincides with the usual Riemann integral. We even have

\[ \left\| \int_I f(t) dt \right\| \leq \int_I \|f(t)\| dt. \]  

In fact, by the triangle inequality this holds for step functions and thus extends to all regulated functions by continuity.
In addition, if \( A \in \mathcal{L}(X,Y) \), then \( f \in R(I,X) \) implies \( Af \in R(I,Y) \) and
\[
A \int_I f(t)dt = \int_I Af(t)dt. \tag{7.7}
\]
Again this holds for step functions and thus extends to all regulated functions by continuity. In particular, if \( \ell \in X^* \) is a continuous linear functional, then
\[
\ell \left( \int_I f(t)dt \right) = \int_I \ell(f(t))dt, \quad f \in R(I,X). \tag{7.8}
\]
Moreover, we will use the usual conventions
\[
\int_{t_1}^{t_2} f(s)ds := \int_I \chi_{(t_1,t_2]}(s)f(s)ds \quad \text{and} \quad \int_{t_1}^{t_2} f(s)ds := -\int_{t_2}^{t_1} f(s)ds.
\]

**Theorem 7.3** (fundamental theorem of calculus). If \( f \in C(I,X) \), then \( F(t) := \int_a^t f(s)ds \in C^1(I,X) \) and \( \dot{F}(t) = f(t) \). Conversely, for any \( F \in C^1(I,X) \) we have
\[
F(t) = F(a) + \int_a^t \dot{F}(s)ds. \tag{7.9}
\]

**Proof.** The first part can be seen from
\[
\left\| \int_a^{t+\varepsilon} f(s)ds - \int_a^t f(s)ds - f(t)\varepsilon \right\| = \left\| \int_t^{t+\varepsilon} (f(s) - f(t))ds \right\| \leq |\varepsilon| \sup_{s \in [t,t+\varepsilon]} \|f(s) - f(t)\|.
\]
The second follows from the first part which implies \( \frac{d}{dt}(F(t) - \int_a^t \dot{F}(s)ds) = 0 \). Hence this difference is constant and equals its value at \( t = a \). \( \square \)

**Problem 7.1** (Product rule). Let \( X \) be a Banach algebra. Show that if \( f, g \in C^1(I,X) \) then \( fg \in C^1(I,X) \) and \( \frac{d}{dt}fg = \dot{f}g + fg \).

**Problem 7.2.** Let \( f \in R(I,X) \) and \( \tilde{I} := I + t_0 \). then \( f(t - t_0) \in R(\tilde{I},X) \) and
\[
\int_I f(t)dt = \int_{t_0}^I f(t)dt.
\]

**Problem 7.3.** Let \( A : \mathcal{D}(A) \subseteq X \to X \) be a closed operator. Show that (7.7) holds for \( f \in C(I,X) \) with \( \text{Ran}(f) \subseteq \mathcal{D}(A) \) and \( Af \in C(I,X) \).

### 7.2. Uniformly continuous operator groups

Now we are ready to apply these ideas to the abstract Cauchy problem
\[
\dot{u} = Au, \quad u(0) = u_0 \tag{7.10}
\]
in some Banach space \( X \). Here \( A \) is some linear operator and we will assume that \( A \in \mathcal{L}(X) \) to begin with. Note that in the simplest case \( X = \mathbb{R}^n \) this
is simply a linear first order system with constant coefficient matrix $A$. In this case the solution is given by

$$u(t) = T(t)u_0,$$

where

$$T(t) := \exp(tA) := \sum_{j=0}^{\infty} \frac{t^j}{j!} A^j$$

is the exponential of $tA$. It is not difficult to see that this also gives the solution in our Banach space setting.

**Theorem 7.4.** Let $A \in \mathcal{L}(X)$. Then the series in $(7.12)$ converges and defines a uniformly continuous operator group:

(i) The map $t \mapsto T(t)$ is continuous, $T \in C(\mathbb{R}, \mathcal{L}(X))$, and commutes with $A$, $AT(t) = T(t)A$.

(ii) $T(0) = I$ and $T(t+s) = T(t)T(s)$ for all $t, s \in \mathbb{R}$.

Moreover, we even have $T \in C^1(\mathbb{R}, \mathcal{L}(X))$ and $T$ is the unique solution of $\dot{T}(t) = AT(t)$ with $T(0) = I$.

**Proof.** Set

$$T_n(t) := \sum_{j=0}^{n} \frac{t^j}{j!} A^j.$$

Then (for $m \leq n$)

$$\|T_n(t) - T_m(t)\| = \left\| \sum_{j=m+1}^{n} \frac{t^j}{j!} A^j \right\| \leq \sum_{j=m+1}^{n} \frac{|t|^j}{j!} \|A\|^j \leq \frac{|t|^{m+1}}{(m+1)!} \|A\|^{m+1} e^{|t|\|A\|}$$

In particular,

$$\|T(t)\| \leq e^{|t|\|A\|}$$

and $AT(t) = \lim_{n \to \infty} AT_n(t) = \lim_{n \to \infty} T_n(t)A = T(t)A$. Furthermore we have $\dot{T}_{n+1} = AT_n$ and thus

$$T_{n+1}(t) = I + \int_0^t AT_n(s)ds.$$

Taking limits shows

$$T(t) = I + \int_0^t AT(s)ds$$

or equivalently $T(t) \in C^1(\mathbb{R}, \mathcal{L}(X))$ and $\dot{T}(t) = AT(t)$, $T(0) = I$.

Suppose $S(t)$ is another solution $\dot{S} = AS$, $S(0) = I$. Then, by the product rule (Problem 7.1), $\frac{d}{dt} T(-t)S(t) = T(-t)AS(t) - AT(-t)S(t) = 0$ implying $T(-t)S(t) = T(0)S(0) = I$. In the special case $T = S$ this shows $T(-t) = T^{-1}(t)$ and in the general case it hence proves uniqueness $S = T$. 

Finally, \( T(t+s) \) and \( T(t)T(s) \) both satisfy our differential equation and coincide at \( t = 0 \). Hence they coincide for all \( t \) by uniqueness. □

Clearly \( A \) is uniquely determined by \( T(t) \) via \( A = T'(0) \). Moreover, from this we also easily get uniqueness for our original Cauchy problem. We will in fact be slightly more general and consider the inhomogeneous problem

\[
\dot{u} = Au + g, \quad u(0) = u_0, \tag{7.13}
\]

where \( g \in C(I, X) \). A solution necessarily satisfies

\[
\frac{d}{dt}T(-t)u(t) = -AT(-t)u(t) + T(-t)\dot{u}(t) = T(-t)g(t)
\]

and integrating this equation (fundamental theorem of calculus) shows

\[
u(t) = T(t) \left( u_0 + \int_0^t T(-s)g(s)ds \right) = T(t)u_0 + \int_0^t T(t-s)g(s)ds. \tag{7.14}
\]

It is straightforward to verify that this is indeed a solution for any given \( g \in C(I, X) \).

**Lemma 7.5.** Let \( A \in \mathcal{L}(X) \) and \( g \in C(I, X) \). Then (7.13) has a unique solution given by (7.14).

**Example.** For example look at the discrete linear wave equation

\[
\ddot{q}_n(t) = k(q_{n+1}(t) - 2q_n(t) + q_{n-1}(t)), \quad n \in \mathbb{Z}.
\]

Factorizing this equation according to

\[
\dot{q}_n(t) = p_n(t), \quad \dot{p}_n(t) = k(q_{n+1}(t) - 2q_n(t) + q_{n-1}(t)),
\]

we can write this as a first order system

\[
\frac{d}{dt} \begin{pmatrix} q_n \\ p_n \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ k & -k \end{pmatrix} \begin{pmatrix} q_n \\ p_n \end{pmatrix}
\]

with the Jacobi operator \( A_0 q_n = q_{n+1} - 2q_n + q_{n-1} \). Since \( A_0 \) is a bounded operator in \( X = \ell^p(\mathbb{Z}) \) we obtain a well-defined uniformly continuous operator group in \( \ell^p(\mathbb{Z}) \oplus \ell^p(\mathbb{Z}) \). □

**Problem 7.4** (Product rule). Suppose \( f \in C^1(I, X) \) and \( T \in C^1(I, \mathcal{L}(X)) \). Show that \( Tf \in C^1(I, X) \) and \( \frac{d}{dt}Tf = \dot{T}f + T\dot{f} \).

### 7.3. Strongly continuous semigroups

In the previous section we have found a quite complete solution of the abstract Cauchy problem (7.13) in the case when \( A \) is bounded. However, since differential operators are typically unbounded this assumption is too strong for applications to partial differential equations. First of all, even rather simple equations like the heat equations are only solvable for positive times...
and hence we will only assume that the solutions give rise to a semigroup. Moreover, continuity in the operator topology is too much to ask for (in fact it is be equivalent to boundedness of $A$ — Problem 7.5) and hence we go for the next best option, namely strong continuity. In this sense, our problem is still well-posed.

A **strongly continuous operator semigroup** (also $C_0$-semigroup) is a family of operators $T(t) \in \mathcal{L}(X)$, $t \geq 0$, such that

1. $T(t)g \in C([0, \infty), X)$ for every $g \in X$ (strong continuity) and
2. $T(0) = I$, $T(t + s) = T(t)T(s)$ for every $t, s \geq 0$ (semigroup property).

If item (ii) holds for all $t, s \in \mathbb{R}$ it is called a **strongly continuous operator group**.

We first note that $\|T(t)\|$ is uniformly bounded on compact time intervals.

**Lemma 7.6.** Let $T(t)$ be a $C_0$-semigroup. Then there are constants $M \geq 1$, $\omega \geq 0$ such that

$$\|T(t)\| \leq Me^{\omega t}, \quad t \geq 0.$$  \hspace{1cm} (7.15)

In case of a $C_0$-group we have $\|T(t)\| \leq Me^{\omega |t|}, t \in \mathbb{R}$.

**Proof.** Since $\|T(.)g\| \in C[0, 1]$ for every $g \in X$ we have $\sup_{t \in [0, 1]} \|T(t)g\| \leq M_f$. Hence by the uniform boundedness principle $\sup_{t \in [0, 1]} \|T(t)\| \leq M$ for some $M \geq 1$. Setting $\omega = \log(M)$ the claim follows by induction using the semigroup property. For the group case apply the semigroup case to both $T(t)$ and $S(t) := T(-t)$. \hfill $\square$

Inspired by the previous section we define the **generator** $A$ of a strongly continuous semigroup as the linear operator

$$Af := \lim_{t \searrow 0} \frac{1}{t}(T(t)f - f),$$  \hspace{1cm} (7.16)

where the domain $\mathcal{D}(A)$ is precisely the set of all $f \in X$ for which the above limit exists. Moreover, a $C_0$-semigroup is the solution of the abstract Cauchy problem associated with its generator $A$:

**Lemma 7.7.** Let $T(t)$ be a $C_0$-semigroup with generator $A$. If $f \in \mathcal{D}(A)$ then $T(t)f \in \mathcal{D}(A)$ and $AT(t)f = T(t)Af$. Moreover, suppose $g \in X$ with $u(t) = T(t)g \in \mathcal{D}(A)$ for $t > 0$. Then $u(t) \in C^1((0, \infty), X) \cap C([0, \infty), X)$ and $u(t)$ is the unique solution of the abstract Cauchy problem

$$\dot{u}(t) = Au(t), \quad u(0) = g.$$  \hspace{1cm} (7.17)

This is, for example, the case if $g \in \mathcal{D}(A)$ in which case we even have $u(t) \in C^1([0, \infty), X)$. 


Similarly, if $T(t)$ is a $C_0$-group and $g \in \mathcal{D}(A)$, then $u(t) := T(t)g \in C^1([\mathbb{R}, X]$ is the unique solution of (7.17) for all $t \in \mathbb{R}$.

**Proof.** Let $f \in \mathcal{D}(A)$ and $t > 0$ (respectively $t \in \mathbb{R}$ for a group), then
\[
\lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} (u(t + \varepsilon) - u(t)) = \lim_{\varepsilon \downarrow 0} T(t) \frac{1}{\varepsilon} (T(\varepsilon) f - f) = T(t) Af.
\]
This shows the first part. To show that $u(t)$ is differentiable it remains to compute
\[
\lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} (u(t - \varepsilon) - u(t)) = \lim_{\varepsilon \downarrow 0} T(t - \varepsilon) \frac{1}{\varepsilon} (T(\varepsilon) f - f)
\]
\[
= \lim_{\varepsilon \downarrow 0} T(t - \varepsilon) (Af + o(1)) = T(t) Af
\]
since $\|T(t)\|$ is bounded on compact $t$ intervals. Hence $u(t) \in C^1([0, \infty), X)$ (respectively $u(t) \in C^1([\mathbb{R}, X]$ for a group) solves (7.17). In the general case $f = T(t_0)g \in \mathcal{D}(A)$ and $u(t) = T(t - t_0)f$ solves our differential equation for every $t > t_0$. Since $t_0 > 0$ is arbitrary it follows that $u(t)$ solves (7.17) by the first part. To see that it is the only solution, let $v(t)$ be a solution corresponding to the initial condition $v(0) = 0$. For $s \leq t$ we have
\[
\frac{d}{ds} T(t-s)v(s) = \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} (T(t-s-\varepsilon)v(s+\varepsilon) - T(\varepsilon) v(s))
\]
\[
= \lim_{\varepsilon \rightarrow 0} T(t-s-\varepsilon) \frac{1}{\varepsilon} (v(s+\varepsilon) - v(s))
\]
\[
- T(t-s) \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} (T(\varepsilon) v(s) - v(s))
\]
\[
=T(t-s)Av(s) - T(t-s)Av(s) = 0.
\]
Whence, $v(t) = T(t-t_0)v(t) = T(t-s)v(s) = T(t)v(0) = 0$. \qed

Before turning to some examples, we establish a useful criterion for a semigroup to be strongly continuous.

**Lemma 7.8.** A (semi)group of bounded operators is strongly continuous if and only if $\limsup_{\varepsilon \downarrow 0} \|T(\varepsilon)g\| < \infty$ for every $g \in X$ and $\lim_{\varepsilon \downarrow 0} T(\varepsilon)f = f$ for $f$ in a dense subset.

**Proof.** First of all, suppose $T(t)$ is not bounded in a small interval $[0, \delta]$, then there exists a sequence $\varepsilon_n \downarrow 0$ with $\|T(\varepsilon_n)\| \rightarrow \infty$. Hence $\|T(\varepsilon_n)g\| \rightarrow \infty$ for some $g$ by the uniform boundedness principle, a contradiction. Thus there exists some $M$ such that $\sup_{t \in [0, \delta]} \|T(t)\| \leq M$. Setting $\omega = \frac{\log(M)}{\delta}$ we even obtain (7.15). Moreover, in combination with boundedness of $T(t)$ this shows that that $\lim_{\varepsilon \downarrow 0} T(\varepsilon)f = f$ for all $f \in X$ by a simple approximation argument.
In case of a group this also shows \( \|T(-t)\| \leq \|T(\delta - t)\| \|T(\delta)\| \leq M\|T(-\delta)\| \) for \( 0 \leq t \leq \delta \). Choosing \( M = \max(M, \|T(-\delta)\|) \) we conclude \( \|T(t)\| \leq \tilde{M} \exp(\tilde{\omega}|t|) \).

Finally, right continuity is immediate from the semigroup property: \( \lim_{\varepsilon \downarrow 0} T(t+\varepsilon)g = T(\varepsilon)T(t)g = T(t)g \). Left continuity follows from \( \|T(t-\varepsilon)g - T(t)g\| = \|T(t-\varepsilon)(T(\varepsilon)g - g)\| \leq \|T(t-\varepsilon)\| \|\|T(\varepsilon)g - g\| \). \( \square \)

**Example.** Let \( X := C_0(\mathbb{R}) \) be the continuous functions vanishing as \( |x| \to \infty \). Then it is straightforward to check that

\[
(T(t)f)(x) := f(x + t)
\]

defines a group of continuous operators on \( X \). Since shifting a function does not alter its supremum we have \( \|T(t)f\|_\infty = \|f\|_\infty \) and hence \( \|T(t)\| = 1 \). Moreover, strong continuity is immediate for uniformly continuous functions. Since every function with compact support is uniformly continuous and since such functions are dense, we get that \( T \) is strongly continuous. Moreover, for \( f \in \mathcal{D}(A) \) we have

\[
\lim_{\varepsilon \to 0} \frac{f(t+\varepsilon) - f(t)}{\varepsilon} = (Af)(t)
\]

uniformly. In particular, \( f \in C^1(\mathbb{R}) \) with \( f,f' \in C_0(\mathbb{R}) \). Conversely, for \( f \in C^1(\mathbb{R}) \) with \( f,f' \in C_0(\mathbb{R}) \) we have

\[
\frac{f(t+\varepsilon) - f(t) - \varepsilon f'(t)}{\varepsilon} = \frac{1}{\varepsilon} \int_0^\varepsilon (f'(t+s) - f'(t)) ds \leq \sup_{0 \leq s \leq \varepsilon} \|T(s)f' - f'\|_\infty
\]

which converges to zero as \( \varepsilon \downarrow 0 \) by strong continuity of \( T \). Whence

\[
A = \frac{d}{dx}, \quad \mathcal{D}(A) = \{ f \in C^1(\mathbb{R}) \cap C_0(\mathbb{R}) | f' \in C_0(\mathbb{R}) \}.
\]

It is not hard to see that \( T \) is not uniformly continuous or, equivalently, that \( A \) is not bounded (cf. Problem 7.5).

Note that this group is not not strongly continuous when considered on \( X := C_b(\mathbb{R}) \). Indeed for \( f(x) = \cos(x^2) \) we can choose \( x_n = \sqrt{2\pi n} \) and \( t_n = \sqrt{2\pi}(\sqrt{n + \frac{1}{4}} - \sqrt{n}) = \frac{1}{4}\sqrt{2\pi} + O(n^{-3/2}) \) such that \( \|T(t_n) f - f\|_\infty \geq |f(x_n + t_n) - f(x_n)| = 1 \).

Next consider

\[
u(t) = T(t)g, \quad v(t) := \int_0^t u(s) ds, \quad g \in X.
\] (7.18)

Then \( v \in C^1([0, \infty), X) \) with \( \dot{v}(t) = u(t) \) and (Problem 7.2)

\[
\lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} (T(\varepsilon)v(t) - v(t)) = \lim_{\varepsilon \downarrow 0} \left( \frac{1}{\varepsilon} v(\varepsilon) + \frac{1}{\varepsilon} (v(t+\varepsilon) - v(t)) \right) = -g + u(t).
\] (7.19)
Consequently $v(t) \in \mathcal{D}(A)$ and $Av(t) = -g + u(t)$ implying that $u(t)$ solves the following integral version of our abstract Cauchy problem

$$u(t) = g + A \int_0^t u(s) ds. \quad (7.20)$$

Note that while in the case of a bounded generator both versions are equivalent, this will not be the case in general. So while $u(t) = T(t)g$ always solves the integral version, it will only solve the differential version if $u(t) \in \mathcal{D}(A)$ for $t > 0$ (which is clearly also necessary for the differential version to make sense).

Two further consequences of these considerations are also worth while noticing:

**Corollary 7.9.** Let $T(t)$ be a $C_0$-semigroup with generator $A$. Then $A$ is a densely defined and closed operator.

**Proof.** Since $v(t) \in \mathcal{D}(A)$ and $\lim_{t \to 0} v(t) = g$ for arbitrary $g$, we see that $\mathcal{D}(A)$ is dense. Moreover, if $f_n \in \mathcal{D}(A)$ and $f_n \to f$, $Af_n \to g$ then

$$T(t)f_n - f_n = \int_0^t T(s)Af_n ds.$$ 

Taking $n \to \infty$ and dividing by $t$ we obtain

$$\frac{1}{t} (T(t)f - f) = \frac{1}{t} \int_0^t T(s)g ds.$$ 

Taking $t \downarrow 0$ finally shows $f \in \mathcal{D}(A)$ and $Af = g$. \hfill $\square$

Note that by the closed graph theorem we have $\mathcal{D}(A) = X$ if and only if $A$ is bounded. Moreover, since a $C_0$ semigroup provides the unique solution of the abstract Cauchy problem for $A$ we obtain

**Corollary 7.10.** A $C_0$-semigroup is uniquely determined by its generator.

**Proof.** Suppose $T$ and $S$ have the same generator $A$. Then by uniqueness for (7.17) we have $T(t)g = S(t)g$ for all $g \in \mathcal{D}(A)$. Since $\mathcal{D}(A)$ is dense this implies $T(t) = S(t)$ as both operators are continuous. \hfill $\square$

Finally, as in the uniformly continuous case, a solution $u(t) \in \mathcal{D}(A)$ of the inhomogeneous problem

$$\dot{u} = Au + f, \quad u(0) = g, \quad (7.21)$$

is necessarily given by

$$u(t) = T(t) \left( g + \int_0^t T(-s)f(s)ds \right) = T(t)g + \int_0^t T(t-s)f(s)ds. \quad (7.22)$$
But now it is not so clear when this will actually be a solution. One obvious condition being \( g \in \mathcal{D}(A) \) and \( f \in C([0, \infty), \mathcal{D}(A)) \).

**Problem 7.5.** Show that a uniformly continuous semigroup has a bounded generator. (Hint: Express \( T(t) \) in terms of \( V(t) := \int_0^t T(s)ds \).)

**Problem 7.6.** Let \( T(t) \) be a \( C_0 \)-semigroup. Show that if \( T(t_0) \) has a bounded inverse for one \( t_0 > 0 \) then it extends to a strongly continuous group.

**Problem 7.7.** Define a semigroup on \( L^1(-1, 1) \) via

\[
(T(t)f)(s) = \begin{cases} 
2f(s - t), & 0 < s \leq t, \\
f(s - t), & \text{else},
\end{cases}
\]

where we set \( f(s) = 0 \) for \( s < 0 \). Show that the estimate from Lemma 7.6 does not hold with \( M < 2 \).

### 7.4. Generator theorems

Of course in practice the abstract Cauchy problem, that is the operator \( A \), is given and the question is if \( A \) generates a corresponding \( C_0 \)-semigroup. Corollary 7.9 already gives us some necessary conditions but this alone is not enough.

It turns out that it is crucial to understand the resolvent of \( A \). As in the case of bounded operators (cf. Section 6.1) we define the resolvent set via

\[
\rho(A) := \{ z \in \mathbb{C} | A - z \text{ is bijective with a bounded inverse} \}
\]

and call

\[
R_A(z) := (A - z)^{-1}, \quad z \in \rho(A)
\]

the resolvent of \( A \). The complement \( \sigma(A) = \mathbb{C} \setminus \rho(A) \) is called the spectrum of \( A \). As in the case of Banach algebras it follows that the resolvent is analytic and that the resolvent set is open (Problem 7.9). However, the spectrum will no longer be bounded in general. Note that if \( A \) is closed, then bijectivity implies boundedness of the inverse (see Corollary 4.9).

Using an operator-valued version of the elementary integral \( \int_0^\infty e^{(a-z)t}dt = -(a-z)^{-1} \) (for \( \text{Re}(a-z) < 0 \)) we can make the connection between the resolvent and the semigroup.

**Lemma 7.11.** Let \( T \) be a semigroup with generator \( A \) satisfying (7.15). Then \( \{ z | \text{Re}(z) > \omega \} \subseteq \rho(A) \) and

\[
R_A(z) = -\int_0^\infty T(t)e^{-zt}dt, \quad \text{Re}(z) > \omega,
\]

were the right-hand side is defined as

\[
\left( \int_0^\infty e^{-zt}T(t)dt \right) f := \lim_{s \to \infty} \int_0^s e^{-zt}T(t)f dt.
\]
Moreover,
\[ \|R_A(z)\| \leq \frac{M}{\Re(z) - \omega}, \quad \Re(z) > \omega. \] (7.27)

**Proof.** Let us abbreviate \( R_s(z)f := -\int_0^s e^{-zt}T(t)f dt \). Then, by virtue of (7.15), \( \|e^{-zt}T(t)f\| \leq Me^{\omega - \Re(z)}\|f\| \) shows that \( R_s(z) \) is a bounded operator satisfying \( \|R_s(z)\| \leq M(\Re(z) - \omega)^{-1} \). Moreover, this estimates also shows that the limit \( R(z) := \lim_{s \to \infty} R_s(z) \) exists (and still satisfies \( \|R(z)\| \leq M(\Re(z) - \omega)^{-1} \)). Next note that \( S(t) = e^{-zt}T(t) \) is a semigroup with generator \( A - z \) and hence for \( f \in \mathcal{D}(A) \) we have
\[ R_s(z)(A - z)f = -\int_0^s S(t)(A - z)f dt = -\int_0^s \dot{S}(t)f dt = f - e^{-zt}T(t)f. \]
In particular, taking the limit \( s \to \infty \), we obtain \( R(z)(A - z)f = f \) for \( f \in \mathcal{D}(A) \). Similarly, by Problem 7.3
\[ (A - z)R_s(z)f = -\int_0^s (A - z)S(t)f dt = -\int_0^s \dot{S}(t)f dt = f - e^{-zt}T(t)f \]
and taking limits, using closedness of \( A \), implies \( (A - z)R(z)f = f \) for \( f \in \mathcal{D}(A) \). Finally, if \( f \in X \) choose \( f_n \in \mathcal{D}(A) \) with \( f_n \to f \). Then \( R(z)f_n \to f \) and \( (A - z)R(z)f_n = f_n \to f \) proving \( (A - z)R(z)f = f \) for \( f \in X \).

**Corollary 7.12.** Let \( T \) be a semigroup with generator \( A \) satisfying (7.15). Then
\[ R_A(z)^{n+1} = \frac{(-1)^{n+1}}{n!} \int_0^\infty t^n e^{-zt}T(t) dt, \quad \Re(z) > \omega, \]
(7.28)
and
\[ \|R_A(z)^n\| \leq \frac{M}{(\Re(z) - \omega)^n}, \quad \Re(z) > \omega, \ n \in \mathbb{N}. \]
(7.29)

**Proof.** Abbreviate \( R_n(z) := \int_0^\infty t^n e^{-zt}T(t) dt \) and note that
\[ \frac{R_n(z + \varepsilon) - R_n(z)}{\varepsilon} = -R_{n+1}(z) + \varepsilon \int_0^\infty t^{n+2} \phi(\varepsilon t)e^{-zt}T(t) dt \]
where \( |\phi(\varepsilon)| \lesssim \frac{1}{\varepsilon}e^{\varepsilon} \) from which we see \( \frac{d}{dz} R_n(z) = -R_{n+1}(z) \) and hence \( \frac{d^n}{dz^n} R_A(z) = -\frac{d^n}{dz^n} R_0(z) = (-1)^n R_n(z) \). Finally the first claim follows using \( R_A(z)^{n+1} = \frac{1}{n!} \frac{d^n}{dz^n} R_A(z) \) (Problem 7.10). Estimating the integral using (7.15) establishes the second claim.

Given these preparations we can now try to answer the question when \( A \) generates a semigroup. In fact, we will be constructive and obtain the
corresponding semigroup by approximation. To this end we introduce the **Yosida approximation**

\[ A_n := -nRAR_{A}(\omega + n) = -n - n(\omega + n)R_{A}(\omega + n) \in \mathcal{L}(A). \] (7.30)

Of course this is motivated by the fact that this is a valid approximation for numbers \( \lim_{n \to \infty} \frac{-n}{\omega - n} = 1 \). That we also get a valid approximation for operators is the content of the next lemma.

**Lemma 7.13.** Suppose \( A \) is a densely defined closed operator with \( (\omega, \infty) \subset \rho(A) \) satisfying

\[ \|R_{A}(\omega + n)\| \leq M n. \] (7.31)

Then

\[ \lim_{n \to \infty} -nR_{A}(\omega + n)f = f, \quad f \in X, \quad \lim_{n \to \infty} A_{n}f = Af, \quad f \in \mathcal{D}(A). \] (7.32)

**Proof.** If \( f \in \mathcal{D}(A) \) we have \( -nR_{A}(\omega + n)f = f - R_{A}(\omega + n)(A - \omega)f \) which shows \( -nR_{A}(\omega + n)f \to f \) if \( f \in \mathcal{D}(A) \). Since \( \mathcal{D}(A) \) is dense and \( \|nR_{A}(\omega + n)\| \leq M \) this even holds for all \( f \in X \). Moreover, for \( f \in \mathcal{D}(A) \) we have \( A_{n}f = -nR_{A}(\omega + n)f = -nR_{A}(\omega + n)(Af) \to Af \) by the first part. \( \square \)

Moreover, \( A_{n} \) can also be used to approximate the corresponding semigroup under suitable assumptions.

**Theorem 7.14 (Feller–Miyadera–Phillips).** A linear operator \( A \) is the generator of a \( C_{0} \)-semigroup satisfying (7.15) if and only if it is densely defined, closed, \( (\omega, \infty) \subset \rho(A) \), and

\[ \|R_{A}(\lambda)^{n}\| \leq \frac{M}{(\lambda - \omega)^{n}}, \quad \lambda > \omega, \ n \in \mathbb{N}. \] (7.33)

**Proof.** Necessity has already been established in Corollaries 7.9 and 7.12.

For the converse we use the semigroups

\[ T_{n}(t) := \exp(tA_{n}) \]

corresponding to the Yosida approximation (7.30). We note (using \( e^{A+B} = e^{A}e^{B} \) for commuting operators \( A, B \))

\[ \|T_{n}(t)\| \leq e^{-nt} \sum_{j=0}^{\infty} \frac{(tn(\omega + n))^{j}}{j!} \|R_{A}(\omega + n)^{j}\| \leq Me^{-tn}e^{t(\omega + n)} = Me^{\omega t}. \]

Moreover, since \( R_{A}(\omega + m) \) and \( R_{A}(\omega + n) \) commute by the first resolvent identity (Problem 7.10), we conclude that the same is true for \( A_{m}, A_{n} \) as well.
as for $T_m(t), T_n(t)$ (by the very definition as a power series). Consequently

$$
\|T_n(t)f - T_m(t)f\| = \left\| \int_0^1 \frac{d}{ds} T_n(st) T_m((1-s)t)f \, ds \right\|
\leq t \int_0^1 \|T_n(st) T_m((1-s)t)(A_n - A_m)f\| \, ds
\leq t M^2 e^{\omega t} \|(A_n - A_m)f\|.
$$

Together with the bound $\|T_n\| \leq M e^{\omega t}$ this shows that $T_n(t)$ is a Cauchy sequence in $C([0, T], \mathfrak{L}(X))$ and hence we can define $T(t) := \lim_{n \to \infty} T_n(t)$. Moreover, this limit preserves the semigroup property and thus $T(t)$ is a semigroup satisfying (7.15). It remains to show that $A$ is its generator. To this end let $f \in D(A)$, then

$$
T(t)f - f = \lim_{n \to \infty} T_n(t)f - f = \lim_{n \to \infty} \int_0^t T_n(s) A_n f \, ds
= \lim_{n \to \infty} \left( \int_0^t T_n(s) A f \, ds + \int_0^t T_n(s) (A_n - A) f \, ds \right)
= \int_0^t T(s) A f \, ds
$$

which shows $\lim_{t \downarrow 0} \frac{1}{t} (T(t)f - f) = Af$ for $f \in D(A)$. Finally, note that the domain of the generator cannot be larger, since $A - \omega - 1$ is bijective and adding a vector to its domain would destroy injectivity. But then $\omega + 1$ would not be in the resolvent set contradicting Lemma 7.11.

Note that in combination with the following lemma this also answers the question when $A$ generates a $C_0$-group.

**Lemma 7.15.** An operator $A$ generates a $C_0$-group if and only if both $A$ and $-A$ generate $C_0$-semigroups.

**Proof.** Clearly, if $A$ generates a $C_0$-group $T(t)$, then $S(t) := T(-t)$ is a $C_0$-group with generator $-A$. Conversely, let $T(t), S(t)$ be the $C_0$-semigroups generated by $A, -A$, respectively. Then a short calculation shows

$$
\frac{d}{dt} T(t) S(t) g = -T(t) A S(t) g + T(t) A S(t) g = 0, \quad t \geq 0.
$$

Consequently, $T(t) S(t) = T(0) S(0) = I$ and similarly $S(t) T(t) = I$, that is, $S(t) = T(t)^{-1}$. Hence it is straightforward to check that $T$ extends to a group via $T(-t) := S(t), t \geq 0$.

The following examples show that the spectral conditions are indeed crucial. Moreover, they also show that an operator might give rise to a Cauchy problem which is uniquely solvable for a dense set of initial conditions, without generating a strongly continuous semigroup.
Example. Let
\[ A = \begin{pmatrix} 0 & A_0 \\ 0 & 0 \end{pmatrix}, \quad \mathcal{D}(A) = X \times \mathcal{D}(A_0). \]

Then \( u(t) = (e^{tA_0}) (f_0) = (f_0 + tA_0 f_1) \) is the unique solution of the corresponding abstract Cauchy problem for given \( f \in \mathcal{D}(A) \). Nevertheless, if \( A_0 \) is unbounded, the corresponding semigroup is not strongly continuous.

Note that in this case we have \( \sigma(A) = \{0\} \) if \( A_0 \) is bounded and \( \sigma(A) = \mathbb{C} \) else. In fact, since \( A \) is not injective we must have \( \{0\} \subseteq \sigma(A) \). For \( z \neq 0 \) the inverse of \( A - z \) is given by
\[
(A - z)^{-1} = -\frac{1}{z} \begin{pmatrix} 1 & \frac{1}{z} A_0 \\ 0 & 1 \end{pmatrix}, \quad \mathcal{D}((A - z)^{-1}) = \text{Ran}(A - z) = X \times \mathcal{D}(A_0),
\]
which is bounded if and only if \( A \) is bounded.

Example. Let \( X_0 = C_0(\mathbb{R}) \) and \( m(x) = ix \). Then we can regard \( m \) as a multiplication operator on \( X_0 \) when defined maximally, that is, \( f \mapsto mf \) with \( \mathcal{D}(m) = \{ f \in X_0 | mf \in X_0 \} \). Note that since \( C_c(\mathbb{R}) \subseteq \mathcal{D}(m) \) we see that \( m \) is densely defined. Moreover, it is easy to check that \( m \) is closed.

Now consider \( X = X_0 \otimes X_0 \) with \( \|f\| = \max(\|f_0\|, \|f_1\|) \) and note that we can also regard \( M \) as an operator on \( X \) by applying it to both components. Then
\[ A = \begin{pmatrix} m & m \\ 0 & m \end{pmatrix}, \quad \mathcal{D}(A) = \mathcal{D}(m) \oplus \mathcal{D}(m), \]
Moreover, for \( z \not\in i\mathbb{R} \) the resolvent is given by the multiplication operator
\[ R_A(z) = -\frac{1}{m - z} \begin{pmatrix} 1 & -\frac{m}{m-z} \\ 0 & 1 \end{pmatrix}. \]

For \( \lambda > 0 \) we compute
\[
\|R_A(\lambda)f\| \leq \left( \sup_{x \in \mathbb{R}} \frac{1}{|ix - \lambda|} + \sup_{x \in \mathbb{R}} \frac{|x|}{|ix - \lambda|^2} \right) \|f\| = \frac{3}{2\lambda} \|f\|
\]
and hence \( A \) satisfies (7.33) with \( M = \frac{3}{2}, \omega = 0 \) and \( n = 1 \). However, by
\[
\|R_A(\lambda + in)\| \geq \|R_A(\lambda + in)(0, f_n)\| \geq \left| \frac{\inf f_n(n)}{(\lambda - in + in)^2} \right| = \frac{n}{\lambda^2},
\]
where \( f_n \) is chosen such that \( f_n(n) = 1 \) and \( \|f_n\|_\infty = 1 \), it does not satisfy (7.29). Hence \( A \) does not generate a \( C_0 \)-semigroup. Indeed, the solution of the corresponding Cauchy problem is
\[ T(t) = e^{tm} \begin{pmatrix} 1 & tm \\ 0 & 1 \end{pmatrix}, \quad \mathcal{D}(T) = X_0 \oplus \mathcal{D}(m), \]
which is unbounded.
Finally we look at the special case of contraction semigroups satisfying
\[ \|T(t)\| \leq 1. \] (7.34)
By a simple transform the case \( M = 1 \) in Lemma 7.6 can always be reduced to this case (Problem 7.8). Moreover, observe, that in the case \( M = 1 \) the estimate (7.27) implies the general estimate (7.29).

**Corollary 7.16** (Hille–Yosida). A linear operator \( A \) is the generator of a contraction semigroup if and only if it is densely defined, closed, \((0, \infty) \subseteq \rho(A)\), and
\[ \|R_A(\lambda)\| \leq \frac{1}{\lambda}, \quad \lambda > 0. \] (7.35)

**Example.** If \( A \) is the generator of a contraction, then clearly all eigenvalues \( z \) must satisfy \( \text{Re}(z) \leq 0 \). Moreover, for
\[ A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \]
we have
\[ R_A(z) = -\frac{1}{z} \begin{pmatrix} 1 & 1/z \\ 0 & 1 \end{pmatrix}, \quad T(t) = \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}, \]
which shows that the bound on the resolvent is crucial. ◊

However, for a given operator even the simple estimate (7.35) might be difficult to establish directly. Hence we outline another criterion.

**Example.** Let \( X \) be a Hilbert space and observe that for a contraction semigroup the expression \( \|T(t)f\| \) must be nonincreasing. Consequently, for \( f \in D(A) \) we must have
\[ \frac{d}{dt}\|T(t)f\|^2\bigg|_{t=0} = 2\text{Re}(\langle f, Af \rangle) \leq 0. \]
Operators satisfying \( \text{Re}(\langle f, Af \rangle) \leq 0 \) are called dissipative and this clearly suggests to replace the resolvent estimate by dissipativity. ◊

To formulate this condition for Banach spaces, we first introduce the **duality set**
\[ \mathcal{J}(x) := \{ x' \in X^* | x'(x) = \|x\|^2 = \|x'\|^2 \} \] (7.36)
of a given vector \( x \in X \). In other words, the elements from \( \mathcal{J}(x) \) are those linear functionals which attain their norm at \( x \) and are normalized to have the same norm as \( x \). As a consequence of the Hahn–Banach theorem (Corollary 4.15) note that \( \mathcal{J}(x) \) is nonempty. Moreover, it is also easy to see that \( \mathcal{J}(x) \) is convex and weak-* closed.

**Example.** Let \( X \) be a Hilbert space and identify \( X \) with \( X^* \) via \( x \mapsto \langle x, . \rangle \) as usual. Then \( \mathcal{J}(x) = \{ x \} \). Indeed since we have equality \( \langle x', x \rangle = \|x'\|\|x\| \)
in the Cauchy–Schwarz inequality, we must have $x' = \alpha x$ for some $\alpha \in \mathbb{C}$ with $|\alpha| = 1$ and $\alpha \|x\|^2 = \langle x', x \rangle = \|x\|^2$ shows $\alpha = 1$.

**Example.** If $X^*$ is strictly convex (cf. Problem 1.12), then the duality set contains only one point. In fact, suppose $x', y' \in \mathcal{J}(x)$, then $z' = \frac{1}{2}(x' + y') \in \mathcal{J}(x)$ and $\frac{1}{2}\|x' + y'\| = \|z'\| = \frac{1}{2}(\|x'\| + \|y'\|)$ implying $x' = y'$ by strict convexity.

**Example.** Let $X = C[0,1]$ and choose $x \in X$. If $t_0$ is chosen such that $|x(t_0)| = \|x\|$, then the functional $y \mapsto x'(y) := x(t_0)y(t_0)$ satisfies $x' \in \mathcal{J}(x)$. Clearly $\mathcal{J}(x)$ will contain more than one element in general.

Now a given operator $\mathcal{D}(A) \subseteq X \to X$ is called **dissipative** if

$$\text{Re}(x'(A(x))) \leq 0 \quad \text{for one } x' \in \mathcal{J}(x) \text{ and all } x \in \mathcal{D}(A).$$  \hspace{1cm} (7.37)

**Lemma 7.17.** Let $x, y \in X$. Then $\|x\| \leq \|x - \alpha y\|$ for all $\alpha > 0$ if and only if there is an $x' \in \mathcal{J}(x)$ such that $\text{Re}(x'(y)) \leq 0$.

**Proof.** Without loss of generality we can assume $x \neq 0$. If $\text{Re}(x'(y)) \leq 0$ for some $x' \in \mathcal{J}(x)$, then for $\alpha > 0$ we have

$$\|x\|^2 = x'(x) \leq \text{Re}(x'(x - \alpha y)) \leq \|x'\| \|x - \alpha y\|$$

implying $\|x\| \leq \|x - \alpha y\|$.

Conversely, if $\|x\| \leq \|x - \alpha y\|$ for all $\alpha > 0$, let $x'_\alpha \in \mathcal{J}(x - \alpha y)$ and set $y'_\alpha = \|x'_\alpha\|^{-1}x'_\alpha$. Then

$$\|x\| \leq \|x - \alpha y\| = y'_\alpha(x - \alpha y) = \text{Re}(y'_\alpha(x)) - \alpha \text{Re}(y'_\alpha(y)) \leq \|x\| - \alpha \text{Re}(y'_\alpha(y)).$$

Now by the Banach–Alaoglu theorem we can choose a subsequence $y'_{i/n_j} \to y_0$ in the weak-* sense. (Note that the use of the Banach–Alaoglu theorem could be avoided by restricting $y'_\alpha$ to the two dimensional subspace spanned by $x, y$, passing to the limit in this subspace and then extending the limit to $X^*$ using Hahn–Banach.) Consequently $\text{Re}(y'_0(y)) \leq 0$ and $\text{Re}(y'_0(x)) \geq \|x\|$. Whence $x'_0 = y'_0 \|x\| \in \mathcal{J}(x)$ and $\text{Re}(x'_0(y)) \leq 0$.

As a straightforward consequence we obtain:

**Corollary 7.18.** A linear operator is dissipative if and only if

$$\|(A - \lambda)f\| \geq \lambda \|f\|, \quad \lambda > 0, \: f \in \mathcal{D}(A).$$  \hspace{1cm} (7.38)

In particular, for a dissipative operator $A - \lambda$ is injective for $\lambda > 0$ and $(A - \lambda)^{-1}$ is bounded with $\|(A - \lambda)^{-1}\| \leq \lambda^{-1}$. However, this does not imply that $\lambda$ is in the resolvent set of $A$ since $\mathcal{D}((A - \lambda)^{-1}) = \text{Ran}(A - \lambda)$ might not be all of $X$. 

Now we are ready to show

**Theorem 7.19** (Lumer–Phillips). A linear operator $A$ is the generator of a contraction semigroup if and only if it is densely defined, dissipative, and $A - \lambda_0$ is surjective for one $\lambda_0 > 0$. Moreover, in this case (7.37) holds for all $x' \in \mathcal{J}(x)$.

**Proof.** Let $A$ generate a contraction semigroup $T(t)$ and let $x \in \mathcal{D}(A)$, $x' \in \mathcal{J}(x)$. Then

$$\text{Re}(x'(T(t)x - x)) \leq |x'(T(t)x)| - \|x\|^2 \leq \|x'||x|| - \|x\|^2 = 0$$

and dividing by $t$ and letting $t \downarrow 0$ shows $\text{Re}(x'(Ax)) \leq 0$. Hence $A$ is dissipative and by Corollary 7.16 $(0, \infty) \subseteq \rho(A)$, that is $A - \lambda$ is bijective for $\lambda > 0$.

Conversely, by Corollary 7.18 $A - \lambda$ has a bounded inverse satisfying $\|(A - \lambda)^{-1}\|$ for all $\lambda > 0$. In particular, for $\lambda_0$ the inverse is defined on all of $X$ and hence closed. Thus $A$ is also closed and $\lambda_0 \in \rho(A)$. Moreover, from $\|R_A(\lambda_0)\| \leq \lambda_0^{-1}$ (cf. Problem 7.9) we even get $(0,2\lambda_0) \subseteq \rho(A)$ and iterating this argument shows $(0, \infty) \subseteq \rho(A)$ as well as $\|R_A(\lambda)\| \leq \lambda^{-1}$, $\lambda > 0$. Hence the requirements from Corollary 7.16 are satisfied. \[\square\]

Note that generators of contraction semigroups are maximal dissipative in the sense that they do not have any dissipative extensions. In fact, if we extend $A$ to a larger domain we must destroy injectivity of $A - \lambda$ and thus the extension cannot be dissipative.

**Example.** Let $X = C[0,1]$ and consider the one-dimensional heat equation

$$\frac{\partial}{\partial t} u(t, x) = \frac{\partial^2}{\partial x^2} u(t, x)$$

on a finite interval $x \in [0,1]$ with the boundary conditions $u(0) = u(1) = 0$ and the initial condition $u(0, x) = u_0(x)$. The corresponding operator is

$$Af = f'', \quad \mathcal{D}(A) = \{f \in C^2[0,1]|f(0) = f(1) = 0\} \subseteq C[0,1].$$

For $\ell \in \mathcal{J}(f)$ we can choose $\ell(g) = f(x_0)^*g(x_0)$ where $x_0$ is chosen such that $|f(x_0)| = \|f\|_{\infty}$. Then $\text{Re}(f(x_0)^*f(x))$ has a global maximum at $x = x_0$ and if $f \in C^2[0,1]$ we must have $\text{Re}(f(x_0)^*f''(x)) \leq 0$ provided this maximum is in the interior of $(0,1)$. Consequently $\text{Re}(f(x_0)^*f''(x_0)) \leq 0$, that is, $A$ is dissipative. That $A - \lambda$ is surjective follows using the Green’s function as in Section 3.3. \[\diamond\]

Finally, we note that the condition that $A - \lambda_0$ is surjective can be weakened to the condition that $\text{Ran}(A - \lambda_0)$ is dense. To this end we need:

**Lemma 7.20.** Suppose $A$ is a densely defined dissipative operator. Then $A$ is closable and the closure $\overline{A}$ is again dissipative.
Proof. Recall that $A$ is closable if and only if for every $x_n \in \mathcal{D}(A)$ with $x_n \to 0$ and $Ax_n \to y$ we have $y = 0$. So let $x_n$ be such a sequence and chose another sequence $y_n \in \mathcal{D}(A)$ such that $y_n \to y$ (which is possible since $\mathcal{D}(A)$ is assumed dense). Then by dissipativity (specifically Corollary 7.18)
\[ \| (A - \lambda)(\lambda x_n - y_m) \| \geq \lambda \|\lambda x_n - y_m\|, \quad \lambda > 0 \]
and letting $n \to \infty$ and dividing by $\lambda$ shows
\[ \| y - (\lambda^{-1} A - 1)y_m \| \geq \| y_m \|. \]
Finally $\lambda \to \infty$ implies $\| y - y_m \| \geq \| y_m \|$ and $m \to \infty$ yields $0 \geq \| y \|$, that is, $y = 0$ and $A$ is closable. To see that $\overline{A}$ is dissipative choose $x \in \mathcal{D}(A)$ and $x_n \in \mathcal{D}(A)$ with $x_n \to x$ and $Ax_n \to \overline{A}x$. Then (again using Corollary 7.18) taking the limit in $\| (A - \lambda)x_n \| \geq \lambda \| x_n \|$ shows $\| (\overline{A} - \lambda)x \| \geq \lambda \| x \|$ as required. \[\square\]

Consequently:

Corollary 7.21. Suppose the linear operator $A$ is densely defined, dissipative, and $\operatorname{Ran}(A - \lambda_0)$ is dense for one $\lambda_0 > 0$. Then $A$ is closable and $\overline{A}$ is the generator of a contraction semigroup.

Proof. By the previous lemma $A$ is closable with $\overline{A}$ again dissipative. In particular, $\overline{A}$ is injective and by Lemma 4.8 we have $(\overline{A} - \lambda_0)^{-1} = (A - \lambda_0)^{-1}$. Since $(A - \lambda_0)^{-1}$ is bounded its closure is defined on the closure of its domain, that is, $\operatorname{Ran}(\overline{A} - \lambda_0) = \operatorname{Ran}(A - \lambda_0) = X$. The rest follows from the Lumer–Phillips theorem. \[\square\]

Problem 7.8. Let $T(t)$ be a $C_0$-semigroup and $\alpha > 0$, $\lambda \in \mathbb{C}$. Show that $S(t) = e^{\alpha t}T(\alpha t)$ is a $C_0$-semigroup with generator $B = \alpha A + \lambda$, $\mathcal{D}(B) = \mathcal{D}(A)$.

Problem 7.9. Let $A$ be a closed operator. Show that if $z_0 \in \rho(A)$, then
\[ R_A(z) = \sum_{n=0}^{\infty} (z - z_0)^n R_A(z_0)^{n+1}, \quad |z - z_0| < \| R_A(z_0) \|^{-1}. \]
In particular, the resolvent is analytic and
\[ \| (A - z)^{-1} \| \geq \frac{1}{\operatorname{dist}(z, \sigma(A))}. \]

Problem 7.10. Let $A$ be a closed operator. Show the first resolvent identity
\[ R_A(z_0) - R_A(z_1) = (z_0 - z_1)R_A(z_0)R_A(z_1) \]
\[ = (z_0 - z_1)R_A(z_1)R_A(z_0), \]
for $z_0, z_1 \in \rho(A)$. Moreover, conclude
\[ \frac{d^n}{dz^n} R_A(z) = n! R_A(z)^{n+1}, \quad \frac{d}{dz} R_A(z)^n = n R_A(z)^{n+1}. \]
Part 2

Real Analysis
8.1. The problem of measuring sets

The Riemann integral starts straight with the definition of the integral by considering functions which can be sandwiched between step functions. This is based on the idea that for a function defined on an interval (or a rectangle in higher dimensions) the domain can be easily subdivided into smaller intervals (or rectangles). Moreover, for nice functions the variation of the values (difference between maximum and minimum) should decrease with the length of the intervals. Of course, this fails for rough functions whose variations cannot be controlled by subdividing the domain into sets of decreasing size. The Lebesgue integral remedies this by subdividing the range of the function. This shifts the problem from controlling the variations of the function to defining the content of the preimage of the subdivisions for the range. Note that this problem does not occur in the Riemann approach since only the length of an interval (or the area of an rectangle) is needed. Consequently, the outset of Lebesgue theory is the problem of defining the content for a sufficiently large class of sets.

The Riemann-style approach to this problem in $\mathbb{R}^n$ is to start with a big rectangle containing the set under consideration and then take subdivisions thereby approximating the measure of the set from the inside and outside by the measure of the rectangles which lie inside and those which cover the set, respectively. If the difference tends to zero, the set is called measurable and the common limit is it measure.

To this end let $\mathcal{S}^n$ be the set of all half-closed rectangles of the form $[a,b] := (a_1, b_1] \times \cdots \times [a_n, b_n] \subseteq \mathbb{R}^n$ with $a < b$ augmented by the empty set. Here $a < b$ should be read as $a_j < b_j$ for all $1 \leq j \leq n$ (and similarly for
Moreover, we allow the intervals to be unbounded, that is, \( a, b \in \mathbb{R}^n \) (with \( \mathbb{R} = \mathbb{R} \cup \{-\infty, +\infty\} \)).

Of course one could as well take open or closed rectangles but our half-closed rectangles have the advantage that they tile nicely. In particular, one can subdivide a half-closed rectangle into smaller ones without gaps or overlap.

This collection has some interesting algebraic properties: A collection of subsets \( S \) of a given set \( X \) is called a **semialgebra** if

- \( \emptyset \in S \),
- \( S \) is closed under finite intersections, and
- the complement of a set in \( S \) can be written as a finite union of sets from \( S \).

A semialgebra \( A \) which is closed under complements is called an **algebra**, that is,

- \( \emptyset \in A \),
- \( A \) is closed under finite intersections, and
- \( A \) is closed under complements.

Note that \( X \in A \) and that \( A \) is also closed under finite unions and relative complements: \( X = \emptyset' \), \( A \cup B = (A' \cap B')' \) (De Morgan), and \( A \setminus B = A \cap B' \), where \( A' = X \setminus A \) denotes the complement.

**Example.** Let \( X := \{1, 2, 3\} \), then \( A := \{\emptyset, \{1\}, \{2, 3\}, X\} \) is an algebra.

In the sequel we will frequently meet unions of disjoint sets and hence we will introduce the following short hand notation for the union of mutually disjoint sets:

\[
\bigcup_{j \in J} A_j := \bigcup_{j \in J} A_j \quad \text{with} \quad A_j \cap A_k = \emptyset \text{ for all } j \neq k.
\]

In fact, considering finite disjoint unions from a semialgebra we always have a corresponding algebra.

**Lemma 8.1.** Let \( S \) be a semialgebra, then the set of all finite disjoint unions 
\( \tilde{S} := \{\bigcup_{j=1}^n A_j \mid A_j \in S\} \) is an algebra.

**Proof.** Suppose \( A = \bigcup_{j=1}^n A_j \in \tilde{S} \) and \( B = \bigcup_{k=1}^m B_k \in \tilde{S} \). Then \( A \cap B = \bigcup_{j,k} (A_j \cap B_k) \in \tilde{S} \). Concerning complements we have \( A' = \bigcap_j A'_j \in \tilde{S} \) since \( A'_j \in \tilde{S} \) by definition of a semialgebra and since \( \tilde{S} \) is closed under finite intersections by the first part.

**Example.** The collection of all intervals (augmented by the empty set) clearly is a semialgebra. Moreover, the collection \( S^1 \) of half-open intervals
of the form \((a, b] \subseteq \mathbb{R}, -\infty \leq a < b \leq \infty\) augmented by the empty set is a semialgebra. Since the product of semialgebras is again a semialgebra (Problem 8.1), the same is true for the collection of rectangles \(S^n\). \(\Diamond\)

The somewhat dissatisfactory situation with the Riemann integral alluded to before led Cantor, Peano, and particularly Jordan to the following attempt of measuring arbitrary sets: Define the measure of a rectangle via

\[(a, b] = \prod_{j=1}^{n} (b_j - a_j). \tag{8.1}\]

Note that the measure will be infinite if the rectangle is unbounded. Furthermore, define the \textbf{inner, outer Jordan content} of a set \(A \subseteq \mathbb{R}^n\) as

\[J_*(A) := \sup \left\{ \sum_{j=1}^{m} |R_j| \left| \bigcup_{j=1}^{m} R_j \subseteq A, R_j \in S^n \right. \right\}, \tag{8.2}\]

\[J^*(A) := \inf \left\{ \sum_{j=1}^{m} |R_j| \left| A \subseteq \bigcup_{j=1}^{m} R_j, R_j \in S^n \right. \right\}, \tag{8.3}\]

respectively. If \(J_*(A) = J^*(A)\) the set \(A\) is called \textbf{Jordan measurable}.

Unfortunately this approach turned out to have several shortcomings (essentially identical to those of the Riemann integral). Its limitation stems from the fact that one only allows finite covers. Switching to countable covers will produce the much more flexible Lebesgue measure.

\textbf{Example.} To understand this limitation let us look at the classical example of a non Riemann integrable function, the characteristic function of the rational numbers inside \([0, 1]\). If we want to cover \(\mathbb{Q} \cap [0, 1]\) by a finite number of intervals, we always end up covering all of \([0, 1]\) since the rational numbers are dense. Conversely, if we want to find the inner content, no single (nontrivial) interval will fit into our set since it has empty interior. In summary, \(J_*(\mathbb{Q} \cap [0, 1]) = 0 \neq 1 = J^*(\mathbb{Q} \cap [0, 1])\).

On the other hand, if we are allowed to take a countable number of intervals, we can enumerate the points in \(\mathbb{Q} \cap [0, 1]\) and cover the \(j\)'th point by an interval of length \(\varepsilon 2^{-j}\) such that the total length of this cover is less than \(\varepsilon\), which can be arbitrarily small. \(\Diamond\)

The previous example also hints at what is going on in general. When computing the outer content you will always end up covering the closure of \(A\) and when computing the inner content you will never get more than the interior of \(A\). Hence a set should be Jordan measurable if the difference between the closure and the interior, which is by definition the boundary \(\partial A = \overline{A} \setminus A^\circ\), is small. However, we do not want to pursue this further at this point and hence we defer it to Appendix 8.7.
Rather we will make the anticipated change and define the **Lebesgue outer measure** via

\[
\lambda^{n,*}(A) := \inf \left\{ \sum_{j=1}^{\infty} |R_j| \left| A \subseteq \bigcup_{j=1}^{\infty} R_j, \, R_j \in S^n \right. \right\}.
\] (8.4)

In particular, we will call \( N \) a Lebesgue **null set** if \( \lambda^{n,*}(N) = 0 \).

Using the fact that \( \tilde{A}_n = A_n \setminus \bigcup_{m=1}^{n-1} A_m \in S^n \) we see that we could even require the covers to be disjoint:

\[
\lambda^{n,*}(A) = \inf \left\{ \sum_{j=1}^{\infty} |R_j| \left| A \subseteq \bigcup_{j=1}^{\infty} R_j, \, R_j \in S^n \right. \right\}.
\] (8.5)

**Example.** As shown in the previous example, the set of rational numbers inside \([0,1]\) is a null set. In fact, the same argument shows that every countable set is a null set.

Consequently we expect the irrational numbers inside \([0,1]\) to be a set of measure one. But if we try to approximate this set from the inside by half-closed intervals we are bound to fail as no single (nonempty) interval will fit into this set. This explains why we did not define a corresponding inner measure.

The construction in (8.4) appears frequently and hence it is well worth to collect some properties in more generality: A function \( \mu^* : \mathcal{P}(X) \to [0, \infty] \) is an **outer measure** if it has the properties

- \( \mu^*(\emptyset) = 0 \),
- \( A \subseteq B \Rightarrow \mu^*(A) \leq \mu^*(B) \) (monotonicity), and
- \( \mu^*(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} \mu^*(A_n) \) (subadditivity).

Here \( \mathcal{P}(X) \) is the power set (i.e., the collection of all subsets) of \( X \). The following lemma shows that Lebesgue outer measure deserves its name.

**Lemma 8.2.** Let \( \mathcal{E} \) be some family of subsets of \( X \) containing \( \emptyset \). Suppose we have a set function \( \rho : \mathcal{E} \to [0, \infty] \) such that \( \rho(\emptyset) = 0 \). Then

\[
\mu^*(A) := \inf \left\{ \sum_{j=1}^{\infty} \rho(A_j) \left| A \subseteq \bigcup_{j=1}^{\infty} A_j, \, A_j \in \mathcal{E} \right. \right\}
\]

is an outer measure. Here the infimum extends over all countable covers from \( \mathcal{E} \) with the convention that the infimum is infinite if no such cover exists.

**Proof.** \( \mu^*(\emptyset) = 0 \) is trivial since we can choose \( A_j = \emptyset \) as a cover.

To see see monotonicity let \( A \subseteq B \) and note that if \( \{A_j\} \) is a cover for \( B \) then it is also a cover for \( A \) (if there is no cover for \( B \), there is nothing to
8.1. The problem of measuring sets

Hence
\[ \mu^*(A) \leq \sum_{j=1}^{\infty} \rho(A_j) \]
and taking the infimum over all covers for \( B \) shows \( \mu^*(A) \leq \mu^*(B) \).

To see subadditivity note that we can assume that all sets \( A_j \) have a cover (otherwise there is nothing to do) \( \{B_{jk}\}_{k=1}^{\infty} \) for \( A_j \) such that
\[ \sum_{k=1}^{\infty} \rho(B_{jk}) \leq \mu^*(A_j) + \frac{\varepsilon}{2^j}. \]
Since \( \{B_{jk}\}_{j,k=1}^{\infty} \) is a cover for \( \bigcup_j A_j \) we obtain
\[ \mu^*(A) \leq \sum_{j,k=1}^{\infty} \rho(B_{jk}) \leq \sum_{j=1}^{\infty} \mu^*(A_j) + \varepsilon \]
and since \( \varepsilon > 0 \) is arbitrary subadditivity follows.

As a consequence note that null sets \( N \) (i.e., \( \mu^*(N) = 0 \)) do not change the outer measure: \( \mu^*(A) \leq \mu^*(A \cup N) \leq \mu^*(A) + \mu^*(N) = \mu^*(A) \).

So we have defined Lebesgue outer measure and we have seen that it has some basic properties. Moreover, there are some further properties. For example, let \( f(x) = Mx + a \) be an affine transformation, then
\[ \lambda^{|n,*}(MA + a) = \det(M) \lambda^{|n,*}(A). \] (8.6)
In fact, that translations do not change the outer measure is immediate since \( S^n \) is invariant under translations and the same is true for \( |R| \). Moreover, every matrix can be written as \( M = O_1 DO_2 \), where \( O_j \) are orthogonal and \( D \) is diagonal (Problem 9.21). So it reduces the problem to showing this for diagonal matrices and for orthogonal matrices. The case of diagonal matrices follows as before but the case of orthogonal matrices is more involved (it can be shown by showing that rectangles can be replaced by open balls in the definition of the outer measure). Hence we postpone this to Section 8.8.

For now we will use this fact only to explain why our outer measure is still not good enough. The reason is that it lacks one key property, namely additivity! Of course this will be crucial for the corresponding integral to be linear and hence is indispensable. Now here comes the bad news: A classical paradox by Banach and Tarski shows that one can break the unit ball in \( \mathbb{R}^3 \) into a finite number of (wild – choosing the pieces uses the Axiom of Choice and cannot be done with a jigsaw;-) pieces, and reassemble them using only rotations and translations to get two copies of the unit ball. Hence our outer measure (as well as any other reasonable notion of size which is translation and rotation invariant) cannot be additive when defined for all sets! If you think that the situation in one dimension is better, I have to disappoint you as well: Problem 8.2.

So our only hope left is that additivity at least holds on a suitable class of sets. In fact, even finite additivity is not sufficient for us since limiting
operations will require that we are able to handle countable operations. To this end we will introduce some abstract concepts first.

A set function \( \mu : \mathcal{A} \rightarrow [0, \infty] \) on an algebra is called a premeasure if it satisfies

- \( \mu(\emptyset) = 0 \),
- \( \mu(\bigcup_{j=1}^{\infty} A_j) = \sum_{j=1}^{\infty} \mu(A_j), \) if \( A_j \in \mathcal{A} \) and \( \bigcup_{j=1}^{\infty} A_j \in \mathcal{A} \) (\( \sigma \)-additivity).

Here the sum is set equal to \( \infty \) if one of the summands is \( \infty \) or if it diverges.

The following lemma gives conditions when the natural extension of a set function on a semialgebra \( \mathcal{S} \) to its associated algebra \( \tilde{\mathcal{S}} \) will be a premeasure.

**Lemma 8.3.** Let \( \mathcal{S} \) be a semialgebra and let \( \mu : \mathcal{S} \rightarrow [0, \infty] \) be additive, that is, \( A = \bigcup_{j=1}^{n} A_j \) with \( A, A_j \in \mathcal{S} \) implies \( \mu(A) = \sum_{j=1}^{n} \mu(A_j) \). Then the natural extension \( \mu : \tilde{\mathcal{S}} \rightarrow [0, \infty] \) given by

\[
\mu(A) := \sum_{j=1}^{n} \mu(A_j), \quad A = \bigcup_{j=1}^{n} A_j,
\]

is (well defined and) additive on \( \tilde{\mathcal{S}} \). Moreover, it will be a premeasure if

\[
\mu(\bigcup_{j=1}^{\infty} A_j) \leq \sum_{j=1}^{\infty} \mu(A_j)
\]

whenever \( \bigcup_{j} A_j \in \mathcal{S} \) and \( A_j \in \mathcal{S} \).

**Proof.** We begin by showing that \( \mu \) is well defined. To this end let \( A = \bigcup_{j=1}^{n} A_j = \bigcup_{k=1}^{m} B_k \) and set \( C_{jk} := A_j \cap B_k \). Then

\[
\sum_{j} \mu(A_j) = \sum_{j} \mu(\bigcup_{k} C_{jk}) = \sum_{j,k} \mu(C_{jk}) = \sum_{k} \mu(\bigcup_{j} C_{jk}) = \sum_{k} \mu(B_k)
\]

by additivity on \( \mathcal{S} \). Moreover, if \( A = \bigcup_{j=1}^{n} A_j \) and \( B = \bigcup_{k=1}^{m} B_k \) are two disjoint sets from \( \tilde{\mathcal{S}} \), then

\[
\mu(A \cup B) = \mu\left( \bigcup_{j=1}^{n} A_j \cup \bigcup_{k=1}^{m} B_k \right) = \sum_{j} \mu(A_j) + \sum_{k} \mu(B_k) = \mu(A) + \mu(B)
\]

which establishes additivity. Finally, let \( A = \bigcup_{j=1}^{\infty} A_j \in \mathcal{S} \) with \( A_j \in \mathcal{S} \) and observe \( B_n = \bigcup_{j=1}^{n} A_j \in \tilde{\mathcal{S}} \). Hence

\[
\sum_{j=1}^{n} \mu(A_j) = \mu(B_n) \leq \mu(B_n) + \mu(A \setminus B_n) = \mu(A)
\]
8.1. The problem of measuring sets

and combining this with our assumption (8.8) shows \( \sigma \)-additivity when all sets are from \( \mathcal{S} \). By finite additivity this extends to the case of sets from \( \bar{\mathcal{S}} \).

\[ \square \]

In fact, our set function \(|R|\) for rectangles is easily seen to be a premeasure.

**Lemma 8.4.** The set function (8.1) for rectangles extends to a premeasure on \( \bar{\mathcal{S}}^n \).

**Proof.** Finite additivity is left at an exercise (see the proof of Lemma 8.12) and it remains to verify (8.8). We can cover each \( A_j := (a_j, b_j] \) by some slightly larger rectangle \( B_j := (a^j, b^j + \delta^j) \) such that \(|B_j| \leq |A_j| + \frac{\varepsilon}{2^j}\). Then for any \( r > 0 \) we can find an \( m \) such that the open intervals \( \{(a^j, b^j + \delta^j)\}_{j=1}^m \) cover the compact set \( A \cap Q_r \), where \( Q_r \) is a half-open cube of side length \( r \). Hence

\[
|A \cap Q_r| \leq \left| \bigcup_{j=1}^m B_j \right| = \sum_{j=1}^m \mu(B_j) \leq \sum_{j=1}^\infty |A_j| + \varepsilon.
\]

Letting \( r \to \infty \) and since \( \varepsilon > 0 \) is arbitrary, we are done. \( \square \)

Note that this shows

\[
\lambda^n(R) = |R|, \quad R \in \bar{\mathcal{S}}^n.
\]

(8.9)

In fact, by intersecting any cover with \( R \) we get a partition which has a smaller measure by monotonicity. But any partition will give the same value \(|R|\) by our lemma.

So the remaining question is if we can extend the family of sets on which \( \sigma \)-additivity holds. A convenient family clearly should be invariant under countable set operations and hence we will call an algebra a \( \sigma \)-algebra if it is closed under countable unions (and hence also under countable intersections by De Morgan’s rules). But how to construct such a \( \sigma \)-algebra?

It was Lebesgue who eventually was successful with the following idea: As pointed out before there is no corresponding inner Lebesgue measure since approximation by intervals from the inside does not work well. However, instead you can try to approximate the complement from the outside thereby setting

\[
\lambda^1_b(A) = (b - a) - \lambda^1([a, b] \setminus A)
\]

(8.10)

for every bounded set \( A \subseteq [a, b] \). Now you can call a bounded set \( A \) measurable if \( \lambda^1_b(A) = \lambda^1(A) \). We will however use a somewhat different approach due to Carathéodory. In this respect note that if we set \( E = [a, b] \) then \( \lambda^1(A) = \lambda^1(A) \) can be written as

\[
\lambda^1(E) = \lambda^1(A \cap E) + \lambda^1((A \cap E)
\]

(8.11)
which should be compared with the Carathéodory condition (8.14).

**Problem 8.1.** Suppose $S_1, S_2$ are semialgebras in $X_1, X_2$. Then $S := S_1 \otimes S_2 := \{A_1 \times A_2 | A_j \in S_j\}$ is a semialgebra in $X := X_1 \times X_2$.

**Problem 8.2 (Vitali set).** Call two numbers $x, y \in [0, 1)$ equivalent if $x - y$ is rational. Construct the set $V$ by choosing one representative from each equivalence class. Show that $V$ cannot be measurable with respect to any nontrivial finite translation invariant measure on $[0, 1)$. (Hint: How can you build up $[0, 1)$ from translations of $V$?)

**Problem 8.3.** show that

$$J_*(A) \leq \lambda^{n,*}(A) \leq J^*(A)$$

and hence $J(A) = \lambda^{n,*}(A)$ for every Jordan measurable set.

**Problem 8.4.** Let $X := \mathbb{N}$ and define the set function

$$\mu(A) := \lim_{N \to \infty} \frac{1}{N} \sum_{n=1}^{N} \chi_A(n) \in [0, 1]$$

on the collection $\mathcal{S}$ of all sets for which the above limit exists. Show that this collection is closed under disjoint unions and complements but not under intersections. Show that there is an extension to the $\sigma$-algebra generated by $\mathcal{S}$ (i.e. to $\mathcal{P}(\mathbb{N})$) which is additive but no extension which is $\sigma$-additive. (Hints: To show that $\mathcal{S}$ is not closed under intersections take a set of even numbers $A_1 \notin \mathcal{S}$ and let $A_2$ be the missing even numbers. Then let $A = A_1 \cup A_2 = 2\mathbb{N}$ and $B = A_1 \cup \tilde{A}_2$, where $\tilde{A}_2 = A_2 + 1$. To obtain an extension to $\mathcal{P}(\mathbb{N})$ consider $\chi_A, A \in \mathcal{S}$, as vectors in $\ell^\infty(\mathbb{N})$ and $\mu$ as a linear functional and see Problem 4.20.)

### 8.2. Sigma algebras and measures

If an algebra is closed under countable intersections, it is called a **$\sigma$-algebra**. Hence a $\sigma$-algebra is a family of subsets $\Sigma$ of a given set $X$ such that

- $\emptyset \in \Sigma$,
- $\Sigma$ is closed under countable intersections, and
- $\Sigma$ is closed under complements.

By De Morgan’s rule $\Sigma$ is also closed under countable unions.

Moreover, the intersection of any family of ($\sigma$-)algebras $\{\Sigma_\alpha\}$ is again a ($\sigma$-)algebra (check this) and for any collection $S$ of subsets there is a unique smallest ($\sigma$-)algebra $\Sigma(S)$ containing $S$ (namely the intersection of all ($\sigma$-)algebras containing $S$). It is called the ($\sigma$-)algebra generated by $S$. 
Example. For a given set $X$ and a subset $A \subseteq X$ we have $\Sigma(\{A\}) = \{\emptyset, A, A', X\}$. Moreover, every finite algebra is also a $\sigma$-algebra and if $S$ is finite, so will be $\Sigma(S)$ (Problem 8.6).

The power set $\mathcal{P}(X)$ is clearly the largest $\sigma$-algebra and $\{\emptyset, X\}$ is the smallest.

If $X$ is a topological space, the **Borel $\sigma$-algebra** $\mathcal{B}(X)$ of $X$ is defined to be the $\sigma$-algebra generated by all open (respectively, all closed) sets. In fact, if $X$ is second countable, any countable base will suffice to generate the Borel $\sigma$-algebra (recall Lemma 0.1). Sets in the Borel $\sigma$-algebra are called **Borel sets**.

Example. In the case $X = \mathbb{R}^n$ the Borel $\sigma$-algebra will be denoted by $\mathcal{B}^n$ and we will abbreviate $\mathcal{B} := \mathcal{B}^1$. Note that in order to generate $\mathcal{B}^n$, open balls with rational center and rational radius suffice. In fact, any base for the topology will suffice. Moreover, since open balls can be written as a countable union of smaller closed balls with increasing radii, we could also use compact balls instead.

Example. If $X$ is a topological space, then any Borel set $Y \subseteq X$ is also a topological space equipped with the relative topology and its Borel $\sigma$-algebra is given by $\mathcal{B}(Y) = \mathcal{B}(X) \cap Y := \{A | A \in \mathcal{B}(X), A \subseteq Y\}$ (show this).

Now let us turn to the definition of a measure: A set $X$ together with a $\sigma$-algebra $\Sigma$ is called a **measurable space**. A **measure** $\mu$ is a map $\mu : \Sigma \to [0, \infty]$ on a $\sigma$-algebra $\Sigma$ such that

- $\mu(\emptyset) = 0,$
- $\mu(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \mu(A_n)$, $A_n \in \Sigma$ ($\sigma$-additivity).

Here the sum is set equal to $\infty$ if one of the summands is $\infty$ or if it diverges.

The measure $\mu$ is called **$\sigma$-finite** if there is a countable cover $\{X_n\}_{n=1}^{\infty}$ of $X$ such that $X_n \in \Sigma$ and $\mu(X_n) < \infty$ for all $n$. (Note that it is no restriction to assume $X_n \subseteq X_{n+1}$.) It is called **finite** if $\mu(X) < \infty$ and a **probability measure** if $\mu(X) = 1$. The sets in $\Sigma$ are called **measurable sets** and the triple $(X, \Sigma, \mu)$ is referred to as a **measure space**.

Example. Take a set $X$ with $\Sigma = \mathcal{P}(X)$ and set $\mu(A)$ to be the number of elements of $A$ (respectively, $\infty$ if $A$ is infinite). This is the so-called **counting measure**. It will be finite if and only if $X$ is finite and $\sigma$-finite if and only if $X$ is countable.

Example. Take a set $X$ and $\Sigma := \mathcal{P}(X)$. Fix a point $x \in X$ and set $\mu(A) = 1$ if $x \in A$ and $\mu(A) = 0$ else. This is the **Dirac measure** centered at $x$. It is also frequently written as $\delta_x$. 
Example. Let \( \mu_1, \mu_2 \) be two measures on \((X, \Sigma)\) and \( \alpha_1, \alpha_2 \geq 0 \). Then 
\[
\mu = \alpha_1 \mu_1 + \alpha_2 \mu_2 \quad \text{defined via}
\]
\[
\mu(A) := \alpha_1 \mu_1(A) + \alpha_2 \mu_2(A)
\]
is again a measure. Furthermore, given a countable number of measures \( \mu_n \) and numbers \( \alpha_n \geq 0 \), then 
\[
\mu := \sum_n \alpha_n \mu_n
\]
is again a measure (show this). \( \diamond \)

Example. Let \( \mu \) be a measure on \((X, \Sigma)\) and \( Y \subseteq X \) a measurable subset. Then 
\[
\nu(A) := \mu(A \cap Y)
\]
is again a measure on \((X, \Sigma)\) (show this). \( \diamond \)

Example. Let \( X \) be some set with a \( \sigma \)-algebra \( \Sigma \). Then every subset \( Y \subseteq X \) has a natural \( \sigma \)-algebra \( \Sigma \cap Y := \{A \cap Y | A \in \Sigma\} \) (show that this is indeed a \( \sigma \)-algebra) known as the relative \( \sigma \)-algebra (also trace \( \sigma \)-algebra).

Note that if \( S \) generates \( \Sigma \), then \( S \cap Y \) generates \( \Sigma \cap Y \): \( \Sigma(S) \cap Y = \Sigma(S \cap Y) \). Indeed, since \( S \cap Y \) is a \( \sigma \)-algebra containing \( S \cap Y \), we have \( \Sigma(S \cap Y) \subseteq \Sigma(S) \cap Y = \Sigma \cap Y \). Conversely, consider \( \{A \in \Sigma | A \cap Y \in \Sigma(S \cap Y)\} \) which is a \( \sigma \)-algebra (check this). Since this last \( \sigma \)-algebra contains \( S \) it must be equal to \( \Sigma = \Sigma(S) \) and thus \( \Sigma \cap Y \subseteq \Sigma(S \cap Y) \). \( \diamond \)

Example. If \( Y \in \Sigma \) we can restrict the \( \sigma \)-algebra \( \Sigma|_Y = \{A \in \Sigma | A \subseteq Y\} \) such that \((Y, \Sigma|_Y, \mu|_Y)\) is again a measurable space. It will be \( \sigma \)-finite if \((X, \Sigma, \mu)\) is. \( \diamond \)

Finally, we will show that \( \sigma \)-additivity implies some crucial continuity properties for measures which eventually will lead to powerful limiting theorems for the corresponding integrals. We will write \( A_n \nearrow A \) if \( A_n \subseteq A_{n+1} \) with \( A = \bigcup_n A_n \) and \( A_n \searrow A \) if \( A_{n+1} \subseteq A_n \) with \( A = \bigcap_n A_n \).

**Theorem 8.5.** Any measure \( \mu \) satisfies the following properties:

(i) \( A \subseteq B \) implies \( \mu(A) \leq \mu(B) \) (monotonicity).

(ii) \( \mu(A_n) \to \mu(A) \) if \( A_n \nearrow A \) (continuity from below).

(iii) \( \mu(A_n) \to \mu(A) \) if \( A_n \searrow A \) and \( \mu(A_1) < \infty \) (continuity from above).

**Proof.** The first claim is obvious from \( \mu(B) = \mu(A) + \mu(B \setminus A) \). To see the second define \( \tilde{A}_1 = A_1, \tilde{A}_n = A_n \setminus A_{n-1} \) and note that these sets are disjoint and satisfy \( A_n = \bigcup_{j=1}^n \tilde{A}_j \). Hence \( \mu(A_n) = \sum_{j=1}^n \mu(\tilde{A}_j) = \sum_{j=1}^\infty \mu(\tilde{A}_j) = \mu(\bigcup_{j=1}^\infty \tilde{A}_j) = \mu(A) \) by \( \sigma \)-additivity. The third follows from the second using \( \tilde{A}_n = A_1 \setminus A_n \nearrow A_1 \setminus A \) implying \( \mu(\tilde{A}_n) = \mu(A_1) - \mu(A_n) \to \mu(A_1 \setminus A) = \mu(A_1) - \mu(A) \). \( \square \)
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Example. Consider the counting measure on \( X = \mathbb{N} \) and let \( A_n = \{ j \in \mathbb{N} | j \geq n \} \), then \( \mu(A_n) = \infty \), but \( \mu(\bigcap_n A_n) = \mu(\emptyset) = 0 \) which shows that the requirement \( \mu(A_1) < \infty \) in item (iii) of Theorem 8.5 is not superfluous. ⋄

Problem 8.5. Find all algebras over \( X := \{1, 2, 3\} \).

Problem 8.6. Let \( \{A_j\}_{j=1}^n \) be a finite family of subsets of a given set \( X \). Show that \( \Sigma(\{A_j\}_{j=1}^n) \) has at most \( 4^n \) elements. (Hint: Let \( X \) have \( 2^n \) elements and look at the case \( n = 2 \) to get an idea. Consider sets of the form \( B_1 \cap \cdots \cap B_n \) with \( B_j \in \{A_j, A'_j\} \).)

Problem 8.7. Show that \( \mathcal{A} := \{ A \subseteq X | A \text{ or } X \setminus A \text{ is finite} \} \) is an algebra (with \( X \) some fixed set). Show that \( \Sigma := \{ A \subseteq X | A \text{ or } X \setminus A \text{ is countable} \} \) is a \( \sigma \)-algebra. (Hint: To verify closedness under unions consider the cases were all sets are finite and where one set has finite complement.)

Problem 8.8. Take some set \( X \) and \( \Sigma := \{ A \subseteq X | A \text{ or } X \setminus A \text{ is countable} \} \). Show that \( \nu(A) := \begin{cases} 0, & \text{if } A \text{ is countable}, \\ 1, & \text{else}. \end{cases} \) is a measure

Problem 8.9. Show that if \( X \) is finite, then every algebra is a \( \sigma \)-algebra. Show that this is not true in general if \( X \) is countable.

8.3. Extending a premeasure to a measure

Now we are ready to show how to construct a measure starting from a premeasure. In fact, we have already seen how to get a premeasure starting from a small collection of sets which generate the \( \sigma \)-algebra and for which it is clear what the measure should be. The crucial requirement for such a collection of sets is the requirement that it is closed under intersections as the following example shows.

Example. Consider \( X := \{1, 2, 3\} \) together with the measures \( \mu(\{1\}) := \mu(\{2\}) := \mu(\{3\}) := \frac{1}{3} \) and \( \nu(\{1\}) := \frac{1}{6}, \nu(\{2\}) := \frac{1}{2}, \nu(\{3\}) := \frac{1}{6} \). Then \( \mu \) and \( \nu \) agree on \( S := \{\{1, 2\}, \{2, 3\}\} \) but not on \( \Sigma(S) = \mathcal{P}(X) \). Note that \( S \) is not closed under intersections. If we take \( S := \{\emptyset, \{1\}, \{2, 3\}\} \), which is closed under intersections, and \( \nu(\{1\}) := \frac{1}{3}, \nu(\{2\}) := \frac{1}{2}, \nu(\{3\}) := \frac{1}{6} \), then \( \mu \) and \( \nu \) agree on \( \Sigma(S) = \{\emptyset, \{1\}, \{2, 3\}, X\} \). But they don’t agree on \( \mathcal{P}(X) \). ⋄

Hence we begin with the question when a family of sets determines a measure uniquely. To this end we need a better criterion to check when a given system of sets is in fact a \( \sigma \)-algebra. In many situations it is easy to show that a given set is closed under complements and under countable
unions of disjoint sets. Hence we call a collection of sets $\mathcal{D}$ with these properties a **Dynkin system** (also $\lambda$-system) if it also contains $X$.

Note that a Dynkin system is closed under proper relative complements since $A, B \in \mathcal{D}$ implies $B \setminus A = (B' \cup A)' \in \mathcal{D}$ provided $A \subseteq B$. Moreover, if it is also closed under finite intersections (or arbitrary finite unions) then it is an algebra and hence also a $\sigma$-algebra. To see the last claim note that if $A = \bigcup_j A_j$ then also $A = \bigcup_j B_j$ where the sets $B_j = A_j \setminus \bigcup_{k<j} A_k$ are disjoint.

**Example.** Let $X := \{1, 2, 3, 4\}$. Then $\mathcal{D} := \{A \subseteq X|\#A \text{ is even}\}$ is a Dynkin system but no algebra.

As with $\sigma$-algebras, the intersection of Dynkin systems is a Dynkin system and every collection of sets $S$ generates a smallest Dynkin system $\mathcal{D}(S)$. The important observation is that if $S$ is closed under finite intersections (in which case it is sometimes called a $\pi$-system), then so is $\mathcal{D}(S)$ and hence will be a $\sigma$-algebra.

**Lemma 8.6 (Dynkin’s $\pi$-$\lambda$ theorem).** Let $S$ be a collection of subsets of $X$ which is closed under finite intersections (or unions). Then $\mathcal{D}(S) = \Sigma(S)$.

**Proof.** It suffices to show that $\mathcal{D} := \mathcal{D}(S)$ is closed under finite intersections. To this end consider the set $D(A) := \{B \in \mathcal{D}|A \cap B \in \mathcal{D}\}$ for $A \in \mathcal{D}$. I claim that $D(A)$ is a Dynkin system.

First of all $X \in D(A)$ since $A \cap X = A \in \mathcal{D}$. Next, if $B \in D(A)$ then $A \cap B' = A \setminus (B \cap A) \in \mathcal{D}$ (since $\mathcal{D}$ is closed under proper relative complements) implying $B' \in D(A)$. Finally if $B = \bigcup_j B_j$ with $B_j \in D(A)$ disjoint, then $A \cap B = \bigcup_j (A \cap B_j) \in \mathcal{D}$ with $A \cap B_j \in \mathcal{D}$ disjoint, implying $B \in D(A)$.

Now if $A \in S$ we have $S \subseteq D(A)$ implying $D(A) = \mathcal{D}$. Consequently $A \cap B \in \mathcal{D}$ if at least one of the sets is in $S$. But this shows $S \subseteq D(A)$ and hence $D(A) = \mathcal{D}$ for every $A \in \mathcal{D}$. So $\mathcal{D}$ is closed under finite intersections and thus a $\sigma$-algebra. The case of unions is analogous.

The typical use of this lemma is as follows: First verify some property for sets in a collection $S$ which is closed under finite intersections and generates the $\sigma$-algebra. In order to show that it holds for every set in $\Sigma(S)$, it suffices to show that the collection of sets for which it holds is a Dynkin system.

As an application we show that a premeasure determines the corresponding measure $\mu$ uniquely (if there is one at all):

**Theorem 8.7 (Uniqueness of measures).** Let $S \subseteq \Sigma$ be a collection of sets which generates $\Sigma$ and which is closed under finite intersections and contains
8.3. Extending a premeasure to a measure

a sequence of increasing sets $X_n \nearrow X$ of finite measure $\mu(X_n) < \infty$. Then
$\mu$ is uniquely determined by the values on $S$.

**Proof.** Let $\tilde{\mu}$ be a second measure and note $\mu(X) = \lim_{n \to \infty} \mu(X_n) = \lim_{n \to \infty} \tilde{\mu}(X_n) = \tilde{\mu}(X)$. We first suppose $\mu(X) < \infty$.

Then

$$D := \{ A \in \Sigma | \mu(A) = \tilde{\mu}(A) \}$$

is a Dynkin system. In fact, by $\mu(A') = \mu(X) - \mu(A) = \tilde{\mu}(X) - \tilde{\mu}(A) = \tilde{\mu}(A')$ for $A \in D$ we see that $D$ is closed under complements. Furthermore, by continuity of measures from below it is also closed under countable disjoint unions. Since $D$ contains $S$ by assumption, we conclude $D = \Sigma(S) = \Sigma$ from Lemma 8.6. This finishes the finite case.

To extend our result to the general case observe that the finite case implies $\mu(A \cap X_j) = \tilde{\mu}(A \cap X_j)$ (just restrict $\mu$, $\tilde{\mu}$ to $X_j$). Hence

$$\mu(A) = \lim_{j \to \infty} \mu(A \cap X_j) = \lim_{j \to \infty} \tilde{\mu}(A \cap X_j) = \tilde{\mu}(A)$$

and we are done. \hfill \Box

**Corollary 8.8.** Let $\mu$ be a $\sigma$-finite premeasure on an algebra $A$. Then there is at most one extension to $\Sigma(A)$.

**Example.** Set $\mu([a,b)) = \infty$ on $S^1$. This determines a unique premeasure $\mu$ on $S^1$. However, the counting measure as well as the measure which assigns every nonempty set the value $\infty$ are two different extensions. Hence the finiteness assumption in the previous theorem/corollary is crucial. \hfill ◦

Now we come to the construction of the extension. For any premeasure $\mu$ we define its corresponding outer measure $\mu^* : \mathcal{P}(X) \to [0, \infty]$ (Lemma 8.2) as

$$\mu^*(A) := \inf \left\{ \sum_{n=1}^{\infty} \mu(A_n) \big| A \subseteq \bigcup_{n=1}^{\infty} A_n, A_n \in \mathcal{A} \right\}, \quad \text{(8.13)}$$

where the infimum extends over all countable covers from $\mathcal{A}$. Replacing $A_n$ by $\tilde{A}_n = A_n \setminus \bigcup_{m=1}^{n-1} A_m$ we see that we could even require the covers to be disjoint. Note that $\mu^*(A) = \mu(A)$ for $A \in \mathcal{A}$ (Problem 8.10).

**Theorem 8.9** (Extensions via outer measures). Let $\mu^*$ be an outer measure. Then the set $\Sigma$ of all sets $A$ satisfying the Carathéodory condition

$$\mu^*(E) = \mu^*(A \cap E) + \mu^*(A' \cap E), \quad \forall E \subseteq \bigcup_{n=1}^{\infty} A_n \in \mathcal{A}$$

(where $A' := X \setminus A$ is the complement of $A$) forms a $\sigma$-algebra and $\mu^*$ restricted to this $\sigma$-algebra is a measure.
Proof. We first show that $\Sigma$ is an algebra. It clearly contains $X$ and is closed under complements. Concerning unions let $A, B \in \Sigma$. Applying Carathéodory’s condition twice shows

$$\mu^*((E) = \mu^*(A \cap B \cap E) + \mu^*(A' \cap B \cap E) + \mu^*(A \cap B' \cap E) \geq \mu^*((A \cup B) \cap E) + \mu^*((A \cup B)' \cap E),$$

where we have used De Morgan and

$$\mu^*(A \cap B \cap E) + \mu^*(A' \cap B \cap E) + \mu^*(A \cap B' \cap E) \geq \mu^*((A \cup B) \cap E)$$

which follows from subadditivity and $(A \cup B) \cap E = (A \cap B \cap E) \cup (A' \cap B \cap E) \cup (A \cap B' \cap E)$. Since the reverse inequality is just subadditivity, we conclude that $\Sigma$ is an algebra.

Next, let $A_n$ be a sequence of sets from $\Sigma$. Without restriction we can assume that they are disjoint (compare the argument for item (ii) in the proof of Theorem 8.5). Abbreviate $\tilde{A}_n = \bigcup_{k \leq n} A_k$, $A = \bigcup_{n} A_n$. Then for every set $E$ we have

$$\mu^*(\tilde{A}_n \cap E) = \mu^*(A_n \cap \tilde{A}_n \cap E) + \mu^*(A_n' \cap \tilde{A}_n \cap E)$$

$$= \mu^*(A_n \cap E) + \mu^*(\tilde{A}_{n-1} \cap E)$$

$$= \ldots = \sum_{k=1}^{n} \mu^*(A_k \cap E).$$

Using $\tilde{A}_n \in \Sigma$ and monotonicity of $\mu^*$, we infer

$$\mu^*(\tilde{A}_n \cap E) = \mu^*(A_n \cap E) + \mu^*(A_n' \cap E)$$

$$\geq \sum_{k=1}^{n} \mu^*(A_k \cap E) + \mu^*(A' \cap E).$$

Letting $n \to \infty$ and using subadditivity finally gives

$$\mu^*(E) \geq \sum_{k=1}^{\infty} \mu^*(A_k \cap E) + \mu^*(A' \cap E)$$

$$\geq \mu^*(A \cap E) + \mu^*(A' \cap E) \geq \mu^*(E) \quad (8.15)$$

and we infer that $\Sigma$ is a $\sigma$-algebra.

Finally, setting $E = A$ in (8.15), we have

$$\mu^*(A) = \sum_{k=1}^{\infty} \mu^*(A_k \cap A) + \mu^*(A' \cap A) = \sum_{k=1}^{\infty} \mu^*(A_k)$$

and we are done.
8.3. Extending a premeasure to a measure

Remark: The constructed measure $\mu$ is complete; that is, for every measurable set $A$ of measure zero, every subset of $A$ is again measurable. In fact, every null set $A$, that is, every set with $\mu^*(A) = 0$, is measurable (Problem 8.11).

The only remaining question is whether there are any nontrivial sets satisfying the Carathéodory condition.

Lemma 8.10. Let $\mu$ be a premeasure on $\mathcal{A}$ and let $\mu^*$ be the associated outer measure. Then every set in $\mathcal{A}$ satisfies the Carathéodory condition.

Proof. Let $A_n \in \mathcal{A}$ be a countable cover for $E$. Then for every $A \in \mathcal{A}$ we have
\[
\sum_{n=1}^{\infty} \mu(A_n) = \sum_{n=1}^{\infty} \mu(A_n \cap A) + \sum_{n=1}^{\infty} \mu(A_n \cap A') \geq \mu^*(E \cap A) + \mu^*(E \cap A')
\]
since $A_n \cap A \in \mathcal{A}$ is a cover for $E \cap A$ and $A_n \cap A' \in \mathcal{A}$ is a cover for $E \cap A'$. Taking the infimum, we have $\mu^*(E) \geq \mu^*(E \cap A) + \mu^*(E \cap A')$, which finishes the proof. \[\square\]

Thus the Lebesgue premeasure on $\mathcal{S}^n$ gives rise to Lebesgue measure $\lambda^n$ when the outer measure $\lambda^{*,n}$ is restricted to the Borel $\sigma$-algebra $\mathcal{B}^n$. In fact, with the very same procedure we can obtain a large class of measures in $\mathbb{R}^n$ as will be demonstrated in the next section.

To end this section, let me emphasize that in our approach we started from a premeasure, which gave rise to an outer measure, which eventually lead to a measure via Carathéodory’s theorem. However, while some effort was required to get the premeasure in our case, an outer measure often can be obtained much easier (recall Lemma 8.2). While this immediately leads again to a measure, one is faced with the problem if any nontrivial sets satisfy the Carathéodory condition.

To address this problem let $(X,d)$ be a metric space and call an outer measure $\mu^*$ on $X$ a metric outer measure if
\[
\mu^*(A_1 \cup A_2) = \mu^*(A_1) + \mu^*(A_2)
\]
whenever
\[
\text{dist}(A_1, A_2) := \inf_{(x_1, x_2) \in A_1 \times A_2} d(x_1, x_2) > 0.
\]

Lemma 8.11. Let $X$ be a metric space. An outer measure is metric if and only if all Borel sets satisfy the Carathéodory condition (8.14).

Proof. To show that all Borel sets satisfy the Carathéodory condition it suffices to show this is true for all closed sets. First of all note that we have $G_n := \{ x \in F' \cap E | d(x, F) \geq \frac{1}{n} \} \not\supset F' \cap E$ since $F$ is closed. Moreover,
8. Measures

d(G_n, F) \geq \frac{1}{n} and hence $\mu^*(F \cap E) + \mu^*(G_n) = \mu^*((E \cap F) \cup G_n) \leq \mu^*(E)$ by the definition of a metric outer measure. Hence it suffices to show $\mu^*(G_n) \to \mu^*(F' \cap E)$. Moreover, we can also assume $\mu^*(E) < \infty$ since otherwise there is noting to show. Now consider $\tilde{G}_n = G_{n+1} \setminus G_n$. Then $d(\tilde{G}_{n+2}, \tilde{G}_n) > 0$ and hence $\sum_{j=1}^{m} \mu^*(\tilde{G}_{2j}) = \mu^*(\bigcup_{j=1}^{m} \tilde{G}_{2j}) \leq \mu^*(E)$ as well as $\sum_{j=1}^{m} \mu^*(\tilde{G}_{2j-1}) = \mu^*(\bigcup_{j=1}^{m} \tilde{G}_{2j-1}) \leq \mu^*(E)$ and consequently $\sum_{j=1}^{\infty} \mu^*(\tilde{G}_{j}) \leq 2\mu^*(E) < \infty$. Now subaditivity implies

$$\mu^*(F' \cap E) \leq \mu(G_n) + \sum_{j \geq n} \mu^*(\tilde{G}_n)$$

and thus

$$\mu^*(F' \cap E) \leq \liminf_{n \to \infty} \mu^*(G_n) \leq \limsup_{n \to \infty} \mu^*(G_n) \leq \mu^*(F' \cap E)$$

as required.

Conversely, suppose $\epsilon := \text{dist}(A_1, A_2) > 0$ and consider $\epsilon$ neighborhood of $A_1$ given by $O_{\epsilon} = \bigcup_{x \in A_1} B_{\epsilon}(x)$. Then $\mu^*(A_1 \cup A_2) = \mu^*(O_{\epsilon} \cap (A_1 \cup A_2)) + \mu^*(O_{\epsilon}' \cap (A_1 \cup A_2)) = \mu^*(A_1) + \mu^*(A_2)$ as required. \qed

**Problem 8.10.** Show that $\mu^*$ defined in (8.13) extends $\mu$. (Hint: For the cover $A_n$ it is no restriction to assume $A_n \cap A_m = \emptyset$ and $A_n \subseteq A$.)

**Problem 8.11.** Show that every null set satisfies the Carathéodory condition (8.14). Conclude that the measure constructed in Theorem 8.9 is complete.

**Problem 8.12 (Completion of a measure).** Show that every measure has an extension which is complete as follows:

Denote by $\mathcal{N}$ the collection of subsets of $X$ which are subsets of sets of measure zero. Define $\tilde{\Sigma} := \{ A \cup N | A \in \Sigma, N \in \mathcal{N} \}$ and $\tilde{\mu}(A \cup N) := \mu(A)$ for $A \cup N \in \tilde{\Sigma}$.

Show that $\tilde{\Sigma}$ is a $\sigma$-algebra and that $\tilde{\mu}$ is a well-defined complete measure. Moreover, show $\mathcal{N} = \{ N \in \Sigma | \tilde{\mu}(N) = 0 \}$ and $\tilde{\Sigma} = \{ B \subseteq X | \exists A_1, A_2 \in \Sigma \text{ with } A_1 \subseteq B \subseteq A_2 \text{ and } \mu(A_2 \setminus A_1) = 0 \}$.

**Problem 8.13.** Let $\mu$ be a finite measure. Show that

$$d(A, B) := \mu(A \Delta B), \quad A \Delta B := (A \cup B) \setminus (A \cap B) \quad (8.16)$$

is a metric on $\Sigma$ if we identify sets differing by sets of measure zero. Show that if $A$ is an algebra, then it is dense in $\Sigma(A)$. (Hint: Show that the sets which can be approximated by sets in $A$ form a Dynkin system.)
8.4. Borel measures

In this section we want to construct a large class of important measures on \( \mathbb{R}^n \). We begin with a few abstract definitions.

Let \( X \) be a topological space. A measure on the Borel \( \sigma \)-algebra is called a **Borel measure** if \( \mu(K) < \infty \) for every compact set \( K \). Note that some authors do not require this last condition.

**Example.** Let \( X := \mathbb{R} \) and \( \Sigma := \mathcal{B} \). The Dirac measure is a Borel measure. The counting measure is no Borel measure since \( \mu([a,b]) = \infty \) for \( a < b \).  

A measure on the Borel \( \sigma \)-algebra is called **outer regular** if

\[
\mu(A) = \inf_{O \supseteq A, O \text{ open}} \mu(O)  \tag{8.17}
\]

and **inner regular** if

\[
\mu(A) = \sup_{K \subseteq A, K \text{ compact}} \mu(K).  \tag{8.18}
\]

It is called **regular** if it is both outer and inner regular.

**Example.** Let \( X := \mathbb{R} \) and \( \Sigma := \mathcal{B} \). The counting measure is inner regular but not outer regular (every nonempty open set has infinite measure). The Dirac measure is a regular Borel measure.

But how can we obtain some more interesting Borel measures? We will restrict ourselves to the case of \( X = \mathbb{R}^n \) and begin with the case of Borel measures on \( X = \mathbb{R} \) which are also known as **Lebesgue–Stieltjes measures**. By what we have seen so far it is clear that our strategy is as follows: Start with some simple sets and then work your way up to all Borel sets. Hence let us first show how we should define \( \mu \) for intervals: To every Borel measure on \( \mathcal{B} \) we can assign its **distribution function**

\[
\mu(x) := \begin{cases} 
-\mu((x,0]), & x < 0, \\
0, & x = 0, \\
\mu((0,x]), & x > 0,
\end{cases}  \tag{8.19}
\]

which is right continuous and nondecreasing as can be easily checked.

**Example.** The distribution function of the Dirac measure centered at 0 is

\[
\mu(x) := \begin{cases} 
0, & x \geq 0, \\
-1, & x < 0.
\end{cases}
\]
For a finite measure the alternate normalization \( \tilde{\mu}(x) = \mu((\infty, x]) \) can be used. The resulting distribution function differs from our above definition by a constant \( \mu(x) = \tilde{\mu}(x) - \mu((\infty, 0]) \). In particular, this is the normalization used for probability measures.

Conversely, to obtain a measure from a nondecreasing function \( m : \mathbb{R} \to \mathbb{R} \) we proceed as follows: Recall that an interval is a subset of the real line of the form

\[
I = (a, b], \quad I = [a, b], \quad I = (a, b), \quad \text{or} \quad I = [a, b),
\]

with \( a \leq b, \ a, b \in \mathbb{R} \cup \{-\infty, \infty\} \). Note that \((a, a), [a, a), \text{ and } (a, a)]\) denote the empty set, whereas \([a, a]\) denotes the singleton \(\{a\}\). For any proper interval with different endpoints (i.e. \(a < b\)) we can define its measure to be

\[
\mu(I) := \begin{cases} 
  m(b+) - m(a+), & I = (a, b], \\
  m(b+) - m(a-), & I = [a, b], \\
  m(b-) - m(a+), & I = (a, b), \\
  m(b-) - m(a-), & I = [a, b),
\end{cases}
\]

(8.21)

where \(m(a\pm) = \lim_{\varepsilon \downarrow 0} m(a \pm \varepsilon)\) (which exist by monotonicity). If one of the endpoints is infinite we agree to use \(m(\pm \infty) = \lim_{x \to \pm \infty} m(x)\). For the empty set we of course set \(\mu(\emptyset) = 0\) and for the singletons we set

\[
\mu(\{a\}) := m(a+) - m(a-)
\]

(8.22)

(which agrees with (8.21) except for the case \(I = (a, a)\) which would give a negative value for the empty set if \(m\) jumps at \(a\)). Note that \(\mu(\{a\}) = 0\) if and only if \(m(x)\) is continuous at \(a\) and that there can be only countably many points with \(\mu(\{a\}) > 0\) since a nondecreasing function can have at most countably many jumps. Moreover, observe that the definition of \(\mu\) does not involve the actual value of \(m\) at a jump. Hence any function \(\tilde{m}\) with \(m(x-) \leq \tilde{m}(x) \leq m(x+)\) gives rise to the same \(\mu\). We will frequently assume that \(m\) is right continuous such that it coincides with the distribution function up to a constant, \(\mu(x) = m(x+) - m(0+)\). In particular, \(\mu\) determines \(m\) up to a constant and the value at the jumps.

Once we have defined \(\mu\) on \(S^1\) we can now show that (8.21) gives a premeasure.

**Lemma 8.12.** Let \(m : \mathbb{R} \to \mathbb{R}\) be right continuous and nondecreasing. Then the set function defined via

\[
\mu((a, b]) := m(b) - m(a)
\]

(8.23)

on \(S^1\) gives rise to a unique \(\sigma\)-finite premeasure on the algebra \(S^1\) of finite unions of disjoint half-open intervals.
Proof. If \( [a, b] = \bigcup_{j=1}^{n} [a_j, b_j] \) then we can assume that the \( a_j \)'s are ordered. Moreover, in this case we must have \( b_j = a_{j+1} \) for \( 1 \leq j < n \) and hence our set function is additive on \( S^1 \): \[ \sum_{j=1}^{n} \mu([a_j, b_j]) = \sum_{j=1}^{n} (\mu(b_j) - \mu(a_j)) = \mu(b_n) - \mu(a_1) = \mu(a) = \mu([a, b]). \]

So by Lemma 8.3 it remains to verify (8.8). By right continuity we can cover each \( A_j := (a_j, b_j] \) by some slightly larger interval \( B_j := (a_j, b_j + \delta_j] \) such that \( \mu(B_j) \leq \mu(A_j) + \frac{\varepsilon}{2j} \). Then for any \( x > 0 \) we can find an \( n \) such that the open intervals \( \{(a_j, b_j + \delta_j)\}_{j=1}^{n} \) cover the compact set \( \overline{A} \cap [-x, x] \) and hence
\[
\mu(A \cap (-x, x]) \leq \mu\left(\bigcup_{j=1}^{n} B_j\right) = \sum_{j=1}^{n} \mu(B_j) \leq \sum_{j=1}^{\infty} \mu(A_j) + \varepsilon.
\]
Letting \( x \to \infty \) and since \( \varepsilon > 0 \) is arbitrary, we are done. \( \square \)

And extending this premeasure to a measure we finally obtain:

**Theorem 8.13.** For every nondecreasing function \( m : \mathbb{R} \to \mathbb{R} \) there exists a unique regular Borel measure \( \mu \) which extends \( (8.21) \). Two different functions generate the same measure if and only if the difference is a constant away from the discontinuities.

**Proof.** Except for regularity existence follows from Theorem 8.9 together with Lemma 8.10 and uniqueness follows from Corollary 8.8. Regularity will be postponed until Section 8.6. \( \square \)

We remark, that in the previous theorem we could as well consider \( m : (a, b) \to \mathbb{R} \) to obtain regular Borel measures on \( (a, b) \).

**Example.** Suppose \( \Theta(x) := 0 \) for \( x < 0 \) and \( \Theta(x) := 1 \) for \( x \geq 0 \). Then we obtain the so-called **Dirac measure** at 0, which is given by \( \Theta(A) = 1 \) if \( 0 \in A \) and \( \Theta(A) = 0 \) if \( 0 \notin A \). \( \diamond \)

**Example.** Suppose \( \lambda(x) := x \). Then the associated measure is the ordinary **Lebesgue measure** on \( \mathbb{R} \). We will abbreviate the Lebesgue measure of a Borel set \( A \) by \( \lambda(A) = |A| \). \( \diamond \)

A set \( A \in \Sigma \) is called a **support** for \( \mu \) if \( \mu(X \setminus A) = 0 \). Note that a support is not unique (see the examples below). If \( X \) is a topological space and \( \Sigma = \mathcal{B}(X) \), one defines the **support** (also **topological support**) of \( \mu \) via
\[
\text{supp}(\mu) := \{ x \in X | \mu(O) > 0 \text{ for every open neighborhood } O \text{ of } x \}.
\]
(8.24)
Equivalently one obtains \( \text{supp}(\mu) \) by removing all points which have an open neighborhood of measure zero. In particular, this shows that \( \text{supp}(\mu) \)
8. Measures

is closed. If \( X \) is second countable, then \( \text{supp}(\mu) \) is indeed a support for \( \mu \): For every point \( x \not\in \text{supp}(\mu) \) let \( O_x \) be an open neighborhood of measure zero. These sets cover \( X \setminus \text{supp}(\mu) \) and by the Lindelöf theorem there is a countable subcover, which shows that \( X \setminus \text{supp}(\mu) \) has measure zero.

**Example.** Let \( X := \mathbb{R}, \Sigma := \mathcal{B} \). The support of the Lebesgue measure \( \lambda \) is all of \( \mathbb{R} \). However, every single point has Lebesgue measure zero and so has every countable union of points (by \( \sigma \)-additivity). Hence any set whose complement is countable is a support. There are even uncountable sets of Lebesgue measure zero (see the Cantor set below) and hence a support might even lack an uncountable number of points.

The support of the Dirac measure centered at 0 is the single point 0. Any set containing 0 is a support of the Dirac measure.

In general, the support of a Borel measure on \( \mathbb{R} \) is given by

\[
\text{supp}(d\mu) = \{ x \in \mathbb{R} | \mu(x - \varepsilon) < \mu(x + \varepsilon), \forall \varepsilon > 0 \}.
\]

Here we have used \( d\mu \) to emphasize that we are interested in the support of the measure \( d\mu \) which is different from the support of its distribution function \( \mu(x) \).

A property is said to hold \( \mu \)-almost everywhere (a.e.) if it holds on a support for \( \mu \) or, equivalently, if the set where it does not hold is contained in a set of measure zero.

**Example.** The set of rational numbers is countable and hence has Lebesgue measure zero, \( \lambda(\mathbb{Q}) = 0 \). So, for example, the characteristic function of the rationals \( \mathbb{Q} \) is zero almost everywhere with respect to Lebesgue measure.

Any function which vanishes at 0 is zero almost everywhere with respect to the Dirac measure centered at 0.

**Example.** The Cantor set is an example of a closed uncountable set of Lebesgue measure zero. It is constructed as follows: Start with \( C_0 := [0,1] \) and remove the middle third to obtain \( C_1 := [0, \frac{1}{3}] \cup [\frac{2}{3}, 1] \). Next, again remove the middle third’s of the remaining sets to obtain \( C_2 := [0, \frac{1}{9}] \cup [\frac{2}{9}, \frac{1}{3}] \cup [\frac{2}{3}, \frac{7}{9}] \cup [\frac{8}{9}, 1] \):

\[
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\end{array}
\]

Proceeding like this, we obtain a sequence of nesting sets \( C_n \) and the limit \( C := \bigcap_n C_n \) is the Cantor set. Since \( C_n \) is compact, so is \( C \). Moreover, \( C_n \) consists of \( 2^n \) intervals of length \( 3^{-n} \), and thus its Lebesgue measure is \( \lambda(C_n) = (2/3)^n \). In particular, \( \lambda(C) = \lim_{n \to \infty} \lambda(C_n) = 0 \). Using the
ternary expansion, it is extremely simple to describe: $C$ is the set of all $x \in [0,1]$ whose ternary expansion contains no one’s, which shows that $C$ is uncountable (why?). It has some further interesting properties: it is totally disconnected (i.e., it contains no subintervals) and perfect (it has no isolated points).

Finally, we show how to extend Theorem 8.13 to $\mathbb{R}^n$. We will write $x \leq y$ if $x_j \leq y_j$ for $1 \leq j \leq n$ and $(a,b) = (a_1,b_1) \times \cdots \times (a_n,b_n)$, $(a,b] = (a_1,b_1] \times \cdots \times (a_n,b_n]$, etc.

The analog of (8.19) is given by

$$
\mu(x) := \text{sign}(x)\mu\left( \prod_{j=1}^{n} \left( \min(0,x_j), \max(0,x_j) \right) \right),
$$

where \text{sign}(x) = \prod_{j=1}^{n} \text{sign}(x_j).

**Example.** The distribution function of the Dirac measure $\mu = \delta_0$ centered at $0$ is

$$
\mu(x) := \begin{cases} 
0, & x \geq 0, \\
-1, & \text{else}.
\end{cases}
$$

Again, for a finite measure the alternative normalization $\tilde{\mu}(x) = \mu((-\infty,x])$ can be used.

To recover a measure $\mu$ from its distribution function we consider the difference with respect to the $j$'th coordinate

$$
\Delta^j_{a^1,a^2}m(x) := m(x_1,\ldots,x_{j-1},a^2,x_{j+1},\ldots,x_n) \\
- m(x_1,\ldots,x_{j-1},a^1,x_{j+1},\ldots,x_n) \\
= \sum_{j \in \{1,2\}} (-1)^j m(x_1,\ldots,x_{j-1},a^j,x_{j+1},\ldots,x_n)
$$

and define

$$
\Delta_{a^1,a^2} := \prod_{j=1}^{n} \Delta^j_{a^1,a^2}m(x) \\
= \sum_{j \in \{1,2\}^n} (-1)^{j_1} \cdots (-1)^{j_n} m(a_1^{j_1},\ldots,a_n^{j_n}).
$$

Note that the above sum is taken over all vertices of the rectangle $(a^1,a^2)$ weighted with $+1$ if the vertex contains an even number of left endpoints and weighted with $-1$ if the vertex contains an odd number of left endpoints.

Then

$$
\mu((a,b]) = \Delta_{a,b}m.
$$

Of course in the case $n = 1$ this reduces to $\mu((a,b]) = m(b) - m(a)$. In the
case $n = 2$ we have $\mu((a, b]) = m(b_1, b_2) - m(b_1, a_2) - m(a_1, b_2) + m(a_1, a_2)$ which (for $0 \leq a \leq b$) is the measure of the rectangle with corners $0, b$, minus the measure of the rectangle on the left with corners $0, (a_1, b_2)$, minus the measure of the rectangle below with corners $0, (b_1, a_2)$, plus the measure of the rectangle with corners $0, a$ which has been subtracted twice. The general case can be handled recursively (Problem 8.14).

Hence we will again assume that a nondecreasing function $m : \mathbb{R}^n \to \mathbb{R}^n$ is given (i.e. $m(a) \leq m(b)$ for $a \leq b$). However, this time monotonicity is not enough as the following example shows.

Example. Let $\mu := \frac{1}{2} \delta_{(0,1)} + \frac{1}{2} \delta_{(1,0)} - \frac{1}{2} \delta_{(1,1)}$. Then the corresponding distribution function is increasing in each coordinate direction as the decrease due to the last term is compensated by the other two. However, $(8.25)$ will give $-\frac{1}{2}$ for any rectangle containing $(1,1)$ but not the other two points $(1,0)$ and $(0,1)$. \hfill \Box

Now we can show how to get a premeasure on $\bar{S}^n$.

Lemma 8.14. Let $m : \mathbb{R}^n \to \mathbb{R}^n$ be right continuous such that $\mu$ defined via $(8.28)$ is a nonnegative set function. Then $\mu$ gives rise to a unique $\sigma$-finite premeasure on the algebra $\bar{S}^n$ of finite unions of disjoint half-open rectangles.

Proof. We first need to show finite additivity. We call $A = \bigcup_k A_k$ a regular partition of $A = (a,b]$ if there are sequences $a_j = c_{j,0} < c_{j,1} < \cdots < c_{j,m_j} = b_j$ such that each rectangle $A_k$ is of the form

$$(c_{1,i-1}, c_{1,i}] \times \cdots \times (c_{n,i-1}, c_{n,i}].$$

That is, the sets $A_k$ are obtained by intersecting $A$ with the hyperplanes $x_j = c_{j,i}, 1 \leq i \leq m_j - 1, 1 \leq j \leq n$. Now let $A$ be bounded. Then additivity holds when partitioning $A$ into two sets by one hyperplane $x_j = c_{j,i}$ (note that in the sum over all vertices, the one containing $c_{j,i}$ instead of $a_j$ cancels with the one containing $c_{j,i}$ instead of $b_j$ as both have opposite signs by the very definition of $\Delta_a \delta m$). Hence applying this case recursively shows that additivity holds for regular partitions. Finally, for every partition we have a
corresponding regular subpartition. Moreover, the sets in this subpartitions can be lumped together into regular subpartions for each of the sets in the original partition. Hence the general case follows from the regular case. Finally, the case of unbounded sets \( A \) follows by taking limits.

The rest follows verbatim as in the previous lemma. \( \square \)

Again this premeasure gives rise to a measure.

**Theorem 8.15.** For every right continuous function \( m : \mathbb{R}^n \to \mathbb{R} \) such that
\[
\mu((a,b]) := \Delta_{a,b} m \geq 0, \quad \forall a \leq b,
\]
(8.29)
there exists a unique regular Borel measure \( \mu \) which extends the above definition.

**Proof.** As in the one-dimensional case existence follows from Theorem 8.9 together with Lemma 8.10 and uniqueness follows from Corollary 8.8. Again regularity will be postponed until Section 8.6. \( \square \)

**Example.** Choosing \( m(x) := \prod_{j=1}^{n} x_j \) we obtain the Lebesgue measure \( \lambda^n \) in \( \mathbb{R}^n \), which is the unique Borel measure satisfying
\[
\lambda^n((a,b]) = \prod_{j=1}^{n} (b_j - a_j).
\]

We collect some simple properties below.

(i) \( \lambda^n \) is regular.

(ii) \( \lambda^n \) is uniquely defined by its values on \( S^n \).

(iii) For every measurable set we have
\[
\lambda^n(A) = \inf \left\{ \sum_{m=1}^{\infty} \lambda^n(A_m) \middle| A \subseteq \bigcup_{m=1}^{\infty} A_m, A_m \in S^n \right\}
\]
where the infimum extends over all countable disjoint covers.

(iv) \( \lambda^n \) is translation invariant and up to normalization the only Borel measure with this property.

(i) and (ii) are part of Theorem 8.15. (iii). This will follow from the construction of \( \lambda^n \) via its outer measure in the following section. (iv). The previous item implies that \( \lambda^n \) is translation invariant. Moreover, let \( \mu \) be a second translation invariant measure. Denote by \( Q_r \) a cube with side length \( r > 0 \). Without loss we can assume \( \mu(Q_1) = 1 \). Since we can split \( Q_1 \) into \( m^n \) cubes of side length \( 1/m \), we see that \( \mu(Q_{1/m}) = m^{-n} \) by translation invariance and additivity. Hence we obtain \( \mu(Q_r) = r^n \) for every rational \( r \) and thus for every \( r \) by continuity from below. Proceeding like this we see that \( \lambda^n \) and \( \mu \) coincide on \( S^n \) and equality follows from item (ii). \( \diamond \)
Example. If \( m_j : \mathbb{R} \to \mathbb{R} \) are nondecreasing right continuous functions, then \( m(x) := \prod_{j=1}^n m_j(x_j) \) satisfies
\[
\Delta_{a,b} m = \prod_{j=1}^n (m_j(b_j) - m_j(a_j))
\]
and hence the requirements of Theorem 8.15 are fulfilled.

Problem 8.14. Let \( \mu \) be a Borel measure on \( \mathbb{R}^n \). For \( a, b \in \mathbb{R}^n \) set
\[
m(a,b) := \text{sign}(b-a)\mu\left( \bigcap_{j=1}^n (\min(a_j,b_j),\max(a_j,b_j)) \right)
\]
and \( m(x) := m(0,x) \). In particular, for \( a \leq b \) we have \( m(a,b) = \mu((a,b]) \).

Show that
\[
m(a,b) = \Delta_{a,b} m(c,\cdot)
\]
for arbitrary \( c \in \mathbb{R}^n \). (Hint: Start with evaluating \( \Delta_{a_j,b_j} m(c,\cdot) \).

Problem 8.15. Let \( \mu \) be a premeasure such that outer regularity (8.17) holds for every set in the algebra. Then the corresponding measure \( \mu \) from Theorem 8.9 is outer regular.

8.5. Measurable functions

The Riemann integral works by splitting the \( x \) coordinate into small intervals and approximating \( f(x) \) on each interval by its minimum and maximum. The problem with this approach is that the difference between maximum and minimum will only tend to zero (as the intervals get smaller) if \( f(x) \) is sufficiently nice. To avoid this problem, we can force the difference to go to zero by considering, instead of an interval, the set of \( x \) for which \( f(x) \) lies between two given numbers \( a < b \). Now we need the size of the set of these \( x \), that is, the size of the preimage \( f^{-1}((a,b)) \). For this to work, preimages of intervals must be measurable.

Let \( (X, \Sigma_X) \) and \( (Y, \Sigma_Y) \) be measurable spaces. A function \( f : X \to Y \) is called \textbf{measurable} if \( f^{-1}(A) \in \Sigma_X \) for every \( A \in \Sigma_Y \). When checking this condition it is useful to note that the collection of sets for which it holds,
\[
\{ A \subseteq Y | f^{-1}(A) \in \Sigma_X \},
\]
forms a \( \sigma \)-algebra on \( Y \) by \( f^{-1}(Y \setminus A) = X \setminus f^{-1}(A) \) and \( f^{-1}(\bigcup_j A_j) = \bigcup_j f^{-1}(A_j) \). Hence it suffices to check this condition for every set \( A \) in a collection of sets which generates \( \Sigma_Y \).

We will be mainly interested in the case where \( (Y, \Sigma_Y) = (\mathbb{R}^n, \mathcal{B}^n) \).

Lemma 8.16. A function \( f : X \to \mathbb{R}^n \) is measurable if and only if
\[
f^{-1}((a,\infty)) \in \Sigma \quad \forall a \in \mathbb{R}^n,
\]
(8.30)
where \((a, \infty) := \bigcap_{j=1}^{n}(a_j, \infty)\). In particular, a function \(f : X \to \mathbb{R}^n\) is measurable if and only if every component is measurable and a complex-valued function \(f : X \to \mathbb{C}^n\) is measurable if and only if both its real and imaginary parts are.

**Proof.** We need to show that \(\mathcal{B}^n\) is generated by rectangles of the above form. The \(\sigma\)-algebra generated by these rectangles also contains all open rectangles of the form \((a, b) := \bigcap_{j=1}^{n}(a_j, b_j)\), which form a base for the topology. \(\Box\)

Clearly the intervals \((a, \infty)\) can also be replaced by \([a, \infty)\), \((-\infty, a)\), or \((-\infty, a]\).

If \(X\) is a topological space and \(\Sigma\) the corresponding Borel \(\sigma\)-algebra, we will also call a measurable function **Borel function**. Note that, in particular,

**Lemma 8.17.** Let \((X, \Sigma_X), (Y, \Sigma_Y), (Z, \Sigma_Z)\) be topological spaces with their corresponding Borel \(\sigma\)-algebras. Any continuous function \(f : X \to Y\) is measurable. Moreover, if \(f : X \to Y\) and \(g : Y \to Z\) are measurable functions, then the composition \(g \circ f\) is again measurable.

The set of all measurable functions forms an algebra.

**Lemma 8.18.** Let \(X\) be a topological space and \(\Sigma\) its Borel \(\sigma\)-algebra. Suppose \(f, g : X \to \mathbb{R}\) are measurable functions. Then the sum \(f + g\) and the product \(fg\) are measurable.

**Proof.** Note that addition and multiplication are continuous functions from \(\mathbb{R}^2 \to \mathbb{R}\) and hence the claim follows from the previous lemma. \(\Box\)

Sometimes it is also convenient to allow \(\pm \infty\) as possible values for \(f\), that is, functions \(f : X \to \overline{\mathbb{R}}, \overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, \infty\}\). In this case \(A \subseteq \overline{\mathbb{R}}\) is called Borel if \(A \cap \mathbb{R}\) is. This implies that \(f : X \to \overline{\mathbb{R}}\) will be Borel if and only if \(f^{-1}(\pm \infty)\) are Borel and \(f : X \setminus f^{-1}(\{-\infty, \infty\}) \to \mathbb{R}\) is Borel. Since

\[
\{\pm \infty\} = \bigcap_n [n, +\infty], \quad \{-\infty\} = \overline{\mathbb{R}} \setminus \bigcup_n (-n, +\infty), \quad (8.31)
\]

we see that \(f : X \to \overline{\mathbb{R}}\) is measurable if and only if

\[
f^{-1}((a, \infty]) \in \Sigma \quad \forall a \in \mathbb{R}. \quad (8.32)
\]

Again the intervals \((a, \infty]\) can also be replaced by \([a, \infty], [-\infty, a]\), or \([-\infty, a]\). Moreover, we can generate a corresponding topology on \(\overline{\mathbb{R}}\) by intervals of the form \([-\infty, b)\), \((a, b)\), and \((a, \infty]\) with \(a, b \in \mathbb{R}\).
Hence it is not hard to check that the previous lemma still holds if one either avoids undefined expressions of the type $\infty - \infty$ and $\pm\infty \cdot 0$ or makes a definite choice, e.g., $\infty - \infty = 0$ and $\pm\infty \cdot 0 = 0$.

Moreover, the set of all measurable functions is closed under all important limiting operations.

**Lemma 8.19.** Suppose $f_n : X \to \mathbb{R}$ is a sequence of measurable functions. Then
\[
\inf_{n \in \mathbb{N}} f_n, \quad \sup_{n \in \mathbb{N}} f_n, \quad \liminf_{n \to \infty} f_n, \quad \limsup_{n \to \infty} f_n
\]
are measurable as well.

**Proof.** It suffices to prove that $\sup f_n$ is measurable since the rest follows from $\inf f_n = -\sup(-f_n)$, $\liminf f_n := \sup_n \inf_{k \geq n} f_k$, and $\limsup f_n := \inf_n \sup_{k \geq n} f_k$. But $(\sup f_n)^{-1}((a, \infty]) = \bigcup_n f_n^{-1}((a, \infty])$ are measurable and we are done. \qed

A few immediate consequences are worthwhile noting: It follows that if $f$ and $g$ are measurable functions, so are $\min(f, g)$, $\max(f, g)$, $|f| = \max(f, -f)$, and $f^\pm = \max(\pm f, 0)$. Furthermore, the pointwise limit of measurable functions is again measurable. Moreover, the set where the limit exists,
\[
\{ x \in X \mid \lim_{n \to \infty} f(x) \text{ exists} \} = \{ x \in X \mid \limsup_{n \to \infty} f(x) - \liminf_{n \to \infty} f(x) = 0 \},
\]
is measurable.

Sometimes the case of arbitrary suprema and infima is also of interest. In this respect the following observation is useful: Recall that a function $f : X \to \mathbb{R}$ is **lower semicontinuous** if the set $f^{-1}((a, \infty])$ is open for every $a \in \mathbb{R}$. Then it follows from the definition that the sup over an arbitrary collection of lower semicontinuous functions
\[
\overline{f}(x) := \sup_{\alpha} f_{\alpha}(x)
\]
is again lower semicontinuous (and hence measurable). Similarly, $f$ is **upper semicontinuous** if the set $f^{-1}((-\infty, a))$ is open for every $a \in \mathbb{R}$. In this case the infimum
\[
\underline{f}(x) := \inf_{\alpha} f_{\alpha}(x)
\]
is again upper semicontinuous. Note that $f$ is lower semicontinuous if and only if $-f$ is upper semicontinuous.

**Problem 8.16 (preimage $\sigma$-algebra).** Let $S \subseteq \mathcal{P}(Y)$. Show that $f^{-1}(S) := \{ f^{-1}(A) \mid A \in S \}$ is a $\sigma$-algebra if $S$ is. Conclude that $f^{-1}(\Sigma_Y)$ is the smallest $\sigma$-algebra on $X$ for which $f$ is measurable.
Problem 8.17. Let \( \{A_n\}_{n \in \mathbb{N}} \) be a partition for \( X, X = \bigcup_{n \in \mathbb{N}} A_n \). Let \( \Sigma = \Sigma(\{A_n\}_{n \in \mathbb{N}}) \) be the \( \sigma \)-algebra generated by these sets. Show that \( f : X \rightarrow \mathbb{R} \) is measurable if and only if it is constant on the sets \( A_n \).

Problem 8.18. Show that the supremum over lower semicontinuous functions is again lower semicontinuous.

Problem 8.19. Show that \( f \) is lower semicontinuous if and only if
\[
\liminf_{x \to x_0} f(x) \geq f(x_0), \quad x_0 \in X.
\]
Similarly, \( f \) is upper semicontinuous if and only if
\[
\limsup_{x \to x_0} f(x) \leq f(x_0), \quad x_0 \in X.
\]
Show that a lower semicontinuous function is also sequentially lower semicontinuous
\[
\liminf_{n \to \infty} f(x_n) \geq f(x_0), \quad x_n \rightarrow x_0, \quad x_0 \in X.
\]
Show that the converse is also true if \( X \) is a metric space. (Hint: Problem 0.14.)

8.6. How wild are measurable objects

In this section we want to investigate how far measurable objects are away from well-understood ones. The situation is intuitively summarized in what is known as Littlewood’s three principles of real analysis:

- Every (measurable) set is nearly a finite union of intervals.
- Every (measurable) function is nearly continuous.
- Every convergent sequence of (measurable) functions is nearly uniformly convergent.

As our first task we want to look at the first and show that measurable sets can be well approximated by using closed sets from the inside and open sets from the outside in nice spaces like \( \mathbb{R}^n \).

Lemma 8.20. Let \( X \) be a metric space and \( \mu \) a finite Borel measure. Then for every \( A \in \mathcal{B}(X) \) and any given \( \varepsilon > 0 \) there exists an open set \( O \) and a closed set \( C \) such that
\[
C \subseteq A \subseteq O \quad \text{and} \quad \mu(O \setminus C) \leq \varepsilon. \tag{8.37}
\]
The same conclusion holds for arbitrary Borel measures if there is a sequence of open sets \( U_n \uparrow X \) such that \( \overline{U}_n \subseteq U_{n+1} \) and \( \mu(U_n) < \infty \) (note that \( \mu \) is also \( \sigma \)-finite in this case).
Proof. To see that (8.37) holds we begin with the case when \( \mu \) is finite. Denote by \( \mathcal{A} \) the set of all Borel sets satisfying (8.37). Then \( \mathcal{A} \) contains every closed set \( C \): Given \( C \) define \( O_n := \{ x \in X | d(x, C) < 1/n \} \) and note that \( O_n \) are open sets which satisfy \( O_n \searrow C \). Thus by Theorem 8.5 (iii) \( \mu(O_n \setminus C) \to 0 \) and hence \( C \in \mathcal{A} \).

Moreover, \( \mathcal{A} \) is even a \( \sigma \)-algebra. That it is closed under complements is easy to see (note that \( \tilde{A} = X \setminus A \)). To see that it is closed under countable unions consider \( A = \bigcup_{n=1}^{\infty} A_n \) with \( A_n \in \mathcal{A} \). Then there are \( C_n, O_n \) such that \( \mu(O_n \setminus C_n) \leq \varepsilon 2^{-n-1} \). Now \( O := \bigcup_{n=1}^{\infty} O_n \) is open and \( C := \bigcup_{n=1}^{N} C_n \) is closed for any finite \( N \). Since \( \mu(A) \) is finite we can choose \( N \) sufficiently large such that \( \mu(\bigcup_{n=N+1}^{\infty} C_n \setminus C) \leq \varepsilon /2 \). Then we have found two sets of the required type: \( \mu(O \setminus C) \leq \sum_{n=1}^{\infty} \mu(O_n \setminus C_n) + \mu(\bigcup_{n=N+1}^{\infty} C_n \setminus C) \leq \varepsilon \). Thus \( \mathcal{A} \) is a \( \sigma \)-algebra containing the open sets, hence it is the entire Borel \( \sigma \)-algebra.

Now suppose \( \mu \) is not finite. Pick some \( x_0 \in X \) and set \( X_1 := U_2 \) and \( X_n := U_{n+1} \setminus \overline{U_{n-1}} \), \( n \geq 2 \). Note that \( X_{n+1} \cap X_n = U_n \setminus \overline{U_{n-1}} \) and \( X_n \cap X_m = \emptyset \) for \( |n-m| > 1 \). Let \( A_n = A \cap X_n \) and note that \( A = \bigcup_{n=0}^{\infty} A_n \).

By the finite case we can choose \( C_n \subseteq A_n \subseteq O_n \subseteq X_n \) such that \( \mu(O_n \setminus C_n) \leq \varepsilon 2^{-n-1} \). Now set \( C := \bigcup_{n} C_n \) and \( O := \bigcup_{n} O_n \) and observe that \( C \) is closed. Indeed, let \( x \in C \) and let \( x_j \) be some sequence from \( C \) converging to \( x \) then \( x \in U_n \) for some \( n \) and hence the sequence must eventually lie in \( C \cap U_n \subseteq \bigcup_{m \leq n} C_m \). Hence \( x \in \bigcup_{m \leq n} C_m = \bigcup_{m \leq n} C_m \subseteq C \). Finally, \( \mu(O \setminus C) \leq \sum_{n=0}^{\infty} \mu(O_n \setminus C_n) \leq \varepsilon \) as required. \( \square \)

This result immediately gives us outer regularity.

Corollary 8.21. Under the assumptions of the previous lemma
\[
\mu(A) = \inf_{O \supseteq A, O \text{ open}} \mu(O) = \sup_{C \subseteq A, C \text{ closed}} \mu(C) \tag{8.38}
\]
and \( \mu \) is outer regular.

Proof. Equation (8.38) follows from \( \mu(A) = \mu(O) - \mu(O \setminus A) = \mu(C) + \mu(A \setminus C) \). \( \square \)

If we strengthen our assumptions, we also get inner regularity. In fact, if we assume the sets \( U_n \) to be relatively compact, then the assumptions for the second case are equivalent to \( X \) being locally compact and separable by Lemma 0.25.

Corollary 8.22. If \( X \) is a \( \sigma \)-compact metric space, then every finite Borel measure is regular. If \( X \) is a locally compact separable metric space, then every Borel measure is regular.
8.6. How wild are measurable objects

Proof. By assumption there is a sequence of compact sets $K_n \nearrow X$ and for every increasing sequence of closed sets $C_n$ with $\mu(C_n) \to \mu(A)$ we also have compact sets $C_n \cap K_n$ with $\mu(C_n \cap K_n) \to \mu(A)$. In the second case we can choose relatively compact open sets $U_n$ as in Lemma 0.25 (iv) such that the assumptions of the previous theorem hold. Now argue as before using $K_n = \overline{U_n}$.

In particular, on a locally compact and separable space every Borel measure is automatically regular and $\sigma$-finite. For example this holds for $X = \mathbb{R}^n$ (or $X = \mathbb{C}^n$).

An inner regular measure on a Hausdorff space which is locally finite (every point has a neighborhood of finite measure) is called a Radon measure. Accordingly every Borel measure on $\mathbb{R}^n$ (or $\mathbb{C}^n$) is automatically a Radon measure.

Example. Since Lebesgue measure on $\mathbb{R}$ is regular, we can cover the rational numbers by an open set of arbitrary small measure (it is also not hard to find such a set directly) but we cannot cover it by an open set of measure zero (since any open set contains an interval and hence has positive measure). However, if we slightly extend the family of admissible sets, this will be possible.

Looking at the Borel $\sigma$-algebra the next general sets after open sets are countable intersections of open sets, known as $G_\delta$ sets (here $G$ and $\delta$ stand for the German words *Gebiet* and *Durchschnitt*, respectively). The next general sets after closed sets are countable unions of closed sets, known as $F_\sigma$ sets (here $F$ and $\sigma$ stand for the French words *fermé* and *somme*, respectively). Of course the complement of a $G_\delta$ set is an $F_\sigma$ set and vice versa.

Example. The irrational numbers are a $G_\delta$ set in $\mathbb{R}$. To see this, let $x_n$ be an enumeration of the rational numbers and consider the intersection of the open sets $O_n := \mathbb{R} \setminus \{x_n\}$. The rational numbers are hence an $F_\sigma$ set.

Corollary 8.23. Suppose $X$ is locally compact and separable and $\mu$ a Borel measure. A set in $X$ is Borel if and only if it differs from a $G_\delta$ set by a Borel set of measure zero. Similarly, a set in $X$ is Borel if and only if it differs from an $F_\sigma$ set by a Borel set of measure zero.

Proof. Since $G_\delta$ sets are Borel, only the converse direction is nontrivial. By Lemma 8.20 we can find open sets $O_n$ such that $\mu(O_n \setminus A) \leq 1/n$. Now let $G := \bigcap_n O_n$. Then $\mu(G \setminus A) \leq \mu(O_n \setminus A) \leq 1/n$ for any $n$ and thus $\mu(G \setminus A) = 0$. The second claim is analogous.

A similar result holds for convergence.
Theorem 8.24 (Egorov). Let $\mu$ be a finite measure and $f_n$ be a sequence of complex-valued measurable functions converging pointwise to a function $f$ for a.e. $x \in X$. Then for every $\varepsilon > 0$ there is a set $A$ of size $\mu(A) < \varepsilon$ such that $f_n$ converges uniformly on $X \setminus A$.

Proof. Let $A_0$ be the set where $f_n$ fails to converge. Set

$$A_{N,k} := \bigcup_{n \geq N} \{ x \in X \mid |f_n(x) - f(x)| \geq \frac{1}{k} \}, \quad A_k := \bigcap_{N \in \mathbb{N}} A_{N,k}$$

and note that $A_{N,k} \downarrow A_k \subseteq A_0$ as $N \to \infty$ (with $k$ fixed). Hence by continuity from above $\mu(A_{N,k}) \to \mu(A_k) = 0$. Hence for every $k$ there is some $N_k$ such that $\mu(A_{N_k,k}) < \frac{\varepsilon}{2^k}$. Then $A = \bigcup_{k \in \mathbb{N}} A_{N_k,k}$ satisfies $\mu(A) < \varepsilon$. Now note that $x \not\in A$ implies that for every $k$ we have $x \not\in A_{N_k,k}$. Thus $|f_n(x) - f(x)| < \frac{1}{k}$ for $n \geq N_k$. Hence $f_n$ converges uniformly away from $A$. $\square$

Example. The example $f_n := \chi_{[n,n+1]} \to 0$ on $X = \mathbb{R}$ with Lebesgue measure shows that the finiteness assumption is important. In fact, suppose there is a set $A$ of size less than 1 (say). Then every interval $[m, m+1]$ contains a point $x_m$ not in $A$ and thus $|f_m(x_m) - 0| = 1$. $\diamond$

To end this section let us briefly discuss the third principle, namely that bounded measurable functions can be well approximated by continuous functions (under suitable assumptions on the measure). We will discuss this in detail in Section 10.4. At this point we only mention that in such a situation Egorov’s theorem implies that the convergence is uniform away from a small set and hence our original function will be continuous restricted to the complement of this small set. This is known as Luzin’s theorem (cf. Theorem 10.16). Note however that this does not imply that measurable functions are continuous at every point of this complement! The characteristic function of the irrational numbers is continuous when restricted to the irrational numbers but it is not continuous at any point when regarded as a function of $\mathbb{R}$.

Problem 8.20. Show directly (without using regularity) that for every $\varepsilon > 0$ there is an open set $O$ of Lebesgue measure $|O| < \varepsilon$ which covers the rational numbers.

Problem 8.21. A finite Borel measure is regular if and only if for every Borel set $A$ and every $\varepsilon > 0$ there is an open set $O$ and a compact set $K$ such that $K \subseteq A \subseteq O$ and $\mu(O \setminus K) < \varepsilon$.

Problem 8.22. A sequence of measurable functions $f_n$ converges in measure to a measurable function $f$ if $\lim_{n \to \infty} \mu(\{ x \mid |f_n(x) - f(x)| \geq \varepsilon \}) = 0$ for every $\varepsilon > 0$. Show that if $\mu$ is finite and $f_n \to f$ a.e. then $f_n \overset{\mu}{\to} f$ in
8.8. Appendix: Equivalent definitions for the outer Lebesgue measure

In this appendix we want to show that the type of sets used in the definition of the outer Lebesgue measure \( \lambda^* \) on \( \mathbb{R}^n \) play no role. You can cover the set...
by half-closed, closed, open rectangles (which is easy to see) or even replace rectangles by balls (which follows from the following lemma). To this end observe that by (8.6)

$$\lambda^n(B_r(x)) = V_n r^n$$

(8.39)

where $V_n := \lambda^n(B_1(0))$ is the volume of the unit ball which is computed explicitly in Section 9.3. Will will write $|A| := \lambda^n(A)$ for the Lebesgue measure of a Borel set for brevity. We first establish a covering lemma which is of independent interest.

**Lemma 8.27** (Vitali covering lemma). Let $O \subseteq \mathbb{R}^n$ be an open set and $\delta > 0$ fixed. Let $C$ be a collection of balls such that every open subset of $O$ contains at least one ball from $C$. Then there exists a countable set of disjoint open balls from $C$ of radius at most $\delta$ such that $O = N \cup \bigcup_j B_j$ with $N$ a Lebesgue null set.

**Proof.** Let $O$ have finite outer measure. Start with all balls which are contained in $O$ and have radius at most $\delta$. Let $R$ be the supremum of the radii of all these balls and take a ball $B_1$ of radius more than $R/2$. Now consider $O \setminus B_1$ and proceed recursively. If this procedure terminates we are done (the missing points must be contained in the boundary of the chosen balls which has measure zero). Otherwise we obtain a sequence of balls $B_j$ whose radii must converge to zero since $\sum_{j=1}^\infty |B_j| \leq |O|$. Now fix $m$ and let $x \in O \setminus \bigcup_{j=1}^m B_j$. Then there must be a ball $B_0 = B_r(x) \subseteq O \setminus \bigcup_{j=1}^m B_j$. Moreover, there must be a first ball $B_k$ with $B_0 \cap B_k \neq \emptyset$ (otherwise all $B_k$ for $k > m$ must have radius larger than $\frac{r}{2}$ violating the fact that they converge to zero). By assumption $k > m$ and hence $r$ must be smaller than two times the radius of $B_k$ (since both balls are available in the $k$th step). So the distance of $x$ to the center of $B_k$ must be less then three times the radius of $B_k$. Now if $B_k$ is a ball with the same center but three times the radius of $B_k$, then $x$ is contained in $B_k$ and hence all missing points from $\bigcup_{j=1}^m B_j$ are either boundary points (which are of measure zero) or contained in $\bigcup_{k>m} B_k$ whose measure $|\bigcup_{k>m} B_k| \leq 3^n \sum_{k>m} |B_k| \to 0$ as $m \to \infty$.

If $|O| = \infty$ consider $O_m = O \cap (B_{m+1}(0) \setminus B_m(0))$ and note that $O = N \cup \bigcup_m O_m$ where $N$ is a set of measure zero. □

Note that in the one-dimensional case open balls are open intervals and we have the stronger result that every open set can be written as a countable union of disjoint intervals (Problem 0.19).

Now observe that in the definition of outer Lebesgue measure we could replace half-open rectangles by open rectangles (show this). Moreover, every open rectangle can be replaced by a disjoint union of open balls up to a set of measure zero by the Vitali covering lemma. Consequently, the Lebesgue
outer measure can be written as
\[ \lambda^n(A) = \inf \left\{ \sum_{k=1}^{\infty} |A_k| \bigg| A \subseteq \bigcup_{k=1}^{\infty} A_k, \ A_k \in \mathcal{C} \right\}, \] (8.40)
where \( \mathcal{C} \) could be the collection of all closed rectangles, half-open rectangles, open rectangles, closed balls, or open balls.
Integration

Now that we know how to measure sets, we are able to introduce the Lebesgue integral. As already mentioned, in the case of the Riemann integral, the domain of the function is split into intervals leading to an approximation by step functions, that is, linear combinations of characteristic functions of intervals. In the case of the Lebesgue integral we split the range into intervals and consider their preimages. This leads to an approximation by simple functions, that is, linear combinations of characteristic functions of arbitrary (measurable) sets.

9.1. Integration — Sum me up, Henri

Throughout this section \((X, \Sigma, \mu)\) will be a measure space. A measurable function \(s : X \to \mathbb{R}\) is called simple if its image is finite; that is, if

\[
s = \sum_{j=1}^{p} \alpha_j \chi_{A_j}, \quad \text{Ran}(s) =: \{\alpha_j\}_{j=1}^{p}, \quad A_j := s^{-1}(\alpha_j) \in \Sigma. \tag{9.1}
\]

Here \(\chi_A\) is the characteristic function of \(A\); that is, \(\chi_A(x) := 1\) if \(x \in A\) and \(\chi_A(x) := 0\) otherwise. Note that \(\bigcup_{j=1}^{p} A_j = X\). Moreover, the set of simple functions \(S(X, \mu)\) is a vector space and while there are different ways of writing a simple function as a linear combination of characteristic functions, the representation (9.1) is unique.

For a nonnegative simple function \(s\) as in (9.1) we define its integral as

\[
\int_A s \, d\mu := \sum_{j=1}^{p} \alpha_j \mu(A_j \cap A). \tag{9.2}
\]

Here we use the convention \(0 \cdot \infty = 0\).
Lemma 9.1. The integral has the following properties:

(i) \( \int_A s \, d\mu = \int_X \chi_A s \, d\mu. \)

(ii) \( \int_{\bigcup_{n=1}^{\infty} A_n} s \, d\mu = \sum_{n=1}^{\infty} \int_{A_n} s \, d\mu. \)

(iii) \( \int_A \alpha s \, d\mu = \alpha \int_A s \, d\mu, \alpha \geq 0. \)

(iv) \( \int_A (s + t) \, d\mu = \int_A s \, d\mu + \int_A t \, d\mu. \)

(v) \( A \subseteq B \Rightarrow \int_A s \, d\mu \leq \int_B s \, d\mu. \)

(vi) \( s \leq t \Rightarrow \int_A s \, d\mu \leq \int_A t \, d\mu. \)

Proof. (i) is clear from the definition. (ii) follows from \( \sigma \)-additivity of \( \mu. \) (iii) is obvious. (iv) Let \( s = \sum_j \alpha_j \chi_{A_j}, \; t = \sum_j \beta_j \chi_{B_j} \) as in (9.1) and abbreviate \( C_{jk} = (A_j \cap B_k) \cap A. \) Note \( \bigcup_{j,k} C_{jk} = A. \) Then by (ii),
\[
\int_A (s + t) \, d\mu = \sum_{j,k} \int_{C_{jk}} (s + t) \, d\mu = \sum_{j,k} (\alpha_j + \beta_k) \mu(C_{jk}) \]
\[
= \sum_{j,k} \left( \int_{C_{jk}} s \, d\mu + \int_{C_{jk}} t \, d\mu \right) = \int_A s \, d\mu + \int_A t \, d\mu. \]

(v) follows from monotonicity of \( \mu. \) (vi) follows since by (iv) we can write \( s = \sum_j \alpha_j \chi_{C_j}, \; t = \sum_j \beta_j \chi_{C_j} \) where, by assumption, \( \alpha_j \leq \beta_j. \)

Our next task is to extend this definition to nonnegative measurable functions by
\[
\int_A f \, d\mu := \sup_{\text{simple functions } s \leq f} \int_A s \, d\mu, \quad (9.3)
\]
where the supremum is taken over all simple functions \( s \leq f. \) By item (vi) from our previous lemma this agrees with (9.2) if \( f \) is simple. Note that, except for possibly (ii) and (iv), Lemma 9.1 still holds for arbitrary nonnegative functions \( s, t. \)

Theorem 9.2 (Monotone convergence, Beppo Levi’s theorem). Let \( f_n \) be a monotone nondecreasing sequence of nonnegative measurable functions, \( f_n \nearrow f. \) Then
\[
\int_A f_n \, d\mu \to \int_A f \, d\mu. \quad (9.4)
\]

Proof. By property (vi), \( \int_A f_n \, d\mu \) is monotone and converges to some number \( \alpha. \) By \( f_n \leq f \) and again (vi) we have
\[
\alpha \leq \int_A f \, d\mu.
\]
To show the converse, let $s$ be simple such that $s \leq f$ and let $\theta \in (0, 1)$. Put $A_n := \{ x \in A | f_n(x) \geq \theta s(x) \}$ and note $A_n \nearrow A$ (show this). Then
\[
\int_A f_n \, d\mu \geq \int_{A_n} f_n \, d\mu \geq \theta \int_{A_n} s \, d\mu.
\]
Letting $n \to \infty$ using (ii), we see
\[
\alpha \geq \theta \int_A s \, d\mu.
\]
Since this is valid for every $\theta < 1$, it still holds for $\theta = 1$. Finally, since $s \leq f$ is arbitrary, the claim follows. \qed

In particular
\[
\int_A f \, d\mu = \lim_{n \to \infty} \int_A s_n \, d\mu,
\]
for every monotone sequence $s_n \nearrow f$ of simple functions. Note that there is always such a sequence, for example,
\[
s_n(x) := \sum_{k=0}^{n^2} \frac{k}{2^n} \chi_{f^{-1}(A_k)}(x), \quad A_k := \left[ \frac{k}{2^n}, \frac{k+1}{2^n} \right), A_{n2^n} := [n, \infty).
\]
By construction $s_n$ converges uniformly if $f$ is bounded, since $0 \leq f(x) - s_n(x) < \frac{1}{2^n}$ if $f(x) \leq n$.

Now what about the missing items (ii) and (iv) from Lemma 9.1? Since limits can be spread over sums, item (iv) holds, and (ii) also follows directly from the monotone convergence theorem. We even have the following result:

**Lemma 9.3.** If $f \geq 0$ is measurable, then $d\nu = f \, d\mu$ defined via
\[
\nu(A) := \int_A f \, d\mu
\]
is a measure such that
\[
\int_A g \, d\nu = \int_A g f \, d\mu
\]
for every measurable function $g$.

**Proof.** As already mentioned, additivity of $\nu$ is equivalent to linearity of the integral and $\sigma$-additivity follows from Lemma 9.1 (ii):
\[
\nu\left( \bigcup_{n=1}^\infty A_n \right) = \int_{\bigcup_{n=1}^\infty A_n} f \, d\mu = \sum_{n=1}^\infty \int_{A_n} f \, d\mu = \sum_{n=1}^\infty \nu(A_n).
\]
The second claim holds for simple functions and hence for all functions by construction of the integral. \qed

If $f_n$ is not necessarily monotone, we have at least
**Theorem 9.4** (Fatou’s lemma). If $f_n$ is a sequence of nonnegative measurable function, then

$$\int_A \liminf_{n \to \infty} f_n \, d\mu \leq \liminf_{n \to \infty} \int_A f_n \, d\mu. \quad (9.9)$$

**Proof.** Set $g_n := \inf_{k \geq n} f_k$ such that $g_n \nearrow \liminf_n f_n$. Then $g_n \leq f_n$ implying

$$\int_A g_n \, d\mu \leq \int_A f_n \, d\mu.$$

Now take the lim inf on both sides and note that by the monotone convergence theorem

$$\liminf_{n \to \infty} \int_A g_n \, d\mu = \lim_{n \to \infty} \int_A g_n \, d\mu = \int_A \lim_{n \to \infty} g_n \, d\mu = \int_A \liminf_{n \to \infty} f_n \, d\mu,$$

proving the claim. □

**Example.** Consider $f_n := \chi_{[n,n+1]}$. Then $\lim_{n \to \infty} f_n(x) = 0$ for every $x \in \mathbb{R}$. However, $\int_{\mathbb{R}} f_n(x) \, dx = 1$. This shows that the inequality in Fatou’s lemma cannot be replaced by equality in general. ◦

If the integral is finite for both the positive and negative part $f^\pm = \max(\pm f, 0)$ of an arbitrary measurable function $f$, we call $f$ *integrable* and set

$$\int_A f \, d\mu := \int_A f^+ \, d\mu - \int_A f^- \, d\mu. \quad (9.10)$$

Similarly, we handle the case where $f$ is complex-valued by calling $f$ integrable if both the real and imaginary part are and setting

$$\int_A f \, d\mu := \int_A \text{Re}(f) \, d\mu + i \int_A \text{Im}(f) \, d\mu. \quad (9.11)$$

Clearly $f$ is integrable if and only if $|f|$ is. The set of all integrable functions is denoted by $L^1(X, d\mu)$.

**Lemma 9.5.** The integral is linear and Lemma 9.1 holds for integrable functions $s,t$.

Furthermore, for all integrable functions $f,g$ we have

$$|\int_A f \, d\mu| \leq \int_A |f| \, d\mu \quad (9.12)$$

and (triangle inequality)

$$\int_A |f + g| \, d\mu \leq \int_A |f| \, d\mu + \int_A |g| \, d\mu. \quad (9.13)$$

In the first case we have equality if and only if $f(x) = e^{i\theta} |f(x)|$ for a.e. $x$ and some real number $\theta$. In the second case we have equality if and only if
9.1. Integration — Sum me up, Henri

\[ f(x) = e^{i\theta(x)}|f(x)|, \quad g(x) = e^{i\theta(x)}|g(x)| \] for a.e. \( x \) and for some real-valued function \( \theta \).

**Proof.** Linearity and Lemma 9.1 are straightforward to check. To see (9.12) put \( \alpha := \frac{z}{|z|} \), where \( z := \int_A f\,d\mu \) (without restriction \( z \neq 0 \)). Then

\[ |\int_A f\,d\mu| = \alpha \int_A f\,d\mu = \int_A \alpha f\,d\mu = \int_A \text{Re}(\alpha f)\,d\mu \leq \int_A |f|\,d\mu, \]

proving (9.12). The second claim follows from \(|f + g| \leq |f| + |g|\). The cases of equality are straightforward to check.

**Lemma 9.6.** Let \( f \) be measurable. Then

\[ \int_X |f|\,d\mu = 0 \quad \iff \quad f(x) = 0 \quad \mu - \text{a.e.} \quad (9.14) \]

Moreover, suppose \( f \) is nonnegative or integrable. Then

\[ \mu(A) = 0 \quad \Rightarrow \quad \int_A f\,d\mu = 0. \quad (9.15) \]

**Proof.** Observe that we have \( A := \{x \mid f(x) \neq 0\} = \bigcup_n A_n \), where \( A_n := \{x \mid |f(x)| \geq \frac{1}{n}\} \). If \( \int_X |f|\,d\mu = 0 \) we must have \( \mu(A_n) \leq n \int_{A_n} |f|\,d\mu = 0 \) for every \( n \) and hence \( \mu(A) = \lim_{n \to \infty} \mu(A_n) = 0 \).

The converse will follow from (9.15) since \( \mu(A) = 0 \) (with \( A \) as before) implies \( \int_X |f|\,d\mu = \int_A |f|\,d\mu = 0 \).

Finally, to see (9.15) note that by our convention \( 0 \cdot \infty = 0 \) it holds for any simple function and hence for any nonnegative \( f \) by definition of the integral (9.3). Since any function can be written as a linear combination of four nonnegative functions this also implies the case when \( f \) is integrable.

Note that the proof also shows that if \( f \) is not 0 almost everywhere, there is an \( \varepsilon > 0 \) such that \( \mu(\{x \mid |f(x)| \geq \varepsilon\}) > 0 \).

In particular, the integral does not change if we restrict the domain of integration to a support of \( \mu \) or if we change \( f \) on a set of measure zero. In particular, functions which are equal a.e. have the same integral.

**Example.** If \( \mu(x) := \Theta(x) \) is the Dirac measure at 0, then

\[ \int_{\mathbb{R}} f(x)\,d\mu(x) = f(0). \]

In fact, the integral can be restricted to any support and hence to \( \{0\} \).

If \( \mu(x) := \sum_n \alpha_n \Theta(x - x_n) \) is a sum of Dirac measures, \( \Theta(x) \) centered at \( x = 0 \), then (Problem 9.2)

\[ \int_{\mathbb{R}} f(x)\,d\mu(x) = \sum_n \alpha_n f(x_n). \]
Hence our integral contains sums as special cases.

Finally, our integral is well behaved with respect to limiting operations. We first state a simple generalization of Fatou’s lemma.

**Lemma 9.7** (generalized Fatou lemma). If \( f_n \) is a sequence of real-valued measurable function and \( g \) some integrable function. Then

\[
\int_A \liminf_{n \to \infty} f_n \, d\mu \leq \liminf_{n \to \infty} \int_A f_n \, d\mu \quad (9.16)
\]

if \( g \leq f_n \) and

\[
\limsup_{n \to \infty} \int_A f_n \, d\mu \leq \int_A \limsup_{n \to \infty} f_n \, d\mu \quad (9.17)
\]

if \( f_n \leq g \).

**Proof.** To see the first apply Fatou’s lemma to \( f_n - g \) and subtract \( \int_A g \, d\mu \) on both sides of the result. The second follows from the first using \( \liminf(-f_n) = -\limsup f_n \).

If in the last lemma we even have \( |f_n| \leq g \), we can combine both estimates to obtain

\[
\int_A \liminf_{n \to \infty} f_n \, d\mu \leq \liminf_{n \to \infty} \int_A f_n \, d\mu \leq \limsup_{n \to \infty} \int_A f_n \, d\mu \leq \int_A \limsup_{n \to \infty} f_n \, d\mu, \quad (9.18)
\]

which is known as **Fatou–Lebesgue theorem**. In particular, in the special case where \( f_n \) converges we obtain

**Theorem 9.8** (Dominated convergence). Let \( f_n \) be a convergent sequence of measurable functions and set \( f := \lim_{n \to \infty} f_n \). Suppose there is an integrable function \( g \) such that \( |f_n| \leq g \). Then \( f \) is integrable and

\[
\lim_{n \to \infty} \int_A f_n \, d\mu = \int_A f \, d\mu. \quad (9.19)
\]

**Proof.** The real and imaginary parts satisfy the same assumptions and hence it suffices to prove the case where \( f_n \) and \( f \) are real-valued. Moreover, since \( \liminf f_n = \limsup f_n = f \) equation (9.18) establishes the claim.

Remark: Since sets of measure zero do not contribute to the value of the integral, it clearly suffices if the requirements of the dominated convergence theorem are satisfied almost everywhere (with respect to \( \mu \)).

**Example.** Note that the existence of \( g \) is crucial: The functions \( f_n(x) := \frac{1}{2n} \chi_{[-n,n]}(x) \) on \( \mathbb{R} \) converge uniformly to 0 but \( \int_{\mathbb{R}} f_n(x) \, dx = 1 \).

In calculus one frequently uses the notation \( \int_a^b f(x) \, dx \). In case of general Borel measures on \( \mathbb{R} \) this is ambiguous and one needs to mention to what
extend the boundary points contribute to the integral. Hence we define
\[
\int_a^b f \, d\mu := \begin{cases} 
\int_{(a,b]} f \, d\mu, & a < b, \\
0, & a = b, \\
-\int_{(b,a]} f \, d\mu, & b < a.
\end{cases}
\] (9.20)
such that the usual formulas
\[
\int_a^b f \, d\mu = \int_a^c f \, d\mu + \int_c^b f \, d\mu
\] (9.21)
remain true. Note that this is also consistent with \(\mu(x) = \int_0^x d\mu\).

**Example.** Let \(f \in C[a,b]\), then the sequence of simple functions
\[
s_n(x) := \sum_{j=1}^n f(x_j) \chi_{(x_{j-1},x_j]}(x), \quad x_j = a + \frac{b-a}{n} j
\]
converges to \(f(x)\) and hence the integral coincides with the limit of the Riemann–Stieltjes sums:
\[
\int_a^b f \, d\mu = \lim_{n \to \infty} \sum_{j=1}^n f(x_j)(\mu(x_j) - \mu(x_{j-1})).
\]
Moreover, the equidistant partition could of course be replaced by an arbitrary partition \(\{x_0 = a < x_1 < \cdots < x_n = b\}\) whose length \(\max_{1 \leq j \leq n}(x_j - x_{j-1})\) tends to 0. In particular, for \(\mu(x) = x\) we get the usual Riemann sums and hence the Lebesgue integral coincides with the Riemann integral at least for continuous functions. Further details on the connection with the Riemann integral will be given in Section 9.5.

Even without referring to the Riemann integral, one can easily identify the Lebesgue integral as an antiderivative: Given a continuous function \(f \in C(a,b)\) which is integrable over \((a,b)\) we can introduce
\[
F(x) := \int_a^x f(y) \, dy, \quad x \in (a,b).
\] (9.22)
Then one has
\[
\frac{F(x+\varepsilon) - F(x)}{\varepsilon} = f(x) + \frac{1}{\varepsilon} \int_x^{x+\varepsilon} (f(y) - f(x)) \, dy
\]
and
\[
\limsup_{\varepsilon \to 0} \frac{1}{\varepsilon} \int_x^{x+\varepsilon} |f(y) - f(x)| \, dy \leq \limsup_{\varepsilon \to 0} \sup_{y \in (x,x+\varepsilon]} |f(y) - f(x)| = 0
\]
by the continuity of \(f\) at \(x\). Thus \(F \in C^1(a,b)\) and
\[
F'(x) = f(x),
\]
which is a variant of the **fundamental theorem of calculus**. This tells us that the integral of a continuous function $f$ can be computed in terms of its antiderivative and, in particular, all tools from calculus like integration by parts or integration by substitution are readily available for the Lebesgue integral on $\mathbb{R}$. A generalization of the fundamental theorem of calculus will be given in Theorem 11.49.

**Example.** Another fact worthwhile mentioning is that integrals with respect to Borel measures $\mu$ on $\mathbb{R}$ can be easily computed if the distribution function is continuously differentiable. In this case $\mu([a, b)) = \mu(b) - \mu(a) = \int_a^b \mu'(x)dx$ implying that $d\mu(x) = \mu'(x)dx$ in the sense of Lemma 9.3. Moreover, it even suffices that the distribution function is piecewise continuously differentiable such that the fundamental theorem of calculus holds. ⋆

**Problem 9.1.** Show the inclusion exclusion principle:

$$\mu(A_1 \cup \cdots \cup A_n) = \sum_{k=1}^n (-1)^{k+1} \sum_{1 \leq i_1 < \cdots < i_k \leq n} \mu(A_{i_1} \cap \cdots \cap A_{i_k}).$$

(Hint: $\chi_{A_1 \cup \cdots \cup A_n} = 1 - \prod_{i=1}^n (1 - \chi_{A_i}).$)

**Problem 9.2.** Consider a countable set of measures $\mu_n$ and numbers $\alpha_n \geq 0$. Let $\mu := \sum \alpha_n \mu_n$ and show

$$\int_A f d\mu = \sum \alpha_n \int_A f d\mu_n \tag{9.23}$$

for any measurable function which is either nonnegative or integrable.

**Problem 9.3** (Fatou for sets). Define

$$\liminf_{n \to \infty} A_n := \bigcup_{k \in \mathbb{N}} \bigcap_{n \geq k} A_n, \quad \limsup_{n \to \infty} A_n := \bigcap_{k \in \mathbb{N}} \bigcup_{n \geq k} A_n.$$

That is, $x \in \liminf A_n$ if $x \in A_n$ eventually and $x \in \limsup A_n$ if $x \in A_n$ infinitely often. In particular, $\liminf A_n \subseteq \limsup A_n$. Show

$$\mu(\liminf A_n) \leq \liminf_{n} \mu(A_n)$$

and

$$\limsup_{n} \mu(A_n) \leq \mu(\limsup A_n), \quad \text{if } \mu(\bigcup A_n) < \infty.$$

(Hint: Show $\liminf \chi_{A_n} = \chi_{\liminf A_n}$ and $\limsup \chi_{A_n} = \chi_{\limsup A_n}$.)

**Problem 9.4** (Borel–Cantelli lemma). Show that $\sum \mu(A_n) < \infty$ implies $\mu(\limsup A_n) = 0$. Give an example which shows that the converse does not hold in general.
Problem 9.5. Show that if $f_n \to f$ in measure (cf. Problem 8.22), then there is a subsequence which converges a.e. (Hint: Choose the subsequence such that the assumptions of the Borel–Cantelli lemma are satisfied.)

Problem 9.6. Consider $X = [0, 1]$ with Lebesgue measure. Show that a.e. convergence does not stem from a topology. (Hint: Start with a sequence which converges in measure to zero but not a.e. By the previous problem you can extract a subsequence which converges a.e. Now compare this with Lemma 0.5.)

Problem 9.7. Let $(X, \Sigma)$ be a measurable space. Show that the set $B(X)$ of bounded measurable functions with the sup norm is a Banach space. Show that the set $S(X)$ of simple functions is dense in $B(X)$. Show that the integral is a bounded linear functional on $B(X)$ if $\mu(X) < \infty$. (Hence Theorem 1.12 could be used to extend the integral from simple to bounded measurable functions.)

Problem 9.8. Show that the monotone convergence holds for nondecreasing sequences of real-valued measurable functions $f_n \nearrow f$ provided $f_1$ is integrable.

Problem 9.9. Show that the dominated convergence theorem implies (under the same assumptions)

$$\lim_{n \to \infty} \int |f_n - f| \, d\mu = 0.$$

Problem 9.10 (Bounded convergence theorem). Suppose $\mu$ is a finite measure and $f_n$ is a bounded sequence of measurable functions which converges in measure to a measurable function $f$ (cf. Problem 8.22). Show that

$$\lim_{n \to \infty} \int f_n \, d\mu = \int f \, d\mu.$$

Problem 9.11. Consider

$$m(x) = \begin{cases} 0, & x < 0, \\ \frac{x}{2}, & 0 \leq x < 1, \\ 1, & 1 \leq x, \end{cases}$$

and let $\mu$ be the associated measure. Compute $\int_{\mathbb{R}} x \, d\mu(x)$.

Problem 9.12. Let $\mu(A) < \infty$ and $f$ be an integrable function satisfying $f(x) \leq M$. Show that

$$\int_{A} f \, d\mu \leq M \mu(A)$$

with equality if and only if $f(x) = M$ for a.e. $x \in A$. 
Problem 9.13. Let $X \subseteq \mathbb{R}$, $Y$ be some measure space, and $f : X \times Y \to \mathbb{C}$. Suppose $y \mapsto f(x, y)$ is measurable for every $x$ and $x \mapsto f(x, y)$ is continuous for every $y$. Show that

$$F(x) := \int_A f(x, y) \, d\mu(y) \quad (9.24)$$

is continuous if there is an integrable function $g(y)$ such that $|f(x, y)| \leq g(y)$.

Problem 9.14. Let $X \subseteq \mathbb{R}$, $Y$ be some measure space, and $f : X \times Y \to \mathbb{C}$. Suppose $y \mapsto f(x, y)$ is integrable for all $x$ and $x \mapsto f(x, y)$ is differentiable for a.e. $y$. Show that

$$F(x) := \int_Y f(x, y) \, d\mu(y) \quad (9.25)$$

is differentiable if there is an integrable function $g(y)$ such that $|\frac{\partial}{\partial x} f(x, y)| \leq g(y)$. Moreover, $y \mapsto \frac{\partial}{\partial x} f(x, y)$ is measurable and

$$F'(x) = \int_Y \frac{\partial}{\partial x} f(x, y) \, d\mu(y) \quad (9.26)$$
in this case. (See Problem 11.44 for an extension.)

9.2. Product measures

Let $\mu_1$ and $\mu_2$ be two measures on $\Sigma_1$ and $\Sigma_2$, respectively. Let $\Sigma_1 \otimes \Sigma_2$ be the $\sigma$-algebra generated by rectangles of the form $A_1 \times A_2$.

Example. Let $\mathcal{B}$ be the Borel sets in $\mathbb{R}$. Then $\mathcal{B}^2 = \mathcal{B} \otimes \mathcal{B}$ are the Borel sets in $\mathbb{R}^2$ (since the rectangles are a basis for the product topology).

Any set in $\Sigma_1 \otimes \Sigma_2$ has the section property; that is,

Lemma 9.9. Suppose $A \in \Sigma_1 \otimes \Sigma_2$. Then its sections

$$A_1(x_2) := \{x_1 | (x_1, x_2) \in A\} \quad \text{and} \quad A_2(x_1) := \{x_2 | (x_1, x_2) \in A\} \quad (9.27)$$

are measurable.

Proof. Denote all sets $A \in \Sigma_1 \otimes \Sigma_2$ with the property that $A_1(x_2) \in \Sigma_1$ by $S$. Clearly all rectangles are in $S$ and it suffices to show that $S$ is a $\sigma$-algebra. Now, if $A \in S$, then $(A')_1(x_2) = (A_1(x_2))' \in \Sigma_1$ and thus $S$ is closed under complements. Similarly, if $A_n \in S$, then $(\bigcup_n A_n)_1(x_2) = \bigcup_n (A_n)_1(x_2)$ shows that $S$ is closed under countable unions.

This implies that if $f$ is a measurable function on $X_1 \times X_2$, then $f(., x_2)$ is measurable on $X_1$ for every $x_2$ and $f(x_1, .)$ is measurable on $X_2$ for every $x_1$ (observe $A_1(x_2) = \{x_1 | f(x_1, x_2) \in B\}$, where $A := \{(x_1, x_2) | f(x_1, x_2) \in B\}$).
9.2. Product measures

Given two measures $\mu_1$ on $\Sigma_1$ and $\mu_2$ on $\Sigma_2$, we now want to construct the **product measure** $\mu_1 \otimes \mu_2$ on $\Sigma_1 \otimes \Sigma_2$ such that

$$\mu_1 \otimes \mu_2(A_1 \times A_2) := \mu_1(A_1)\mu_2(A_2), \quad A_j \in \Sigma_j, \quad j = 1, 2. \quad (9.28)$$

Since the rectangles are closed under intersection, Theorem 8.7 implies that there is at most one measure on $\Sigma_1 \otimes \Sigma_2$ provided $\mu_1$ and $\mu_2$ are $\sigma$-finite.

**Theorem 9.10.** Let $\mu_1$ and $\mu_2$ be two $\sigma$-finite measures on $\Sigma_1$ and $\Sigma_2$, respectively. Let $A \in \Sigma_1 \otimes \Sigma_2$. Then $\mu_2(A_2(x_1))$ and $\mu_1(A_1(x_2))$ are measurable and

$$\int_{X_1} \mu_2(A_2(x_1))d\mu_1(x_1) = \int_{X_2} \mu_1(A_1(x_2))d\mu_2(x_2). \quad (9.29)$$

**Proof.** As usual, we begin with the case where $\mu_1$ and $\mu_2$ are finite. Let $\mathcal{D}$ be the set of all subsets for which our claim holds. Note that $\mathcal{D}$ contains at least all rectangles. Thus it suffices to show that $\mathcal{D}$ is a Dynkin system by Lemma 8.6. To see this, note that measurability and equality of both integrals follow from $A_1(x_2)' = A_1'(x_2)$ (implying $\mu_1(A_1'(x_2)) = \mu_1(X_1)$) for complements and from the monotone convergence theorem for disjoint unions of sets.

If $\mu_1$ and $\mu_2$ are $\sigma$-finite, let $X_{i,j} \uplus X_i$ with $\mu_i(X_{i,j}) < \infty$ for $i = 1, 2$. Now $\mu_2((A \cap X_{1,j} \times X_{2,j})2(x_1)) = \mu_2(A_2(x_1) \cap X_{2,j})\chi_{X_{1,j}}(x_1)$ and similarly with 1 and 2 exchanged. Hence by the finite case

$$\int_{X_1} \mu_2(A_2 \cap X_{2,j})\chi_{X_{1,j}}d\mu_1 = \int_{X_2} \mu_1(A_1 \cap X_{1,j})\chi_{X_{2,j}}d\mu_2$$

and the $\sigma$-finite case follows from the monotone convergence theorem. \qed

Hence for given $A \in \Sigma_1 \otimes \Sigma_2$ we can define

$$\mu_1 \otimes \mu_2(A) := \int_{X_1} \mu_2(A_2(x_1))d\mu_1(x_1) = \int_{X_2} \mu_1(A_1(x_2))d\mu_2(x_2) \quad (9.30)$$

or equivalently, since $\chi_{A_1(x_2)}(x_1) = \chi_{A_2(x_1)}(x_2) = \chi_A(x_1, x_2),$

$$\mu_1 \otimes \mu_2(A) = \int_{X_1} \left( \int_{X_2} \chi_A(x_1, x_2)d\mu_2(x_2) \right) d\mu_1(x_1)$$

$$= \int_{X_2} \left( \int_{X_1} \chi_A(x_1, x_2)d\mu_1(x_1) \right) d\mu_2(x_2). \quad (9.31)$$

Then $\mu_1 \otimes \mu_2$ gives rise to a unique measure on $A \in \Sigma_1 \otimes \Sigma_2$ since $\sigma$-additivity follows from the monotone convergence theorem.

**Example.** Let $X_1 = X_2 = [0, 1]$ with $\mu_1$ Lebesgue measure and $\mu_2$ the counting measure. Let $A = \{(x, x)\mid x \in [0, 1]\}$ such that $\mu_2(A_2(x_1)) = 1$ and
\[ \mu_1(A_1(x_2)) = 0 \] implying
\[ 1 = \int_{X_1} \mu_2(A_2(x_1)) d\mu_1(x_1) \neq \int_{X_2} \mu_1(A_1(x_2)) d\mu_2(x_2) = 0. \]

Hence the theorem can fail if one of the measures is not \( \sigma \)-finite. Note that it is still possible to define a product measure without \( \sigma \)-finiteness (Problem 9.16), but, as the example shows, it will lack some nice properties. \( \diamond \)

Finally we have

**Theorem 9.11** (Fubini). *Let \( f \) be a measurable function on \( X_1 \times X_2 \) and let \( \mu_1, \mu_2 \) be \( \sigma \)-finite measures on \( X_1, X_2 \), respectively.*

(i) If \( f \geq 0 \), then \( \int f(.,x_2) d\mu_2(x_2) \) and \( \int f(x_1,.) d\mu_1(x_1) \) are both measurable and
\[
\int \int_{X_1 \times X_2} f(x_1,x_2) d\mu_1 \otimes d\mu_2(x_1,x_2) = \int_{X_2} \left( \int_{X_1} f(x_1,x_2) d\mu_1(x_1) \right) d\mu_2(x_2)
= \int_{X_1} \left( \int_{X_2} f(x_1,x_2) d\mu_2(x_2) \right) d\mu_1(x_1). \tag{9.32}
\]

(ii) If \( f \) is complex-valued, then
\[
\int_{X_1} |f(x_1,x_2)| d\mu_1(x_1) \in \mathcal{L}^1(X_2,d\mu_2), \tag{9.33}
\]
respectively,
\[
\int_{X_2} |f(x_1,x_2)| d\mu_2(x_2) \in \mathcal{L}^1(X_1,d\mu_1), \tag{9.34}
\]
if and only if \( f \in \mathcal{L}^1(X_1 \times X_2, d\mu_1 \otimes d\mu_2) \). In this case (9.32) holds.

**Proof.** By Theorem 9.10 and linearity the claim holds for simple functions. To see (i), let \( s_n \nearrow f \) be a sequence of nonnegative simple functions. Then it follows by applying the monotone convergence theorem (twice for the double integrals).

For (ii) we can assume that \( f \) is real-valued by considering its real and imaginary parts separately. Moreover, splitting \( f = f^+ - f^- \) into its positive and negative parts, the claim reduces to (i). \( \square \)

In particular, if \( f(x_1,x_2) \) is either nonnegative or integrable, then the order of integration can be interchanged. The case of nonnegative functions is also called **Tonelli’s theorem.** In the general case the integrability condition is crucial, as the following example shows.
Example. Let $X := [0, 1] \times [0, 1]$ with Lebesgue measure and consider

$$f(x, y) = \frac{x - y}{(x + y)^3}.$$ 

Then

$$\int_0^1 \int_0^1 f(x, y) dx \, dy = -\int_0^1 \frac{1}{(1 + y)^2} dy = -\frac{1}{2},$$

but (by symmetry)

$$\int_0^1 \int_0^1 f(x, y) dy \, dx = \int_0^1 \frac{1}{(1 + x)^2} dx = \frac{1}{2}.$$ 

Consequently $f$ cannot be integrable over $X$ (verify this directly).  

Lemma 9.12. If $\mu_1$ and $\mu_2$ are outer regular measures, then so is $\mu_1 \otimes \mu_2$.

Proof. Outer regularity holds for every rectangle and hence also for the algebra of finite disjoint unions of rectangles (Problem 9.15). Thus the claim follows from Problem 8.15. \hfill \Box

In connection with Theorem 8.7 the following observation is of interest:

Lemma 9.13. If $S_1$ generates $\Sigma_1$ and $S_2$ generates $\Sigma_2$, then $S_1 \times S_2 := \{ A_1 \times A_2 | A_j \in S_j, j = 1, 2 \}$ generates $\Sigma_1 \otimes \Sigma_2$.

Proof. Denote the $\sigma$-algebra generated by $S_1 \times S_2$ by $\Sigma$. Consider the set $\{ A_1 \in \Sigma_1 | A_1 \times X_2 \in \Sigma \}$ which is clearly a $\sigma$-algebra containing $S_1$ and thus equal to $\Sigma_1$. In particular, $\Sigma_1 \times X_2 \subset \Sigma$ and similarly $X_1 \times \Sigma_2 \subset \Sigma$. Hence also $(\Sigma_1 \times X_2) \cap (X_1 \times \Sigma_2) = \Sigma_1 \times \Sigma_2 \subset \Sigma$. \hfill \Box

Finally, note that we can iterate this procedure.

Lemma 9.14. Suppose $(X_j, \Sigma_j, \mu_j)$, $j = 1, 2, 3$, are $\sigma$-finite measure spaces. Then $(\Sigma_1 \otimes \Sigma_2) \otimes \Sigma_3 = \Sigma_1 \otimes (\Sigma_2 \otimes \Sigma_3)$ and

$$\mu_1 \otimes (\mu_2 \otimes \mu_3) = (\mu_1 \otimes \mu_2) \otimes \mu_3.$$ \hfill (9.35)

Proof. First of all note that $(\Sigma_1 \otimes \Sigma_2) \otimes \Sigma_3 = \Sigma_1 \otimes (\Sigma_2 \otimes \Sigma_3)$ is the sigma algebra generated by the rectangles $A_1 \times A_2 \times A_3$ in $X_1 \times X_2 \times X_3$. Moreover, since

$$(\mu_1 \otimes \mu_2) \otimes \mu_3)(A_1 \times A_2 \times A_3) = \mu_1(A_1) \mu_2(A_2) \mu_3(A_3)$$

$$= (\mu_1 \otimes (\mu_2 \otimes \mu_3))(A_1 \times A_2 \times A_3),$$

the two measures coincide on rectangles and hence everywhere by Theorem 8.7. \hfill \Box
Hence we can take the product of finitely many measures. The case of infinitely many measures requires a bit more effort and will be discussed in Section 11.5.

Example. If $\lambda$ is Lebesgue measure on $\mathbb{R}$, then $\lambda^n = \lambda \otimes \cdots \otimes \lambda$ is Lebesgue measure on $\mathbb{R}^n$. In fact, it satisfies $\lambda^n((a,b]) = \prod_{j=1}^n (b_j - a_j)$ and hence must be equal to Lebesgue measure which is the unique Borel measure with this property. \hfill \diamond

Example. If $X_1$, $X_2$ are topological spaces, then $\mathcal{B}(X_1 \times X_2) = \mathcal{B}(X_1) \otimes \mathcal{B}(X_2)$ since open rectangles are a base for the product topology. Moreover, if $\mu_1$, $\mu_2$ are Borel measures and both $X_1$ and $X_1$ are locally compact, then $\mu_1 \otimes \mu_2$ is also a Borel measure. Indeed, let $K \subseteq X_1 \times X_2$ be compact. Then for every point in $K$ there is a relatively compact open rectangle containing this point. By compactness finitely many of them suffice to cover $K$, that is $K \subseteq \bigcup_{j=1}^n K_{1,j} \times K_{2,j}$ implying $\mu_1 \otimes \mu_2(K) \leq \sum_{j=1}^n \mu(K_{1,j}) \mu(K_{2,j}) < \infty$. \hfill \diamond

Problem 9.15. Show that the set of all finite union of measurable rectangles $A_1 \times A_2$ forms an algebra. Moreover, every set in this algebra can be written as a finite union of disjoint rectangles.

Problem 9.16. Given two measure spaces $(X_1, \Sigma_1, \mu_1)$ and $(X_2, \Sigma_2, \mu_2)$ let $\mathcal{R} = \{A_1 \times A_2 | A_i \in \Sigma_j, j = 1, 2\}$ be the collection of measurable rectangles. Define $\rho : \mathcal{R} \to [0, \infty]$, $\rho(A_1 \times A_2) = \mu_1(A_1) \mu_2(A_2)$. Then

$$(\mu_1 \otimes \mu_2^*)(A) := \inf \left\{ \sum_{j=1}^\infty \rho(A_j) \left| A \subseteq \bigcup_{j=1}^\infty A_j, A_j \in \mathcal{R} \right. \right\}$$

is an outer measure on $X_1 \times X_2$. Show that this constructions coincides with (9.30) for $A \in \Sigma_1 \otimes \Sigma_2$ in case $\mu_1$ and $\mu_2$ are $\sigma$-finite.

Problem 9.17. Let $P : \mathbb{R}^n \to \mathbb{C}$ be a nonzero polynomial. Show that $N := \{x \in \mathbb{R}^n | P(x) = 0\}$ is a Borel set of Lebesgue zero. (Hint: Induction using Fubini.)

Problem 9.18. Let $U \subseteq \mathbb{C}$ be a domain, $Y$ be some measure space, and $f : U \times Y \to \mathbb{C}$. Suppose $y \mapsto f(z, y)$ is measurable for every $z$ and $z \mapsto f(z, y)$ is holomorphic for every $y$. Show that

$$F(z) := \int_Y f(z, y) \, d\mu(y)$$

is holomorphic if for every compact subset $V \subseteq U$ there is an integrable function $g(y)$ such that $|f(z, y)| \leq g(y)$, $z \in V$. (Hint: Use Fubini and Morera.)

Problem 9.19. Suppose $\phi : X \to [0, \infty)$ is integrable over every compact interval and set $\Phi(r) = \int_0^r \phi(s) \, ds$. Let $f : X \to \mathbb{C}$ be measurable and
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introduce its distribution function

\[ E_f(r) := \mu(\{x \in X| |f(x)| > r\}) \]

Show that

\[ \int_X \Phi(|f|) d\mu = \int_0^\infty \phi(r) E_f(r) dr. \]

Moreover, show that if \( f \) is integrable, then the set of all \( \alpha \in \mathbb{C} \) for which \( \mu(\{x \in X| f(x) = \alpha\}) > 0 \) is countable.

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Finally we want to transform measures. Let \( f : X \to Y \) be a measurable function. Given a measure \( \mu \) on \( X \) we can introduce a measure \( f_*\mu \) on \( Y \) via

\[ (f_*\mu)(A) := \mu(f^{-1}(A)). \]  \hspace{1cm} (9.36)

It is straightforward to check that \( f_*\mu \) is indeed a measure. Moreover, note that \( f_*\mu \) is supported on the range of \( f \).

**Theorem 9.15.** Let \( f : X \to Y \) be measurable and let \( g : Y \to \mathbb{C} \) be a Borel function. Then the Borel function \( g \circ f : X \to \mathbb{C} \) is a.e. nonnegative or integrable if and only if \( g \) is and in both cases

\[ \int_Y g d(f_*\mu) = \int_X g \circ f d\mu. \]  \hspace{1cm} (9.37)

**Proof.** In fact, it suffices to check this formula for simple functions \( g \), which follows since \( \chi_A \circ f = \chi_{f^{-1}(A)} \). \qed

**Example.** Suppose \( f : X \to Y \) and \( g : Y \to Z \). Then

\[ (g \circ f)_* = g_*(f_*\mu). \]

since \( (g \circ f)^{-1} = f^{-1} \circ g^{-1} \). \odo

**Example.** Let \( f(x) = Mx + a \) be an affine transformation, where \( M : \mathbb{R}^n \to \mathbb{R}^n \) is some invertible matrix. Then Lebesgue measure transforms according to

\[ f_*\lambda^n = \frac{1}{|\det(M)|} \lambda^n. \]

To see this, note that \( f_*\lambda^n \) is translation invariant and hence must be a multiple of \( \lambda^n \). Moreover, for an orthogonal matrix this multiple is one (since an orthogonal matrix leaves the unit ball invariant) and for a diagonal matrix it must be the absolute value of the product of the diagonal elements (consider a rectangle). Finally, since every matrix can be written as \( M = O_1DO_2 \), where \( O_j \) are orthogonal and \( D \) is diagonal (Problem 9.21), the claim follows.
As a consequence we obtain
\[
\int_A g(Mx + a) d^n x = \frac{1}{|\det(M)|} \int_{MA + a} g(y) d^n y,
\]
which applies, for example, to shifts \( f(x) = x + a \) or scaling transforms \( f(x) = \alpha x \).

This result can be generalized to diffeomorphisms (one-to-one \( C^1 \) maps with inverse again \( C^1 \)):

**Lemma 9.16.** Let \( U, V \subseteq \mathbb{R}^n \) and suppose \( f \in C^1(U, V) \) is a diffeomorphism. Then
\[
(f^{-1})_* d^n x = |J_f(x)| d^n x,
\]
where \( J_f = \det(\frac{df}{dx}) \) is the Jacobi determinant of \( f \). In particular,
\[
\int_U g(f(x)) |J_f(x)| d^n x = \int_V g(y) d^n y.
\]

**Proof.** It suffices to show
\[
\int_{f(R)} d^n y = \int_R |J_f(x)| d^n x
\]
for every bounded open rectangle \( R \subseteq U \). By Theorem 8.7 it will then follow for characteristic functions and thus for arbitrary functions by the very definition of the integral.

To this end we consider the integral
\[
I_\varepsilon := \int_{f(R)} \int_R |J_f(f^{-1}(y))| \varphi_\varepsilon(f(z) - y) d^n z d^n y.
\]
Here \( \varphi := V_n^{-1} \chi_{B_1(0)} \) and \( \varphi_\varepsilon(y) := \varepsilon^{-n} \varphi(\varepsilon^{-1} y) \), where \( V_n \) is the volume of the unit ball (cf. below), such that \( \int \varphi_\varepsilon(x) d^n x = 1 \).

We will evaluate this integral in two ways. To begin with we consider the inner integral
\[
h_\varepsilon(y) := \int_R \varphi_\varepsilon(f(z) - y) d^n z.
\]
For \( \varepsilon < \varepsilon_0 \) the integrand is nonzero only for \( z \in K = f^{-1}(B_{\varepsilon_0}(y)) \), where \( K \) is some compact set containing \( x = f^{-1}(y) \). Using the affine change of coordinates \( z = x + \varepsilon w \) we obtain
\[
h_\varepsilon(y) = \int_{W_\varepsilon(x)} \varphi \left( \frac{f(x + \varepsilon w) - f(x)}{\varepsilon} \right) d^n w, \quad W_\varepsilon(x) = \frac{1}{\varepsilon}(K - x).
\]
By
\[
\left| \frac{f(x + \varepsilon w) - f(x)}{\varepsilon} \right| \geq \frac{1}{C} |w|, \quad C := \sup_K \|df^{-1}\|,
\]
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the integrand is nonzero only for \( w \in BC(0) \). Hence, as \( \varepsilon \to 0 \) the domain \( W_\varepsilon(x) \) will eventually cover all of \( BC(0) \) and dominated convergence implies

\[
\lim_{\varepsilon \downarrow 0} h_\varepsilon(y) = \int_{BC(0)} \varphi(df(x)w)dw = |J_f(x)|^{-1}.
\]

Consequently, \( \lim_{\varepsilon \downarrow 0} I_\varepsilon = |f(R)| \) again by dominated convergence. Now we use Fubini to interchange the order of integration

\[
I_\varepsilon = \int_R \int_{f(R)} |J_f(f^{-1}(y))|\varphi_\varepsilon(f(z) - y)d^nyd^nz.
\]

Since \( f(z) \) is an interior point of \( f(R) \) continuity of \( |J_f(f^{-1}(y))| \) implies

\[
\lim_{\varepsilon \downarrow 0} \int_{f(R)} |J_f(f^{-1}(y))|\varphi_\varepsilon(f(z) - y)d^ny = |J_f(f^{-1}(f(z)))| = |J_f(z)|
\]

and hence dominated convergence shows \( \lim_{\varepsilon \downarrow 0} I_\varepsilon = \int_R |J_f(z)|d^n z. \)

**Example.** For example, we can consider polar coordinates \( T_2 : [0, \infty) \times [0, 2\pi) \to \mathbb{R}^2 \) defined by

\[
T_2(\rho, \varphi) := (\rho \cos(\varphi), \rho \sin(\varphi)).
\]

Then

\[
\det \frac{\partial T_2}{\partial (\rho, \varphi)} = \det \begin{vmatrix} \cos(\varphi) & -\rho \sin(\varphi) \\ \sin(\varphi) & \rho \cos(\varphi) \end{vmatrix} = \rho
\]

and one has

\[
\int_U f(\rho \cos(\varphi), \rho \sin(\varphi))\rho d(\rho, \varphi) = \int_{T_2(U)} f(x)d^2x.
\]

Note that \( T_2 \) is only bijective when restricted to \((0, \infty) \times [0, 2\pi)\). However, since the set \( \{0\} \times [0, 2\pi) \) is of measure zero, it does not contribute to the integral on the left. Similarly, its image \( T_2(\{0\} \times [0, 2\pi)) = \{0\} \) does not contribute to the integral on the right.

**Example.** We can use the previous example to obtain the transformation formula for spherical coordinates in \( \mathbb{R}^n \) by induction. We illustrate the process for \( n = 3 \). To this end let \( x = (x_1, x_2, x_3) \) and start with spherical coordinates in \( \mathbb{R}^2 \) (which are just polar coordinates) for the first two components:

\[
x = (\rho \cos(\varphi), \rho \sin(\varphi), x_3), \quad \rho \in [0, \infty), \ \varphi \in [0, 2\pi).
\]

Next use polar coordinates for \( (\rho, x_3) \):

\[
(\rho, x_3) = (r \sin(\theta), r \cos(\theta)), \quad r \in [0, \infty), \ \theta \in [0, \pi].
\]
Note that the range for $\theta$ follows since $\rho \geq 0$. Moreover, observe that $r^2 = \rho^2 + x_3^2 = x_1^2 + x_2^2 + x_3^2 = |x|^2$ as already anticipated by our notation.

In summary, $$x = T_3(r, \varphi, \theta) := (r \sin(\theta) \cos(\varphi), r \sin(\theta) \sin(\varphi), r \cos(\theta)).$$

Furthermore, since $T_3$ is the composition with $T_2$ acting on the first two coordinates with the last unchanged and polar coordinates $P$ acting on the first and last coordinate, the chain rule implies

\[
\det \frac{\partial T_3}{\partial (r, \varphi, \theta)} = \det \left. \frac{\partial T_2}{\partial (\rho, \varphi, x_3)} \right|_{\rho = r \sin(\theta)} \det \frac{\partial P}{\partial (r, \varphi, \theta)} = r^2 \sin(\theta).
\]

Hence one has

\[
\int_U f(T_3(r, \varphi, \theta)) r^2 \sin(\theta) d(r, \varphi, \theta) = \int_{T_3(U)} f(x) d^3x.
\]

Again $T_3$ is only bijective on $(0, \infty) \times [0, 2\pi) \times (0, \pi)$.

It is left as an exercise to check that the extension $T_n : [0, \infty) \times [0, 2\pi) \times [0, \pi]^{n-2} \to \mathbb{R}^n$ is given by

\[
x = T_n(r, \varphi, \theta_1, \ldots, \theta_{n-2})
\]

with

\[
\begin{align*}
x_1 &= r \cos(\varphi) \sin(\theta_1) \sin(\theta_2) \cdots \sin(\theta_{n-2}), \\
x_2 &= r \sin(\varphi) \sin(\theta_1) \sin(\theta_2) \cdots \sin(\theta_{n-2}), \\
x_3 &= r \cos(\theta_1) \sin(\theta_2) \cdots \sin(\theta_{n-2}), \\
x_4 &= r \cos(\theta_2) \cdots \sin(\theta_{n-2}), \\
&\vdots \\
x_{n-1} &= r \cos(\theta_{n-3}) \sin(\theta_{n-2}), \\
x_n &= r \cos(\theta_{n-2}).
\end{align*}
\]

The Jacobi determinant is given by

\[
\det \frac{\partial T_n}{\partial (r, \varphi, \theta_1, \ldots, \theta_{n-2})} = r^{n-1} \sin(\theta_1) \sin(\theta_2)^2 \cdots \sin(\theta_{n-2})^{n-2}.
\]

Another useful consequence of Theorem 9.15 is the following rule for integrating radial functions.

**Lemma 9.17.** There is a measure $\sigma^{n-1}$ on the unit sphere $S^{n-1} := \partial B_1(0) = \{x \in \mathbb{R}^n | |x| = 1\}$, which is rotation invariant and satisfies

\[
\int_{\mathbb{R}^n} g(x) d^n x = \int_0^\infty \int_{S^{n-1}} g(r\omega) r^{n-1} d\sigma^{n-1}(\omega) dr,
\]

for every integrable (or positive) function $g$.

Moreover, the surface area of $S^{n-1}$ is given by

\[
S_n := \sigma^{n-1}(S^{n-1}) = nV_n,
\]
where \( V_n := \lambda^n(B_1(0)) \) is the volume of the unit ball in \( \mathbb{R}^n \), and if \( g(x) = \tilde{g}(|x|) \) is radial we have
\[
\int_{\mathbb{R}^n} g(x) d^n x = S_n \int_0^\infty \tilde{g}(r) r^{n-1} dr.
\] (9.42)

**Proof.** Consider the measurable transformation \( f : \mathbb{R}^n \to [0, \infty) \times S^{n-1}, x \mapsto (|x|, \frac{x}{|x|}) \) (with \( \frac{0}{0} = 1 \)). Let \( d\mu(r) := r^{n-1} dr \) and
\[
\sigma^{n-1}(A) := n \lambda^n(f^{-1}([0,1) \times A))
\] (9.43)
for every \( A \in \mathcal{B}(S^{n-1}) = \mathcal{B}^n \cap S^{n-1} \). Note that \( \sigma^{n-1} \) inherits the rotation invariance from \( \lambda^n \). By Theorem 9.15 it suffices to show \( f_* \lambda^n = \mu \otimes \sigma^{n-1} \).

This follows from
\[
(f_* \lambda^n)([0,r) \times A) = \lambda^n(f^{-1}([0,r) \times A)) = r^n \lambda^n(f^{-1}([0,1) \times A)) = \mu([0,r)) \sigma^{n-1}(A).
\]
since these sets determine the measure uniquely. \( \square \)

**Example.** Let us compute the volume of a ball in \( \mathbb{R}^n \):
\[
V_n(r) := \int_{\mathbb{R}^n} \chi_{B_r(0)} d^n x.
\]
By the simple scaling transform \( f(x) = rx \) we obtain \( V_n(r) = V_n(1)r^n \) and hence it suffices to compute \( V_n := V_n(1) \).

To this end we use (Problem 9.22)
\[
\pi^n = \int_{\mathbb{R}^n} e^{-|x|^2} d^n x = nV_n \int_0^\infty e^{-r^2} r^{n-1} dr = \frac{nV_n}{2} \int_0^\infty e^{-s} s^{n/2-1} ds
= \frac{nV_n}{2} \Gamma\left(\frac{n}{2}\right) = \frac{V_n}{2} \Gamma\left(\frac{n}{2} + 1\right)
\]
where \( \Gamma \) is the gamma function (Problem 9.23). Hence
\[
V_n = \frac{\pi^{n/2}}{\Gamma\left(\frac{n}{2} + 1\right)}.
\] (9.44)

By \( \Gamma\left(\frac{1}{2}\right) = \sqrt{\pi} \) (see Problem 9.24) this coincides with the well-known values for \( n = 1, 2, 3 \). \( \diamond \)

**Example.** The above lemma can be used to determine when a radial function is integrable. For example, we obtain
\[
|x|^\alpha \in L^1(B_1(0)) \iff \alpha > -n, \quad |x|^\alpha \in L^1(\mathbb{R}^n \setminus B_1(0)) \iff \alpha < -n.
\] \( \diamond \)
Problem 9.20. Let $\lambda$ be Lebesgue measure on $\mathbb{R}$, and let $f$ be a strictly increasing function with $\lim_{x \to \pm \infty} f(x) = \pm \infty$. Show that
\[
d(f_\ast \lambda) = d(f^{-1}),
\]
where $f^{-1}$ is the inverse of $f$ extended to all of $\mathbb{R}$ by setting $f^{-1}(y) = x$ for $y \in [f(x-), f(x+)]$ (note that $f^{-1}$ is continuous).

Moreover, if $f \in C^1(\mathbb{R})$ with $f' > 0$, then
\[
d(f_\ast \lambda) = \frac{1}{f'(f^{-1})} d\lambda.
\]

Problem 9.21. Show that every invertible matrix $M$ can be written as $M = O_1 DO_2$, where $D$ is diagonal and $O_j$ are orthogonal. (Hint: The matrix $M^* M$ is nonnegative and hence there is an orthogonal matrix $U$ which diagonalizes $M^* M = UD^2 U^*$. Then one can choose $O_1 = MU D^{-1}$ and $O_2 = U^*$.)

Problem 9.22. Show
\[
I_n := \int_{\mathbb{R}^n} e^{-|x|^2} d^n x = \pi^{n/2}.
\]
(Hint: Use Fubini to show $I_n = I_1^n$ and compute $I_2$ using polar coordinates.)

Problem 9.23. The gamma function is defined via
\[
\Gamma(z) := \int_0^\infty x^{z-1} e^{-x} dx, \quad \text{Re}(z) > 0.
\]
Verify that the integral converges and defines an analytic function in the indicated half-plane (cf. Problem 9.18). Use integration by parts to show
\[
\Gamma(z+1) = z\Gamma(z), \quad \Gamma(1) = 1.
\]
Conclude $\Gamma(n) = (n-1)!$ for $n \in \mathbb{N}$. Show that the relation $\Gamma(z) = \Gamma(z+1)/z$ can be used to define $\Gamma(z)$ for all $z \in \mathbb{C}\setminus\{0, -1, -2, \ldots\}$. Show that near $z = -n$, $n \in \mathbb{N}_0$, the Gamma functions behaves like $\Gamma(z) = \frac{(-1)^n}{n!(z+n)} + O(1)$.

Problem 9.24. Show that $\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$. Moreover, show
\[
\Gamma\left(n + \frac{1}{2}\right) = \frac{(2n)!}{4^n n! \sqrt{\pi}}
\]
(Hint: Use the change of coordinates $x = t^2$ and then use Problem 9.22.)

Problem 9.25. Establish
\[
\left(\frac{d}{dz}\right)^j \Gamma(z) = \int_0^\infty \log(x)^j x^{z-1} e^{-x} dx, \quad \text{Re}(z) > 0,
\]
and show that $\Gamma$ is log-convex, that is, $\log(\Gamma(x))$ is convex. (Hint: Regard the first derivative as a scalar product and apply Cauchy–Schwarz.)

**Problem 9.26.** Show that the Beta function satisfies

$$B(u, v) := \int_0^1 t^{u-1}(1 - t)^{v-1} dt = \frac{\Gamma(u)\Gamma(v)}{\Gamma(u + v)}, \quad \text{Re}(u) > 0, \text{Re}(v) > 0.$$  

Use this to establish Euler’s reflection formula

$$\Gamma(z)\Gamma(1 - z) = \frac{\pi}{\sin(\pi z)}.$$  

Conclude that the Gamma function has no zeros on $\mathbb{C}$.

(Hint: Start with $\Gamma(u)\Gamma(v)$ and make a change of variables $x = ts$, $y = t(1 - s)$. For the reflection formula evaluate $B(z, 1 - z)$ using Problem 9.27.)

**Problem 9.27.** Show

$$\int_{-\infty}^{\infty} \frac{e^{zx}}{1 + e^x} dx = \frac{\pi}{\sin(z\pi)}, \quad 0 < \text{Re}(z) < 1.$$  

(Hint: To compute the integral, use a contour consisting of the straight lines connecting the points $-R, R, R+2\pi i, -R+2\pi i$. Evaluate the contour integral using the residue theorem and let $R \to \infty$. Show that the contributions from the vertical lines vanish in the limit and relate the integrals along the horizontal lines.)

**Problem 9.28.** Let $U \subseteq \mathbb{R}^m$ be open and let $f : U \to \mathbb{R}^n$ be locally Lipschitz (i.e., for every compact set $K \subset U$ there is some constant $L$ such that $|f(x) - f(y)| \leq L|x - y|$ for all $x, y \in K$). Show that if $A \subset U$ has Lebesgue measure zero, then $f(A)$ is contained in a set of Lebesgue measure zero. (Hint: By Lindelöf it is no restriction to assume that $A$ is contained in a compact ball contained in $U$. Now approximate $A$ by a union of rectangles.)

### 9.4. Appendix: Transformation of Lebesgue–Stieltjes integrals

In this section we will look at Borel measures on $\mathbb{R}$. In particular, we want to derive a generalized substitution rule.

As a preparation we will need a generalization of the usual inverse which works for arbitrary nondecreasing functions. Such a generalized inverse arises, for example, as quantile functions in probability theory.

So we look at nondecreasing functions $f : \mathbb{R} \to \mathbb{R}$. By monotonicity the limits from left and right exist at every point and we will denote them by

$$f(x\pm) := \lim_{\varepsilon \downarrow 0} f(x \pm \varepsilon). \quad (9.47)$$
Clearly we have \( f(x- \leq f(x+) \) and a strict inequality can occur only at a countable number of points. By monotonicity the value of \( f \) has to lie between these two values \( f(x- \leq f(x) \leq f(x+) \). It will also be convenient to extend \( f \) to a function on the extended reals \( \mathbb{R} \cup \{-\infty, +\infty\} \). Again by monotonicity the limits \( f(\pm \infty) = \lim_{x \to \pm \infty} f(x) \) exist and we will set \( f(\pm \infty) = f(\pm \infty) \).

If we want to define an inverse, problems will occur at points where \( f \) jumps and on intervals where \( f \) is constant. Informally speaking, if \( f \) jumps, then the corresponding jump will be missing in the domain of the inverse and if \( f \) is constant, the inverse will be multivalued. For the first case there is a natural fix by choosing the inverse to be constant along the missing interval. In particular, observe that this natural choice is independent of the actual value of \( f \) at the jump and hence the inverse loses this information. The second case will result in a jump for the inverse function and here there is no natural choice for the value at the jump (except that it must be between the left and right limits such that the inverse is again a nondecreasing function).

To give a precise definition it will be convenient to look at relations instead of functions. Recall that a (binary) relation \( R \) on \( \mathbb{R} \) is a subset of \( \mathbb{R} \times \mathbb{R} \).

To every nondecreasing function \( f \) associate the relation
\[
\Gamma(f) := \{(x,y) | y \in [f(x-),f(x+)]\}. \tag{9.48}
\]
Note that \( \Gamma(f) \) does not depend on the values of \( f \) at a discontinuity and \( f \) can be partially recovered from \( \Gamma(f) \) using \( f(x-) = \inf \Gamma(f)(x) \) and \( f(x+) = \sup \Gamma(f)(x) \), where \( \Gamma(f)(x) := \{y | (x,y) \in \Gamma(f)\} = [f(x-),f(x+)] \). Moreover, the relation \( \Gamma(f) \) is nondecreasing in the sense that \( x_1 < x_2 \) implies \( y_1 \leq y_2 \) for \( (x_1,y_1),(x_2,y_2) \in \Gamma(f) \) (just note \( y_1 \leq f(x_1+) \leq f(x_2+) \leq y_2 \)). It is uniquely defined as the largest relation containing the graph of \( f \) with this property.

The graph of any reasonable inverse should be a subset of the inverse relation
\[
\Gamma(f)^{-1} := \{(y,x) | (x,y) \in \Gamma(f)\} \tag{9.49}
\]
and we will call any function \( f^{-1} \) whose graph is a subset of \( \Gamma(f)^{-1} \) a **generalized inverse** of \( f \). Note that any generalized inverse is again nondecreasing since a pair of points \((y_1,x_1),(y_2,x_2) \in \Gamma(f)^{-1} \) with \( y_1 < y_2 \) and \( x_1 > x_2 \) would contradict the fact that \( \Gamma(f) \) is nondecreasing. Moreover, since \( \Gamma(f)^{-1} \) and \( \Gamma(f^{-1}) \) are two nondecreasing relations containing the graph of \( f^{-1} \), we conclude
\[
\Gamma(f^{-1}) = \Gamma(f)^{-1} \tag{9.50}
\]
since both are maximal. In particular, it follows that if \( f^{-1} \) is a generalized inverse of \( f \) then \( f \) is a generalized inverse of \( f^{-1} \).
There are two particular choices, namely the left continuous version $f_-(y) := \inf \Gamma(f)^{-1}(y)$ and the right continuous version $f_+(y) := \sup \Gamma(f)^{-1}(y)$. It is straightforward to verify that they can be equivalently defined via

\begin{align}
f_-(y) &:= \inf f^{-1}([y, \infty)) = \sup f^{-1}((-\infty, y)), \\
f_+(y) &:= \inf f^{-1}((y, \infty)) = \sup f^{-1}((-\infty, y]).
\end{align}

(9.51)

For example, $\inf f^{-1}([y, \infty)) = \inf \{x | (x, \tilde{y}) \in \Gamma(f), \tilde{y} \geq y\} = \inf \Gamma(f)^{-1}(y)$. The first one is typically used in probability theory, where it corresponds to the quantile function of a distribution.

If $f$ is strictly increasing the generalized inverse $f^{-1}$ extends the usual inverse by setting it constant on the gaps missing in the range of $f$. In particular we have $f^{-1}(x)) = x$ and $f(f^{-1}(y)) = y$ for $y$ in the range of $f$. The purpose of the next lemma is to investigate to what extend this remains valid for a generalized inverse.

Note that for every $y$ there is some $x$ with $y \in [f(x-), f(x+)]$. Moreover, if we can find two values, say $x_1$ and $x_2$, with this property, then $f(x) = y$ is constant for $x \in (x_1, x_2)$. Hence, the set of all such $x$ is an interval which is closed since at the left, right boundary point the left, right limit equals $y$, respectively.

We collect a few simple facts for later use.

**Lemma 9.18.** Let $f$ be nondecreasing.

(i) $f_-(y) \leq x$ if and only if $y \leq f(x+)$.

(i') $f_+(y) \geq x$ if and only if $y \geq f(x-)$. 

(ii) $f_-(f(x)) \leq x \leq f_+(f(x))$ with equality on the left, right iff $f$ is not constant to the right, left of $x$, respectively.

(iii) $f(f_-(y)-) \leq y \leq f(f_+(y)+)$ with equality on the left, right iff $f^{-1}$ is not constant to right, left of $y$, respectively.

**Proof.** Item (i) follows since both claims are equivalent to $y \leq f(\tilde{x})$ for all $\tilde{x} > x$. Similarly for (i'). Item (ii) follows from $f_-(f(x)) = \inf f^{-1}([f(x), \infty)) = \inf \{\tilde{x} | f(\tilde{x}) \geq f(x)\} \leq x$ with equality iff $f(\tilde{x}) < f(x)$ for $\tilde{x} < x$. Similarly for the other inequality. Item (iii) follows by reversing the roles of $f$ and $f^{-1}$ in (ii). \hfill $\square$

In particular, $f(f_+(y)) = y$ if $f$ is continuous. We will also need the set

$L(f) := \{y | f^{-1}((y, \infty)) = (f_+^{-1}(y), \infty)\}.$

(9.52)

Note that $y \not\in L(f)$ if and only if there is some $x$ such that $y \in [f(x-), f(x))$.

**Lemma 9.19.** Let $m : \mathbb{R} \to \mathbb{R}$ be a nondecreasing function on $\mathbb{R}$ and $\mu$ its associated measure via (8.21). Let $f(x)$ be a nondecreasing function on $\mathbb{R}$
such that \( \mu((0, \infty)) < \infty \) if \( f \) is bounded above and \( \mu((\infty, 0)) < \infty \) if \( f \) is bounded below.

Then \( f_*\mu \) is a Borel measure whose distribution function coincides up to a constant with \( m_+ \circ f_+^{-1} \) at every point \( y \) which is in \( L(f) \) or satisfies \( \mu(\{f_+^{-1}(y)\}) = 0 \). If \( y \in [f(x-), f(x)] \) and \( \mu(\{f_+^{-1}(y)\}) > 0 \), then \( m_+ \circ f_+^{-1} \) jumps at \( f(x-) \) and \( (f_*\mu)(y) \) jumps at \( f(x) \).

**Proof.** First of all note that the assumptions in case \( f \) is bounded from above or below ensure that \( (f_*\mu)(K) < \infty \) for any compact interval. Moreover, we can assume \( m = m_+ \) without loss of generality. Now note that we have \( f^{-1}((y, \infty)) = (f^{-1}(y), \infty) \) for \( y \in L(f) \) and \( f^{-1}((y, \infty)) = [f^{-1}(y), \infty) \) else. Hence
\[
(f_*\mu)((y_0, y_1]) = \mu(f^{-1}((y_0, y_1])) = \mu((f^{-1}(y_0), f^{-1}(y_1)]
\]
\[
= m(f_+^{-1}(y_1)) - m(f_+^{-1}(y_0)) = (m \circ f_+^{-1})(y_1) - (m \circ f_+^{-1})(y_0)
\]
if \( y_j \) is either in \( L(f) \) or satisfies \( \mu(\{f_+^{-1}(y_j)\}) = 0 \). For the last claim observe that \( f^{-1}((y, \infty)) \) will jump from \((f_+^{-1}(y), \infty)\) to \([f_+^{-1}(y), \infty)\) at \( y = f(x) \). \( \square \)

**Example.** For example, consider \( f(x) = \chi_{[0, \infty)}(x) \) and \( \mu = \Theta \), the Dirac measure centered at 0 (note that \( \Theta(x) = f(x) \)). Then
\[
f_+^{-1}(y) = \begin{cases} +\infty, & 1 \leq y, \\ 0, & 0 \leq y < 1, \\ -\infty, & y < 0, \end{cases}
\]
and \( L(f) = (-\infty, 0) \cup [1, \infty) \). Moreover, \( \mu(f_+^{-1}(y)) = \chi_{[0, \infty)}(y) \) and \( (f_*\mu)(y) = \chi_{[1, \infty)}(y) \). If we choose \( g(x) = \chi_{(0, \infty)}(x) \), then \( g_+^{-1}(y) = f_+^{-1}(y) \) and \( L(g) = \mathbb{R} \). Hence \( \mu(g_+^{-1}(y)) = \chi_{[0, \infty)}(y) = (g_*\mu)(y) \).

For later use it is worth while to single out the following consequence:

**Corollary 9.20.** Let \( m, f \) be as in the previous lemma and denote by \( \mu, \nu_\pm \) the measures associated with \( m, m_+ \circ f^{-1} \), respectively. Then, \( (f_+)_*\mu = \nu_\pm \) and hence
\[
\int g \, d(m_+ \circ f^{-1}) = \int (g \circ f_+) \, dm.
\] (9.53)

In the special case where \( \mu \) is Lebesgue measure this reduces to a way of expressing the Lebesgue–Stieltjes integral as a Lebesgue integral via
\[
\int g \, dh = \int g(h^{-1}(y)) \, dy.
\] (9.54)

If we choose \( f \) to be the distribution function of \( \mu \) we get the following generalization of the integration by substitution rule. To formulate it
we introduce
\[ i_m(y) := m(m^{-1}(y)) \quad (9.55) \]
Note that \( i_m(y) = y \) if \( m \) is continuous. By \( \text{hull}(\text{Ran}(m)) \) we denote the convex hull of the range of \( m \).

**Corollary 9.21.** Suppose \( m, n \) are two nondecreasing functions on \( \mathbb{R} \) with \( n \) right continuous. Then we have
\[ \int_{\mathbb{R}} (g \circ m) d(n \circ m) = \int_{\text{hull}(\text{Ran}(m))} (g \circ i_m) d\text{m} \quad (9.56) \]
for any Borel function \( g \) which is either nonnegative or for which one of the two integrals is finite. Similarly, if \( n \) is left continuous and \( i_m \) is replaced by \( m(m^{-1}(y)) \).

Hence the usual \( \int_{\mathbb{R}} (g \circ m) d(n \circ m) = \int_{\text{Ran}(m)} g d\text{m} \) only holds if \( m \) is continuous. In fact, the right-hand side loses all point masses of \( \mu \). The above formula fixes this problem by rendering \( g \) constant along a gap in the range of \( m \) and includes the gap in the range of integration such that it makes up for the lost point mass. It should be compared with the previous example!

If one does not want to bother with \( i_m \) one can at least get inequalities for monotone \( g \).

**Corollary 9.22.** Suppose \( m, n \) are nondecreasing functions on \( \mathbb{R} \) and \( g \) is monotone. Then we have
\[ \int_{\mathbb{R}} (g \circ m) d(n \circ m) \leq \int_{\text{hull}(\text{Ran}(m))} g d\text{m} \quad (9.57) \]
if \( m, n \) are right continuous and \( g \) nonincreasing or \( m, n \) left continuous and \( g \) nondecreasing. If \( m, n \) are right continuous and \( g \) nondecreasing or \( m, n \) left continuous and \( g \) nonincreasing the inequality has to be reversed.

**Proof.** Immediate from the previous corollary together with \( i_m(y) \leq y \) if \( y = f(x) = f(x+) \) and \( i_m(y) \geq y \) if \( y = f(x) = f(x-) \) according to Lemma 9.18. \( \square \)

**Problem 9.29.** Show (9.51).

**Problem 9.30.** Show that \( \Gamma(f) \circ \Gamma(f^{-1}) = \{(y, z) | y, z \in [f(x-), f(x+)] \text{ for some } x \} \) and \( \Gamma(f^{-1}) \circ \Gamma(f) = \{(y, z) | f(y+) > f(z-) \text{ or } f(y-) < f(z+) \} \).

**Problem 9.31.** Let \( d\mu(\lambda) := \chi_{[0,1]}(\lambda)d\lambda \) and \( f(\lambda) := \chi_{(-\infty,t]}(\lambda), t \in \mathbb{R} \). Compute \( f_*\mu \).
9.5. Appendix: The connection with the Riemann integral

In this section we want to investigate the connection with the Riemann integral. We restrict our attention to compact intervals \([a, b]\) and bounded real-valued functions \(f\). A partition of \([a, b]\) is a finite set \(P = \{x_0, \ldots, x_n\}\) with

\[
a = x_0 < x_1 < \cdots < x_{n-1} < x_n = b.
\]  

(9.58)

The number

\[
\|P\| := \max_{1 \leq j \leq n} x_j - x_{j-1}
\]  

(9.59)

is called the norm of \(P\). Given a partition \(P\) and a bounded real-valued function \(f\), we can define

\[
s_{P, f, -}(x) := \sum_{j=1}^{n} m_j \chi_{[x_{j-1}, x_j)}(x), \quad m_j := \inf_{x \in [x_{j-1}, x_j]} f(x),
\]  

(9.60)

\[
s_{P, f, +}(x) := \sum_{j=1}^{n} M_j \chi_{[x_{j-1}, x_j)}(x), \quad M_j := \sup_{x \in [x_{j-1}, x_j]} f(x).
\]  

(9.61)

Hence \(s_{P, f, -}(x)\) is a step function approximating \(f\) from below and \(s_{P, f, +}(x)\) is a step function approximating \(f\) from above. In particular,

\[
m \leq s_{P, f, -}(x) \leq f(x) \leq s_{P, f, +}(x) \leq M, \quad m := \inf_{x \in [a, b]} f(x), M := \sup_{x \in [a, b]} f(x).
\]  

(9.62)

Moreover, we can define the upper and lower Riemann sum associated with \(P\) as

\[
L(P, f) := \sum_{j=1}^{n} m_j (x_j - x_{j-1}), \quad U(P, f) := \sum_{j=1}^{n} M_j (x_j - x_{j-1}).
\]  

(9.63)

Of course, \(L(f, P)\) is just the Lebesgue integral of \(s_{P, f, -}\) and \(U(f, P)\) is the Lebesgue integral of \(s_{P, f, +}\). In particular, \(L(P, f)\) approximates the area under the graph of \(f\) from below and \(U(P, f)\) approximates this area from above.

By the above inequality

\[
m (b - a) \leq L(P, f) \leq U(P, f) \leq M (b - a).
\]  

(9.64)

We say that the partition \(P_2\) is a refinement of \(P_1\) if \(P_1 \subseteq P_2\) and it is not hard to check, that in this case

\[
s_{P_1, f, -}(x) \leq s_{P_2, f, -}(x) \leq f(x) \leq s_{P_2, f, +}(x) \leq s_{P_1, f, +}(x)
\]  

(9.65)

as well as

\[
L(P_1, f) \leq L(P_2, f) \leq U(P_2, f) \leq U(P_1, f).
\]  

(9.66)
Hence we define the lower, upper Riemann integral of \( f \) as
\[
\int f(x) \, dx := \sup_P L(P, f), \quad \overline{\int} f(x) \, dx := \inf_P U(P, f),
\]
respectively. Since for arbitrary partitions \( P \) and \( Q \) we have
\[
L(P, f) \leq L(P \cup Q, f) \leq U(P \cup Q, f) \leq U(Q, f).
\]
we obtain
\[
m(b - a) \leq \int f(x) \, dx \leq \overline{\int} f(x) \, dx \leq M(b - a).
\]
We will call \( f \) Riemann integrable if both values coincide and the common value will be called the Riemann integral of \( f \).

Example. Let \([a, b] := [0, 1]\) and \( f(x) := \chi_Q(x) \). Then \( s_{P,L}(x) = 0 \) and \( s_{P,U}(x) = 1 \). Hence \( \int f(x) \, dx = 0 \) and \( \overline{\int} f(x) \, dx = 1 \) and \( f \) is not Riemann integrable.

On the other hand, every continuous function \( f \in C[a, b] \) is Riemann integrable (Problem 9.32).

Example. Let \( f \) nondecreasing, then \( f \) is integrable. In fact, since \( m_j = f(x_{j-1}) \) and \( M_j = f(x_j) \) we obtain
\[
U(f, P) - L(f, P) \leq \|P\| \sum_{j=1}^{n} (f(x_j) - f(x_{j-1})) = \|P\|(f(b) - f(a))
\]
and the claim follows (cf. also the next lemma). Similarly nonincreasing functions are integrable.

Lemma 9.23. A function \( f \) is Riemann integrable if and only if there exists a sequence of partitions \( P_n \) such that
\[
\lim_{n \to \infty} L(P_n, f) = \lim_{n \to \infty} U(P_n, f).
\]
In this case the above limits equal the Riemann integral of \( f \) and \( P_n \) can be chosen such that \( P_n \subseteq P_{n+1} \) and \( \|P_n\| \to 0 \).

Proof. If there is such a sequence of partitions then \( f \) is integrable by \( \lim_{n} L(P_n, f) \leq \sup_P L(P, f) \leq \inf_P U(P, f) \leq \lim_{n} U(P_n, f) \).

Conversely, given an integrable \( f \), there is a sequence of partitions \( P_{L,n} \) such that \( \int f(x) \, dx = \lim_{n} L(P_{L,n}, f) \) and a sequence \( P_{U,n} \) such that \( \overline{\int} f(x) \, dx = \lim_{n} U(P_{U,n}, f) \). By (9.66) the common refinement \( P_n = P_{L,n} \cup P_{U,n} \) is the partition we are looking for. Since, again by (9.66), any refinement will also work, the last claim follows.
Note that when computing the Riemann integral as in the previous lemma one could choose instead of $m_j$ or $M_j$ any value in $[m_j, M_j]$ (e.g. $f(x_{j-1})$ or $f(x_j)$).

With the help of this lemma we can give a characterization of Riemann integrable functions and show that the Riemann integral coincides with the Lebesgue integral.

**Theorem 9.24 (Lebesgue).** A bounded measurable function $f : [a, b] \to \mathbb{R}$ is Riemann integrable if and only if the set of its discontinuities is of Lebesgue measure zero. Moreover, in this case the Riemann and the Lebesgue integral of $f$ coincide.

**Proof.** Suppose $f$ is Riemann integrable and let $P_j$ be a sequence of partitions as in Lemma 9.23. Then $s_{f, P_j, -}(x)$ will be monotone and hence converge to some function $s_{f, -}(x) \leq f(x)$. Similarly, $s_{f, P_j, +}(x)$ will converge to some function $s_{f, +}(x) \geq f(x)$. Moreover, by dominated convergence

$$0 = \lim_j \int (s_{f, P_j, +}(x) - s_{f, P_j, -}(x)) \, dx = \int (s_{f, +}(x) - s_{f, -}(x)) \, dx$$

and thus by Lemma 9.6 $s_{f, +}(x) = s_{f, -}(x)$ almost everywhere. Moreover, $f$ is continuous at every $x$ at which equality holds and which is not in any of the partitions. Since the first as well as the second set have Lebesgue measure zero, $f$ is continuous almost everywhere and

$$\lim_j L(P_j, f) = \lim_j U(P_j, f) = \int s_{f, \pm}(x) \, dx = \int f(x) \, dx.$$

Conversely, let $f$ be continuous almost everywhere and choose some sequence of partitions $P_j$ with $\|P_j\| \to 0$. Then at every $x$ where $f$ is continuous we have $\lim_j s_{f, P_j, \pm}(x) = f(x)$ implying

$$\lim_j L(P_j, f) = \int s_{f, -}(x) \, dx = \int f(x) \, dx = \int s_{f, +}(x) \, dx = \lim_j U(P_j, f)$$

by the dominated convergence theorem. \hfill \Box

Note that if $f$ is not assumed to be measurable, the above proof still shows that $f$ satisfies $s_{f, -} \leq f \leq s_{f, +}$ for two measurable functions $s_{f, \pm}$ which are equal almost everywhere. Hence if we replace the Lebesgue measure by its completion, we can drop this assumption.

Finally, recall that if one endpoint is unbounded or $f$ is unbounded near one endpoint, one defines the **improper Riemann integral** by taking limits towards this endpoint. More specifically, if $f$ is Riemann integrable for every $(a, c) \subset (a, b)$ one defines

$$\int_a^b f(x) \, dx := \lim_{c \uparrow b} \int_a^c f(x) \, dx \tag{9.71}$$
with an analogous definition if \( f \) is Riemann integrable for every \((c, b) \subset (a, b)\). Note that in this case improper integrability no longer implies Lebesgue integrability unless \(|f(x)|\) has a finite improper integral. The prototypical example being the Dirichlet integral

\[
\int_0^\infty \frac{\sin(x)}{x} \, dx = \lim_{c \to \infty} \int_0^c \frac{\sin(x)}{x} \, dx = \frac{\pi}{2}
\] (9.72)

(cf. Problem 13.24) which does not exist as a Lebesgue integral since

\[
\int_0^\infty \frac{\sin(x)}{x} \, dx \geq \sum_{k=0}^{\infty} \int_{\pi/4}^{3\pi/4} \frac{\sin(k\pi + x)}{k\pi + 3/4} \, dx \geq \frac{1}{2\sqrt{2}} \sum_{k=1}^{\infty} \frac{1}{k} = \infty. \quad (9.73)
\]

**Problem 9.32.** Show that for any function \( f \in C[a, b] \) we have

\[
\lim_{\|P\| \to 0} L(P, f) = \lim_{\|P\| \to 0} U(P, f).
\]

In particular, \( f \) is Riemann integrable.

**Problem 9.33.** Prove that the Riemann integral is linear: If \( f, g \) are Riemann integrable and \( \alpha \in \mathbb{R} \), then \( \alpha f \) and \( f + g \) are Riemann integrable with

\[
\int (f + g) \, dx = \int f \, dx + \int g \, dx \quad \text{and} \quad \int \alpha f \, dx = \alpha \int f \, dx.
\]

**Problem 9.34.** Show that if \( f, g \) are Riemann integrable, so is \( fg \). (Hint: First show that \( f^2 \) is integrable and then reduce it to this case).

**Problem 9.35.** Let \( \{q_n\}_{n \in \mathbb{N}} \) be an enumeration of the rational numbers in [0, 1). Show that

\[
f(x) := \sum_{n \in \mathbb{N} : q_n < x} \frac{1}{2^n}
\]

is discontinuous at every \( q_n \) but still Riemann integrable.
The Lebesgue spaces $L^p$

10.1. Functions almost everywhere

We fix some measure space $(X, \Sigma, \mu)$ and define the $L^p$ norm by

$$\|f\|_p := \left(\int_X |f|^p \, d\mu\right)^{1/p}, \quad 1 \leq p,$$

and denote by $L^p(X, d\mu)$ the set of all complex-valued measurable functions for which $\|f\|_p$ is finite. First of all note that $L^p(X, d\mu)$ is a vector space, since $|f + g|^p \leq 2^p \max(|f|^p, |g|^p) = 2^p \max(|f|^p, |g|^p) \leq 2^p(|f|^p + |g|^p)$. Of course our hope is that $L^p(X, d\mu)$ is a Banach space. However, Lemma 9.6 implies that there is a small technical problem (recall that a property is said to hold almost everywhere if the set where it fails to hold is contained in a set of measure zero):

Lemma 10.1. Let $f$ be measurable. Then

$$\int_X |f|^p \, d\mu = 0$$

if and only if $f(x) = 0$ almost everywhere with respect to $\mu$.

Thus $\|f\|_p = 0$ only implies $f(x) = 0$ for almost every $x$, but not for all! Hence $\|\|_p$ is not a norm on $L^p(X, d\mu)$. The way out of this misery is to identify functions which are equal almost everywhere: Let

$$\mathcal{N}(X, d\mu) := \{f | f(x) = 0 \text{ } \mu\text{-almost everywhere}\}.$$
Then $\mathcal{N}(X, d\mu)$ is a linear subspace of $L^p(X, d\mu)$ and we can consider the quotient space

$$L^p(X, d\mu) := L^p(X, d\mu)/\mathcal{N}(X, d\mu).$$

(10.4)

If $d\mu$ is the Lebesgue measure on $X \subseteq \mathbb{R}^n$, we simply write $L^p(X)$. Observe that $\|f\|_p$ is well defined on $L^p(X, d\mu)$ and hence we have a normed space.

Even though the elements of $L^p(X, d\mu)$ are, strictly speaking, equivalence classes of functions, we will still treat them functions for notational convenience. However, if we do so it is important to ensure that every statement made does not depend on the representative in the equivalence classes. In particular, note that for $f \in L^p(X, d\mu)$ the value $f(x)$ is not well defined (unless there is a continuous representative and continuous functions with different values are in different equivalence classes, e.g., in the case of Lebesgue measure).

With this modification we are back in business since $L^p(X, d\mu)$ turns out to be a Banach space. We will show this in the following sections. Moreover, note that $L^2(X, d\mu)$ is a Hilbert space with scalar product given by

$$\langle f, g \rangle := \int_X f(x)^* g(x) d\mu(x).$$

(10.5)

But before that let us also define $L^\infty(X, d\mu)$. It should be the set of bounded measurable functions $B(X)$ together with the sup norm. The only problem is that if we want to identify functions equal almost everywhere, the supremum is no longer independent of the representative in the equivalence class. The solution is the essential supremum

$$\|f\|_\infty := \inf\{C \mid \mu(\{x \mid |f(x)| > C\}) = 0\}. \quad (10.6)$$

That is, $C$ is an essential bound if $|f(x)| \leq C$ almost everywhere and the essential supremum is the infimum over all essential bounds.

**Example.** If $\lambda$ is the Lebesgue measure, then the essential sup of $\chi_Q$ with respect to $\lambda$ is 0. If $\Theta$ is the Dirac measure centered at 0, then the essential sup of $\chi_Q$ with respect to $\Theta$ is 1 (since $\chi_Q(0) = 1$, and $x = 0$ is the only point which counts for $\Theta$).

As before we set

$$L^\infty(X, d\mu) := B(X)/\mathcal{N}(X, d\mu)$$

(10.7)

and observe that $\|f\|_\infty$ is independent of the representative from the equivalence class.

If you wonder where the $\infty$ comes from, have a look at Problem 10.2.

Since the support of a function in $L^p$ is also not well defined one uses the essential support in this case:

$$\text{supp}(f) = X \setminus \bigcup\{O \mid f = 0 \text{ } \mu\text{-almost everywhere on } O \subseteq X \text{ open}\}. \quad (10.8)$$
In other words, $x$ is in the essential support if for every neighborhood the set of points where $f$ does not vanish has positive measure. Here we use the same notation as for functions and it should be understood from the context which one is meant. Note that the essential support is always smaller than the support (since we get the latter if we require $f$ to vanish everywhere on $O$ in the above definition).

**Example.** The support of $\chi_{\mathbb{Q}}$ is $\overline{\mathbb{Q}} = \mathbb{R}$ but the essential support with respect to Lebesgue measure is $\emptyset$ since the function is 0 a.e. \hfill \diamond

If $X$ is a locally compact Hausdorff space (together with the Borel sigma algebra), a function is called **locally integrable** if it is integrable when restricted to any compact subset $K \subseteq X$. The set of all (equivalence classes of) locally integrable functions will be denoted by $L^1_{loc}(X,d\mu)$. We will say that $f_n \to f$ in $L^1_{loc}(X,d\mu)$ if this holds on $L^1(K,d\mu)$ for all compact subsets $K \subseteq X$. Of course this definition extends to $L^p$ for any $1 \leq p \leq \infty$.

**Problem 10.1.** Let $\| \cdot \|$ be a seminorm on a vector space $X$. Show that $N := \{ x \in X | \| x \| = 0 \}$ is a vector space. Show that the quotient space $X/N$ is a normed space with norm $\| x + N \| := \| x \|$. 

**Problem 10.2.** Suppose $\mu(X) < \infty$. Show that $L^\infty(X,d\mu) \subseteq L^p(X,d\mu)$ and
\[
\lim_{p \to \infty} \| f \|_p = \| f \|_\infty, \quad f \in L^\infty(X,d\mu).
\]

**Problem 10.3.** Construct a function $f \in L^p(0,1)$ which has a singularity at every rational number in $[0,1]$ (such that the essential supremum is infinite on every open subinterval). (Hint: Start with the function $f_0(x) = |x|^{-\alpha}$ which has a single singularity at 0, then $f_j(x) = f_0(x-x_j)$ has a singularity at $x_j$.)

**Problem 10.4.** Show that for a continuous function on $\mathbb{R}^n$ the support and the essential support with respect to Lebesgue measure coincide.

### 10.2. Jensen $\leq$ Hölder $\leq$ Minkowski

As a preparation for proving that $L^p$ is a Banach space, we will need Hölder’s inequality, which plays a central role in the theory of $L^p$ spaces. In particular, it will imply Minkowski’s inequality, which is just the triangle inequality for $L^p$. Our proof is based on Jensen’s inequality and emphasizes the connection with convexity. In fact, the triangle inequality just states that a norm is convex:
\[
\|(1-\lambda)f + \lambda g\| \leq \lambda \|f\| + \lambda \|g\|, \quad \lambda \in (0,1). \quad (10.9)
\]
Recall that a real function $\varphi$ defined on an open interval $(a, b)$ is called convex if
\[
\varphi((1 - \lambda)x + \lambda y) \leq (1 - \lambda)\varphi(x) + \lambda \varphi(y), \quad \lambda \in (0, 1),
\]
that is, on $(x, y)$ the graph of $\varphi(x)$ lies below or on the line connecting $(x, \varphi(x))$ and $(y, \varphi(y))$:

If the inequality is strict, then $\varphi$ is called strictly convex. It is not hard to see (use $z = (1 - \lambda)x + \lambda y$) that the definition implies
\[
\frac{\varphi(z) - \varphi(x)}{z - x} \leq \frac{\varphi(y) - \varphi(x)}{y - x} \leq \frac{\varphi(y) - \varphi(z)}{y - z}, \quad x < z < y,
\]
where the inequalities are strict if $\varphi$ is strictly convex. A function $\varphi$ is concave if $-\varphi$ is convex.

**Lemma 10.2.** Let $\varphi : (a, b) \to \mathbb{R}$ be convex. Then

(i) $\varphi$ is locally Lipschitz continuous.

(ii) The left/right derivatives $\varphi'_\pm(x) = \lim_{\varepsilon \downarrow 0} \frac{\varphi(x \pm \varepsilon) - \varphi(x)}{\pm \varepsilon}$ exist and are monotone nondecreasing. Moreover, $\varphi'$ exists except at a countable number of points.

(iii) For fixed $x$ we have $\varphi(y) \geq \varphi(x) + \alpha(y - x)$ for every $\alpha$ with $\varphi'_-(x) \leq \alpha \leq \varphi'_+(x)$. The inequality is strict for $y \neq x$ if $\varphi$ is strictly convex.

**Proof.** Abbreviate $D(x, y) = D(y, x) := \frac{\varphi(y) - \varphi(x)}{y - x}$ and observe that (10.11) implies
\[
D(x, z) \leq D(y, z) \quad \text{for } x < z < y.
\]
Hence $\varphi'_\pm(x)$ exist and we have $\varphi'_-(x) \leq \varphi'_+(x) \leq \varphi'_-(y) \leq \varphi'_+(y)$ for $x < y$. So (ii) follows after observing that a monotone function can have at most a countable number of jumps. Next
\[
\varphi'_+(x) \leq D(y, x) \leq \varphi'_-(x)
\]
shows $\varphi(y) \geq \varphi(x) + \varphi'_+(x)(y - x)$ if $\pm(y - x) > 0$ and proves (iii). Moreover, $\varphi'_+(z) \leq D(y, x) \leq \varphi'_-(\tilde{z})$ for $z < x, y < \tilde{z}$ proves (i). \qed
10.2. Jensen ≤ Hölder ≤ Minkowski

Remark: It is not hard to see that \( \varphi \in C^1 \) is convex if and only if \( \varphi'(x) \) is monotone nondecreasing (e.g., \( \varphi'' \geq 0 \) if \( \varphi \in C^2 \)) — Problem 10.5.

With these preparations out of the way we can show

**Theorem 10.3** (Jensen’s inequality). Let \( \varphi : (a, b) \to \mathbb{R} \) be convex (\( a = -\infty \) or \( b = \infty \) being allowed). Suppose \( \mu \) is a finite measure satisfying \( \mu(X) = 1 \) and \( f \in L^1(X, d\mu) \) with \( a < f(x) < b \). Then the negative part of \( \varphi \circ f \) is integrable and

\[
\varphi\left(\int_X f \, d\mu\right) \leq \int_X (\varphi \circ f) \, d\mu.
\]

(10.12)

For \( \varphi \geq 0 \) nondecreasing and \( f \geq 0 \) the requirement that \( f \) is integrable can be dropped if \( \varphi(b) \) is understood as \( \lim_{x \to b} \varphi(x) \).

**Proof.** By (iii) of the previous lemma we have

\[
\varphi(f(x)) \geq \varphi(I) + \alpha(f(x) - I), \quad I = \int_X f \, d\mu \in (a, b).
\]

This shows that the negative part of \( \varphi \circ f \) is integrable and integrating over \( X \) finishes the proof in the case \( f \in L^1 \). If \( f \geq 0 \) we note that for \( X_n = \{ x \in X | f(x) \leq n \} \) the first part implies

\[
\varphi\left(\frac{1}{\mu(X_n)} \int_{X_n} f \, d\mu\right) \leq \frac{1}{\mu(X_n)} \int_{X_n} \varphi(f) \, d\mu.
\]

Taking \( n \to \infty \) the claim follows from \( X_n \nearrow X \) and the monotone convergence theorem.

Observe that if \( \varphi \) is strictly convex, then equality can only occur if \( f \) is constant.

Now we are ready to prove

**Theorem 10.4** (Hölder’s inequality). Let \( p \) and \( q \) be dual indices; that is,

\[
\frac{1}{p} + \frac{1}{q} = 1
\]

(10.13)

with \( 1 \leq p \leq \infty \). If \( f \in L^p(X, d\mu) \) and \( g \in L^q(X, d\mu) \), then \( fg \in L^1(X, d\mu) \) and

\[
\|fg\|_1 \leq \|f\|_p \|g\|_q.
\]

(10.14)

**Proof.** The case \( p = 1, q = \infty \) (respectively \( p = \infty, q = 1 \)) follows directly from the properties of the integral and hence it remains to consider \( 1 < p, q < \infty \).

First of all it is no restriction to assume \( \|g\|_q = 1 \). Let \( A = \{ x \mid |g(x)| > 0 \} \), then (note \( (1 - q)p = -q \))

\[
\|fg\|_1^p \leq \int_A |f| |g|^{1-q} |g|^q d\mu \leq \int_A (|f| |g|^{1-q})^p |g|^q d\mu = \int_A |f|^p d\mu \leq \|f\|_p^p.
\]
where we have used Jensen’s inequality with \( \varphi(x) = |x|^p \) applied to the function \( h = |f| |g|^{1-q} \) and measure \( d\nu = |g|^q d\mu \) (note \( \nu(X) = \int |g|^q d\mu = \|g\|_q^q = 1 \)).

Note that in the special case \( p = 2 \) we have \( q = 2 \) and Hölder’s inequality reduces to the Cauchy–Schwarz inequality. For a generalization see Problem 10.8. Moreover, in the case \( 1 < p < \infty \) the function \( x^p \) is strictly convex and equality will occur precisely if \( |f| \) is a multiple of \( |g|^{q-1} \) or \( g \) is trivial. This gives us a

**Corollary 10.5.** Consider \( f \in L^p(X, d\mu) \) with if \( 1 \leq p < \infty \) and let \( q \) be the corresponding dual index, \( \frac{1}{p} + \frac{1}{q} = 1 \). Then

\[
\|f\|_p = \sup_{\|g\|_q = 1} \left| \int_X fg \, d\mu \right|.
\] (10.15)

If every set of infinite measure has a subset of finite positive measure (e.g. if \( \mu \) is \( \sigma \)-finite), then the claim also holds for \( p = \infty \).

**Proof.** In the case \( 1 < p < \infty \) equality is attained for \( g = c^{-1} \text{sign}(f^*)|f|^{p-1} \), where \( c = \|f|^{p-1}\|_q \) (assuming \( c > 0 \) w.l.o.g.). In the case \( p = 1 \) equality is attained for \( g = \text{sign}(f^*) \). Now let us turn to the case \( p = \infty \). For every \( \varepsilon > 0 \) the set \( A_\varepsilon = \{ x \mid |f(x)| \geq \|f\|_\infty - \varepsilon \} \) has positive measure. Moreover, by assumption on \( \mu \) we can choose a subset \( B_\varepsilon \subseteq A_\varepsilon \) with finite positive measure. Then \( g_\varepsilon = \text{sign}(f^*)\chi_{B_\varepsilon}/\mu(B_\varepsilon) \) satisfies \( \int_X fg_\varepsilon \, d\mu \geq \|f\|_\infty - \varepsilon \). \( \square \)

Of course it suffices to take the sup in (10.15) over a dense set of \( L^q \) (e.g. integrable simple functions — see Problem 10.15). Moreover, note that the extra assumption for \( p = \infty \) is crucial since if there is a set of infinite measure which has no subset with finite positive measure, then every integrable function must vanish on this subset.

If it is not a priori known that \( f \in L^p \) the following generalization will be useful.

**Lemma 10.6.** Suppose \( \mu \) is \( \sigma \)-finite. Consider \( L^p(X, d\mu) \) with if \( 1 \leq p \leq \infty \) and let \( q \) be the corresponding dual index, \( \frac{1}{p} + \frac{1}{q} = 1 \). If \( fs \in L^1 \) for every simple function \( s \in L^q \), then

\[
\|f\|_p = \sup_{s \text{ simple}, \|s\|_q = 1} \left| \int_X fs \, d\mu \right|.
\]

**Proof.** If \( \|f\|_p < \infty \) the claim follows from the previous corollary since simple functions are dense (Problem 10.15). Conversely, if \( \|f\|_p = \infty \) recall that \( f = (f_1 - f_2) + i(f_2 - f_4) \) can be decomposed into four nonnegative functions at least one of which, say the positive part of the real part \( f_1 \), has infinite norm. Now choose \( s_n \nearrow f_1 \) as in (9.6). Moreover, since \( \mu \) is \( \sigma \)-finite
we can find \( X_n \neq X \) with \( \mu(X_n) < \infty \). Then \( \tilde{s}_n = \chi_{X_n} s_n \) will be in \( L^p \) and will still satisfy \( \tilde{s}_n \nearrow f_1 \). Now if \( 1 \leq p < \infty \) choose \( \hat{s}_n = \text{sign}(\tilde{s}_n) \tilde{s}_n^{1-p} \) such that \( \hat{s}_n f_1 \nearrow |f_1|^p \) and use monotone convergence to conclude that the sup is infinite. If \( p = \infty \) there is some \( m \) such that \( \mu(A_n) > 0 \), where \( A_n = \{ X_m | f_1 \geq n \} \) and use \( s_n = \mu(A_n)^{-1} \chi_{A_n} \) to conclude that the sup is infinite.

Again note that if there is a set of infinite measure which has no subset with finite positive measure, then every integrable function must vanish on this subset and hence the above lemma cannot work in such a situation.

As another consequence we get

**Theorem 10.7** (Minkowski’s integral inequality). Suppose, \( \mu \) and \( \nu \) are two \( \sigma \)-finite measures and \( f \) is \( \mu \otimes \nu \) measurable. Let \( 1 \leq p \leq \infty \). Then

\[
\left\| \int_Y f(., y) d\nu(y) \right\|_p \leq \int_Y \| f(., y) \|_p d\nu(y), \tag{10.16}
\]

where the \( p \)-norm is computed with respect to \( \mu \).

**Proof.** Let \( g \in L^q(X, d\mu) \) with \( g \geq 0 \) and \( \|g\|_q = 1 \). Then using Fubini

\[
\int_X g(x) \int_Y |f(x, y)| d\nu(y) d\mu(x) = \int_Y \int_X |f(x, y)| g(x) d\mu(x) d\nu(y)
\]

\[
\leq \int_Y \| f(., y) \|_p d\nu(y)
\]

and the claim follows from Lemma 10.6. \( \square \)

In the special case where \( \nu \) is supported on two points this reduces to the triangle inequality (our proof inherits the assumption that \( \mu \) is \( \sigma \)-finite, but this can be avoided – Problem 10.7).

**Corollary 10.8** (Minkowski’s inequality). Let \( f, g \in L^p(X, d\mu) \), \( 1 \leq p \leq \infty \). Then

\[
\| f + g \|_p \leq \| f \|_p + \| g \|_p. \tag{10.17}
\]

This shows that \( L^p(X, d\mu) \) is a normed vector space.

Note that Fatou’s lemma implies that the norm is lower semi continuous \( \| f \|_p \leq \lim \inf_{n \to \infty} \| f_n \|_p \) with respect to pointwise convergence (a.e.). The next lemma sheds some light on the missing part.

**Lemma 10.9** (Brezis–Lieb). Let \( 1 \leq p < \infty \) and let \( f_n \in L^p(X, d\mu) \) be a sequence which converges pointwise a.e. to \( f \) such that \( \| f_n \|_p \leq C \) if \( p > 1 \). Then \( f \in L^p(X, d\mu) \) and

\[
\lim_{n \to \infty} \left( \| f_n \|_p^p - \| f_n - f \|_p^p \right) = \| f \|_p^p. \tag{10.18}
\]

In the case \( p = 1 \) we can replace \( \| f_n \|_1 \leq C \) by \( f \in L^1(X, d\mu) \).
Proof. As pointed out before \( \|f\|_p \leq \liminf_{n \to \infty} \|f_n\|_p \leq C \) which shows \( f \in L^p(X, d\mu) \). Moreover, one easily checks the elementary inequality
\[
|s + t|^p - |t|^p - |s|^p| \leq \|s + t|^p - |t|^p - |s|^p| \leq \varepsilon |t|^p + C_\varepsilon |s|^p
\]
(note that by scaling it suffices to consider the case \( s = 1 \)). Setting \( t = f_n - f \) and \( s = f \), bringing everything to the right-hand-side and applying Fatou gives
\[
C_\varepsilon \|f\|_p^p \leq \liminf_{n \to \infty} \int_X (\varepsilon |f_n - f|^p + C_\varepsilon |f|^p - |f_n|^p - |f - f_n|^p - |f|^p) \, d\mu
\]
\[
\leq \varepsilon (2C)^p + C_\varepsilon \|f\|_p^p - \limsup_{n \to \infty} \int_X |f_n|^p - |f - f_n|^p - |f|^p |d\mu.
\]
Since \( \varepsilon > 0 \) is arbitrary the claim follows. Finally, note that in the case \( p = 1 \) we can choose \( \varepsilon = 0 \) and \( C_\varepsilon = 2 \). \( \square \)

It might be more descriptive to write the conclusion of the lemma as
\[
\|f_n\|_p^p = \|f\|_p^p + \|f_n - f\|_p^p + o(1) \tag{10.19}
\]
which shows an important consequence:

Corollary 10.10. Let \( 1 \leq p < \infty \) and let \( f_n \in L^p(X, d\mu) \) be a sequence which converges pointwise a.e. to \( f \) such that \( \|f_n\|_p \leq C \) if \( p > 1 \) or \( f \in L^1(X, d\mu) \) if \( p = 1 \). Then \( \|f_n - f\|_p \to 0 \) if and only if \( \|f_n\|_p \to \|f\|_p \).

If \( n \) is large enough then \( \|f_n\|_p \leq C \) and \( C_\varepsilon \|f\|_p \leq \varepsilon (2C)^p + C_\varepsilon \|f\|_p^p - \limsup_{n \to \infty} \int_X |f_n|^p - |f_n - f|^p - |f|^p |d\mu.\)

Note that it even suffices to show \( \limsup_{n \to \infty} \|f_n\|_p \leq \|f\|_p \) since \( \|f\|_p \leq \liminf_{n \to \infty} \|f_n\|_p \) comes for free from Fatou as pointed out before.

Note that a similar conclusion holds for weakly convergent sequences by the Radon–Riesz theorem (Theorem 5.19) since \( L^p \) is uniformly convex for \( 1 < p < \infty \).

Theorem 10.11 (Clarkson). Suppose \( 1 < p < \infty \), then \( L^p(X, d\mu) \) is uniformly convex.

Proof. As a preparation we note that strict convexity of \(|.|^p\) implies that
\[
|\frac{t + s}{2}|^p \leq \frac{|t|^p + |s|^p}{2} < \frac{|t|^p + |s|^p}{2}
\]
and hence
\[
\rho(\varepsilon) := \min \left\{ \frac{|t|^p + |s|^p}{2} - \frac{|t|^p + |s|^p}{2} \middle| |t| + |s| = 2, \ \frac{|t - s|^p}{2} \geq \varepsilon \right\} > 0.
\]
Hence, by scaling,
\[
|\frac{t - s}{2}|^p \geq \varepsilon \frac{|t|^p + |s|^p}{2} \Rightarrow \frac{|t|^p + |s|^p}{2} - \rho(\varepsilon) \leq \frac{|t|^p + |s|^p}{2} - \frac{|t + s|^p}{2}.
\]
Now given \( f, g \) with \( \|f\|_p = \|g\|_p = 1 \) and \( \varepsilon > 0 \) we need to find a \( \delta > 0 \) such that \( \|\frac{t + s}{2}\|_p > 1 - \delta \) implies \( \|f - g\|_p < 2\varepsilon \). Introduce
\[
M := \left\{ x \in X \middle| \left| \frac{f(x) - g(x)}{2} \right|^p \geq \varepsilon \frac{|f(x)|^p + |g(x)|^p}{2} \right\}.
\]
Then
\[ \int_X |\frac{f-g}{2}|^p d\mu = \int_{X \setminus M} |\frac{f-g}{2}|^p d\mu + \int_M |\frac{f-g}{2}|^p d\mu \]
\[ \leq \varepsilon \int_{X \setminus M} |f|^p + |g|^p d\mu + \int_M |f|^p + |g|^p d\mu \]
\[ \leq \varepsilon \int_{X \setminus M} |f|^p + |g|^p d\mu + \frac{1}{\rho} \int_M \left( |f|^p + |g|^p - |\frac{f+g}{2}|^p \right) d\mu \]
\[ \leq \varepsilon + \frac{1 - (1 - \delta)^p}{\rho} < 2\varepsilon \]
provided \( \delta < (1 + \varepsilon \rho)^{1/p} - 1. \)

\[ \square \]

Problem 10.5. Show that \( \varphi \in C^1(a,b) \) is (strictly) convex if and only if \( \varphi' \) is (strictly) increasing. Moreover, \( \varphi \in C^2(a,b) \) is (strictly) convex if and only if \( \varphi'' \geq 0 \) (\( \varphi'' > 0 \) a.e.).

Problem 10.6. Prove
\[ \prod_{k=1}^n x_k^{\alpha_k} \leq \sum_{k=1}^n \alpha_k x_k, \quad \text{if} \quad \sum_{k=1}^n \alpha_k = 1, \]
for \( \alpha_k > 0, x_k > 0. \) (Hint: Take a sum of Dirac-measures and use that the exponential function is convex.)

Problem 10.7. Show Minkowski’s inequality directly from Hölder’s inequality. Show that \( L^p(X,d\mu) \) is strictly convex for \( 1 < p < \infty \) but not for \( p = 1, \infty \) if \( X \) contains two disjoint subsets of positive finite measure. (Hint: Start from \( |f + g|^p \leq |f|^p |f + g|^{p-1} + |g|^p |f + g|^{p-1}. \))

Problem 10.8. Show the generalized Hölder’s inequality:
\[ \|fg\|_r \leq \|f\|_p \|g\|_q, \quad \frac{1}{p} + \frac{1}{q} = \frac{1}{r}. \quad (10.20) \]

Problem 10.9. Show the iterated Hölder’s inequality:
\[ \|f_1 \cdots f_m\|_r \leq \prod_{j=1}^m \|f_j\|_{p_j}, \quad \frac{1}{p_1} + \cdots + \frac{1}{p_m} = \frac{1}{r}. \quad (10.21) \]

Problem 10.10. Suppose \( \mu \) is finite. Show that \( L^{p_0} \subseteq L^p \) and
\[ \|f\|_{p_0} \leq \mu(X)^{\frac{1}{p_0} - \frac{1}{p}} \|f\|_p, \quad 1 \leq p_0 \leq p. \]
(Hint: Hölder’s inequality.)
Problem 10.11. Let $1 < p < \infty$ and $\mu$ $\sigma$-finite. Let $f_n \in L^p(X, d\mu)$ be a sequence which converges pointwise a.e. to $f$ such that $\|f_n\|_p \leq C$. Then

$$\int_X f_n g \, d\mu \to \int_X f g \, d\mu$$

for every $g \in L^q(X, d\mu)$. By Theorem 12.1 this implies that $f_n$ converges weakly to $f$. (Hint: Theorem 8.24 and Problem 4.32.)

10.3. Nothing missing in $L^p$

Finally it remains to show that $L^p(X, d\mu)$ is complete.

Theorem 10.12 (Riesz–Fischer). The space $L^p(X, d\mu)$, $1 \leq p \leq \infty$, is a Banach space.

Proof. We begin with the case $1 \leq p < \infty$. Suppose $f_n$ is a Cauchy sequence. It suffices to show that some subsequence converges (show this). Hence we can drop some terms such that

$$\|f_{n+1} - f_n\|_p \leq \frac{1}{2^n}.$$

Now consider $g_n = f_n - f_{n-1}$ (set $f_0 = 0$). Then

$$G(x) = \sum_{k=1}^\infty |g_k(x)|$$

is in $L^p$. This follows from

$$\left\| \sum_{k=1}^n |g_k| \right\|_p \leq \sum_{k=1}^n \|g_k\|_p \leq \|f_1\|_p + 1$$

using the monotone convergence theorem. In particular, $G(x) < \infty$ almost everywhere and the sum

$$\sum_{n=1}^\infty g_n(x) = \lim_{n \to \infty} f_n(x)$$

is absolutely convergent for those $x$. Now let $f(x)$ be this limit. Since $|f(x) - f_n(x)|^p$ converges to zero almost everywhere and $|f(x) - f_n(x)|^p \leq (2G(x))^p \in L^1$, dominated convergence shows $\|f - f_n\|_p \to 0$.

In the case $p = \infty$ note that the Cauchy sequence property $|f_n(x) - f_m(x)| < \varepsilon$ for $n, m > N$ holds except for sets $A_{m,n}$ of measure zero. Since $A = \bigcup_{n,m} A_{n,m}$ is again of measure zero, we see that $f_n(x)$ is a Cauchy sequence for $x \in X \setminus A$. The pointwise limit $f(x) = \lim_{n \to \infty} f_n(x)$, $x \in X \setminus A$, is the required limit in $L^\infty(X, d\mu)$ (show this).

In particular, in the proof of the last theorem we have seen:
Corollary 10.13. If $\|f_n - f\|_p \to 0$, $1 \leq p \leq \infty$, then there is a subsequence $f_{n_j}$ (of representatives) which converges pointwise almost everywhere and a function $g \in L^p(X,d\mu)$ such that $f_{n_j} \leq g$ almost everywhere.

Consequently, if $f_n \in L^{p_0} \cap L^{p_1}$ converges in both $L^{p_0}$ and $L^{p_1}$, then the limits will be equal a.e. Be warned that the statement is not true in general without passing to a subsequence (Problem 10.12).

It even turns out that $L^p$ is separable.

Lemma 10.14. Suppose $X$ is a second countable topological space (i.e., it has a countable basis) and $\mu$ is an outer regular Borel measure. Then $L^p(X,d\mu)$, $1 \leq p < \infty$, is separable. In particular, for every countable base the set of characteristic functions $\chi_O(x)$ with $O$ in this base is total.

Proof. The set of all characteristic functions $\chi_A(x)$ with $A \in \Sigma$ and $\mu(A) < \infty$ is total by construction of the integral (Problem 10.15). Now our strategy is as follows: Using outer regularity, we can restrict $A$ to open sets and using the existence of a countable base, we can restrict $A$ to open sets from this base.

Fix $A$. By outer regularity, there is a decreasing sequence of open sets $O_n \supseteq A$ such that $\mu(O_n) \to \mu(A)$. Since $\mu(A) < \infty$, it is no restriction to assume $\mu(O_n) < \infty$, and thus $\|\chi_A - \chi_{O_n}\|_p = \mu(O_n \setminus A) = \mu(O_n) - \mu(A) \to 0$.

Thus the set of all characteristic functions $\chi_O(x)$ with $O$ open and $\mu(O) < \infty$ is total. Finally let $\mathcal{B}$ be a countable base for the topology. Then, every open set $O$ can be written as $O = \bigcup_{j=1}^{\infty} \tilde{O}_j$ with $\tilde{O}_j \in \mathcal{B}$. Moreover, by considering the set of all finite unions of elements from $\mathcal{B}$, it is no restriction to assume $\bigcup_{j=1}^{n} \tilde{O}_j \in \mathcal{B}$. Hence there is an increasing sequence $\tilde{O}_n \nearrow O$ with $\tilde{O}_n \in \mathcal{B}$. By monotone convergence, $\|\chi_O - \chi_{\tilde{O}_n}\|_p \to 0$ and hence the set of all characteristic functions $\chi_{\tilde{O}}$ with $\tilde{O} \in \mathcal{B}$ is total. \hfill \Box

Problem 10.12. Find a sequence $f_n$ which converges to 0 in $L^p([0,1],dx)$, $1 \leq p < \infty$, but for which $f_n(x) \to 0$ for a.e. $x \in [0,1]$ does not hold.

(Hint: Every $n \in \mathbb{N}$ can be uniquely written as $n = 2^m + k$ with $0 \leq m$ and $0 \leq k < 2^m$. Now consider the characteristic functions of the intervals $I_{m,k} = [k2^{-m}, (k+1)2^{-m})$.)


Problem 10.14. Let $X_1$, $X_2$ be second countable topological spaces and let $\mu_1$, $\mu_2$ be outer regular $\sigma$-finite Borel measures. Let $\mathcal{B}_1$, $\mathcal{B}_2$ be bases for $X_1$, $X_2$, respectively. Show that the set of all functions $\chi_{O_1 \times O_2}(x_1,x_2) = \chi_{O_1}(x_1)\chi_{O_2}(x_2)$ for $O_1 \in \mathcal{B}_1$, $O_2 \in \mathcal{B}_2$ is total in $L^2(X_1 \times X_2, \mu_1 \otimes \mu_2)$.

(Hint: Lemma 9.12 and 10.14.)
Problem 10.15. Show that for any \( f \in L^p(X, d\mu) \), \( 1 \leq p \leq \infty \) there exists a sequence of simple functions \( s_n \) such that \( |s_n| \leq |f| \) and \( s_n \to f \) in \( L^p(X, d\mu) \). If \( p < \infty \) then \( s_n \) will be integrable. (Hint: Split \( f \) into the sum of four nonnegative functions and use (9.6).)

10.4. Approximation by nicer functions

Since measurable functions can be quite wild they are sometimes hard to work with. In fact, in many situations some properties are much easier to prove for a dense set of nice functions and the general case can then be reduced to the nice case by an approximation argument. But for such a strategy to work one needs to identify suitable sets of nice functions which are dense in \( L^p \).

Theorem 10.15. Let \( X \) be a locally compact metric space and let \( \mu \) be a regular Borel measure. Then the set \( C_c(X) \) of continuous functions with compact support is dense in \( L^p(X, d\mu) \), \( 1 \leq p < \infty \).

Proof. As in the proof of Lemma 10.14 the set of all characteristic functions \( \chi_K(x) \) with \( K \) compact is total (using inner regularity). Hence it suffices to show that \( \chi_K(x) \) can be approximated by continuous functions. By outer regularity there is an open set \( O \supset K \) such that \( \mu(O \setminus K) \leq \varepsilon \). By Urysohn’s lemma (Lemma 0.27) there is a continuous function \( f_\varepsilon : X \to [0, 1] \) with compact support which is 1 on \( K \) and 0 outside \( O \). Since

\[
\int_X |\chi_K - f_\varepsilon|^p d\mu = \int_{O \setminus K} |f_\varepsilon|^p d\mu \leq \mu(O \setminus K) \leq \varepsilon,
\]

we have \( \|f_\varepsilon - \chi_K\|_p \to 0 \) and we are done. \( \square \)

Clearly this result has to fail in the case \( p = \infty \) (in general) since the uniform limit of continuous functions is again continuous. In fact, the closure of \( C_c(\mathbb{R}^n) \) in the infinity norm is the space \( C_0(\mathbb{R}^n) \) of continuous functions vanishing at \( \infty \) (Problem 1.42). Another variant of this result is

Theorem 10.16 (Luzin). Let \( X \) be a locally compact metric space and let \( \mu \) be a finite regular Borel measure. Let \( f \) be integrable. Then for every \( \varepsilon > 0 \) there is an open set \( O_\varepsilon \) with \( \mu(O_\varepsilon) < \varepsilon \) and \( X \setminus O_\varepsilon \) compact and a continuous function \( g \) which coincides with \( f \) on \( X \setminus O_\varepsilon \).

Proof. From the proof of the previous theorem we know that the set of all characteristic functions \( \chi_K(x) \) with \( K \) compact is total. Hence we can restrict our attention to a sufficiently large compact set \( K \). By Theorem 10.15 we can find a sequence of continuous functions \( f_n \) which converges to \( f \) in \( L^1 \). After passing to a subsequence we can assume that \( f_n \) converges a.e. and by Egorov’s theorem there is a subset \( A_\varepsilon \) on which the convergence is
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uniform. By outer regularity we can replace $A_\varepsilon$ by a slightly larger open set $O_\varepsilon$ such that $C = K \setminus O_\varepsilon$ is compact. Now Tietze’s extension theorem implies that $f|_C$ can be extended to a continuous function $g$ on $X$. □

If $X$ is some subset of $\mathbb{R}^n$, we can do even better and approximate integrable functions by smooth functions. The idea is to replace the value $f(x)$ by a suitable average computed from the values in a neighborhood. This is done by choosing a nonnegative bump function $\phi$, whose area is normalized to 1, and considering the convolution

$$ (\phi * f)(x) := \int_{\mathbb{R}^n} \phi(x-y)f(y)dy = \int_{\mathbb{R}^n} \phi(y)f(x-y)dy. \quad (10.22) $$

For example, if we choose $\phi_r = |B_r(0)|^{-1} \chi_{B_r(0)}$ to be the characteristic function of a ball centered at 0, then $(\phi_r * f)(x)$ will be precisely the average of the values of $f$ in the ball $B_r(x)$. In the general case we can think of $(\phi * f)(x)$ as an weighted average. Moreover, if we choose $\phi$ differentiable, we can interchange differentiation and integration to conclude that $\phi * f$ will also be differentiable. Iterating this argument shows that $\phi * f$ will have as many derivatives as $\phi$. Finally, if the set over which the average is computed (i.e., the support of $\phi$) shrinks, we expect $(\phi * f)(x)$ to get closer and closer to $f(x)$.

To make these ideas precise we begin with a few properties of the convolution.

**Lemma 10.17.** The convolution has the following properties:

(i) $f(x)g(.)$ is integrable if and only if $f(.)g(x - .)$ is and

$$ (f * g)(x) = (g * f)(x) \quad (10.23) $$

in this case.

(ii) Suppose $\phi \in C^k_c(\mathbb{R}^n)$ and $f \in L^1_{loc}(\mathbb{R}^n)$, then $\phi * f \in C^k(\mathbb{R}^n)$ and

$$ \partial_\alpha(\phi * f) = (\partial_\alpha \phi) * f \quad (10.24) $$

for any partial derivative of order at most $k$.

(iii) We have $\text{supp}(f * g) \subseteq \text{supp}(f) + \text{supp}(g)$. In particular, if $\phi \in C^k_c(\mathbb{R}^n)$ and $f \in L^1_{loc}(\mathbb{R}^n)$, then $\phi * f \in C^k_c(\mathbb{R}^n)$.

(iv) Suppose $\phi \in L^1(\mathbb{R}^n)$ and $f \in L^p(\mathbb{R}^n)$, $1 \leq p \leq \infty$, then their convolution is in $L^p(\mathbb{R}^n)$ and satisfies Young’s inequality

$$ \|\phi * f\|_p \leq \|\phi\|_1\|f\|_p. \quad (10.25) $$

(v) Suppose $\phi \geq 0$ with $\|\phi\|_1 = 1$ and $f \in L^\infty(\mathbb{R}^n)$ real-valued, then

$$ \inf_{x \in \mathbb{R}^n} f(x) \leq (\phi * f)(x) \leq \sup_{x \in \mathbb{R}^n} f(x). \quad (10.26) $$
The standard mollifier (also Friedrichs mollifier) is \( \phi \in C_c^\infty(\mathbb{R}^n) \) with support \( \text{supp} \phi = \{ x \in \mathbb{R}^n : \phi(x) \neq 0 \} \) such that \( \int_{\mathbb{R}^n} \phi(x) dx = 1 \) and \( \phi(x) \) is smooth and rapidly decreasing. The mollifier is used to approximate smooth functions with integrable functions.

**Proof.** (i) is a simple affine change of coordinates. (ii) follows by interchanging differentiation with the integral using Problems 9.13 and 9.14. (iii) If \( x \notin \text{supp}(f) + \text{supp}(g) \), then \( x - y \notin \text{supp}(f) \) for \( y \in \text{supp}(g) \) and hence \( f(x - y)g(y) \) vanishes on \( \text{supp}(g) \). This establishes the claim about the support and the rest follows from the previous item. (iv) The case \( p = \infty \) follows from Hölder’s inequality and we can assume \( 1 \leq p < \infty \). Without loss of generality let \( \| \phi \|_1 = 1 \). Then

\[
\| \phi * f \|_p^p \leq \int_{\mathbb{R}^n} \left| \int_{\mathbb{R}^n} |f(y - x)| \phi(y) \, d^n y \right|^p d^n x 
\leq \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} |f(y - x)|^p |\phi(y)| \, d^n y \, d^n x = \| f \|_p^p,
\]

where we have use Jensen’s inequality with \( \varphi(x) = |x|^p, \, d\mu = |\phi| d^n y \) in the first and Fubini in the second step. (v) Immediate from integrating \( \phi(y) \inf_{x \in \mathbb{R}^n} f(x) \leq \phi(y)f(x - y) \leq \phi(y) \sup_{x \in \mathbb{R}^n} f(x) \). \( \square \)

Note that (iv) also extends to Hölder continuous functions since it is straightforward to show

\[
[\phi * f]_\gamma \leq \| \phi \|_1 [f]_\gamma. \tag{10.27}
\]

**Example.** For \( f := \chi_{[0,1]} - \chi_{[-1,0]} \) and \( g := \chi_{[-2,2]} \) we have that \( f * g \) is the difference of two triangles supported on \([1,3]\) and \([-3,-1]\) whereas \( \text{supp}(f) + \text{supp}(g) = [-3,3] \), which shows that the inclusion in (iii) is strict in general. \( \diamond \)

Next we turn to approximation of \( f \). To this end we call a family of integrable functions \( \phi_\varepsilon, \, \varepsilon \in (0,1] \), an approximate identity if it satisfies the following three requirements:

(i) \( \| \phi_\varepsilon \|_1 \leq C \) for all \( \varepsilon > 0 \).

(ii) \( \int_{\mathbb{R}^n} \phi_\varepsilon(x) d^n x = 1 \) for all \( \varepsilon > 0 \).

(iii) For every \( r > 0 \) we have \( \lim_{\varepsilon \downarrow 0} \int_{|x| \geq r} \phi_\varepsilon(x) d^n x = 0 \).

Moreover, a nonnegative function \( \phi \in C_c^\infty(\mathbb{R}^n) \) satisfying \( \| \phi \|_1 = 1 \) is called a mollifier. Note that if the support of \( \phi \) is within a ball of radius \( s \), then the support of \( \phi * f \) will be within \( \{ x \in \mathbb{R}^n : \text{dist}(x, \text{supp}(f)) \leq s \} \).

**Example.** The standard mollifier (also Friedrichs mollifier) is \( \phi(x) := c^{-1} \exp \left( \frac{1}{|x|^2 - 1} \right) \) for \( |x| < 1 \) and \( \phi(x) = 0 \) otherwise. Here the constant \( c := \int_{B(1)} \exp \left( \frac{1}{|x|^2 - 1} \right) d^n x = S_n \int_0^1 \exp \left( \frac{1}{r^2 - 1} \right) r^{n-1} dr \) is chosen such that \( \| \phi \|_1 = 1 \). To show that this function is indeed smooth it suffices to show that all left derivatives of \( f(r) = \exp \left( \frac{1}{r^2 - 1} \right) \) at \( r = 1 \) vanish, which can be done using l’Hôpital’s rule. \( \diamond \)
Example. Scaling a mollifier according to \( \phi_{\varepsilon}(x) = \varepsilon^{-n} \phi(\frac{x}{\varepsilon}) \) such that its mass is preserved (\( \| \phi_{\varepsilon} \|_1 = 1 \)) and it concentrates more and more around the origin as \( \varepsilon \downarrow 0 \) we obtain an approximate identity:

\[
\phi_{\varepsilon}
\]

In fact, (i), (ii) are obvious from \( \| \phi_{\varepsilon} \|_1 = 1 \) and the integral in (iii) will be identically zero for \( \varepsilon \geq \frac{r}{s} \), where \( s \) is chosen such that \( \text{supp} \phi \subseteq B_s(0) \).

Now we are ready to show that an approximate identity deserves its name.

Lemma 10.18. Let \( \phi_{\varepsilon} \) be an approximate identity. If \( f \in L^p(\mathbb{R}^n) \) with \( 1 \leq p < \infty \). Then

\[
\lim_{\varepsilon \downarrow 0} \phi_{\varepsilon} * f = f
\]

with the limit taken in \( L^p \). In the case \( p = \infty \) the claim holds for \( f \in C_0(\mathbb{R}^n) \).

Proof. We begin with the case where \( f \in C_c(\mathbb{R}^n) \). Fix some small \( \delta > 0 \). Since \( f \) is uniformly continuous we know \( |f(x - y) - f(x)| \to 0 \) as \( y \to 0 \) uniformly in \( x \). Since the support of \( f \) is compact, this remains true when taking the \( L^p \) norm and thus we can find some \( r \) such that

\[
\| f(\cdot - y) - f(\cdot) \|_p \leq \frac{\delta}{2C}, \quad |y| \leq r.
\]

(Here the \( C \) is the one for which \( \| \phi_{\varepsilon} \|_1 \leq C \) holds.) Now we use

\[
(\phi_{\varepsilon} * f)(x) - f(x) = \int_{\mathbb{R}^n} \phi_{\varepsilon}(y)(f(x - y) - f(x))d^ny
\]

Splitting the domain of integration according to \( \mathbb{R}^n = \{ y| |y| \leq r \} \cup \{ y| |y| > r \} \), we can estimate the \( L^p \) norms of the individual integrals using the Minkowski inequality as follows:

\[
\left\| \int_{|y| \leq r} \phi_{\varepsilon}(y)(f(x - y) - f(x))d^ny \right\|_p \leq \int_{|y| \leq r} |\phi_{\varepsilon}(y)||f(\cdot - y) - f(\cdot)||d^ny \leq \frac{\delta}{2}
\]
and
\[
\left\| \int_{|y|>r} \phi_{\varepsilon}(y)(f(x-y) - f(x))d^ny \right\|_p \leq 2\|f\|_p \int_{|y|>r} |\phi_{\varepsilon}(y)|d^ny \leq \frac{\delta}{2}
\]
provided \(\varepsilon\) is sufficiently small such that the integral in (iii) is less than \(\delta/2\).

This establishes the claim for \(f \in C_c(\mathbb{R}^n)\). Since these functions are dense in \(L^p\) for \(1 \leq p < \infty\) and in \(C_0(\mathbb{R}^n)\) for \(p = \infty\) the claim follows from Lemma 4.33 and Young’s inequality. \(\square\)

Note that in case of a mollifier with support in \(B_r(0)\) this result implies a corresponding local version since the value of \((\phi_{\varepsilon} * f)(x)\) is only affected by the values of \(f\) on \(B_{\varepsilon r}(x)\). The question when the pointwise limit exists will be addressed in Problem 10.23.

**Example.** The Fejér kernel introduced in (2.52) is an approximate identity if we set it equal to 0 outside \([-\pi,\pi]\) (see the proof of Theorem 2.19). Then taking a \(2\pi\) periodic function and setting it equal to 0 outside \([-2\pi,2\pi]\) Lemma 10.18 shows that for \(f \in L^p(-\pi,\pi)\) the mean values of the partial sums of a Fourier series \(\bar{S}_n(f)\) converge to \(f\) in \(L^p(-\pi,\pi)\) for every \(1 \leq p < \infty\). For \(p = \infty\) we recover Theorem 2.19. Note also that this shows that the map \(f \mapsto \hat{f}\) is injective on \(L^p\).

Another classical example it the Poisson kernel, see Problem 10.22. Note that the Dirichlet kernel \(D_n\) from (2.46) is no approximate identity since \(\|D_n\|_1 \to \infty\) as was shown in the example on page 127. \(\diamond\)

Now we are ready to prove

**Theorem 10.19.** If \(X \subseteq \mathbb{R}^n\) is open and \(\mu\) is a regular Borel measure, then the set \(C_c^\infty(X)\) of all smooth functions with compact support is dense in \(L^p(X,d\mu)\), \(1 \leq p < \infty\).

**Proof.** By Theorem 10.15 it suffices to show that every continuous function \(f(x)\) with compact support can be approximated by smooth ones. By setting \(f(x) = 0\) for \(x \notin X\), it is no restriction to assume \(X = \mathbb{R}^n\). Now choose a mollifier \(\phi\) and observe that \(\phi_{\varepsilon} * f\) has compact support inside \(X\) for \(\varepsilon\) sufficiently small (since the distance from \(\text{supp}(f)\) to the boundary \(\partial X\) is positive by compactness). Moreover, \(\phi_{\varepsilon} * f \to f\) uniformly by the previous lemma and hence also in \(L^p(X,d\mu)\). \(\square\)

Note that we can also use mollification to establish existence of smooth partitions of unity.
Lemma 10.20. Let $O \subseteq \mathbb{R}^n$ be an open set and $\{O_j\}$ a locally finite countable open cover. Moreover, let $\{O'_j\}$ be another open cover of relatively compact sets with $\overline{O}_j \subset O_j$. Then there is a smooth partition of unity subordinate to this cover; that is, there are functions $h_j \in C_c^\infty(O_j, [0,1])$ such that $\sum_j h_j(x) = 1$.

Proof. Since $\overline{O}_j'$ is compact we know that $\varepsilon_j := \text{dist}(\overline{O}_j', O \setminus O_j) > 0$ and we consider $\tilde{h}_j := \phi_{\varepsilon_j/2} \ast \chi_{O_j'}$ with $\phi$ the standard mollifier. By construction $\tilde{h}_j \in C_c^\infty(O_j, [0,1])$. Moreover, $h := \sum_j \tilde{h}_j$ is smooth (in the neighborhood of every point only finitely many terms are nonzero) and positive (since $x \in O'_j$ for at least one $j_0$ and hence also $\tilde{h}_{j_0}(x) > 0$). Finally, $h^{-1}\tilde{h}_j$ has all the required properties. \qed

Note that given $O \subseteq \mathbb{R}^n$ a required cover can be constructed as follows. Start with the bounded open sets $U_j := \{x \in O \mid \text{dist}(x, \mathbb{R}^n \setminus O) > \frac{1}{j} \text{ and } |x| \leq j\}$ (cf. also Lemma 0.25 (iv)) such that $U_j \subseteq U_{j+1}$ and $U_j \not\supset O$. Then $O_0 := U_3$, $O_j := U_{j+3} \setminus U_j$ and $O_0' := U_2$, $O_j' := U_{j+2} \setminus U_j$ is a required cover which is locally finite since every compact subset of $O$ is contained in some $U_{j_0}$ and hence does not meet the $O_j$ for $j \geq j_0$.

Our final result is known as the fundamental lemma of the calculus of variations.

Lemma 10.21. Suppose $\mu$ is a Borel measure on an open set $X \subseteq \mathbb{R}^n$ and $f \in L^1_{\text{loc}}(X)$. (i) If $f$ is real-valued then

$$
\int_X \varphi(x)f(x)d\mu x \geq 0, \quad \forall \varphi \in C_c^\infty(X), \varphi \geq 0, \quad (10.29)
$$

if and only if $f(x) \geq 0$ (a.e.). (ii) Moreover,

$$
\int_X \varphi(x)f(x)d\mu x = 0, \quad \forall \varphi \in C_c^\infty(X), \varphi \geq 0, \quad (10.30)
$$

if and only if $f(x) = 0$ (a.e.).

Proof. (i) Choose a compact set $K \subset X$ and some $\varepsilon_0 > 0$ such that $K_{\varepsilon_0} := K + B_{\varepsilon_0}(0) \subseteq U$. Set $\tilde{f} := f|_{K_{\varepsilon_0}}$ and let $\phi$ be the standard mollifier. Then $(\phi_{\varepsilon} \ast \tilde{f})(x) = (\phi_{\varepsilon} \ast f)(x) \geq 0$ for $x \in K$, $\varepsilon < \varepsilon_0$ and since $\phi_{\varepsilon} \ast \tilde{f} \rightarrow \tilde{f}$ in $L^1(X)$ we have $(\phi_{\varepsilon} \ast \tilde{f})(x) \rightarrow f(x) \geq 0$ for a.e. $x \in K$ for an appropriate subsequence. Since $K \subset X$ is arbitrary the first claim follows. (ii) The first part shows that $\text{Re}(f) \geq 0$ as well as $-\text{Re}(f) \geq 0$ and hence $\text{Re}(f) = 0$. Applying the same argument to $\text{Im}(f)$ establishes the claim. \qed

The following variant is also often useful
Lemma 10.22 (du Bois-Reymond). Suppose $U \subseteq \mathbb{R}^n$ is open and connected. If $f \in L^1_{\text{loc}}(U)$ with
\[
\int_U f(x) \partial_j \varphi(x) d^n x = 0, \quad \forall \varphi \in C_c^\infty(U), \ 1 \leq j \leq n, \quad (10.31)
\]
then $f$ is constant a.e. on $U$.

**Proof.** Choose a compact set $K \subset U$ and $\tilde{f}, \phi$ as in the proof of the previous lemma but additionally assume that $K$ is connected. Then by Lemma 10.17 (ii)
\[
\partial_j (\phi \ast \tilde{f})(x) = ((\partial_j \phi) \ast \tilde{f})(x) = (\partial_j f)(x) = 0, \quad x \in K, \ \varepsilon \leq \varepsilon_0.
\]
Hence $(\phi \ast \tilde{f})(x) = c_\varepsilon$ for $x \in K$ and as $\varepsilon \to 0$ there is a subsequence which converges a.e. on $K$. Clearly this limit function must also be constant: $(\phi \ast \tilde{f})(x) = c_\varepsilon \to f(x) = c$ for a.e. $x \in K$. Now write $U$ as a countable union of open balls whose closure is contained in $U$. If the corresponding constants for these balls were not all the same, we could find a partition into two union of open balls which were disjoint. This contradicts that $U$ is connected. \(\square\)

Of course the last result can be extended to higher derivatives. For the one-dimensional case this is outlined in Problem 10.26.

**Problem 10.16** (Smooth Urysohn lemma). Suppose $K$ and $C$ are disjoint closed subsets of $\mathbb{R}^n$ with $K$ compact. Then there is a smooth function $f \in C_c^\infty(\mathbb{R}^n, [0, 1])$ such that $f$ is zero on $C$ and one on $K$.

**Problem 10.17.** Let $f \in L^p(\mathbb{R}^n)$ and consider the translation operator $T_a(f)(x) = f(x - a)$, for fixed $a \in \mathbb{R}^n$. Show that $T_a(f) \to f$ in $L^p$ as $a \to 0$. (Hint: Start with $f \in C_c(\mathbb{R}^n)$.)

**Problem 10.18.** Let $f \in L^p(\mathbb{R}^n)$ and $g \in L^q(\mathbb{R}^n)$ with $\frac{1}{p} + \frac{1}{q} = 1$. Show that $f \ast g \in C_0(\mathbb{R}^n)$ with
\[
\|f \ast g\|_{\infty} \leq \|f\|_p \|g\|_q.
\]

**Problem 10.19.** Show that the convolution on $L^1(\mathbb{R}^n)$ is associative. Conclude that $L^1(\mathbb{R}^n)$ together with convolution as a product is a commutative Banach algebra (without identity). (Hint: It suffices to verify associativity for nice functions.)

**Problem 10.20.** Let $\mu$ be a finite measure on $\mathbb{R}$. Then the set of all exponentials $\{e^{itx}\}_{t \in \mathbb{R}}$ is total in $L^p(\mathbb{R}, d\mu)$ for $1 \leq p < \infty$.

**Problem 10.21.** Let $\phi$ be integrable and normalized such that $\int_{\mathbb{R}^n} \phi(x) d^n x = 1$. Show that $\phi_\varepsilon(x) = \varepsilon^{-n} \phi(\frac{x}{\varepsilon})$ is an approximate identity.
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Problem 10.22. Show that the Poisson kernel

\[ P_\varepsilon(x) := \frac{\varepsilon}{\pi x^2 + \varepsilon^2} \]

is an approximate identity on \( \mathbb{R} \).

Show that the Cauchy transform (also Borel transform)

\[ F(z) := \frac{1}{\pi} \int_{\mathbb{R}} \frac{f(\lambda)}{\lambda - z} d\lambda \]

of a real-valued function \( f \in L^p(\mathbb{R}) \) is analytic in the upper half-plane with imaginary part given by

\[ \text{Im}(F(x + iy)) = (P_y * f)(x). \]

In particular, by Young’s inequality \( \| \text{Im}(F(. + iy)) \|_p \leq \| f \|_p \) and thus \( \sup_{y>0} \| \text{Im}(F(. + iy)) \|_p = \| f \|_p \). Such analytic functions are said to be in the Hardy space \( H^p(\mathbb{C}+) \).

(Hint: To see analyticity of \( F \) use Problem 9.18 plus the estimate

\[ \left| \frac{1}{\lambda - z} \right| \leq \frac{1}{1 + |\lambda| |\text{Im}(z)|} \]

Problem 10.23. Let \( \phi \) be bounded with support in \( \overline{B_1(0)} \) and normalized such that \( \int_{\mathbb{R}^n} \phi(x) dx = 1 \). Set \( \phi_\varepsilon(x) = \varepsilon^{-n} \phi(\frac{x}{\varepsilon}) \).

For \( f \) locally integrable show

\[ |(\phi_\varepsilon * f)(x) - f(x)| \leq \frac{V_n \| \phi \|_\infty}{|B_\varepsilon(x)|} \int_{B_\varepsilon(x)} |f(y) - f(x)| d^n y. \]

Hence at every Lebesgue point (cf. Theorem 11.6) \( x \) we have

\[ \lim_{\varepsilon \downarrow 0} (\phi_\varepsilon * f)(x) = f(x). \]

If \( f \) is uniformly continuous then the above limit will be uniform. See Problem 14.9 for the case when \( \phi \) is not compactly supported.

Problem 10.24. Let \( f, g \) be integrable (or nonnegative). Show that

\[ \int_{\mathbb{R}^n} (f * g)(x) d^n x = \int_{\mathbb{R}^n} f(x) d^n x \int_{\mathbb{R}^n} g(x) d^n x. \]

Problem 10.25. Let \( \mu, \nu \) be two complex measures on \( \mathbb{R}^n \) and set \( S : \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}^n, (x, y) \mapsto x + y \). Define the convolution of \( \mu \) and \( \nu \) by

\[ \mu * \nu := S_*)(\mu \otimes \nu). \]

Show
• $\mu \ast \nu$ is a complex measure given by

$$(\mu \ast \nu)(A) = \int_{\mathbb{R}^n \times \mathbb{R}^n} \chi_A(x + y) d\mu(x) d\nu(y) = \int_{\mathbb{R}^n} \mu(A - y) d\nu(y)$$

and satisfying $|\mu \ast \nu|(\mathbb{R}^n) \leq |\mu|(\mathbb{R}^n)|\nu|(\mathbb{R}^n)$ with equality for positive measures.

• $\mu \ast \nu = \nu \ast \mu$.

• If $d\nu(x) = g(x) dx$ then $d(\mu \ast \nu)(x) = h(x) dx$ with $h(x) = \int_{\mathbb{R}^n} g(x - y) d\mu(y)$.

In particular the last item shows that this definition agrees with our definition for functions if both measures have a density.

Problem 10.26 (du Bois-Reymond lemma). Let $f$ be a locally integrable function on the interval $(a, b)$ and let $n \in \mathbb{N}_0$. If

$$\int_a^b f(x) \varphi^{(n)}(x) dx = 0, \quad \forall \varphi \in C_\infty^0(a, b),$$

then $f$ is a polynomial of degree at most $n - 1$ a.e. (Hint: Begin by showing that there exists $\{\phi_{n,j}\}_{0 \leq j \leq n} \subseteq C_\infty^0(a, b)$ such that

$$\int_a^b x^k \phi_{n,j}(x) dx = \delta_{j,k}, \quad 0 \leq j, k \leq n.$$  

The case $n = 0$ is easy and for the general case note that one can choose $\phi_{n+1,n+1} = (n + 1)^{-1} \phi_{n,n}$. Then, for given $\phi \in C_\infty^0(a, b)$, look at $\varphi(x) = \frac{1}{n!} \int_a^b (\phi(y) - \tilde{\phi}(y))(x - y)^n dy$ where $\tilde{\phi}$ is chosen such that this function is in $C_\infty^0(a, b)$.

10.5. Integral operators

Using Hölder’s inequality, we can also identify a class of bounded operators from $L^p(Y, d\nu)$ to $L^p(X, d\mu)$. We will assume all measures to be $\sigma$-finite throughout this section.

Lemma 10.23 (Schur criterion). Let $\mu, \nu$ be measures on $X, Y$, respectively, and let $\frac{1}{p} + \frac{1}{q} = 1$. Suppose that $K(x, y)$ is measurable and there are nonnegative measurable functions $K_1(x, y)$, $K_2(x, y)$ such that $|K(x, y)| \leq K_1(x, y)K_2(x, y)$ and

$$\|K_1(x, \cdot)\|_{L^q(Y, d\nu)} \leq C_1, \quad \|K_2(\cdot, y)\|_{L^p(X, d\mu)} \leq C_2$$  \hspace{1cm} (10.32)

for $\mu$-almost every $x$, respectively, for $\nu$-almost every $y$. Then the operator $K : L^p(Y, d\nu) \to L^p(X, d\mu)$, defined by

$$(Kf)(x) := \int_Y K(x, y) f(y) d\nu(y),$$  \hspace{1cm} (10.33)

for $\mu$-almost every $x$ is bounded with $\|K\| \leq C_1C_2$. 

Proof. We assume $1 < p < \infty$ for simplicity and leave the cases $p = 1, \infty$ to the reader. Choose $f \in L^p(Y, d\nu)$. By Fubini’s theorem $\int_Y |K(x, y)f(y)|d\nu(y)$ is measurable and by Hölder’s inequality we have

$$
\int_Y |K(x, y)f(y)|d\nu(y) \leq \int_Y K_1(x, y)K_2(x, y)|f(y)|d\nu(y)
$$

$$
\leq \left(\int_Y K_1(x, y)^q d\nu(y)\right)^{1/q} \left(\int_Y |K_2(x, y)f(y)|^p d\nu(y)\right)^{1/p}
$$

$$
\leq C_1 \left(\int_Y |K_2(x, y)f(y)|^p d\nu(y)\right)^{1/p}
$$

for $\mu$ a.e. $x$ (if $K_2(x, .)f(.) \notin L^p(X, d\mu)$, the inequality is trivially true). Now take this inequality to the $p$’th power and integrate with respect to $x$ using Fubini

$$
\int_X \left(\int_Y |K(x, y)f(y)|d\nu(y)\right)^p d\mu(x) \leq C_1^p \int_X \int_Y |K_2(x, y)f(y)|^p d\nu(y)d\mu(x)
$$

$$
= C_1^p \int_X \int_Y |K_2(x, y)f(y)|^p d\mu(x)d\nu(y) \leq C_1^p C_2^p \|f\|_p^p.
$$

Hence $\int_Y |K(x, y)f(y)|d\nu(y) \in L^p(X, d\mu)$ and, in particular, it is finite for $\mu$-almost every $x$. Thus $K(x, .)f(.)$ is $\nu$ integrable for $\mu$-almost every $x$ and $\int_Y K(x, y)f(y)d\nu(y)$ is measurable. \hfill \Box

Note that the assumptions are, for example, satisfied if $\|K(., .)\|_{L^1(Y, d\omega)} \leq C$ and $\|K(., .)\|_{L^{1}(X, d\mu)} \leq C$ which follows by choosing $K_1(x, y) = |K(x, y)|^{1/q}$ and $K_2(x, y) = |K(x, y)|^{1/p}$. For related results see also Problems 10.29 and 14.3.

Another case of special importance is the case of integral operators

$$(Kf)(x) := \int_X K(x, y)f(y)d\mu(y), \quad f \in L^2(X, d\mu), \quad (10.34)$$

where $K(x, y) \in L^2(X \times X, d\mu \otimes d\mu)$. Such an operator is called a Hilbert–Schmidt operator.

Lemma 10.24. Let $K$ be a Hilbert–Schmidt operator in $L^2(X, d\mu)$. Then

$$
\int_X \int_X |K(x, y)|^2 d\mu(x)d\mu(y) = \sum_{j \in J} \|Ku_j\|^2 \quad (10.35)
$$

for every orthonormal basis $\{u_j\}_{j \in J}$ in $L^2(X, d\mu)$.

Proof. Since $K(x, .) \in L^2(X, d\mu)$ for $\mu$-almost every $x$ we infer from Parseval’s relation

$$
\sum_j \left|\int_X K(x, y)u_j(y)d\mu(y)\right|^2 = \int_X |K(x, y)|^2 d\mu(y)
$$

...
for $\mu$-almost every $x$ and thus
\[
\sum_j \|Ku_j\|^2 = \sum_j \int_X \left| \int_X K(x,y)u_j(y)d\mu(y) \right|^2 d\mu(x)
\]
\[
= \int_X \sum_j \left| \int_X K(x,y)u_j(y)d\mu(y) \right|^2 d\mu(x)
\]
\[
= \int_X \int_X |K(x,y)|^2 d\mu(x)d\mu(y)
\]
as claimed. \hfill \Box

Hence in combination with Lemma 3.23 this shows that our definition for integral operators agrees with our previous definition from Section 3.6. In particular, this gives us an easy to check test for compactness of an integral operator.

**Example.** Let $[a,b]$ be some compact interval and suppose $K(x,y)$ is bounded. Then the corresponding integral operator in $L^2(a,b)$ is Hilbert–Schmidt and thus compact. This generalizes Lemma 3.4. \hfill \diamond

In combination with the spectral theorem for compact operators (in particular Corollary 3.8) we obtain the classical Hilbert–Schmidt theorem:

**Theorem 10.25 (Hilbert–Schmidt).** Let $K$ be a self-adjoint Hilbert–Schmidt operator in $L^2(X,d\mu)$. Let $\{u_j\}$ be an orthonormal set of eigenfunctions with corresponding nonzero eigenvalues $\{\kappa_j\}$ from the spectral theorem for compact operators (Theorem 3.7). Then
\[
K(x,y) = \sum_j \kappa_j u_j(x)^* u_j(y), \quad (10.36)
\]
where the sum converges in $L^2(X \times X, d\mu \otimes d\mu)$.

In this context the above theorem is known as second Hilbert–Schmidt theorem and the spectral theorem for compact operators is known as first Hilbert–Schmidt theorem.

If an integral operator is positive we can say more. But first we will discuss two equivalent definitions of positivity in this context. First of all recall that an operator $K \in \mathcal{L}(L^2(X,d\mu))$ is positive if $\langle f, Kf \rangle \geq 0$ for all $f \in L^2(X,d\mu)$. Secondly we call a continuous kernel positive semidefinite on $U \subseteq X$ if
\[
\sum_{j,k=1}^n \alpha_j^* \alpha_k K(x_j,x_k) \geq 0 \quad (10.37)
\]
for all $(\alpha_1, \ldots, \alpha_n) \in \mathbb{C}^n$ and $\{x_j\}_{j=1}^n \subseteq U$.  

Both conditions have their advantages. For example, note that for a positive semidefinite kernel the case \( n = 1 \) shows that \( K(x, x) \geq 0 \) for \( x \in U \) and the case \( n = 2 \) shows (look at the determinant) \( |K(x, y)|^2 \leq K(x, x)K(y, y) \) for \( x, y \in U \). On the other hand, note that for a positive operator all eigenvalues are nonnegative.

**Lemma 10.26.** Suppose \( X \) a locally compact metric space and \( \mu \) a Borel measure. Let \( K \in \mathcal{L}(L^2(X, d\mu)) \) an integral operator with a continuous kernel. Then, if \( K \) is positive, its kernel is positive semidefinite on \( \text{supp}(\mu) \).

If \( \mu \) is regular, the converse is also true.

**Proof.** Let \( x_0 \in \text{supp}(\mu) \) and consider \( \delta_{x_0, \varepsilon} = \mu(B_{\varepsilon}(x_0))^{-1}\chi_{B_{\varepsilon}(x_0)} \). Then for \( f_\varepsilon = \sum_{j=1}^n \alpha_j \delta_{x_0, \varepsilon} \)

\[
0 \leq \lim_{\varepsilon \downarrow 0} \langle f_\varepsilon, Kf_\varepsilon \rangle = \lim_{\varepsilon \downarrow 0} \sum_{j,k=1}^n \alpha_j^* \alpha_k \int_{B_{\varepsilon}(x_j)} \int_{B_{\varepsilon}(x_k)} K(x, y) \frac{d\mu(x)}{\mu(B_{\varepsilon}(x_j))} \frac{d\mu(y)}{\mu(B_{\varepsilon}(x_k))} = \sum_{j,k=1}^n \alpha_j^* \alpha_k K(x_j, x_k).
\]

Conversely, let \( f \in C_c(X) \) and let \( S := \text{supp}(f) \cap \text{supp}(\mu) \). Then the function \( f(x)^* K(x, y) f(y) \in C_c(X \times X) \) is uniformly continuous and for every \( \varepsilon > 0 \) we can partition the compact set \( S \) into a finite number of sets \( U_j \) which are contained in a ball \( B_{\delta}(x_j) \) such that

\[
|f(x)^* K(x, y) f(y) - \sum_{j,k} \chi_{U_j}(x) f(x_j)^* K(x_j, x_k) f(x_k) \chi_{U_k}(y)| \leq \varepsilon, \quad x, y \in S.
\]

Hence

\[
\left| \langle f, Kf \rangle - \sum_{j,k} \mu(U_j) f(x_j)^* K(x_j, x_k) f(x_k) \mu(U_k) \right| \leq \varepsilon \mu(S)^2
\]

and since \( \varepsilon > 0 \) is arbitrary we have \( \langle f, Kf \rangle \geq 0 \). Since \( C_c(X) \) is dense in \( L^2(X, d\mu) \) if \( \mu \) is regular by Theorem 10.15 we get this for all \( f \in L^2(X, d\mu) \) by taking limits.

Now we are ready for the following classical result:

**Theorem 10.27** (Mercer). Let \( K \) be a positive integral operator with a continuous kernel on \( L^2(X, d\mu) \) with \( X \) a locally compact metric space. Let \( \mu \) a Borel measure such that the diagonal \( K(x, x) \) is integrable. Then \( K \) is trace class, all eigenfunctions \( u_j \) corresponding to positive eigenvalues \( \kappa_j \) are continuous, and (10.36) converges uniformly on the support of \( \mu \). Moreover,

\[
\text{tr}(K) = \sum_j \kappa_j = \int_X K(x, x) d\mu(x).
\]
\textbf{Proof.} Define }k(x)^2 := K(x, x) \text{ such that } |K(x, y)| \leq k(x)k(y) \text{ for } x, y \in \text{supp}(\mu). \text{ Since by assumption } k \in L^2(X, d\mu) \text{ we see that } K \text{ is Hilbert–Schmidt and we have the representation (10.36) with } \kappa_j > 0. \text{ Moreover, dominated convergence shows that } u_j(x) = \kappa_j^{-1} \int_X K(x, y)u_j(y)d\mu(y) \text{ is continuous.}

Now note that the operator corresponding to the kernel
\[ K_n(x, y) = K(x, y) - \sum_{j \leq n} \kappa_j u_j(x)^*u_j(y) = \sum_{j > n} \kappa_j u_j(x)^*u_j(y) \]
is also positive (its nonzero eigenvalues are } \kappa_j, j > n)\text{. Hence } K_n(x, x) \geq 0 \text{ implying }
\[ \sum_{j \leq n} \kappa_j |u_j(x)|^2 \leq k(x)^2 \]
for } x \in \text{supp}(\mu) \text{ and hence (10.36) converges uniformly on the support of } \mu. \text{ Finally, integrating (10.36) for } x = y \text{ using } \|u_j\| = 1 \text{ shows the last claim.} \quad \square

\textbf{Example.} Let } k \text{ be a periodic function which is square integrable over } [-\pi, \pi]. \text{ Then the integral operator }
\[ (Kf)(x) = \frac{1}{2\pi} \int_{-\pi}^{\pi} k(y - x)f(y)dy \]
has the eigenfunctions } u_j(x) = (2\pi)^{-1/2}e^{-ijx} \text{ with corresponding eigenvalues } \hat{k}_j, j \in \mathbb{Z}, \text{ where } \hat{k}_j \text{ are the Fourier coefficients of } k. \text{ Since } \{u_j\}_{j \in \mathbb{Z}} \text{ is an ONB we have found all eigenvalues. Moreover, in this case (10.36) is just the Fourier series }
k(y - x) = \sum_{j \in \mathbb{Z}} \hat{k}_j e^{ij(y-x)}.
Choosing a continuous function for which the Fourier series does not converge absolutely shows that the positivity assumption in Mercer’s theorem is crucial. \hfill \diamond

Of course given a kernel this raises the question if it is positive semi-definite. This can often be done by reverse engineering based on constructions which produce new positive semi-definite kernels out of old ones.

\textbf{Lemma 10.28.} Let } K_j(x, y) \text{ be positive semi-definite kernels on } X. \text{ Then the following kernels are also positive semi-definite.}

(i) } \alpha_1 K_1 + \alpha_2 K_2 \text{ provided } \alpha_j \geq 0.

(ii) } \lim_j K_j(x, y) \text{ provided the limit exists pointwise.}

(iii) } K_1(\phi(x), \phi(y)) \text{ for } \phi : X \rightarrow X.

(iv) } f(x)^* K_1(x, y)f(y) \text{ for } f : X \rightarrow \mathbb{C}.
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(v) \(K_1(x, y)K_2(x, y)\).

**Proof.** Only the last item is not straightforward. However, it boils down to the Schur product theorem from linear algebra given below. \(\square\)

**Theorem 10.29** (Schur). Let \(A\) and \(B\) be two \(n \times n\) matrices and let \(A \circ B\) be their Hadamard product defined by multiplying the entries pointwise: \((A \circ B)_{jk} := A_{jk}B_{jk}\). If \(A\) and \(B\) are positive (semi-) definite, then so is \(A \circ B\).

**Proof.** By the spectral theorem we can write \(A = \sum_{j=1}^{n} \alpha_j \langle u_j, \cdot \rangle u_j\), where \(\alpha_j > 0\) (\(\alpha_j \geq 0\)) are the eigenvalues and \(u_j\) are a corresponding ONB of eigenvectors. Similarly \(B = \sum_{j=1}^{n} \beta_j \langle v_j, \cdot \rangle v_j\). Then \(A \circ B = \sum_{j,k=1}^{n} \alpha_j \beta_k \langle u_j \circ v_k, \cdot \rangle u_j \circ v_k\), which proves the claim. \(\square\)

**Example.** Let \(X = \mathbb{R}^n\). The most basic kernel is \(K(x, y) = x^* \cdot y\) which is positive semi-definite since

\[
\sum_{j,k} \alpha_j^* \alpha_k K(x_j, x_k) = \left| \sum_j \alpha_j x_j \right|^2.
\]

Slightly more general is \(K(x, y) = \langle x, Ay \rangle\), where \(A\) is a positive semi-definite matrix. Indeed writing \(A = B^2\) this follows from the previous observation using item (iii) with \(\phi = B\). Another famous kernel is the Gaussian kernel

\[
K(x, y) = e^{-|x-y|^2/\sigma}, \quad \sigma > 0.
\]

To see that it is positive semi-definite start by observing that \(K_1(x, y) = \exp(2x^* \cdot y/\sigma)\) is by items (i) and (ii) since the Taylor series of the exponential function has positive coefficients. Finally observe that our kernel is of the form (vi) with \(f(x) = \exp(-|x|^2/\sigma)\). \(\diamond\)

**Example.** A reproducing kernel Hilbert space is a Hilbert space \(\mathcal{H}\) of complex-valued functions \(X \rightarrow \mathbb{C}\) such that point evaluations are continuous linear functionals. In this case the Riesz lemma implies that for every \(x \in X\) there is a corresponding function \(K_x \in \mathcal{H}\) such that

\[
f(x) = \langle K_x, f \rangle.
\]

Applying this to \(f = K_y\) we get \(K_y(x) = \langle K_x, K_y \rangle\) which suggests the more symmetric notation

\[
K(x, y) := \langle K_x, K_y \rangle.
\]

The kernel \(K\) is called the **reproducing kernel** for \(\mathcal{H}\). A short calculation

\[
\sum_{j,k} \alpha_j^* \alpha_k K(x_j, x_k) = \sum_{j,k} \alpha_j^* \alpha_k \langle K_{x_j}, K_{x_k} \rangle = \left\| \sum_j \alpha_j K_{x_j} \right\|^2
\]
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verifies that it is a positive semi-definite kernel. An explicit example is given by the quadratic form domains of positive Sturm–Liouville operators; Problem 3.9.

Problem 10.27. Suppose $K$ is a Hilbert–Schmidt operator in $L^2(X,d\mu)$ with kernel $K(x,y)$. Show that the adjoint operator is given by

$$(K^* f)(x) = \int_X K(y,x)^* f(y) d\mu(y), \quad f \in L^2(X,d\mu).$$

Problem 10.28. Obtain Young’s inequality from Schur’s criterion.

Problem 10.29 (Schur test). Let $K(x,y)$ be given and suppose there are positive measurable functions $a(x)$ and $b(y)$ such that

$$\|K(.,b(\cdot))\|_{L^1(Y,d\nu)} \leq C_1 a(x), \quad \|a(\cdot)K(.,y)\|_{L^1(X,d\mu)} \leq C_2 b(y).$$

Then the operator $K : L^2(Y,d\nu) \to L^2(X,d\mu)$, defined by

$$(Kf)(x) := \int_Y K(x,y)f(y)d\nu(y),$$

for $\mu$-almost every $x$ is bounded with $\|K\| \leq \sqrt{C_1 C_2}$. (Hint: Estimate $|(Kf)(x)|^2 = \|\int_Y K(x,y)b(y)f(y)b(y)^{-1}d\nu(y)|^2$ using Cauchy–Schwarz and integrate the result with respect to $x$.)

Problem 10.30. Let $K$ be an self-adjoint integral operator with continuous kernel satisfying the estimate $|K(x,y)| \leq k(x)k(y)$ with $k \in L^2(X,d\mu) \cap L^{\infty}(X,d\mu)$ (e.g. $X$ is compact). Show that the conclusion from Mercer’s theorem still hold if $K$ has only a finite number of negative eigenvalues.

Problem 10.31. Show that the Fourier transform of a finite (positive) measure

$$\hat{\mu}(p) := \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ipx} d\mu(x)$$

gives rise to a positive semi-definite kernel $K(x,y) = \hat{\mu}(x-y)$.

10.6. Sobolev spaces

Let $U \subseteq \mathbb{R}^n$ be nonempty and open. Our aim is to extended the Lebesgue spaces to include derivatives. To this end we call an element $\alpha \in \mathbb{N}_0^n$ a multi-index and $|\alpha|$ its order. For $f \in C^k(U)$ and $\alpha \in \mathbb{N}_0^n$ with $|\alpha| \leq k$ we set

$$\partial_\alpha f = \frac{\partial^{|\alpha|} f}{\partial x_1^{\alpha_1} \cdots \partial x_n^{\alpha_n}}, \quad x^\alpha = x_1^{\alpha_1} \cdots x_n^{\alpha_n}, \quad |\alpha| = \alpha_1 + \cdots + \alpha_n. \quad (10.39)$$
10.6. Sobolev spaces

Then for locally integrable function \( f \in L^1_{\text{loc}}(U) \) a function \( h \in L^1_{\text{loc}}(U) \) satisfying
\[
\int_U \varphi(x) h(x) d^n x = (-1)^{|\alpha|} \int_U (\partial_\alpha \varphi)(x) f(x) d^n x, \quad \forall \varphi \in C^\infty_c(U),
\]
(10.40)
is called the \textit{weak derivative} or the derivative in the sense of distributions of \( f \). Note that by Lemma 10.21 such a function is unique if it exists.

Moreover, if \( f \in C^k(U) \) then integration by parts shows that the weak derivative coincides with the usual derivative.

\textbf{Example.} Consider \( U := \mathbb{R} \). \( f(x) := |x| \), then \( \partial f(x) = \text{sign}(x) \) as a weak derivative. If we try to take another derivative we are lead to
\[
\int_\mathbb{R} \varphi(x) h(x) dx = - \int_\mathbb{R} \varphi'(x) \text{sign}(x) dx = 2 \varphi(0)
\]
and it is easy to see that no locally integrable function can satisfy this requirement.

In fact, in one dimension the class of weakly differentiable functions can be identified with the class of absolutely continuous functions (which is discussed in detail in Section 11.8 — see also Problem 13.19). Note that in higher dimensions weakly differentiable might not be continuous — Problem 10.33.

Now we can define the \textbf{Sobolev space} \( W^{k,p}(U) \) as the set of all functions in \( L^p(U) \) which have weak derivatives up to order \( k \) in \( L^p(U) \). Clearly \( W^{k,p}(U) \) is a linear space since \( f, g \in W^{k,p}(U) \) implies \( af + bg \in W^{k,p}(U) \) for \( a, b \in \mathbb{C} \) and \( \partial_\alpha (af + bg) = a \partial_\alpha f + b \partial_\alpha g \) for all \( |\alpha| \leq k \). Moreover, for \( f \in W^{k,p}(U) \) we define its norm
\[
\|f\|_{k,p} := \begin{cases} \left( \sum_{|\alpha| \leq k} \|\partial_\alpha f\|_p^p \right)^{1/p}, & 1 \leq p < \infty, \\ \sum_{|\alpha| \leq k} \|\partial_\alpha f\|_\infty, & p = \infty. \end{cases}
\]
(10.41)

It is easy to check that with this definition \( W^{k,p} \) becomes a normed linear space. Of course for \( p = 2 \) we have a corresponding scalar product
\[
\langle f, g \rangle_{W^{k,2}} = \sum_{|\alpha| \leq k} \langle \partial_\alpha f, \partial_\alpha g \rangle_{L^2}.
\]
(10.42)
and one reserves the special notation \( H^k(U) := W^{k,2}(U) \) for this case. Similarly we define local versions of these spaces \( W^{k,p}_{\text{loc}}(U) \) as the set of all functions in \( L^p_{\text{loc}}(U) \) which have weak derivatives up to order \( k \) in \( L^p_{\text{loc}}(U) \).

\textbf{Theorem 10.30.} For each \( k \in \mathbb{N}_0, 1 \leq p \leq \infty \) the Sobolev space \( W^{k,p}(U) \) is a Banach space. It is separable for \( 1 \leq p < \infty \) and reflexive for \( 1 < p < \infty \).
Proof. Let \( f_m \) be a Cauchy sequence in \( W^{k,p} \). Then \( \partial_\alpha f_m \) is a Cauchy sequence in \( L^p \) for every \( |\alpha| \leq k \). Consequently \( \partial_\alpha f_m \to f_\alpha \) in \( L^p \). Moreover, letting \( m \to \infty \) in
\[
\int_U \varphi f_\alpha d^n x = \lim_{m \to \infty} \int_U \varphi(\partial_\alpha f_m) d^n x = \lim_{m \to \infty} (-1)^{|\alpha|} \int_U (\partial_\alpha \varphi) f_m d^n x
\]
shows \( f_0 \in W^{k,p} \) with \( \partial_\alpha f_0 = f_\alpha \) for \( |\alpha| \leq k \). By construction \( f_m \to f_0 \) in \( W^{k,p} \) which implies that \( W^{k,p} \) is complete.

Concerning the last claim note that \( W^{k,p}(U) \) can regarded a subspace of \( X_{p,|\alpha|\leq k} L^p(U) \) which has the claimed properties. \( \square \)

For \( L^p \) we know that smooth functions with compact support are dense. This is no longer true in general for \( W^{k,p} \) since convergence of derivatives enforces that the vanishing of the function on the boundary is preserved in the limit. However, making this precise requires some additional effort. So for now we will just give the closure of \( C^\infty_c(U) \) in \( W^{k,p}(U) \) a special name \( W^{k,p}_0(U) \). It is easy to see that \( C^\infty_c(U) \subseteq W^{k,p}_0(U) \) for every \( 1 \leq p \leq \infty \) (mollify and use \( \partial_\alpha (\phi_x * f) = \phi_x * (\partial_\alpha f) \)).

Next we collect some simple properties.

Lemma 10.31. Let \( U \subseteq \mathbb{R}^n \) be open and \( 1 \leq p \leq \infty \).

(i) The operator \( \partial_\alpha : W^{k,p}(U) \to W^{k-|\alpha|,p}(U) \) is a bounded linear map and \( \partial_\beta \partial_\alpha f = \partial_\alpha \partial_\beta f = \partial_{\alpha+\beta} f \) for \( f \in W^{k,p} \) and all multi-indices \( \alpha, \beta \) with \( |\alpha| + |\beta| \leq k \).

(ii) We have
\[
\int_U g(\partial_\alpha f) d^n x = (-1)^{|\alpha|} \int_U (\partial_\alpha g) f d^n x, \quad g \in W^{k,q}_0(U), \quad f \in W^{k,p}(U),
\]
for all \( |\alpha| \leq k \), \( \frac{1}{p} + \frac{1}{q} = 1 \).

(iii) Suppose \( f \in W^{k,p}(U) \) and \( g \in W^{k,q}(U) \). Then \( f \cdot g \in W^{k,r}(U) \), where \( \frac{1}{r} = \frac{1}{p} + \frac{1}{q} \), and we have the product rule
\[
\partial_j (f \cdot g) = (\partial_j f) g + f (\partial_j g), \quad 1 \leq j \leq n.
\]

(iv) Suppose \( g \in C^1_c(\mathbb{R}) \) and \( f \in W^{1,1}_{loc}(U) \) is real-valued. Then \( g \circ f \in W^{1,1}_{loc}(U) \) and we have the chain rule \( \partial_j (g \circ f) = g'(f) \partial_j f \). If in addition \( g(0) = 0 \) and \( f \in W^{1,p}(U) \) then \( g \circ f \in W^{1,p}(U) \).

Proof. (i) Problem 10.35. (ii) Take limits in (10.40) using Hölder’s inequality. (iii) First of all note that if \( \phi, \varphi \in C^\infty_c(U) \), then \( \phi \varphi \in C^\infty_c(U) \) and hence using the ordinary product rule for smooth functions and rearranging
Proof. Choose $O_j$ as in the discussion after Lemma 10.20 and let $h_j$ be the corresponding smooth partition of unity. Let $f \in W^{k,p}(U)$ be given and fix $\delta > 0$. Choose $\varepsilon_i > 0$ sufficiently small such that $f_j := \phi_{\varepsilon_j} \ast (h_j f)$ (with $\phi$ the standard mollifier) satisfies

$$
\|f - h_j f\|_{W^{k,p}} \leq \frac{\delta}{2^j + 1} \quad \text{and} \quad \text{supp}(f_j) \subseteq O_j.
$$

Theorem 10.32 (Meyers–Serrin). Let $U \subseteq \mathbb{R}^n$ be open and $1 \leq p < \infty$. Then $C^\infty(U) \cap W^{k,p}(U)$ is dense in $W^{k,p}(U)$. 

Proof. Choose $O_j$ as in the discussion after Lemma 10.20 and let $h_j$ be the corresponding smooth partition of unity. Let $f \in W^{k,p}(U)$ be given and fix $\delta > 0$. Choose $\varepsilon_i > 0$ sufficiently small such that $f_j := \phi_{\varepsilon_j} \ast (h_j f)$ (with $\phi$ the standard mollifier) satisfies

$$
\|f - h_j f\|_{W^{k,p}} \leq \frac{\delta}{2^j + 1} \quad \text{and} \quad \text{supp}(f_j) \subseteq O_j.
$$
Then \( f_\delta = \sum_j f_j \in C^\infty(U) \) since our cover is locally finite. Moreover, for every relatively compact set \( V \subseteq U \) we have
\[
\|f_\delta - f\|_{W^{k,p}(V)} = \|\sum_j (f_j - h_j f)\|_{W^{k,p}(V)} \leq \sum_j \|f_j - h_j f\|_{W^{k,p}(V)} \leq \delta
\]
and letting \( V \nearrow U \) we get \( f_\delta \in W^{k,p}(U) \) as well as \( \|f_\delta - f\|_{W^{k,p}(U)} \leq \delta \). \( \square \)

**Problem 10.32.** Consider \( f(x) = \sqrt{x}, \ U = (0,1) \). Compute the weak derivative. For which \( p \) is \( f \in W^{1,p}(U) \)?

**Problem 10.33.** Consider \( U := B_1(0) \subset \mathbb{R}^n \) and \( f(x) := |x|^{\alpha} \). Since \( f \) is smooth away from the origin we expect
\[
\partial_j f(x) = -\frac{\alpha x_j}{|x|^{\alpha+2}}
\]
which is integrable if \( \alpha < n-1 \). Show that this is indeed the weak derivative of \( f \). (Hint: Use integration by parts on a domain which excludes \( B_\varepsilon(0) \) and let \( \varepsilon \to 0 \).)

**Problem 10.34.** Suppose \( V \subseteq U \) is nonempty and open. Then \( f \in W^{k,p}(U) \) implies \( f \in W^{k,p}(V) \).

**Problem 10.35.** Show Lemma 10.31 (i).

**Problem 10.36.** Suppose \( f \in W^{k,p}(U) \) and \( h \in C^k_b(U) \). Then \( h \cdot f \in W^{k,p}(U) \) and we have Leibniz’ rule
\[
\partial_\alpha(h \cdot f) = \sum_{\beta \leq \alpha} \binom{\alpha}{\beta} (\partial_\beta h)(\partial_{\alpha-\beta} f), \tag{10.45}
\]
where \( \binom{\alpha}{\beta} = \frac{\alpha!}{\beta!(\alpha-\beta)!} \), \( \alpha! = \prod_{j=1}^m (\alpha_j!) \), and \( \beta \leq \alpha \) means \( \beta_j \leq \alpha_j \) for \( 1 \leq j \leq m \).

**Problem 10.37.** Suppose \( f \in W^{1,p}(U) \) satisfies \( \partial_j u = 0 \) for \( 1 \leq j \leq n \). Show that \( f \) is constant if \( U \) is connected.

**Problem 10.38.** Suppose \( f \in W^{1,p}(U) \) is real-valued. Show that \( f_\pm := \max(0, \pm f) \in W^{1,p}(U) \) as well as \( |f| \in W^{1,p}(U) \) with
\[
\partial_j f_\pm(x) = \begin{cases} \pm \partial_j f(x), & \pm f(x) > 0, \\ 0, & \text{else}, \end{cases} \quad \partial_j |f|(x) = \begin{cases} \partial_j f(x), & \pm f(x) > 0, \\ -\partial_j f(x), & \pm f(x) < 0, \\ 0, & \text{else}. \end{cases}
\]
(Hint: \( f_+ = \lim_{\varepsilon \to 0} g_\varepsilon(f) \) with \( g_\varepsilon(x) = \sqrt{x^2 + \varepsilon^2} - \varepsilon \) if \( x > 0 \) and \( g_\varepsilon(x) = 0 \) if \( x \leq 0 \).)
More measure theory

11.1. Decomposition of measures

Let $\mu$, $\nu$ be two measures on a measurable space $(X, \Sigma)$. They are called mutually singular (in symbols $\mu \perp \nu$) if they are supported on disjoint sets. That is, there is a measurable set $N$ such that $\mu(N) = 0$ and $\nu(X \setminus N) = 0$.

**Example.** Let $\lambda$ be the Lebesgue measure and $\Theta$ the Dirac measure (centered at 0). Then $\lambda \perp \Theta$: Just take $N = \{0\}$; then $\lambda(\{0\}) = 0$ and $\Theta(\mathbb{R} \setminus \{0\}) = 0$. \hfill \ding{51}

On the other hand, $\nu$ is called absolutely continuous with respect to $\mu$ (in symbols $\nu \ll \mu$) if $\mu(A) = 0$ implies $\nu(A) = 0$.

**Example.** The prototypical example is the measure $d\nu := f d\mu$ (compare Lemma 9.3). Indeed by Lemma 9.6 $\mu(A) = 0$ implies

$$\nu(A) = \int_A f d\mu = 0$$

(11.1)

and shows that $\nu$ is absolutely continuous with respect to $\mu$. In fact, we will show below that every absolutely continuous measure is of this form. \hfill \ding{51}

The two main results will follow as simple consequence of the following result:

**Theorem 11.1.** Let $\mu$, $\nu$ be $\sigma$-finite measures. Then there exists a nonnegative function $f$ and a set $N$ of $\mu$ measure zero, such that

$$\nu(A) = \nu(A \cap N) + \int_A f d\mu.$$ 

(11.2)
Proof. We first assume \( \mu, \nu \) to be finite measures. Let \( \alpha := \mu + \nu \) and consider the Hilbert space \( L^2(X, d\alpha) \). Then
\[
\ell(h) := \int_X h \, d\nu
\]
is a bounded linear functional on \( L^2(X, d\alpha) \) by Cauchy–Schwarz:
\[
|\ell(h)|^2 = \left| \int_X 1 \cdot h \, d\nu \right|^2 \leq \left( \int |1|^2 \, d\nu \right) \left( \int |h|^2 \, d\nu \right) \\
\leq \nu(X) \left( \int |h|^2 \, d\alpha \right) = \nu(X) \|h\|^2.
\]
Hence by the Riesz lemma (Theorem 2.10) there exists a \( g \in L^2(X, d\alpha) \) such that
\[
\ell(h) = \int_X h g \, d\alpha.
\]
By construction
\[
\nu(A) = \int \chi_A \, d\nu = \int \chi_A g \, d\alpha = \int_A g \, d\alpha.
\]  
(11.3)
In particular, \( g \) must be positive a.e. (take \( A \) the set where \( g \) is negative).
Moreover,
\[
\mu(A) = \alpha(A) - \nu(A) = \int_A (1 - g) \, d\alpha
\]
which shows that \( g \leq 1 \) a.e. Now chose \( N = \{x|g(x) = 1\} \) such that \( \mu(N) = 0 \) and set
\[
f = \frac{g}{1-g} \chi_N, \quad N' = X \setminus N.
\]
Then, since (11.3) implies \( d\nu = g \, d\alpha \), respectively, \( d\mu = (1-g) \, d\alpha \), we have
\[
\int_A f \, d\mu = \int \chi_A \frac{g}{1-g} \chi_N \, d\mu = \int \chi_A \cap N' g \, d\alpha = \nu(A \cap N')
\]
as desired.
To see the \( \sigma \)-finite case, observe that \( Y_n \not
rightarrow X, \mu(Y_n) < \infty \) and \( Z_n \not
rightarrow X, \nu(Z_n) < \infty \) implies \( X_n := Y_n \cap Z_n \not
rightarrow X \) and \( \alpha(X_n) < \infty \). Now we set \( \tilde{X}_n := X_n \setminus X_{n-1} \) (where \( X_0 = \emptyset \)) and consider \( \mu_n(A) := \mu(A \cap \tilde{X}_n) \) and \( \nu_n(A) := \nu(A \cap \tilde{X}_n) \). Then there exist corresponding sets \( N_n \) and functions \( f_n \) such that
\[
\nu_n(A) = \nu_n(A \cap N_n) + \int_A f_n \, d\mu_n = \nu(A \cap N_n) + \int_A f_n \, d\mu,
\]
where for the last equality we have assumed \( N_n \subseteq \tilde{X}_n \) and \( f_n(x) = 0 \) for \( x \in \tilde{X}_n \) without loss of generality. Now set \( N := \bigcup_n N_n \) as well as
11.1. Decomposition of measures

$$f := \sum_n f_n$$, then $$\mu(N) = 0$$ and
$$\nu(A) = \sum_n \nu_n(A) = \sum_n \nu(A \cap N_n) + \sum_n \int_A f_n d\mu = \nu(A \cap N) + \int_A f d\mu,$$
which finishes the proof. \square

Note that another set $$\tilde{N}$$ will give the same decomposition as long as $$\mu(\tilde{N}) = 0$$ and $$\nu(\tilde{N}' \cap N) = 0$$ since in this case $$\nu(A) = \nu(A \cap \tilde{N}) + \nu(A \cap \tilde{N}') = \nu(A \cap N) + \nu(A \cap \tilde{N} \cap N) + \int_{A \cap \tilde{N}'} f d\mu = \nu(A \cap N) + \int_A f d\mu.$$ Hence we can increase $$N$$ by sets of $$\mu$$ measure zero and decrease $$\tilde{N}$$ by sets of $$\nu$$ measure zero.

Now the anticipated results follow with no effort:

Theorem 11.2 (Radon–Nikodym). Let $$\mu$$, $$\nu$$ be two $$\sigma$$-finite measures on a measurable space $$(X, \Sigma)$$. Then $$\nu$$ is absolutely continuous with respect to $$\mu$$ if and only if there is a nonnegative measurable function $$f$$ such that
$$\nu(A) = \int_A f \, d\mu$$
for every $$A \in \Sigma$$. The function $$f$$ is determined uniquely a.e. with respect to $$\mu$$ and is called the Radon–Nikodym derivative $$d\nu/d\mu$$ of $$\nu$$ with respect to $$\mu$$.

Proof. Just observe that in this case $$\nu(A \cap N) = 0$$ for every $$A$$. Uniqueness will be shown in the next theorem. \square

Example. Take $$X := \mathbb{R}$$. Let $$\mu$$ be the counting measure and $$\nu$$ Lebesgue measure. Then $$\nu \ll \mu$$ but there is no $$f$$ with $$d\nu = f \, d\mu$$. If there were such an $$f$$, there must be a point $$x_0 \in \mathbb{R}$$ with $$f(x_0) > 0$$ and we have $$0 = \nu(\{x_0\}) = \int_{\{x_0\}} f \, d\mu = f(x_0) > 0$$, a contradiction. Hence the Radon–Nikodym theorem can fail if $$\mu$$ is not $$\sigma$$-finite. \square

Theorem 11.3 (Lebesgue decomposition). Let $$\mu$$, $$\nu$$ be two $$\sigma$$-finite measures on a measurable space $$(X, \Sigma)$$. Then $$\nu$$ can be uniquely decomposed as $$\nu = \nu_{\text{sing}} + \nu_{\text{ac}}$$, where $$\mu$$ and $$\nu_{\text{sing}}$$ are mutually singular and $$\nu_{\text{ac}}$$ is absolutely continuous with respect to $$\mu$$.

Proof. Taking $$\nu_{\text{sing}}(A) := \nu(A \cap N)$$ and $$d\nu_{\text{ac}} := f \, d\mu$$ from the previous theorem, there is at least one such decomposition. To show uniqueness assume there is another one, $$\nu = \tilde{\nu}_{\text{ac}} + \tilde{\nu}_{\text{sing}}$$, and let $$\tilde{N}$$ be such that $$\mu(\tilde{N}) = 0$$ and $$\tilde{\nu}_{\text{sing}}(\tilde{N}') = 0$$. Then $$\nu_{\text{sing}}(A) - \tilde{\nu}_{\text{sing}}(A) = \int_A (\tilde{f} - f) \, d\mu$$. In particular, $$\int_{A \cap N \cap \tilde{N}} (\tilde{f} - f) \, d\mu = 0$$ and hence, since $$A$$ is arbitrary, $$\tilde{f} = f$$ a.e. away from $$N \cup \tilde{N}$$. Since $$\mu(N \cup \tilde{N}) = 0$$, we have $$\tilde{f} = f$$ a.e. and hence $$\tilde{\nu}_{\text{ac}} = \nu_{\text{ac}}$$ as well as $$\tilde{\nu}_{\text{sing}} = \nu - \tilde{\nu}_{\text{ac}} = \nu - \nu_{\text{ac}} = \nu_{\text{sing}}$$. \square
Problem 11.1. Let $\mu$ be a Borel measure on $\mathcal{B}$ and suppose its distribution function $\mu(x)$ is continuously differentiable. Show that the Radon–Nikodym derivative equals the ordinary derivative $\mu'(x)$.

Problem 11.2. Suppose $\nu$ is an inner regular measures. Show that $\nu \ll \mu$ if and only if $\mu(K) = 0$ implies $\nu(K) = 0$ for every compact set.

Problem 11.3. Suppose $\nu(A) \leq C \mu(A)$ for all $A \in \Sigma$. Then $d\nu = f \, d\mu$ with $0 \leq f \leq C$ a.e.

Problem 11.4. Let $d\nu = f \, d\mu$. Suppose $f > 0$ a.e. with respect to $\mu$. Then $\mu \ll \nu$ and $d\mu = f^{-1} \, d\nu$.

Problem 11.5 (Chain rule). Show that $\nu \ll \mu$ is a transitive relation. In particular, if $\omega \ll \nu \ll \mu$, show that
\[
\frac{d\omega}{d\mu} = \frac{d\omega}{d\nu} \frac{d\nu}{d\mu}.
\]

Problem 11.6. Suppose $\nu \ll \mu$. Show that for every measure $\omega$ we have
\[
\frac{d\omega}{d\mu} \, d\mu = \frac{d\omega}{d\nu} \, d\nu + d\zeta,
\]
where $\zeta$ is a positive measure (depending on $\omega$) which is singular with respect to $\nu$. Show that $\zeta = 0$ if and only if $\mu \ll \nu$.

11.2. Derivatives of measures

If $\mu$ is a Borel measure on $\mathcal{B}$ and its distribution function $\mu(x)$ is continuously differentiable, then the Radon–Nikodym derivative is just the ordinary derivative $\mu'(x)$ (Problem 11.1). Our aim in this section is to generalize this result to arbitrary Borel measures on $\mathcal{B}^n$.

Let $\mu$ be a Borel measure on $\mathbb{R}^n$. We call
\[
(D\mu)(x) := \lim_{\varepsilon \downarrow 0} \frac{\mu(B_{\varepsilon}(x))}{|B_{\varepsilon}(x)|}
\]
the derivative of $\mu$ at $x \in \mathbb{R}^n$ provided the above limit exists. (Here $B_r(x) \subset \mathbb{R}^n$ is a ball of radius $r$ centered at $x \in \mathbb{R}^n$ and $|A|$ denotes the Lebesgue measure of $A \in \mathcal{B}^n$.)

Example. Consider a Borel measure on $\mathcal{B}$ and suppose its distribution $\mu(x)$ (as defined in (8.19)) is differentiable at $x$. Then
\[
(D\mu)(x) = \lim_{\varepsilon \downarrow 0} \frac{\mu((x+\varepsilon, x-\varepsilon))}{2\varepsilon} = \lim_{\varepsilon \downarrow 0} \frac{\mu(x+\varepsilon) - \mu(x-\varepsilon)}{2\varepsilon} = \mu'(x).
\]
To compute the derivative of $\mu$, we introduce the upper and lower derivative,

$$
(D\mu)(x) := \limsup_{\varepsilon \downarrow 0} \frac{\mu(B_\varepsilon(x))}{|B_\varepsilon(x)|} \quad \text{and} \quad (D\mu)(x) := \liminf_{\varepsilon \downarrow 0} \frac{\mu(B_\varepsilon(x))}{|B_\varepsilon(x)|}.
$$

(11.6)

Clearly $\mu$ is differentiable at $x$ if $(D\mu)(x) = (D\mu)(x) < \infty$. Next note that they are measurable: In fact, this follows from

$$
(D\mu)(x) = \lim_{n \to \infty} \sup_{0 < \varepsilon < 1/n} \frac{\mu(B_\varepsilon(x))}{|B_\varepsilon(x)|}
$$

(11.7)

since the supremum on the right-hand side is lower semicontinuous with respect to $x$ (cf. Problem 8.18) as $x \mapsto \mu(B_\varepsilon(x))$ is lower semicontinuous (Problem 11.7). Similarly for $(D\mu)(x)$.

Next, the following geometric fact of $\mathbb{R}^n$ will be needed.

**Lemma 11.4** (Wiener covering lemma). Given open balls $B_1 := B_{r_1}(x_1)$, $\ldots$, $B_m := B_{r_m}(x_m)$ in $\mathbb{R}^n$, there is a subset of disjoint balls $B_{j_1}, \ldots, B_{j_k}$ such that

$$
\bigcup_{j=1}^m B_j \subseteq \bigcup_{\ell=1}^k B_{3r_{j_\ell}}(x_{j_\ell})
$$

(11.8)

**Proof.** Assume that the balls $B_j$ are ordered by decreasing radius. Start with $B_{j_1} = B_1$ and remove all balls from our list which intersect $B_{j_1}$. Observe that the removed balls are all contained in $B_{3r_1}(x_1)$. Proceeding like this, we obtain the required subset. \hfill $\Box$

The upshot of this lemma is that we can select a disjoint subset of balls which still controls the Lebesgue volume of the original set up to a universal constant $3^n$ (recall $|B_{3r}(x)| = 3^n|B_r(x)|$).

Now we can show

**Lemma 11.5.** Let $\alpha > 0$. For every Borel set $A$ we have

$$
|\{x \in A \mid (D\mu)(x) > \alpha\}| \leq 3^n \frac{\mu(A)}{\alpha}
$$

(11.9)

and

$$
|\{x \in A \mid (D\mu)(x) > 0\}| = 0, \text{ whenever } \mu(A) = 0.
$$

(11.10)

**Proof.** Let $A_\alpha = \{x \in A \mid (D\mu)(x) > \alpha\}$. We will show

$$
|K| \leq 3^n \frac{\mu(O)}{\alpha}
$$

for every open set $O$ with $A \subseteq O$ and every compact set $K \subseteq A_\alpha$. The first claim then follows from outer regularity of $\mu$ and inner regularity of the Lebesgue measure.
Given fixed $K$, $O$, for every $x \in K$ there is some $r_x$ such that $B_{r_x}(x) \subseteq O$ and $|B_{r_x}(x)| < \alpha^{-1}\mu(B_{r_x}(x))$. Since $K$ is compact, we can choose a finite subcover of $K$ from these balls. Moreover, by Lemma 11.4 we can refine our set of balls such that

$$|K| \leq 3^n \sum_{i=1}^{k} |B_{r_i}(x_i)| < \frac{3^n}{\alpha} \sum_{i=1}^{k} \mu(B_{r_i}(x_i)) \leq 3^n \frac{\mu(O)}{\alpha}.$$ 

To see the second claim, observe that $A_0 = \bigcup_{j=1}^{\infty} A_{1/j}$ and by the first part $|A_{1/j}| = 0$ for every $j$ if $\mu(A) = 0$. □

**Theorem 11.6** (Lebesgue). Let $f$ be (locally) integrable, then for a.e. $x \in \mathbb{R}^n$ we have

$$\lim_{r \downarrow 0} \frac{1}{|B_r(x)|} \int_{B_r(x)} |f(y) - f(x)| d^n y = 0. \quad (11.11)$$

The points where (11.11) holds are called **Lebesgue points** of $f$.

**Proof.** Decompose $f$ as $f = g + h$, where $g$ is continuous and $\|h\|_1 < \varepsilon$ (Theorem 10.15) and abbreviate

$$D_r(f)(x) := \frac{1}{|B_r(x)|} \int_{B_r(x)} |f(y) - f(x)| d^n y.$$ 

Then, since $\lim D_r(g)(x) = 0$ (for every $x$) and $D_r(f) \leq D_r(g) + D_r(h)$, we have

$$\limsup_{r \downarrow 0} D_r(f)(x) \leq \limsup_{r \downarrow 0} D_r(h)(x) \leq (\mathcal{D}\mu)(x) + |h(x)|,$$

where $d\mu = |h| d^n x$. This implies

$$\{x \mid \limsup_{r \downarrow 0} D_r(f)(x) \geq 2\alpha\} \subseteq \{x \mid (\mathcal{D}\mu)(x) \geq \alpha\} \cup \{x \mid |h(x)| \geq \alpha\}$$

and using the first part of Lemma 11.5 plus $|\{x \mid |h(x)| \geq \alpha\}| \leq \alpha^{-1}\|h\|_1$ (Problem 11.10), we see

$$|\{x \mid \limsup_{r \downarrow 0} D_r(f)(x) \geq 2\alpha\}| \leq (3^n + 1) \frac{\varepsilon}{\alpha}.$$ 

Since $\varepsilon$ is arbitrary, the Lebesgue measure of this set must be zero for every $\alpha$. That is, the set where the lim sup is positive has Lebesgue measure zero. □

**Example.** It is easy to see that every point of continuity of $f$ is a Lebesgue point. However, the converse is not true. To see this consider a function $f : \mathbb{R} \to [0, 1]$ which is given by a sum of non-overlapping spikes centered at $x_j = 2^{-j}$ with base length $2b_j$ and height 1. Explicitly

$$f(x) = \sum_{j=1}^{\infty} \max(0, 1 - b_j^{-1}|x - x_j|)$$
with \( b_{j+1} + b_j \leq 2^{-j-1} \) (such that the spikes don’t overlap). By construction \( f(x) \) will be continuous except at \( x = 0 \). However, if we let the \( b_j \)'s decay sufficiently fast such that the area of the spikes inside \((-r, r)\) is \( o(r) \), the point \( x = 0 \) will nevertheless be a Lebesgue point. For example, \( b_j = 2^{-2j-1} \) will do.

Note that the balls can be replaced by more general sets: A sequence of sets \( A_j(x) \) is said to shrink to \( x \) nicely if there are balls \( B_{r_j}(x) \) with \( r_j \to 0 \) and a constant \( \varepsilon > 0 \) such that \( A_j(x) \subseteq B_{r_j}(x) \) and \( |A_j| \geq \varepsilon |B_{r_j}(x)| \). For example, \( A_j(x) \) could be some balls or cubes (not necessarily containing \( x \)). However, the portion of \( B_{r_j}(x) \) which they occupy must not go to zero! For example, the rectangles \((0, \frac{1}{2}) \times (0, \frac{2}{2^j}) \subseteq \mathbb{R}^2 \) do shrink nicely to 0, but the rectangles \((0, \frac{1}{2}) \times (0, \frac{2}{2^j}) \) do not.

**Lemma 11.7.** Let \( f \) be (locally) integrable. Then at every Lebesgue point we have

\[
f(x) = \lim_{j \to \infty} \frac{1}{|A_j(x)|} \int_{A_j(x)} f(y)d^n y
\]  

whenever \( A_j(x) \) shrinks to \( x \) nicely.

**Proof.** Let \( x \) be a Lebesgue point and choose some nicely shrinking sets \( A_j(x) \) with corresponding \( B_{r_j}(x) \) and \( \varepsilon \). Then

\[
\frac{1}{|A_j(x)|} \int_{A_j(x)} |f(y) - f(x)|d^n y \leq\frac{1}{\varepsilon |B_{r_j}(x)|} \int_{B_{r_j}(x)} |f(y) - f(x)|d^n y
\]

and the claim follows. \( \square \)

**Corollary 11.8.** Let \( \mu \) be a Borel measure on \( \mathbb{R} \) which is absolutely continuous with respect to Lebesgue measure. Then its distribution function is differentiable a.e. and \( d\mu(x) = \mu'(x)dx \).

**Proof.** By assumption \( d\mu(x) = f(x)dx \) for some locally integrable function \( f \). In particular, the distribution function \( \mu(x) = \int_0^x f(y)dy \) is continuous. Moreover, since the sets \((x, x+r)\) shrink nicely to \( x \) as \( r \to 0 \), Lemma 11.7 implies

\[
\lim_{r \to 0} \frac{\mu((x, x+r))}{r} = \lim_{r \to 0} \frac{\mu(x+r) - \mu(x)}{r} = f(x)
\]

at every Lebesgue point of \( f \). Since the same is true for the sets \((x-r, x)\), \( \mu(x) \) is differentiable at every Lebesgue point and \( \mu'(x) = f(x) \). \( \square \)

As another consequence we obtain

**Theorem 11.9.** Let \( \mu \) be a Borel measure on \( \mathbb{R}^n \). The derivative \( D\mu \) exists a.e. with respect to Lebesgue measure and equals the Radon–Nikodym derivative of the absolutely continuous part of \( \mu \) with respect to Lebesgue measure;
that is,
\[ \mu_{ac}(A) = \int_A (D\mu)(x)d^n x. \] (11.13)

**Proof.** If \( d\mu = f d^n x \) is absolutely continuous with respect to Lebesgue measure, then \( (D\mu)(x) = f(x) \) at every Lebesgue point of \( f \) by Lemma 11.7 and the claim follows from Theorem 11.6. To see the general case, use the Lebesgue decomposition of \( \mu \) and let \( N \) be a support for the singular part with \( |N| = 0 \). Then \( (D\mu_{sing})(x) = 0 \) for a.e. \( x \in \mathbb{R}^n \setminus N \) by the second part of Lemma 11.5.

In particular, \( \mu \) is singular with respect to Lebesgue measure if and only if \( D\mu = 0 \) a.e. with respect to Lebesgue measure.

Using the upper and lower derivatives, we can also give supports for the absolutely and singularly continuous parts.

**Theorem 11.10.** The set \( \{ x | 0 < (D\mu)(x) < \infty \} \) is a support for the absolutely continuous and \( \{ x | (D\mu)(x) = \infty \} \) is a support for the singular part.

**Proof.** The first part is immediate from the previous theorem. For the second part first note that by \( (D\mu)(x) \geq (D\mu_{sing})(x) \) we can assume that \( \mu \) is purely singular. It suffices to show that the set \( A_k := \{ x | (D\mu)(x) < k \} \) satisfies \( \mu(A_k) = 0 \) for every \( k \in \mathbb{N} \).

Let \( K \subset A_k \) be compact, and let \( V_j \supseteq K \) be some open set such that \( |V_j \setminus K| \leq \frac{1}{j} \). For every \( x \in K \) there is some \( \varepsilon = \varepsilon(x) \) such that \( B_{\varepsilon}(x) \subseteq V_j \) and \( \mu(B_{3\varepsilon}(x)) \leq k|B_{3\varepsilon}(x)| \). By compactness, finitely many of these balls cover \( K \) and hence
\[
\mu(K) \leq \sum_i \mu(B_{\varepsilon_i}(x_i)).
\]
Selecting disjoint balls as in Lemma 11.4 further shows
\[
\mu(K) \leq \sum_{i} \mu(B_{3\varepsilon_i}(x_i)) \leq k3^n \sum_{i} |B_{\varepsilon_i}(x_i)| \leq k3^n |V_j|.
\]
Letting \( j \to \infty \), we see \( \mu(K) \leq k3^n |K| \) and by regularity we even have \( \mu(A) \leq k3^n |A| \) for every \( A \subseteq A_k \). Hence \( \mu \) is absolutely continuous on \( A_k \) and since we assumed \( \mu \) to be singular, we must have \( \mu(A_k) = 0 \).

Finally, we note that these supports are minimal. Here a support \( M \) of some measure \( \mu \) is called a minimal support (it is sometimes also called an essential support) if every subset \( M_0 \subseteq M \) which does not support \( \mu \) (i.e., \( \mu(M_0) = 0 \)) has Lebesgue measure zero.

**Example.** Let \( X := \mathbb{R} \), \( \Sigma := \mathcal{B} \). If \( d\mu(x) := \sum_n \alpha_n d\theta(x - x_n) \) is a sum of Dirac measures, then the set \( \{ x_n \} \) is clearly a minimal support for \( \mu \). Moreover, it is clearly the smallest support as none of the \( x_n \) can be
removed. If we choose \( \{x_n\} \) to be the rational numbers, then \( \text{supp}(\mu) = \mathbb{R} \), but \( \mathbb{R} \) is not a minimal support, as we can remove the irrational numbers.

On the other hand, if we consider the Lebesgue measure \( \lambda \), then \( \mathbb{R} \) is a minimal support. However, the same is true if we remove any set of measure zero, for example, the Cantor set. In particular, since we can remove any single point, we see that, just like supports, minimal supports are not unique.

\[\text{Lemma 11.11.} \] The set \( M_{ac} := \{x|0 < (D\mu)(x) < \infty\} \) is a minimal support for \( \mu_{ac} \).

\[\text{Proof.} \] Suppose \( M_0 \subseteq M_{ac} \) and \( \mu_{ac}(M_0) = 0 \). Set \( M_\varepsilon = \{x \in M_0|\varepsilon < (D\mu)(x)\} \) for \( \varepsilon > 0 \). Then \( M_\varepsilon \nrightarrow M_0 \) and

\[|M_\varepsilon| = \int_{M_\varepsilon} d^n x \leq \frac{1}{\varepsilon} \int_{M_\varepsilon} (D\mu)(x) dx = \frac{1}{\varepsilon} \mu_{ac}(M_\varepsilon) \leq \frac{1}{\varepsilon} \mu_{ac}(M_0) = 0\]

shows \( |M_0| = \lim_{\varepsilon \downarrow 0} |M_\varepsilon| = 0 \). \( \square \)

Note that the set \( M = \{x|0 < (D\mu)(x)\} \) is a minimal support of \( \mu \) (Problem 11.8).

\[\text{Example.} \] The \textbf{Cantor function} is constructed as follows: Take the sets \( C_n \) used in the construction of the Cantor set \( C \): \( C_n \) is the union of \( 2^n \) closed intervals with \( 2^n - 1 \) open gaps in between. Set \( f_n \) equal to \( j/2^n \) on the \( j \)'th gap of \( C_n \) and extend it to \( [0,1] \) by linear interpolation. Note that, since we are creating precisely one new gap between every old gap when going from \( C_n \) to \( C_{n+1} \), the value of \( f_{n+1} \) is the same as the value of \( f_n \) on the gaps of \( C_n \). Explicitly, we have \( f_0(x) = x \) and \( f_{n+1} = K(f_n) \), where

\[K(f)(x) := \begin{cases} \frac{1}{3}f(3x), & 0 \leq x \leq \frac{1}{3}, \\ \frac{1}{2}, & \frac{1}{3} \leq x \leq \frac{2}{3}, \\ \frac{1}{2}(1 + f(3x - 2)), & \frac{2}{3} \leq x \leq 1. \end{cases}\]

Since \( \|f_{n+1} - f_n\|_\infty \leq \frac{1}{2}\|f_{n+1} - f_n\|_\infty \) we can define the Cantor function as \( f = \lim_{n \to \infty} f_n \). By construction \( f \) is a continuous function which is constant on every subinterval of \([0,1] \setminus C\) Since \( C \) is of Lebesgue measure zero, this set is of full Lebesgue measure and hence \( f' = 0 \) a.e. in \([0,1] \). In particular, the corresponding measure, the \textbf{Cantor measure}, is supported on \( C \) and is purely singular with respect to Lebesgue measure. \( \diamond \)

\[\text{Problem 11.7.} \] Show that

\[\mu(B_\varepsilon(x)) \leq \liminf_{y \to x} \mu(B_\varepsilon(y)) \leq \limsup_{y \to x} \mu(B_\varepsilon(y)) \leq \mu(B_\varepsilon(x)).\]

In particular, conclude that \( x \mapsto \mu(B_\varepsilon(x)) \) is lower semicontinuous for \( \varepsilon > 0 \).
Problem 11.8. Show that \( M := \{ x | 0 < (D\mu)(x) \} \) is a minimal support of \( \mu \).

Problem 11.9. Suppose \( D\mu \leq \alpha \). Show that \( d\mu = f \, d^n \) with \( 0 \leq f \leq \alpha \).

Problem 11.10 (Markov (also Chebyshev) inequality). For \( f \in L^1(\mathbb{R}^n) \) and \( \alpha > 0 \) show
\[
|\{ x \in A | |f(x)| \geq \alpha \}| \leq \frac{1}{\alpha} \int_A |f(x)| \, d^n x.
\]

Somewhat more general, assume \( g(x) \geq 0 \) is nondecreasing and \( g(\alpha) > 0 \). Then
\[
\mu(\{ x \in A | f(x) \geq \alpha \}) \leq \frac{1}{g(\alpha)} \int_A g \circ f \, d\mu.
\]

Problem 11.11. Show that the Cantor function is Hölder continuous \( |f(x) - f(y)| \leq |x - y|^\alpha \) with exponent \( \alpha = \log_3(2) \). (Hint: Show that if a bijection \( g : [0, 1] \to [0, 1] \) satisfies a Hölder estimate \( |g(x) - g(y)| \leq M|x - y|^\alpha \), then so does \( K(g) : |K(g)(x) - K(g)(y)| \leq \frac{3^\alpha}{2} M|x - y|^\alpha \).)

11.3. Complex measures

Let \( (X, \Sigma) \) be some measurable space. A map \( \nu : \Sigma \to \mathbb{C} \) is called a complex measure if it is \( \sigma \)-additive:
\[
\nu(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \nu(A_n), \quad A_n \in \Sigma.
\]
Choosing \( A_n = \emptyset \) for all \( n \) in (11.14) shows \( \nu(\emptyset) = 0 \).

Note that a positive measure is a complex measure only if it is finite (the value \( \infty \) is not allowed for complex measures). Moreover, the definition implies that the sum is independent of the order of the sets \( A_n \), that is, it converges unconditionally and thus absolutely by the Riemann series theorem.

Example. Let \( \mu \) be a positive measure. For every \( f \in L^1(X, d\mu) \) we have that \( f \, d\mu \) is a complex measure (compare the proof of Lemma 9.3 and use dominated in place of monotone convergence). In fact, we will show that every complex measure is of this form. \( \diamond \)

Example. Let \( \nu_1, \nu_2 \) be two complex measures and \( \alpha_1, \alpha_2 \) two complex numbers. Then \( \alpha_1 \nu_1 + \alpha_2 \nu_2 \) is again a complex measure. Clearly we can extend this to any finite linear combination of complex measures. \( \diamond \)

When dealing with complex functions \( f \) an important object is the positive function \( |f| \). Given a complex measure \( \nu \) it seems natural to consider the set function \( A \mapsto |\nu(A)| \). However, considering the simple example
\( d\nu(x) := \text{sign}(x)dx \) on \( X := [-1,1] \) one sees that this set function is not additive and this simple approach does not provide a positive measure associated with \( \nu \). However, using \( |\nu(A \cap [-1,0])| + |\nu(A \cap [0,1])| \) we do get a positive measure. Motivated by this we introduce the total variation of a measure defined as

\[
|\nu|(A) := \sup \left\{ \sum_{k=1}^{n} |\nu(A_k)| \mid A_k \in \Sigma \text{ disjoint}, \ A = \bigcup_{k=1}^{n} A_k \right\}. \tag{11.15}
\]

Note that by construction we have

\[
|\nu|(A) \leq |\nu|(A). \tag{11.16}
\]

Moreover, the total variation is monotone \( |\nu|(A) \leq |\nu|(B) \) if \( A \subseteq B \) and for a positive measure \( \mu \) we have of course \( |\mu|(A) = \mu(A) \).

**Theorem 11.12.** The total variation \( |\nu| \) of a complex measure \( \nu \) is a finite positive measure.

**Proof.** We begin by showing that \( |\nu| \) is a positive measure. We need to show \( |\nu|(A) = \sum_{n=1}^{\infty} |\nu|(A_n) \) for any partition of \( A \) into disjoint sets \( A_n \). If \( |\nu|(A_n) = \infty \) for some \( n \) it is not hard to see that \( |\nu|(A) = \infty \) and hence we can assume \( |\nu|(A_n) < \infty \) for all \( n \).

Let \( \epsilon > 0 \) be fixed and for each \( A_n \) choose a disjoint partition \( B_{n,k} \) of \( A_n \) such that

\[
|\nu|(A_n) \leq \sum_{k=1}^{m} |\nu(B_{n,k})| + \frac{\epsilon}{2^n}.
\]

Then

\[
\sum_{n=1}^{N} |\nu|(A_n) \leq \sum_{n=1}^{N} \sum_{k=1}^{m} |\nu(B_{n,k})| + \epsilon \leq |\nu|\left( \bigcup_{n=1}^{N} A_n \right) + \epsilon \leq |\nu|(A) + \epsilon
\]

since \( \bigcup_{n=1}^{N} \bigcup_{k=1}^{m} B_{n,k} = \bigcup_{n=1}^{N} A_n \). Since \( \epsilon \) was arbitrary this shows \( |\nu|(A) \geq \sum_{n=1}^{\infty} |\nu|(A_n) \).

Conversely, given a finite partition \( B_k \) of \( A \), then

\[
\sum_{k=1}^{m} |\nu(B_k)| = \sum_{k=1}^{m} \left| \sum_{n=1}^{\infty} \nu(B_k \cap A_n) \right| \leq \sum_{k=1}^{m} \sum_{n=1}^{\infty} |\nu(B_k \cap A_n)|
\]

\[
= \sum_{n=1}^{\infty} \sum_{k=1}^{m} |\nu(B_k \cap A_n)| \leq \sum_{n=1}^{\infty} |\nu|(A_n).
\]

Taking the supremum over all partitions \( B_k \) shows \( |\nu|(A) \leq \sum_{n=1}^{\infty} |\nu|(A_n) \).

Hence \( |\nu| \) is a positive measure and it remains to show that it is finite. Splitting \( \nu \) into its real and imaginary part, it is no restriction to assume that \( \nu \) is real-valued since \( |\nu|(A) \leq |\text{Re}(\nu)|(A) + |\text{Im}(\nu)|(A) \).
The idea is as follows: Suppose we can split any given set \( A \) with \( |\nu(A)| = \infty \) into two subsets \( B \) and \( A \setminus B \) such that \( |\nu(B)| \geq 1 \) and \( |\nu(A \setminus B)| = \infty \). Then we can construct a sequence \( B_n \) of disjoint sets with \( |\nu(B_n)| \geq 1 \) for which
\[
\sum_{n=1}^{\infty} \nu(B_n)
\]
diverges (the terms of a convergent series must converge to zero). But \( \sigma \)-additivity requires that the sum converges to \( \nu(\bigcup_n B_n) \), a contradiction.

It remains to show existence of this splitting. Let \( A \) with \( |\nu(A)| = \infty \) be given. Then there are disjoint sets \( A_j \) such that
\[
\sum_{j=1}^{n} |\nu(A_j)| \geq 2 + |\nu(A)|.
\]
Now let \( A_+ = \bigcup\{A_j|\nu(A_j) \geq 0\} \) and \( A_- = A \setminus A_+ = \bigcup\{A_j|\nu(A_j) < 0\} \). Then the above inequality reads \( \nu(A_+) + |\nu(A_-)| \geq 2 + |\nu(A_+) - |\nu(A_-)|| \)
implying (show this) that for both of them we have \( |\nu(A_\pm)| \geq 1 \) and by \( |\nu|(A) = |\nu|(A_+) + |\nu|(A_-) \) either \( A_+ \) or \( A_- \) must have infinite \( |\nu| \) measure.

Note that this implies that every complex measure \( \nu \) can be written as a linear combination of four positive measures. In fact, first we can split \( \nu \) into its real and imaginary part
\[
\nu = \nu_r + i\nu_i, \quad \nu_r(A) = \text{Re}(\nu(A)), \quad \nu_i(A) = \text{Im}(\nu(A)). \quad (11.17)
\]
Second we can split every real (also called **signed**) measure according to
\[
\nu = \nu_+ - \nu_- , \quad \nu_{\pm}(A) = \frac{|\nu|(A) \pm \nu(A)}{2} \quad (11.18)
\]
By (11.16) both \( \nu_- \) and \( \nu_+ \) are positive measures. This splitting is also known as **Jordan decomposition** of a signed measure. In summary, we can split every complex measure \( \nu \) into four positive measures
\[
\nu = \nu_{r,+} - \nu_{r,-} + i(\nu_{i,+} - \nu_{i,-}) \quad (11.19)
\]
which is also known as Jordan decomposition.

Of course such a decomposition of a signed measure is not unique (we can always add a positive measure to both parts), however, the Jordan decomposition is unique in the sense that it is the smallest possible decomposition.

**Lemma 11.13.** Let \( \nu \) be a complex measure and \( \mu \) a positive measure satisfying \( |\nu(A)| \leq \mu(A) \) for all measurable sets \( A \). Then \( |\nu| \leq \mu \). (Here \( |\nu| \leq \mu \) has to be understood as \( |\nu|(A) \leq \mu(A) \) for every measurable set \( A \).)
Theorem 11.14. The set of all complex measures \( \mathcal{M}(X) \) together with the norm \( \|\nu\| := |\nu|(X) \) is a Banach space.

**Proof.** Clearly \( \mathcal{M}(X) \) is a vector space and it is straightforward to check that \( |\nu|(X) \) is a norm. Hence it remains to show that every Cauchy sequence \( \nu_k \) has a limit.

First of all, by \( |\nu_k(A) - \nu_j(A)| = |(\nu_k - \nu_j)(A)| \leq |\nu_k - \nu_j|(A) \leq \|\nu_k - \nu_j\| \), we see that \( \nu_k(A) \) is a Cauchy sequence in \( \mathbb{C} \) for every \( A \in \Sigma \) and we can define

\[
\nu(A) := \lim_{k \to \infty} \nu_k(A).
\]

Moreover, \( C_j := \sup_{k \geq j} \|\nu_k - \nu_j\| \to 0 \) as \( j \to \infty \) and we have

\[
|\nu_j(A) - \nu(A)| \leq C_j.
\]

Next we show that \( \nu \) satisfies (11.14). Let \( A_m \) be given disjoint sets and set \( \tilde{A}_n := \bigcup_{m=1}^n A_m, A := \bigcup_{m=1}^\infty A_m \). Since we can interchange limits with finite sums, (11.14) holds for finitely many sets. Hence it remains to show \( \nu(\tilde{A}_n) -> \nu(A) \). This follows from

\[
|\nu(\tilde{A}_n) - \nu(A)| \leq |\nu(\tilde{A}_n) - \nu(\tilde{A}_n)| + |\nu(\tilde{A}_n) - \nu(A)| + |\nu(A) - \nu(A)| \leq 2C_k + |\nu_k(\tilde{A}_n) - \nu_k(A)|.
\]

Finally, \( \nu_k \to \nu \) since \( |\nu_k(A) - \nu(A)| \leq C_k \) implies \( \|\nu_k - \nu\| \leq 4C_k \) (Problem 11.16).

If \( \mu \) is a positive and \( \nu \) a complex measure we say that \( \nu \) is absolutely continuous with respect to \( \mu \) if \( \mu(A) = 0 \) implies \( \nu(A) = 0 \). We say that \( \nu \) is singular with respect to \( \mu \) if \( |\nu| \) is, that is, there is a measurable set \( N \) such that \( \mu(N) = 0 \) and \( |\nu|(X \setminus N) = 0 \).

**Lemma 11.15.** If \( \mu \) is a positive and \( \nu \) a complex measure then \( \nu \ll \mu \) if and only if \( |\nu| \ll \mu \).

**Proof.** If \( \nu \ll \mu \), then \( \mu(A) = 0 \) implies \( \mu(B) = 0 \) for every \( B \subset A \) and hence \( |\nu|(A) = 0 \). Conversely, if \( |\nu| \ll \mu \), then \( \mu(A) = 0 \) implies \( |\nu(A)| \leq |\nu|(A) = 0 \). \( \square \)
Now we can prove the complex version of the Radon–Nikodym theorem:

**Theorem 11.16** (Complex Radon–Nikodym). Let $(X, \Sigma)$ be a measurable space, $\mu$ a positive $\sigma$-finite measure and $\nu$ a complex measure. Then there exists a function $f \in L^1(X, \mu)$ and a set $N$ of $\mu$ measure zero, such that

$$\nu(A) = \nu(A \cap N) + \int_A f \, d\mu.$$  \hspace{1cm} (11.20)

The function $f$ is determined uniquely a.e. with respect to $\mu$ and is called the Radon–Nikodym derivative $d\nu/d\mu$ of $\nu$ with respect to $\mu$.

In particular, $\nu$ can be uniquely decomposed as $\nu = \nu_{\text{sing}} + \nu_{\text{ac}}$, where $\nu_{\text{sing}}(A) := \nu(A \cap N)$ is singular and $d\nu_{\text{ac}} := f \, d\mu$ is absolutely continuous with respect to $\mu$.

**Proof.** We start with the case where $\nu$ is a signed measure. Let $\nu = \nu_+ - \nu_-$ be its Jordan decomposition. Then by Theorem 11.1 there are sets $N_{\pm}$ and functions $f_{\pm}$ such that $\nu_{\pm}(A) = \nu_{\pm}(A \cap N_{\pm}) + \int_A f_{\pm} \, d\mu$. Since $\nu_{\pm}$ are finite measures we must have $\int_X f_{\pm} \, d\mu \leq \nu_{\pm}(X)$ and hence $f_{\pm} \in L^1(X, \mu)$. Moreover, since $N = N_- \cup N_+$ has $\mu$ measure zero the remark after Theorem 11.1 implies $\nu_{\pm}(A) = \nu_{\pm}(A \cap N) + \int_A f_{\pm} \, d\mu$ and hence $\nu(A) = \nu(A \cap N) + \int_A f \, d\mu$ where $f = f_+ - f_- \in L^1(X, \mu)$. If $\nu$ is complex we can split it into real and imaginary part and use the same reasoning to reduce it to the signed case. If $\nu$ is absolutely continuous we have $|\nu|(N) = 0$ and hence $\nu(A \cap N) = 0$. Uniqueness of the decomposition (and hence of $f$) follows literally as in the proof of Theorem 11.3. \hfill \Box

If $\nu$ is absolutely continuous with respect to $\mu$ the total variation of $d\nu = f \, d\mu$ is just $d|\nu| = |f| \, d\mu$.

**Lemma 11.17.** Let $d\nu = d\nu_{\text{sing}} + f \, d\mu$ be the Lebesgue decomposition of a complex measure $\nu$ with respect to a positive $\sigma$-finite measure $\mu$. Then

$$|\nu|(A) = |\nu_{\text{sing}}|(A) + \int_A |f| \, d\mu.$$  \hspace{1cm} (11.21)

**Proof.** We first show $|\nu| = |\nu_{\text{sing}}| + |\nu_{\text{ac}}|$. Let $A$ be given and let $A_k$ be a partition of $A$ as in (11.15). By the definition of the total variation we can find a partition $A_{\text{sing},k}$ of $A \cap N$ such that $\sum_k |\nu(A_{\text{sing},k})| \geq |\nu_{\text{sing}}|(A) - \frac{\varepsilon}{2}$ for arbitrary $\varepsilon > 0$ (note that $\nu_{\text{sing}}(A_{\text{sing},k}) = \nu(A_{\text{sing},k})$ as well as $\nu_{\text{sing}}(A \cap N') = 0$). Similarly, there is such a partition $A_{\text{ac},k}$ of $A \cap N'$ such that $\sum_k |\nu(A_{\text{ac},k})| \geq |\nu_{\text{ac}}|(A) - \frac{\varepsilon}{2}$. Then combing both partitions into a partition $A_k$ for $A$ we obtain $|\nu|(A) \geq \sum_k |\nu(A_k)| \geq |\nu_{\text{sing}}|(A) + |\nu_{\text{ac}}|(A) - \varepsilon$. Since $\varepsilon > 0$ is arbitrary we conclude $|\nu|(A) \geq |\nu_{\text{sing}}|(A) + |\nu_{\text{ac}}|(A)$ and as the converse inequality is trivial the first claim follows.
It remains to show \( d|\nu_{ac}| = |f| \, d\mu \). If \( A_n \) are disjoint sets and \( A = \bigcup_n A_n \) we have
\[
\sum_n |\nu(A_n)| = \sum_n \int_{A_n} |f| \, d\mu \leq \sum_n \int_{A_n} |f| \, d\mu = \int_A |f| \, d\mu.
\]
Hence \( |\nu|(A) \leq \int_A |f| \, d\mu \). To show the converse define
\[
A^n_k = \{ x \in A \mid \frac{k-1}{n} < \frac{\arg(f(x)) + \pi}{2\pi} \leq \frac{k}{n} \}, \quad 1 \leq k \leq n.
\]
Then the simple functions
\[
s_n(x) = \sum_{k=1}^n e^{-2\pi i \frac{k-1}{n} + i\pi} \chi_{A^n_k}(x)
\]
converge to \( \text{sign}(f(x)^*) \) for every \( x \in A \) and hence
\[
\lim_{n \to \infty} \int_A s_n f \, d\mu = \int_A |f| \, d\mu
\]
by dominated convergence. Moreover,
\[
\left| \int_A s_n f \, d\mu \right| \leq \sum_{k=1}^n \left| \int_{A^n_k} s_n f \, d\mu \right| = \sum_{k=1}^n |\nu(A^n_k)| \leq |\nu|(A)
\]
shows \( \int_A |f| \, d\mu \leq |\nu|(A) \). \( \square \)

As a consequence we obtain (Problem 11.12):

**Corollary 11.18.** If \( \nu \) is a complex measure, then \( d\nu = h \, d|\nu| \), where \( |h| = 1 \).

If \( \nu \) is a signed measure, then \( h \) is real-valued and we obtain:

**Corollary 11.19.** If \( \nu \) is a signed measure, then \( d\nu = h \, d|\nu| \), where \( h^2 = 1 \). In particular, \( d\nu_{\pm} = \chi_{X_{\pm}} \, d|\nu| \), where \( X_{\pm} := h^{-1}(\{-1,1\}) \).

The decomposition \( X = X_+ \cup X_- \) from the previous corollary is known as **Hahn decomposition** and it is characterized by the property that \( \pm \nu(A) \geq 0 \) if \( A \subseteq X_\pm \). This decomposition is not unique since we can shift sets of \( |\nu| \) measure zero from one to the other.

We also briefly mention that the concept of regularity generalizes in a straightforward manner to complex Borel measures. If \( X \) is a topological space with its Borel \( \sigma \)-algebra we call \( \nu \) **(outer/inner) regular** if \( |\nu| \) is. It is not hard to see (Problem 11.17):

**Lemma 11.20.** A complex measure is regular if and only if all measures in its Jordan decomposition are.
The subspace of regular Borel measures will be denoted by \( M_{\text{reg}}(X) \). Note that it is closed and hence again a Banach space (Problem 11.18).

Clearly we can use Corollary 11.18 to define the integral of a bounded function \( f \) with respect to a complex measure \( d\nu = h\, d|\nu| \) as

\[
\int f\, d\nu := \int fh\, d|\nu|.
\]  

(11.22)

In fact, it suffices to assume that \( f \) is integrable with respect to \( d|\nu| \) and we obtain

\[
\left| \int f\, d\nu \right| \leq \int |f|\, d|\nu|.
\]  

(11.23)

For bounded functions this implies

\[
\left| \int_A f\, d\nu \right| \leq \|f\|_{\infty}|\nu|(A).
\]  

(11.24)

Finally, there is an interesting equivalent definition of absolute continuity:

**Lemma 11.21.** If \( \mu \) is a positive and \( \nu \) a complex measure then \( \nu \ll \mu \) if and only if for every \( \varepsilon > 0 \) there is a corresponding \( \delta > 0 \) such that

\[
\mu(A) < \delta \Rightarrow |\nu(A)| < \varepsilon, \quad \forall A \in \Sigma.
\]  

(11.25)

**Proof.** Suppose \( \nu \ll \mu \) implying that it is of the from \( d\nu = f\, d\mu \). Let \( X_n = \{ x \in X | |f(x)| \leq n \} \) and note that \( |\nu|(X \setminus X_n) \to 0 \) since \( X_n \nearrow X \) and \( |\nu|(X) < \infty \). Given \( \varepsilon > 0 \) we can choose \( n \) such that \( |\nu|(X \setminus X_n) \leq \frac{\varepsilon}{2} \) and \( \delta = \frac{\varepsilon}{2n} \). Then, if \( \mu(A) < \delta \) we have

\[
|\nu(A)| \leq |\nu|(A \cap X_n) + |\nu|(X \setminus X_n) \leq n\mu(A) + \frac{\varepsilon}{2} < \varepsilon.
\]

The converse direction is obvious. \( \square \)

It is important to emphasize that the fact that \( |\nu|(X) < \infty \) is crucial for the above lemma to hold. In fact, it can fail for positive measures as the simple counterexample \( d\nu(\lambda) = \lambda^2 d\lambda \) on \( \mathbb{R} \) shows.

**Problem 11.12.** Prove Corollary 11.18. (Hint: Use the complex Radon–Nikodym theorem to get existence of \( h \). Then show that \( 1 - |h| \) vanishes a.e.)

**Problem 11.13** *(Markov inequality).* Let \( \nu \) be a complex and \( \mu \) a positive measure. If \( f \) denotes the Radon–Nikodym derivative of \( \nu \) with respect to \( \mu \), then show that

\[
\mu(\{ x \in A | |f(x)| \geq \alpha \}) \leq \frac{|\nu|(A)}{\alpha}.
\]
Problem 11.14. Let $\nu$ be a complex and $\mu$ a positive measure and suppose $|\nu(A)| \leq C \mu(A)$ for all $A \in \Sigma$. Then $d\nu = f \, d\mu$ with $\|f\|_{\infty} \leq C$. (Hint: First show $|\nu|(A) \leq C \mu(A)$ and then use Problem 11.3.)

Problem 11.15. Let $\nu$ be a signed measure and $\nu_{\pm}$ its Jordan decomposition. Show $\nu_+(A) = \max_{B \in \Sigma, B \subseteq A} \nu(B)$, $\nu_-(A) = -\min_{B \in \Sigma, B \subseteq A} \nu(B)$.

Problem 11.16. Let $\nu$ be a complex measure with Jordan decomposition (11.19). Show the estimate
\[ \frac{1}{\sqrt{2}} \nu_s(A) \leq |\nu|(A) \leq \nu_s(A), \quad \nu_s = \nu_{r,+} + \nu_{r,-} + \nu_{i,+} + \nu_{i,-}. \]
Show that $|\nu(A)| \leq C$ for all measurable sets $A$ implies $\|\nu\| \leq 4C$.

Problem 11.17. Show Lemma 11.20. (Hint: Problems 8.21 and 11.16.)

Problem 11.18. Let $X$ be a topological space. Show that $M_{\text{reg}}(X) \subseteq M(X)$ is a closed subspace.

Problem 11.19. Define the convolution of two complex Borel measures $\mu$ and $\nu$ on $\mathbb{R}^n$ via
\[(\mu \ast \nu)(A) := \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \chi_A(x+y) d\mu(x) d\nu(y).\]
Note $|\mu \ast \nu|(\mathbb{R}^n) \leq |\mu|_{\mathbb{R}^n} |\nu|_{\mathbb{R}^n}$. Show that this implies
\[ \int_{\mathbb{R}^n} h(x) d(\nu \ast \nu)(x) = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} h(x+y) d\mu(x) d\nu(y) \]
for any bounded measurable function $h$. Conclude that it coincides with our previous definition in case $\mu$ and $\nu$ are absolutely continuous with respect to Lebesgue measure.

11.4. Hausdorff measure

Throughout this section we will assume that $(X,d)$ is a metric space. Recall that the diameter of a subset $A \subseteq X$ is defined by $\text{diam}(A) := \sup_{x,y \in A} d(x,y)$ with the convention that $\text{diam}(\emptyset) = 0$. A cover $\{A_j\}$ of $A$ is called a $\delta$-cover if it is countable and if $\text{diam}(A_j) \leq \delta$ for all $j$.

For $A \subseteq X$ and $\alpha \geq 0$, $\delta > 0$ we define
\[ h_\delta^{\alpha,*}(A) := \inf \left\{ \sum_j \text{diam}(A_j)^\alpha \left| \{A_j\} \text{ is a } \delta\text{-cover of } A \right. \right\} \in [0, \infty]. \quad (11.26) \]
which is an outer measure by Lemma 8.2. In the case $\alpha = 0$ and $A = \emptyset$ we also regard the empty cover as a valid cover such that $h_\delta^{0,*}(\emptyset) = 0$. 

As $\delta$ decreases the number of admissible covers decreases and hence $h_\delta^\alpha(A)$ increases as a function of $\delta$. Thus the limit
\[ h^{\alpha,*}(A) := \lim_{\delta \to 0} h_\delta^{\alpha,*}(A) = \sup_{\delta > 0} h_\delta^{\alpha,*}(A) \] (11.27)
exists. Moreover, it is not hard to see that it is again an outer measure (Problem 11.20) and by Theorem 8.9 we get a measure. To show that the $\sigma$-algebra from Theorem 8.9 contains all Borel sets it suffices to show that $\mu^*$ is a metric outer measure (cf. Lemma 8.11).

Now if $A_1, A_2$ with $\text{dist}(A_1, A_2) > 0$ are given and $\delta < \text{dist}(A_1, A_2)$ then every set from a cover for $A_1 \cup A_2$ can have nonempty intersection with at most one of both sets. Consequently $h_\delta^{\alpha,*}(A_1 \cup A_2) = h_\delta^{\alpha,*}(A_1) + h_\delta^{\alpha,*}(A_2)$ for $\delta < \text{dist}(A_1, A_2)$ implying $h^{\alpha,*}(A_1 \cup A_2) = h^{\alpha,*}(A_1) + h^{\alpha,*}(A_2)$. Hence $h^{\alpha,*}$ is a metric outer measure and the resulting measure $h^\alpha$ on the Borel $\sigma$-algebra is called the $\alpha$-dimensional Hausdorff measure. Note that if $X$ is a vector space with a translation invariant metric, then the diameter of a set is translation invariant and so will be $h^\alpha$.

**Example.** For example, consider the case $\alpha = 0$. Suppose $A = \{x, y\}$ consists of two points. Then $h_\delta^0(A) = 1$ for $\delta \geq d(x, y)$ and $h_\delta^0(A) = 2$ for $\delta < |x - y|$. In particular, $h^0(A) = 2$. Similarly, it is not hard to see (show this) that $h^0(A)$ is just the number of points in $A$, that is, $h^0$ is the counting measure on $X$. ♦

**Example.** At the other extreme, if $X := \mathbb{R}^n$, we have $h^n(A) = c_n |A|$, where $|A|$ denotes the Lebesgue measure of $A$. In fact, since the square $(0, 1]$ has $\text{diam}((0, 1]) = \sqrt{n}$ and we can cover it by $k^n$ squares of side length $\frac{1}{k}$, we see $h^n((0, 1]) \leq n^{n/2}$. Thus $h^n$ is a translation invariant Borel measure which hence must be a multiple of Lebesgue measure. To identify $c_n$ we will need the isodiametric inequality. ♦

For the rest of this section we restrict ourselves to the case $X = \mathbb{R}^n$. Note that while in dimension more than one it is not true that a set of diameter $d$ is contained in a ball of diameter $d$ (a counter example is an equilateral triangle), we at least have the following:

**Lemma 11.22** (Isodiametric inequality). *For every Borel set $A \in \mathcal{B}^n$ we have
\[ |A| \leq \frac{V_n}{2^n} \text{diam}(A)^n. \]
In other words, a ball is the set with the largest volume when the diameter is kept fixed.*

**Proof.** The trick is to transform $A$ to a set with smaller diameter but same volume via Steiner symmetrization. To this end we build up $A$ from slices
obtained by keeping the first coordinate fixed: \( A(y) = \{ x \in \mathbb{R} | (x, y) \in A \} \).
Now we build a new set \( \tilde{A} \) by replacing \( A(y) \) with a symmetric interval with the same measure, that is, \( \tilde{A} = \{ (x, y) | |x| \leq |A(y)|/2 \} \). Note that by Theorem 9.10 \( \tilde{A} \) is measurable with \( |A| = |\tilde{A}| \). Hence the same is true for \( \tilde{A} = \{ (x, y) | |x| \leq |A(y)|/2 \} \setminus \{ (0, y) | A(y) = \emptyset \} \) if we look at the complete Lebesgue measure, since the set we subtract is contained in a set of measure zero.

Moreover, if \( \tilde{I}, \tilde{J} \) are closed intervals, then \( \sup_{x_1 \in \tilde{I}, x_2 \in \tilde{J}} |x_2 - x_1| \geq \frac{|	ilde{I}|}{2} + \frac{|	ilde{J}|}{2} \) (without loss both intervals are compact and \( i \leq j \), where \( i,j \) are the midpoints of \( \tilde{I}, \tilde{J} \); then the sup is at least \( (j + \frac{|	ilde{J}|}{2}) - (i - \frac{|	ilde{I}|}{2}) \)). If \( I, J \) are Borel sets in \( \mathcal{B}^1 \) and \( \tilde{I}, \tilde{J} \) are their respective closed convex hulls, then \( \sup_{x_1 \in \tilde{I}, x_2 \in \tilde{J}} |x_2 - x_1| = \sup_{x_1 \in I, x_2 \in J} |x_2 - x_1| \geq \frac{|I|}{2} + \frac{|J|}{2} \geq \frac{|I|}{2} + \frac{|J|}{2} \). Thus for \( (\tilde{x}_1, y_1), (\tilde{x}_2, y_2) \in \tilde{A} \) we can find \( (x_1, y_1), (x_2, y_2) \in A \) with \( |x_1 - x_2| \geq |\tilde{x}_1 - \tilde{x}_2| \) implying \( \text{diam}(\tilde{A}) \leq \text{diam}(A) \).

In addition, if \( A \) is symmetric with respect to \( x_j \mapsto -x_j \) for some \( 2 \leq j \leq n \), then so is \( \tilde{A} \).

Now repeat this procedure with the remaining coordinate directions, to obtain a set \( \tilde{A} \) which is symmetric with respect to reflection \( x \mapsto -x \) and satisfies \( |A| = |\tilde{A}| \), \( \text{diam}(\tilde{A}) \leq \text{diam}(A) \). By symmetry \( \tilde{A} \) is contained in a ball of diameter \( \text{diam}(\tilde{A}) \) and the claim follows.

**Lemma 11.23.** For every Borel set \( A \in \mathcal{B}^n \) we have

\[
h^n(A) = \frac{2^n}{V_n} |A|.
\]

**Proof.** Using (8.40) (for \( C \) choose the collection of all open balls of radius at most \( \delta \) one infers \( h^n_\delta(A) \leq \frac{2^n}{V_n} |A| \) implying \( c_n \leq 2^n/V_n \). The converse inequality \( h^n_\delta(A) \geq \frac{2^n}{V_n} |A| \) follows from the isodiametric inequality.

We have already noted that the Hausdorff measure is translation invariant. Similarly, it is also invariant under orthogonal transformations (since the diameter is). Moreover, using the fact that for \( \lambda > 0 \) the map \( \lambda : x \mapsto \lambda x \) gives rise to a bijection between \( \delta \)-covers and \( (\delta/\lambda) \)-covers, we easily obtain the following scaling property of Hausdorff measures.

**Lemma 11.24.** Let \( \lambda > 0, d \in \mathbb{R}^n, O \in O(n) \) an orthogonal matrix, and \( A \) be a Borel set of \( \mathbb{R}^n \). Then

\[
h^n(\lambda OA + d) = \lambda^n h^n(A).
\] (11.28)

Moreover, Hausdorff measures also behave nicely under uniformly Hölder continuous maps.
Lemma 11.25. Suppose \( f : A \to \mathbb{R}^n \) is uniformly Hölder continuous with exponent \( \gamma > 0 \), that is,
\[
|f(x) - f(y)| \leq c|x - y|^\gamma \quad \text{for all } x, y \in A.
\] (11.29)
Then
\[
h^\alpha(f(A)) \leq c^\alpha h^\alpha(A). \tag{11.30}
\]
Proof. A simple consequence of the fact that for every \( \delta \)-cover \( \{A_j\} \) of a Borel set \( A \), the set \( \{f(A \cap A_j)\} \) is a \((c\delta^\gamma)\)-cover for the Borel set \( f(A) \). \( \Box \)

Now we are ready to define the Hausdorff dimension. First note that \( h^\alpha_\delta \) is non increasing with respect to \( \alpha \) for \( \delta < 1 \) and hence the same is true for \( h^\alpha \). Moreover, for \( \alpha \leq \beta \) we have \( \sum_j \text{diam}(A_j)^\beta \leq \delta^{\beta-\alpha} \sum_j \text{diam}(A_j)^\alpha \) and hence
\[
h^\beta_\delta(A) \leq \delta^{\beta-\alpha} h^\alpha_\delta(A) \leq \delta^{\beta-\alpha} h^\alpha(A). \tag{11.31}
\]
Thus if \( h^\alpha(A) \) is finite, then \( h^\beta(A) = 0 \) for every \( \beta > \alpha \). Hence there must be one value of \( \alpha \) where the Hausdorff measure of a set jumps from \( \infty \) to 0. This value is called the Hausdorff dimension
\[
\dim_H(A) = \inf \{ \alpha | h^\alpha(A) = 0 \} = \sup \{ \alpha | h^\alpha(A) = \infty \}. \tag{11.32}
\]
It is also not hard to see that we have \( \dim_H(A) \leq n \) (Problem 11.22).

The following observations are useful when computing Hausdorff dimensions. First the Hausdorff dimension is monotone, that is, for \( A \subseteq B \) we have \( \dim_H(A) \leq \dim_H(B) \). Furthermore, if \( A_j \) is a (countable) sequence of Borel sets we have \( \dim_H(\bigcup_j A_j) = \sup_j \dim_H(A_j) \) (show this).

Using Lemma 11.25 it is also straightforward to show

Lemma 11.26. Suppose \( f : A \to \mathbb{R}^n \) is uniformly Hölder continuous with exponent \( \gamma > 0 \), that is,
\[
|f(x) - f(y)| \leq c|x - y|^\gamma \quad \text{for all } x, y \in A,
\] (11.33)
then
\[
\dim_H(f(A)) \leq \frac{1}{\gamma} \dim_H(A). \tag{11.34}
\]
Similarly, if \( f \) is bi-Lipschitz, that is,
\[
a|x - y| \leq |f(x) - f(y)| \leq b|x - y| \quad \text{for all } x, y \in A,
\] (11.35)
then
\[
\dim_H(f(A)) = \dim_H(A). \tag{11.36}
\]

Example. The Hausdorff dimension of the Cantor set (see the example on page 258) is
\[
\dim_H(C) = \frac{\log(2)}{\log(3)}. \tag{11.37}
\]
To see this let \( \delta = 3^{-n} \). Using the \( \delta \)-cover given by the intervals forming \( C_n \) used in the construction of \( C \) we see \( h^\alpha_\delta(C) \leq \left( \frac{2}{3} \right)^n \). Hence for \( \alpha = d = \log(2)/\log(3) \) we have \( h^\alpha_\delta(C) \leq 1 \) implying \( \dim_H(C) \leq d \).

The reverse inequality is a little harder. Let \( \{A_j\} \) be a cover and \( \delta < \frac{1}{3} \). It is clearly no restriction to assume that all \( V_j \) are open intervals. Moreover, finitely many of these sets cover \( C \) by compactness. Drop all others and fix \( j \). Furthermore, increase each interval \( A_j \) by at most \( \varepsilon \).

For \( V_j \) there is a \( k \) such that
\[
\frac{1}{3^{k+1}} \leq |A_j| < \frac{1}{3^k}.
\]
Since the distance of two intervals in \( C_k \) is at least \( 3^{-k} \) we can intersect at most one such interval. For \( n \geq k \) we see that \( V_j \) intersects at most \( 2^{n-k} = 2^n(3^{-k})^d \leq 2^n 3^d |A_j|^d \) intervals of \( C_n \).

Now choose \( n \) larger than all \( k \) (for all \( A_j \)). Since \( \{A_j\} \) covers \( C \), we must intersect all \( 2^n \) intervals in \( C_n \). So we end up with
\[
2^n \leq \sum_j 2^n 3^d |A_j|^d,
\]
which together with our first estimate yields
\[
\frac{1}{2} \leq h^d(C) \leq 1.
\]
Observe that this result can also formally be derived from the scaling property of the Hausdorff measure by solving the identity
\[
h^\alpha(C) = h^\alpha(C \cap [0, \frac{1}{3}]) + h^\alpha(C \cap [\frac{2}{3}, 1]) = 2 h^\alpha(C \cap [0, \frac{1}{3}]))
= \frac{2}{3^\alpha} h^\alpha(3(C \cap [0, \frac{1}{3}])) = \frac{2}{3^\alpha} h^\alpha(C)
\]
for \( \alpha \). However, this is possible only if we already know that \( 0 < h^\alpha(C) < \infty \) for some \( \alpha \). \( \diamond \)

**Problem 11.20.** Suppose \( \{\mu^*_\alpha\}_\alpha \) is a family of outer measures on \( X \). Then \( \mu^* = \sup_\alpha \mu^*_\alpha \) is again an outer measure.

**Problem 11.21.** Let \( L = [0, 1] \times \{0\} \subseteq \mathbb{R}^2 \). Show that \( h^1(L) = 1 \).

**Problem 11.22.** Show that \( \dim_H(U) \leq n \) for every \( U \subseteq \mathbb{R}^n \).

### 11.5. Infinite product measures

In Section 9.2 we have dealt with finite products of measures. However, in some situations even infinite products are of interest. For example, in probability theory one describes a single random experiment is described by a probability measure and performing \( n \) independent trials is modeled
by taking the $n$-fold product. If one is interested in the behavior of certain quantities in the limit as $n \to \infty$ one is naturally lead to an infinite product.

Hence our goal is to to define a probability measure on the product space $\mathbb{R}^N = \times_N \mathbb{R}$. We can regard $\mathbb{R}^N$ as the set of all sequences $x = (x_j)_{j \in \mathbb{N}}$. A cylinder set in $\mathbb{R}^N$ is a set of the form $A \times \mathbb{R}^N \subseteq \mathbb{R}^N$ with $A \subseteq \mathbb{R}^n$ for some $n$. We equip $\mathbb{R}^N$ with the product topology, that is, the topology generated by open cylinder sets with $A$ open (which are a base for the product topology since they are closed under intersections — note that the cylinder sets are precisely the finite intersections of preimages of projections). Then the Borel $\sigma$-algebra $\mathcal{B}^N$ on $\mathbb{R}^N$ is the $\sigma$-algebra generated by cylinder sets with $A \in \mathcal{B}^n$.

Now suppose we have probability measures $\mu_n$ on $(\mathbb{R}^n, \mathcal{B}^n)$ which are consistent in the sense that $\mu_{n+1}(A \times \mathbb{R}^n) = \mu_n(A)$, $A \in \mathcal{B}^n$. (11.39)

Example. The prototypical example would be the case where $\mu$ is a probability measure on $(\mathbb{R}, \mathcal{B})$ and $\mu_n = \mu \otimes \cdots \otimes \mu$ is the $n$-fold product. Slightly more general, one could even take probability measures $\nu_j$ on $(\mathbb{R}, \mathcal{B})$ and consider $\mu_n = \nu_1 \otimes \cdots \otimes \nu_n$.

Theorem 11.27 (Kolmogorov extension theorem). Suppose that we have a consistent family of probability measures $(\mathbb{R}^n, \mathcal{B}^n, \mu_n)$, $n \in \mathbb{N}$. Then there exists a unique probability measure $\mu$ on $(\mathbb{R}^N, \mathcal{B}^N)$ such that $\mu(A \times \mathbb{R}^N) = \mu_n(A)$ for all $A \in \mathcal{B}^n$.

Proof. Consider the algebra $\mathcal{A}$ of all Borel cylinder sets which generates $\mathcal{B}^N$ as noted above. Then $\mu(A \times \mathbb{R}^N) = \mu_n(A)$ for $A \times \mathbb{R}^N \in \mathcal{A}$ defines an additive set function on $\mathcal{A}$. Indeed, by our consistency assumption different representations of a cylinder set will give the same value and (finite) additivity follows from additivity of $\mu_n$. Hence it remains to verify that $\mu$ is a premeasure such that we can apply the extension results from Section 8.3.

Now in order to show $\sigma$-additivity it suffices to show continuity from above, that is, for given sets $A_n \in \mathcal{A}$ with $A_n \searrow \emptyset$ we have $\mu(A_n) \searrow 0$. Suppose to the contrary that $\mu(A_n) \searrow \epsilon > 0$. Moreover, by repeating sets in the sequence $A_n$ if necessary, we can assume without loss of generality that $A_n = \tilde{A}_n \times \mathbb{R}^n$ with $\tilde{A}_n \subseteq \mathbb{R}^n$. Next, since $\mu_n$ is inner regular, we can find a compact set $\tilde{K}_n \subseteq \tilde{A}_n$ such that $\mu_n(\tilde{K}_n) \geq \frac{\epsilon}{2}$. Furthermore, since $A_n$ is decreasing we can arrange $\tilde{K}_n = \tilde{K}_n \times \mathbb{R}^n$ to be decreasing as well: $K_n \searrow \emptyset$. However, by compactness of $\tilde{K}_n$ we can find a sequence with $x \in K_n$ for all $n$ (Problem 11.23), a contradiction. $\square$
Example. The simplest example for the use of this theorem is a discrete random walk in one dimension. So we suppose we have a fictitious particle confined to the lattice $\mathbb{Z}$ which starts at 0 and moves one step to the left or right depending on whether a fictitious coin gives head or tail (the imaginative reader might also see the price of share here). Somewhat more formal, we take $\mu_1 = (1-p)\delta_{-1} + p\delta_1$ with $p \in [0,1]$ being the probability of moving right and $1-p$ the probability of moving left. Then the infinite product will give a probability measure for the sets of all paths $x \in \{-1,1\}^\mathbb{N}$ and one might try to answer questions like if the location of the particle at step $n$, $s_n = \sum_{j=1}^n x_j$ remains bounded for all $n \in \mathbb{N}$, etc.

Example. Another classical example is the Anderson model. The discrete one-dimensional Schrödinger equation for a single electron in an external potential is given by the difference operator $(Hu)_n = u_{n+1} + u_{n-1} + q_n u_n$, $u \in \ell^2(\mathbb{Z})$, where the potential $q_n$ is a bounded real-valued sequence. A simple model for an electron in a crystal (where the atoms are arranged in a periodic structure) is hence the case when $q_n$ is periodic. But what happens if you introduce some random impurities (known as doping in the context of semiconductors)? This can be modeled by $q_n = q_0^n + x_n(q_1^n - q_0^n)$ where $x \in \{0,1\}^\mathbb{N}$ and we can take $\mu_1 = (1-p)\delta_0 + p\delta_1$ with $p \in [0,1]$ the probability of an impurity being present.

Problem 11.23. Suppose $K_n \subseteq \mathbb{R}^n$ is a sequence of nonempty compact sets which are nesting in the sense that $K_{n+1} \subseteq K_n \times \mathbb{R}$. Show that there is a sequence $x = (x_j)_{j \in \mathbb{N}}$ with $(x_1, \ldots, x_n) \in K_n$ for all $n$. (Hint: Choose $x_m$ by considering the projection of $K_n$ onto the $m$’th coordinate and using the finite intersection property of compact sets.)

11.6. The Bochner integral

In this section we want to extend the Lebesgue integral to the case of functions with values in a normed space. This extension is known as Bochner integral. Since a normed space has no order we cannot use monotonicity and hence are restricted to finite values for the integral. Other than that, we only need some small adaptions.

Let $(X, \Sigma, \mu)$ be a measure space and $Y$ a Banach space equipped with the Borel $\sigma$-algebra $\mathcal{B}(Y)$. As in (9.1), a measurable function $s : X \to Y$ is called simple if its image is finite; that is, if

$$s = \sum_{j=1}^p \alpha_j \chi_{A_j}, \quad \text{Ran}(s) =: \{\alpha_j\}_{j=1}^p, \quad A_j := s^{-1}(\alpha_j) \in \Sigma.$$ (11.40)
Also the integral of a simple function can be defined as in (9.2) provided we ensure that it is finite. To this end we call \( s \) **integrable** if \( \mu(A_j) < \infty \) for all \( j \) with \( \alpha_j \neq 0 \). Now, for an integrable simple function \( s \) as in (11.40) we define its **Bochner integral** as

\[
\int_A s \, d\mu := \sum_{j=1}^{p} \alpha_j \mu(A_j \cap A).
\]  

(11.41)

As before we use the convention \( 0 \cdot \infty = 0 \) (where the 0 on the left is the zero vector from \( Y \)).

**Lemma 11.28.** The integral has the following properties:

(i) \( \int_A s \, d\mu = \int_X \chi_A s \, d\mu \).

(ii) \( \int_{\bigcup_{n=1}^{\infty} A_n} s \, d\mu = \sum_{n=1}^{\infty} \int_{A_n} s \, d\mu \).

(iii) \( \int_A \alpha s \, d\mu = \alpha \int_A s \, d\mu, \alpha \in \mathbb{C} \).

(iv) \( \int_A (s + t) \, d\mu = \int_A s \, d\mu + \int_A t \, d\mu \).

(v) \( \| \int_A s \, d\mu \| \leq \int_A \| s \| \, d\mu \).

**Proof.** The first four items follow literally as in Lemma 9.1. (v) follows from the triangle inequality. \( \square \)

Now we extend the integral via approximation by simple functions. However, while a complex-valued measurable function can always be approximated by simple functions, this might no longer be true in our present setting. In fact, note that a sequence of simple functions involves only a countable number of values from \( Y \) and since the limit must be in the closure of the span of these values, the range of \( f \) must be separable. Moreover, we also need to ensure finiteness of the integral.

If \( \mu \) is finite, the latter requirement is easily satisfied by considering only bounded functions. Consequently we could equip the space of simple functions \( S(X, \mu, Y) \) with the supremum norm \( \| s \|_{\infty} = \sup_{x \in X} \| s(x) \| \) and use the fact that the integral is a bounded linear functional,

\[
\| \int_A s \, d\mu \| \leq \mu(A) \| s \|_{\infty},
\]  

(11.42)

to extend it to the completion of the simple functions, known as the regulated functions \( R(X, \mu, Y) \). Hence the integrable functions will be the bounded functions which are uniform limits of simple functions. While this gives a theory suitable for many cases we want to do better and look at functions which are pointwise limits of simple functions.
11.6. The Bochner integral

Consequently, we call a function $f$ integrable if there is a sequence of integrable simple functions $s_n$ which converges pointwise to $f$ such that

$$\int_X \|f - s_n\| d\mu \to 0.$$  \hspace{1cm} (11.43)

In this case item (v) from Lemma 11.28 ensures that $\int_X s_n d\mu$ is a Cauchy sequence and we can define the **Bochner integral** of $f$ to be

$$\int_X f d\mu := \lim_{n \to \infty} \int_X s_n d\mu.$$  \hspace{1cm} (11.44)

If there are two sequences of integrable simple functions as in the definition, we could combine them into one sequence (taking one as even and the other one as odd elements) to conclude that the limit of the first two sequences equals the limit of the third sequence. In other words, the definition of the integral is independent of the sequence chosen.

**Lemma 11.29.** The integral is linear and Lemma 11.28 holds for integrable functions $s, t$.

**Proof.** All items except for (ii) are immediate. (ii) is also immediate for finite unions. The general case will follow from the dominated convergence theorem to be shown below. \hfill \Box

Before we proceed, we try to shed some light on when a function is integrable.

**Lemma 11.30.** A function $f : X \to Y$ is the pointwise limit of simple functions if and only if it is measurable and its range is separable. If its range is compact it is the uniform limit of simple functions. Moreover, the sequence $s_n$ can be chosen such that $\|s_n(x)\| \leq 2\|f(x)\|$ for every $x \in X$ and $\text{Ran}(s_n) \subseteq \text{Ran}(f) \cup \{0\}$.

**Proof.** Let $\{y_j\}_{j \in \mathbb{N}}$ be a dense set for the range. Note that the balls $B_{1/m}(y_j)$ will cover the range for every fixed $m \in \mathbb{N}$. Furthermore, we will augment this set by $y_0 = 0$ to make sure that any value less than $1/m$ is replaced by 0 (since otherwise one might destroy properties of $f$). By iteratively removing what has already been covered we get a disjoint cover $A_j^m \subseteq B_{1/m}(y_j)$ (some sets might be empty) such that $A_{n,m} := \bigcup_{j \leq n} A_j^m = \bigcup_{j \leq n} B_{1/m}(y_j)$. Now set $A_n := A_{n,n}$ and consider the simple function $s_n$ defined as follows

$$s_n = \begin{cases} y_j, & \text{if } f(x) \in A_j^m \setminus \bigcup_{m<k \leq n} A_k, \\ 0, & \text{else}. \end{cases}$$

That is, we first search for the largest $m \leq n$ with $f(x) \in A_m$ and look for the unique $j$ with $f(x) \in A_j^m$. If such an $m$ exists we have $s_n(x) = y_j$
and otherwise \( s_n(x) = 0 \). By construction \( s_n \) is measurable and converges pointwise to \( f \). Moreover, to see \( \|s_n(x)\| \leq 2\|f(x)\| \) we consider two cases. If \( s_n(x) = 0 \) then the claim is trivial. If \( s_n(x) \neq 0 \) then \( f(x) \in A^m_j \) with \( j > 0 \) and hence \( \|f(x)\| \geq 1/m \) implying

\[
\|s_n(x)\| = \|y_j\| \leq \|y_j - f(x)\| + \|f(x)\| \leq \frac{1}{m} + \|f(x)\| \leq 2\|f(x)\|.
\]

If the range of \( f \) is compact, we can find a finite index \( N_m \) such that \( A^m_{N_m,m} \) covers the whole range. Now our definition simplifies since the largest \( m \leq n \) with \( f(x) \in A_m \) is always \( n \). In particular, \( \|s_n(x) - f(x)\| \leq \frac{1}{n} \) for all \( x \).

Functions \( f : X \to Y \) which are the pointwise limit of simple functions are also called **strongly measurable**. Since the limit of measurable functions is again measurable (Problem 11.24), the limit of strongly measurable functions is strongly measurable. Moreover, our lemma also shows that continuous functions on separable spaces are strongly measurable (recall Lemma 8.17 and Problem 0.13).

Now we are in the position to give a useful characterization of integrability.

**Lemma 11.31.** A function \( f : X \to Y \) is integrable if and only if it is strongly measurable and \( \|f(x)\| \) is integrable.

**Proof.** We have already seen that an integrable function has the stated properties. Conversely, the sequence \( s_n \) from the previous lemma satisfies (11.43) by the dominated convergence theorem. \( \square \)

Another useful observation is the fact that the integral behaves nicely under bounded linear transforms.

**Theorem 11.32** (Hille). Let \( A : \mathcal{D}(A) \subseteq Y \to Z \) be closed. Suppose \( f : X \to Y \) is integrable with \( \text{Ran}(f) \subseteq \mathcal{D}(A) \) and \( Af \) is also integrable. Then

\[
A \int_X f \, d\mu = \int_X (Af) \, d\mu, \quad f \in \mathcal{D}(A).
\]  

(11.45)

If \( A \in \mathcal{L}(Y,Z) \), then \( f \) integrable implies \( Af \) is integrable.

**Proof.** By assumption \( (f,Af) : X \to Y \times Z \) is integrable and by the proof of Lemma 11.31 the sequence of simple functions can be chosen to have its range in the graph of \( A \). In other words, there exists a sequence of simple functions \( (s_n, As_n) \) such that

\[
\int_X s_n \, d\mu \to \int_X f \, d\mu, \quad A \int_X s_n \, d\mu = \int_X As_n \, d\mu \to \int_X Af \, d\mu.
\]

Hence the claim follows from closedness of \( A \).
11.6. The Bochner integral

If $A$ is bounded and $s_n$ is a sequence of simple functions satisfying (11.43), then $t_n = As_n$ is a corresponding sequence of simple functions for $Af$ since $\|t_n - Af\| \leq \|A\| \|s_n - f\|$. This shows the last claim. \qed

Next, we note that the dominated convergence theorem holds for the Bochner integral.

**Theorem 11.33.** Suppose $f_n$ are integrable and $f_n \to f$ pointwise. Moreover suppose there is an integrable function $g$ with $\|f_n(x)\| \leq \|g(x)\|$. Then $f$ is integrable and

$$\lim_{n \to \infty} \int_X f_n d\mu = \int_X f d\mu \quad (11.46)$$

**Proof.** If $s_{n,m}(x) \to f_n(x)$ as $m \to \infty$ then $s_{n,n}(x) \to f(x)$ as $n \to \infty$ which together with $\|f(x)\| \leq \|g(x)\|$ shows that $f$ is integrable. Moreover, the usual dominated convergence theorem shows $\int_X \|f_n - f\| d\mu \to 0$ from which the claim follows. \qed

There is also yet another useful characterization of strong measurability. To this end we call a function $f : X \to Y$ **weakly measurable** if $\ell \circ f$ is measurable for every $\ell \in Y^*$.

**Theorem 11.34 (Pettis).** A function $f : X \to Y$ is strongly measurable if and only if it is weakly measurable and its range is separable.

**Proof.** Since every measurable function is weakly measurable, Lemma 11.30 shows one direction. To see the converse direction let $\{y_k\}_{k \in \mathbb{N}} \subseteq f(X)$ be dense. Define $s_n : f(X) \to \{y_1, \ldots, y_n\}$ via $s_n(y) = y_k$, where $k = k_{y,n}$ is the smallest $k$ such that $\|y - y_k\| = \min_{1 \leq j \leq n} \|y - y_j\|$. By density we have $\lim_{n \to \infty} s_n(y) = y$ for every $y \in f(X)$. Now set $f_n = s_n \circ f$ and note that $f_n \to f$ pointwise. Moreover, for $1 \leq k \leq n$ we have

$$f_n^{-1}(y_k) = \{x \in X : \|f(x) - y_k\| = \min_{1 \leq j \leq n} \|f(x) - y_j\|\} \cap \{x \in X : \|f(x) - y_l\| < \min_{1 \leq j \leq n} \|f(x) - y_j\|, 1 \leq l < k\}$$

and $f_n$ will be measurable once we show that $\|f - y\|$ is measurable for every $y \in f(Y)$. To show this last claim choose (Problem 11.25) a countable set $\{y'_k\} \in Y^*$ of unit vectors such that $\|y\| = \sup_k y'_k(y)$. Then $\|f - y\| = \sup_k y'_k(f - y)$ from which the claim follows. \qed

We close with the remark that since the integral does not see null sets, one could also work with functions which satisfy the above requirements only away from null sets.
Finally, we briefly discuss the associated Lebesgue spaces. As in the complex-valued case we define the $L^p$ norm by

$$\|f\|_p := \left( \int_X \|f\|^p \, d\mu \right)^{1/p}, \quad 1 \leq p,$$

and denote by $L^p(X, d\mu, Y)$ the set of all strongly measurable functions for which $\|f\|_p$ is finite. Note that $L^p(X, d\mu, Y)$ is a vector space, since

$$\|f + g\|^p \leq 2^p \max(\|f\|^p, \|g\|^p) = 2^p \max(\|f\|^p, \|g\|^p) \leq 2^p (\|f\|^p + \|g\|^p).$$

Again Lemma 9.6 (which still holds in this situation) implies that we need to identify functions which are equal almost everywhere: Let

$$N(X, d\mu, Y) := \{ f \text{ strongly measurable} | f(x) = 0 \ \mu-\text{almost everywhere} \}$$

and consider the quotient space

$$L^p(X, d\mu, Y) := L^p(X, d\mu, Y) / N(X, d\mu, Y).$$

If $d\mu$ is the Lebesgue measure on $X \subseteq \mathbb{R}^n$, we simply write $L^p(X, Y)$. Observe that $\|f\|_p$ is well defined on $L^p(X, d\mu, Y)$ and hence we have a normed space (the triangle will be established below).

**Lemma 11.35.** The integrable simple functions are dense in $L^p(X, d\mu, Y)$.

**Proof.** Let $f \in L^p(X, d\mu, Y)$. By Lemma 11.30 there is a sequence of simple functions such that $s_n(x) \to f(x)$ pointwise and $\|s_n(x)\| \leq 2\|f(x)\|$. In particular, $s_n \in L^p(X, d\mu, Y)$ and thus integrable since $\|s_n\|^p = \|s_n\|_1$. Moreover, $\|f(x) - s_n(x)\|^p \leq 3^p \|f(x)\|^p$ and thus $s_n \to f$ in $L^p(X, d\mu, Y)$ by dominated convergence. \qed

Similarly we define $L^\infty(X, d\mu, Y)$ together with the essential supremum

$$\|f\|_\infty := \inf \{ C | \mu(\{x | \|f(x)\| > C\}) = 0 \}.$$

From Hölder’s inequality for complex-valued functions we immediately get the following generalized version:

**Theorem 11.36** (Hölder’s inequality). Let $p$ and $q$ be dual indices, $\frac{1}{p} + \frac{1}{q} = 1$, with $1 \leq p \leq \infty$. If

(i) $f \in L^p(X, d\mu, Y^*)$ and $g \in L^q(X, d\mu, Y)$, or

(ii) $f \in L^p(X, d\mu)$ and $g \in L^q(X, d\mu, Y)$, or

(iii) $f \in L^p(X, d\mu, Y)$ and $g \in L^q(X, d\mu, Y)$ and $Y$ is a Banach algebra,

then $fg$ is integrable and

$$\|fg\|_1 \leq \|f\|_p \|g\|_q.$$

As a consequence we get
Corollary 11.37 (Minkowski’s inequality). Let \( f, g \in L^p(X, d\mu, Y) \), \( 1 \leq p \leq \infty \). Then
\[
\| f + g \|_p \leq \| f \|_p + \| g \|_p.
\] (11.52)

Proof. Since the cases \( p = 1, \infty \) are straightforward, we only consider \( 1 < p < \infty \). Using \( \| f(x) + g(x) \|_p \leq \| f(x) \|_p \| f(x) + g(x) \|^{p-1} + \| g(x) \| \| f(x) + g(x) \|^{p-1} \) we obtain from Hölder’s inequality (note \( (p-1)q = p \))
\[
\| f + g \|_p^p \leq \| f \|_p \| f + g \|^{p-1}_q + \| g \|_p \| f + g \|^{p-1}_q = (\| f \|_p + \| g \|_p) \| f + g \|^{p-1}_p.
\]
\( \square \)

Moreover, literally the same proof as for the complex-valued case shows:

Theorem 11.38 (Riesz–Fischer). The space \( L^p(X, d\mu, Y) \), \( 1 \leq p \leq \infty \), is a Banach space.

Of course, if \( Y \) is a Hilbert space, then \( L^2(X, d\mu, Y) \) is a Hilbert space with scalar product
\[
\langle f, g \rangle = \int_X \langle f(x), g(x) \rangle d\mu(x).
\] (11.53)

Problem 11.24. Let \( Y \) be a metric space equipped with the Borel \( \sigma \)-algebra. Show that the pointwise limit \( f : X \to Y \) of measurable functions \( f_n : X \to Y \) is measurable. (Hint: Show that for \( U \subseteq Y \) open we have that \( f^{-1}(U) = \bigcup_{m=1}^{\infty} \bigcap_{n=1}^{\infty} f_k^{-1}(U_m) \), where \( U_m := \{ y \in U : d(y, Y \setminus U) > \frac{1}{m} \} \).)

Problem 11.25. Let \( X \) be a separable Banach space. Show that there is a countable set \( \ell_k \in X^* \) such that \( \| x \| = \sup_k |\ell_k(x)| \) for all \( x \).

Problem 11.26. Let \( Y = C[a,b] \) and \( f : [0,1] \to Y \). Compute the Bochner integral \( \int_0^1 f(x) dx \).

11.7. Weak and vague convergence of measures

In this section \( X \) will be a metric space equipped with the Borel \( \sigma \)-algebra. We say that a sequence of finite Borel measures \( \mu_n \) converges weakly to a finite Borel measure \( \mu \) if
\[
\int_X f d\mu_n \to \int_X f d\mu \quad (11.54)
\]
for every \( f \in C_b(X) \). Since by the Riesz representation theorem the set of (complex) measures is the dual of \( C(X) \) (under appropriate assumptions on \( X \)), this is what would be denoted by weak*-convergence in functional analysis. However, we will not need this connection here. Nevertheless we remark that the weak limit is unique. To see this let \( C \) be a nonempty closed set and consider
\[
f_n(x) := \max(0, 1 - n \text{dist}(x, C)).
\] (11.55)
Then $f \in C_b(X)$, $f_n \downarrow \chi_C$, and (dominated convergence)

$$
\mu(C) = \lim_{n \to \infty} \int_X g_n \, d\mu
$$

(11.56)

shows that $\mu$ is uniquely determined by the integral for continuous functions (recall Lemma 8.20 and its corollary). For later use observe that $f_n$ is even Lipschitz continuous, $|f_n(x) - f_n(y)| \leq n \, d(x, y)$ (cf. Lemma 0.26).

Moreover, choosing $f \equiv 1$ shows

$$
\mu(X) = \lim_{n \to \infty} \mu_n(X).
$$

(11.57)

However, this might not be true for arbitrary sets in general. To see this look at the case $X = \mathbb{R}$ with $\mu_n = \delta_{1/n}$. Then $\mu_n$ converges weakly to $\mu = \delta_0$. But $\mu_n((0, 1)) = 1 \not\to 0 = \mu((0, 1))$ as well as $\mu_n([-1, 0]) = 0 \not\to 1 = \mu([-1, 0])$. So mass can appear/disappear at boundary points but this is the worst that can happen:

**Theorem 11.39 (Portmanteau).** Let $X$ be a metric space and $\mu_n, \mu$ finite Borel measures. The following are equivalent:

(i) $\mu_n$ converges weakly to $\mu$.

(ii) $\int_X f \, d\mu_n \to \int_X f \, d\mu$ for every bounded Lipschitz continuous $f$.

(iii) $\limsup_n \mu_n(C) \leq \mu(C)$ for all closed sets $C$ and $\mu_n(X) \to \mu(X)$.

(iv) $\liminf_n \mu_n(O) \geq \mu(O)$ for all open sets $O$ and $\mu_n(X) \to \mu(X)$.

(v) $\mu_n(A) \to \mu(A)$ for all Borel sets $A$ with $\mu(\partial A) = 0$.

(vi) $\int_X f \, d\mu_n \to \int_X f \, d\mu$ for all bounded functions $f$ which are continuous at $\mu$-a.e. $x \in X$.

**Proof.** (i) $\Rightarrow$ (ii) is trivial. (ii) $\Rightarrow$ (iii): Define $f_n$ as in (11.55) and start by observing

$$
\limsup_n \mu_n(C) = \limsup_n \int_X \chi_F \mu_n \leq \limsup_n \int_X f_n \, d\mu_n = \int_X f_n \, d\mu.
$$

Now taking $m \to \infty$ establishes the claim. Moreover, $\mu_n(X) \to \mu(X)$ follows choosing $f \equiv 1$. (iii) $\Leftrightarrow$ (iv): Just use $O = X \setminus C$ and $\mu_n(O) = \mu_n(X) - \mu_n(C)$. (iii) and (iv) $\Rightarrow$ (v): By $A^\circ \subseteq A \subseteq \overline{A}$ we have

$$
\limsup_n \mu_n(A) \leq \limsup_n \mu_n(\overline{A}) \leq \mu(\overline{A})
$$

$$
= \mu(A^\circ) \leq \liminf_n \mu_n(A^\circ) \leq \liminf_n \mu_n(A)
$$

provided $\mu(\partial A) = 0$. (v) $\Rightarrow$ (vi): By considering real and imaginary parts separately we can assume $f$ to be real-valued. moreover, adding an appropriate constant we can even assume $0 \leq f_n \leq M$. Set $A_r = \{x \in X | f(x) > r\}$ and denote by $D_f$ the set of discontinuities of $f$. Then $\partial A_r \subseteq D_f \cup \{x \in$
X|f(x) = r}. Now the first set has measure zero by assumption and the second set is countable by Problem 9.19. Thus the set of all \( r \) with \( \mu(\partial A_r) > 0 \) is countable and thus of Lebesgue measure zero. Then by Problem 9.19 (with \( \phi(r) = 1 \)) and dominated convergence
\[
\int_X f \, d\mu_n = \int_0^M \mu_n(A_r) \, dr \to \int_0^M \mu(A_r) \, dr = \int_X f \, d\mu
\]
Finally, (vi) \( \Rightarrow \) (i) is trivial.

Next we want to extend our considerations to unbounded measures. In this case boundedness of \( f \) will not be sufficient to guarantee integrability and hence we will require \( f \) to have compact support. If \( x \in X \) is such that \( f(x) \neq 0 \), then \( f^{-1}(B_r(f(x))) \) will be a relatively compact neighborhood of \( x \) whenever \( 0 < r < |f(x)| \). Hence \( C_b(X) \) will not have sufficiently many functions with compact support unless we assume \( X \) to be locally compact, which we will do for the remainder of this section.

Let \( \mu_n \) be a sequence of Borel measures on a locally compact metric space \( X \). We will say that \( \mu_n \) converges vaguely to a Borel measure \( \mu \) if
\[
\int_X f \, d\mu_n \to \int_X f \, d\mu \quad (11.58)
\]
for every \( f \in C_c(X) \). As with weak convergence (cf. Problem 12.3) we can conclude that the integral over functions with compact supports determines \( \mu(K) \) for every compact set. Hence the vague limit will be unique if \( \mu \) is inner regular (which we already know to always hold if \( X \) is locally compact and separable by Corollary 8.22).

We first investigate the connection with weak convergence.

**Lemma 11.40.** Let \( X \) be a locally compact separable metric space and suppose \( \mu_n \to \mu \) vaguely. Then \( \mu(X) \leq \liminf_n \mu_n(X) \) and (11.58) holds for every \( f \in C_0(X) \) in case \( \mu_n(X) \leq M \). If in addition \( \mu_n(X) \to \mu(X) \), then (11.58) holds for every \( f \in C_b(X) \).

**Proof.** For every compact set \( K_m \) we can find a nonnegative function \( g_m \in C_c(X) \) which is one on \( K_m \) by Urysohn’s lemma. Hence \( \mu(K_m) \leq \int g_m \, d\mu = \lim_n \int g_m \, d\mu_n \leq \liminf_n \mu_n(X) \). Letting \( K_m \nearrow X \) shows \( \mu(X) \leq \liminf_n \mu_n(X) \).

Next, let \( f \in C_0(X) \) and fix \( \varepsilon > 0 \). Then there is a compact set \( K \) such that \( |f(x)| \leq \varepsilon \) for \( x \in X \setminus K \). Choose \( g \) for \( K \) as before and set \( f = f_1 + f_2 \) with \( f_1 = gf \). Then \( |\int f \, d\mu - \int f \, d\mu_n| \leq |\int f_1 \, d\mu - \int f_1 \, d\mu_n| + 2\varepsilon M \) and the first claim follows.

Similarly, for the second claim, let \( |f| \leq C \) and choose a compact set \( K \) such that \( \mu(X \setminus K) < \frac{\varepsilon}{2} \). Then we have \( \mu_n(X \setminus K) < \varepsilon \) for \( n \geq N \). Choose \( g \) for \( K \) as before and set \( f = f_1 + f_2 \) with \( f_1 = gf \). Then \( |\int f \, d\mu - \int f \, d\mu_n| \leq |\int f_1 \, d\mu - \int f_1 \, d\mu_n| + 2\varepsilon C \) and the second claim follows. \( \square \)
Example. The example $X = \mathbb{R}$ with $\mu_n = \delta_n$ shows that in the first claim $f$ cannot be replaced by a bounded continuous function. Moreover, the example $\mu_n = n \delta_n$ also shows that the uniform bound cannot be dropped. ○

The analog of Theorem 11.39 reads as follows.

**Theorem 11.41.** Let $X$ be a locally compact metric space and $\mu_n, \mu$ Borel measures. The following are equivalent:

1. $\mu_n$ converges vaguely to $\mu$.
2. $\int_X f \, d\mu_n \to \int_X f \, d\mu$ for every Lipschitz continuous $f$ with compact support.
3. $\limsup_n \mu_n(K) \leq \mu(K)$ for all compact sets $K$ and $\liminf_n \mu_n(O) \geq \mu(O)$ for all relatively compact open sets $O$.
4. $\mu_n(A) \to \mu(A)$ for all relative compact Borel sets $A$ with $\mu(\partial A) = 0$.
5. $\int_X f \, d\mu_n \to \int_X f \, d\mu$ for all bounded functions $f$ with compact support which are continuous at $\mu$-a.e. $x \in X$.

**Proof.** (i) $\Rightarrow$ (ii) is trivial. (ii) $\Rightarrow$ (iii): The claim for compact sets follows as in the proof of Theorem 11.39. To see the case of open sets let $K_n = \{x \in X \mid \text{dist}(x, X \setminus O) \geq n^{-1}\}$. Then $K_n \subseteq O$ is compact and we can look at

$$g_n(x) := \frac{\text{dist}(x, X \setminus O)}{\text{dist}(x, X \setminus O) + \text{dist}(x, K_n)}.$$ 

Then $g_n$ is supported on $O$ and $g_n \rightharpoonup \chi_O$. Then

$$\liminf_n \mu_n(O) = \liminf_n \int \chi_O \mu_n \geq \liminf_n \int g_m \mu_n = \int g_m \mu.$$ 

Now taking $m \to \infty$ establishes the claim.

The remaining directions follow literally as in the proof of Theorem 11.39 (concerning (iv) $\Rightarrow$ (v) note that $A_r$ is relatively compact).

Finally we look at Borel measures on $\mathbb{R}$. In this case, we have the following equivalent characterization of vague convergence.

**Lemma 11.42.** Let $\mu_n$ be a sequence of Borel measures on $\mathbb{R}$. Then $\mu_n \to \mu$ vaguely if and only if the distribution functions (normalized at a point of continuity of $\mu$) converge at every point of continuity of $\mu$.

**Proof.** Suppose $\mu_n \to \mu$ vaguely. Then the distribution functions converge at every point of continuity of $\mu$ by item (iv) of Theorem 11.41.

Conversely, suppose that the distribution functions converge at every point of continuity of $\mu$. To see that in fact $\mu_n \to \mu$ vaguely, let $f \in C_c(\mathbb{R})$. Fix some $\varepsilon > 0$ and note that, since $f$ is uniformly continuous, there is a
11.7. Weak and vague convergence of measures

\[ \delta > 0 \text{ such that } |f(x) - f(y)| \leq \varepsilon \text{ whenever } |x - y| \leq \delta. \]
Next, choose some points \( x_0 < x_1 < \cdots < x_k \) such that \( \text{supp}(f) \subset (x_0, x_k) \), \( \mu \) is continuous at \( x_j \), and \( x_j - x_{j-1} \leq \delta \) (recall that a monotone function has at most countable discontinuities). Furthermore, there is some \( N \) such that \( |\mu_n(x_j) - \mu(x_j)| \leq \frac{\varepsilon}{2k} \) for all \( j \) and \( n \geq N \). Then

\[
\left| \int f \, d\mu_n - \int f \, d\mu \right| \leq \sum_{j=1}^{k} \int_{(x_{j-1}, x_j]} |f(x) - f(x_j)| \, d\mu_n(x) \\
+ \sum_{j=1}^{k} |f(x_j)||\mu((x_{j-1}, x_j]) - \mu_n((x_{j-1}, x_j])| \\
+ \sum_{j=1}^{k} \int_{(x_{j-1}, x_j]} |f(x) - f(x_j)| \, d\mu(x).
\]

Now, for \( n \geq N \), the first and the last terms on the right-hand side are both bounded by \( (\mu((x_0, x_k]) + \frac{\varepsilon}{k})\varepsilon \) and the middle term is bounded by \( \max |f|\varepsilon \).
Thus the claim follows. \( \square \)

Moreover, every bounded sequence of measures has a vaguely convergent subsequence (this is a special case of Helly’s selection theorem — a generalization will be provided in Theorem 12.11).

**Lemma 11.43.** Suppose \( \mu_n \) is a sequence of finite Borel measures on \( \mathbb{R} \) such that \( \mu_n(\mathbb{R}) \leq M \). Then there exists a subsequence \( n_j \) which converges vaguely to some measure \( \mu \) with \( \mu(\mathbb{R}) \leq \lim \inf \mu_{n_j}(\mathbb{R}) \).

**Proof.** Let \( \mu_n(x) = \mu_n((-\infty, x]) \) be the corresponding distribution functions. By \( 0 \leq \mu_n(x) \leq M \) there is a convergent subsequence for fixed \( x \). Moreover, by the standard diagonal series trick (cf. Theorem 0.37), we can assume that \( \mu_n(y) \) converges to some number \( \mu(y) \) for each rational \( y \). For irrational \( x \) we set \( \mu(x) = \inf \{ \mu(y) | x < y \in \mathbb{Q} \} \). Then \( \mu(x) \) is monotone, \( 0 \leq \mu(x_1) \leq \mu(x_2) \leq M \) for \( x_1 < x_2 \). Indeed for \( x_1 \leq y_1 < x_2 \leq y_2 \) with \( y_j \in \mathbb{Q} \) we have \( \mu(x_1) \leq \mu(y_1) = \lim n \mu_n(y_1) \leq \lim n \mu_n(y_2) = \mu(y_2) \). Taking the infimum over \( y_2 \) gives the result.

Furthermore, for every \( \varepsilon > 0 \) and \( x - \varepsilon < y_1 \leq x \leq y_2 < x + \varepsilon \) with \( y_j \in \mathbb{Q} \) we have

\[
\mu(x-\varepsilon) \leq \lim_n \mu_n(y_1) \leq \lim \inf \mu_n(x) \leq \lim \sup \mu_n(x) \leq \lim_n \mu_n(y_2) \leq \mu(x+\varepsilon)
\]
and thus

\[
\mu(x-) \leq \lim \inf \mu_n(x) \leq \lim \sup \mu_n(x) \leq \mu(x+)
\]
which shows that \( \mu_n(x) \to \mu(x) \) at every point of continuity of \( \mu \). So we can redefine \( \mu \) to be right continuous without changing this last fact. The
bound for \( \mu(\mathbb{R}) \) follows since for every point of continuity \( x \) of \( \mu \) we have \( \mu(x) = \lim_n \mu_n(x) \leq \liminf_n \mu_n(\mathbb{R}) \).

**Example.** The example \( d\mu_n(x) = d\Theta(x - n) \) for which \( \mu_n(x) = \Theta(x - n) \to 0 \) shows that we can have \( \mu(\mathbb{R}) = 0 < 1 = \mu_n(\mathbb{R}) \).

**Problem 11.27.** Suppose \( \mu_n \to \mu \) vaguely and let \( I \) be a bounded interval with boundary points \( x_0 \) and \( x_1 \). Then

\[
\limsup_n \left| \int_I f d\mu_n - \int_I f d\mu \right| \leq |f(x_1)|\mu(\{x_1\}) + |f(x_0)|\mu(\{x_0\})
\]

for any \( f \in C([x_0, x_1]) \).

**Problem 11.28.** Let \( \mu_n(X) \leq M \) and suppose (11.58) holds for all \( f \in U \subseteq C(X) \). Then (11.58) holds for all \( f \) in the closed span of \( U \).

**Problem 11.29.** Let \( X \) be a proper metric space. A sequence of finite measures is called **tight** if for every \( \varepsilon > 0 \) there is a compact set \( K \subseteq X \) such that \( \sup_n \mu_n(X \setminus K) \leq \varepsilon \). Show that a vaguely convergent sequence is tight if and only if \( \mu_n(X) \to \mu(X) \).

**Problem 11.30.** Let \( \mu_n(\mathbb{R}), \mu(\mathbb{R}) \leq M \) and suppose the Cauchy transforms converge

\[
\int_{\mathbb{R}} \frac{1}{x-z}d\mu_n(x) \to \int_{\mathbb{R}} \frac{1}{x-z}d\mu(x)
\]

for \( z \in U \), where \( U \subseteq \mathbb{C} \setminus \mathbb{R} \) is a set which has a limit point. Then \( \mu_n \to \mu \) vaguely. (Hint: Problem 0.30.)

### 11.8. Appendix: Functions of bounded variation and absolutely continuous functions

Let \([a, b] \subseteq \mathbb{R}\) be some compact interval and \( f : [a, b] \to \mathbb{C} \). Given a partition \( P = \{a = x_0, \ldots, x_n = b\} \) of \([a, b]\) we define the **variation** of \( f \) with respect to the partition \( P \) by

\[
V(P, f) := \sum_{k=1}^{n} |f(x_k) - f(x_{k-1})|.
\]  

(11.59)

Note that the triangle inequality implies that adding points to a partition increases the variation: if \( P_1 \subseteq P_2 \) then \( V(P_1, f) \leq V(P_2, f) \). The supremum over all partitions

\[
V_a^b(f) := \sup_{\text{partitions } P \text{ of } [a, b]} V(P, f)
\]

(11.60)

is called the **total variation** of \( f \) over \([a, b] \). If the total variation is finite, \( f \) is called of **bounded variation.** Since we clearly have

\[
V_a^b(\alpha f) = |\alpha|V_a^b(f), \quad V_a^b(f + g) \leq V_a^b(f) + V_a^b(g)
\]

(11.61)
the space $BV[a,b]$ of all functions of finite total variation is a vector space. However, the total variation is not a norm since (consider the partition $P = \{a, x, b\}$)

$$V^b_a(f) = 0 \iff f(x) \equiv c.$$  \hfill (11.62)

Moreover, any function of bounded variation is in particular bounded (consider again the partition $P = \{a, x, b\}$)

$$\sup_{x \in [a,b]} |f(x)| \leq |f(a)| + V^b_a(f).$$  \hfill (11.63)

**Theorem 11.44.** The functions of bounded variation $BV[a,b]$ together with the norm

$$\|f\|_{BV} := |f(a)| + V^b_a(f)$$  \hfill (11.64)

are a Banach space. Moreover, by (11.63) we have $\|f\|_{\infty} \leq \|f\|_{BV}$.

**Proof.** By (11.62) we have $\|f\|_{BV} = 0$ if and only if $f$ is constant and $|f(a)| = 0$, that is $f = 0$. Moreover, by (11.61) the norm is homogenous and satisfies the triangle inequality. So let $f_n$ be a Cauchy sequence. Then $f_n$ converges uniformly and pointwise to some bounded function $f$. Moreover, choose $N$ such that $\|f_n - f_m\|_{BV} < \varepsilon$ whenever $m, n \geq N$. Then for $n \geq N$ and for any fixed partition

$$|f(a) - f_n(a)| + V(P, f - f_n) = \lim_{m \to \infty} \left( |f_m(a) - f_n(a)| + V(P, f_m - f_n) \right)$$

$$\leq \sup_{m \geq N} \|f_n - f_m\|_{BV} < \varepsilon.$$

Consequently $\|f - f_n\|_{BV} < \varepsilon$ which shows $f \in BV[a,b]$ as well as $f_n \to f$ in $BV[a,b]$. \hfill \Box

Observe $V^a_a(f) = 0$ as well as (Problem 11.32)

$$V^b_a(f) = V^c_a(f) + V^b_c(f), \quad c \in [a,b],$$  \hfill (11.65)

and it will be convenient to set

$$V^a_b(f) = -V^b_a(f).$$  \hfill (11.66)

**Example.** Every Lipschitz continuous function is of bounded variation. In fact, if $|f(x) - f(y)| \leq L|x - y|$ for $x, y \in [a,b]$, then $V^b_a(f) \leq L(b - a)$. However, (Hölder) continuity is not sufficient (cf. Problems 11.33 and 11.34). \hfill \diamond

**Example.** By the inverse triangle inequality we have

$$V^b_a(|f|) \leq V^b_a(f)$$

whereas the converse is not true: The function $f : [0,1] \to \{-1,1\}$ which is 1 on the rational and $-1$ on the irrational numbers satisfies $V^1_0(f) = \infty$ (show this) and $V^1_0(|f|) = V^1_0(1) = 0$. 

From $2^{-1/2}(|\operatorname{Re}(z)| + |\operatorname{Im}(z)|) \leq |z| \leq |\operatorname{Re}(z)| + |\operatorname{Im}(z)|$ we infer
$2^{-1/2} (V^b_a(\operatorname{Re}(f)) + V^a_b(\operatorname{Im}(f))) \leq V^b_a(f) \leq V^b_a(\operatorname{Re}(f)) + V^a_b(\operatorname{Im}(f))$
which shows that $f$ is of bounded variation if and only if $\operatorname{Re}(f)$ and $\operatorname{Im}(f)$ are.

**Example.** Any real-valued nondecreasing function $f$ is of bounded variation with variation given by $V^b_a(f) = f(b) - f(a)$. Similarly, every real-valued nonincreasing function $g$ is of bounded variation with variation given by $V^a_b(g) = g(a) - g(b)$. Moreover, the sum $f + g$ is of bounded variation with variation given by $V^b_a(f + g) \leq V^b_a(f) + V^a_b(g)$. The following theorem shows that the converse is also true.

**Theorem 11.45** (Jordan). Let $f : [a, b] \to \mathbb{R}$ be of bounded variation, then $f$ can be decomposed as

$$f(x) = f_+(x) - f_-(x), \quad f_\pm(x) := \frac{1}{2} (V^x_a(f) \pm f(x)),$$

(11.67)

where $f_\pm$ are nondecreasing functions. Moreover, $V^b_a(f_\pm) \leq V^b_a(f)$.

**Proof.** From

$$f(y) - f(x) \leq |f(y) - f(x)| \leq V^y_x(f) = V^y_x(f) - V^x_a(f)$$

for $x \leq y$ we infer $V^x_a(f) - f(x) \leq V^y_x(f) - f(y)$, that is, $f_+$ is nondecreasing. Moreover, replacing $f$ by $-f$ shows that $f_-$ is nondecreasing and the claim follows.

In particular, we see that functions of bounded variation have at most countably many discontinuities and at every discontinuity the limits from the left and right exist.

For functions $f : (a, b) \to \mathbb{C}$ (including the case where $(a, b)$ is unbounded) we will set

$$V^b_a(f) := \lim_{c \downarrow a, d \uparrow b} V^d_c(f).$$

(11.68)

In this respect the following lemma is of interest (whose proof is left as an exercise):

**Lemma 11.46.** Suppose $f \in BV[a, b]$. We have $\lim_{c \uparrow b} V^c_a(f) = V^b_a(f)$ if and only if $f(b) = f(b-)$ and $\lim_{c \downarrow a} V^b_c(f) = V^b_a(f)$ if and only if $f(a) = f(a+)$. In particular, $V^x_a(f)$ is left, right continuous if and only $f$ is.

If $f : \mathbb{R} \to \mathbb{C}$ is of bounded variation, then we can write it as a linear combination of four nondecreasing functions and hence associate a complex measure $df$ with $f$ via Theorem 8.13 (since all four functions are bounded, so are the associated measures).
Theorem 11.47. There is a one-to-one correspondence between functions in $f \in BV(\mathbb{R})$ which are right continuous and normalized by $f(0) = 0$ and complex Borel measures $\nu$ on $\mathbb{R}$ such that $f$ is the distribution function of $\nu$ as defined in (8.19). Moreover, in this case the distribution function of the total variation of $\nu$ is $|\nu|(x) = V^x_0(f)$.

Proof. We have already seen how to associate a complex measure $d\nu$ with a function of bounded variation. If $f$ is right continuous and normalized, it will be equal to the distribution function of $d\nu$ by construction. Conversely, let $d\nu$ be a complex measure with distribution function $\nu$. Then for every $a < b$ we have

$$V^b_a(\nu) = \sup_{P = \{a=x_0,...,x_n=b\}} V(P, \nu)$$

$$= \sup_{P = \{a=x_0,...,x_n=b\}} \sum_{k=1}^n |\nu((x_{k-1}, x_k])| \leq |\nu|(a,b]$$

and thus the distribution function is of bounded variation. Furthermore, consider the measure $\mu$ whose distribution function is $\mu(x) = V^x_0(\nu)$. Then we see $|\nu((a,b]| = |\nu(b) - \nu(a)| \leq V^b_a(\nu) = \mu((a,b]) \leq |\nu|(a,b])$. Hence we obtain $|\nu(A)| \leq \mu(A) \leq |\nu|(A)$ for all intervals $A$, thus for all open sets (by Problem 0.19), and thus for all Borel sets by outer regularity. Hence Lemma 11.13 implies $\mu = |\nu|$ and hence $|\nu|(x) = V^x_0(f)$. □

We will call a function $f : [a,b] \to \mathbb{C}$ absolutely continuous if for every $\varepsilon > 0$ there is a corresponding $\delta > 0$ such that

$$\sum_k |y_k - x_k| < \delta \implies \sum_k |f(y_k) - f(x_k)| < \varepsilon$$

(11.69)

for every countable collection of pairwise disjoint intervals $(x_k, y_k) \subset [a,b]$. The set of all absolutely continuous functions on $[a,b]$ will be denoted by $AC[a,b]$. The special choice of just one interval shows that every absolutely continuous function is (uniformly) continuous, $AC[a,b] \subset C[a,b]$.

Example. Every Lipschitz continuous function is absolutely continuous. In fact, if $|f(x) - f(y)| \leq L|x - y|$ for $x, y \in [a,b]$, then we can choose $\delta = \frac{\varepsilon}{L}$. In particular, $C^1[a,b] \subset AC[a,b]$. Note that Hölder continuity is neither sufficient (cf. Problem 11.34 and Theorem 11.49 below) nor necessary (cf. Problem 11.45).

Theorem 11.48. A complex Borel measure $\nu$ on $\mathbb{R}$ is absolutely continuous with respect to Lebesgue measure if and only if its distribution function is locally absolutely continuous (i.e., absolutely continuous on every compact subinterval). Moreover, in this case the distribution function $\nu(x)$ is
differentiable almost everywhere and
\[ \nu(x) = \nu(0) + \int_0^x \nu'(y) dy \]  \hspace{1cm} (11.70)

with \( \nu' \) integrable, \( \int_{\mathbb{R}} |\nu'(y)| dy = |\nu|(\mathbb{R}) \).

**Proof.** Suppose the measure \( \nu \) is absolutely continuous. Since we can write \( \nu \) as a sum of four positive measures, we can suppose \( \nu \) is positive. Now (11.69) follows from (11.25) in the special case where \( A \) is a union of pairwise disjoint intervals.

Conversely, suppose \( \nu(x) \) is absolutely continuous on \([a, b] \). We will verify (11.25). To this end fix \( \varepsilon \) and choose \( \delta \) such that \( \nu(x) \) satisfies (11.69). By outer regularity it suffices to consider the case where \( A \) is open. Moreover, by Problem 0.19, every open set \( O \subset (a, b) \) can be written as a countable union of disjoint intervals \( I_k = (x_k, y_k) \) and thus \( |O| = \sum_k |y_k - x_k| \leq \delta \) implies

\[ \nu(O) = \sum_k (\nu(y_k) - \nu(x_k)) \leq \sum_k |\nu(y_k) - \nu(x_k)| \leq \varepsilon \]

as required.

The rest follows from Corollary 11.8. \( \Box \)

As a simple consequence of this result we can give an equivalent definition of absolutely continuous functions as precisely the functions for which the **fundamental theorem of calculus** holds.

**Theorem 11.49.** A function \( f : [a, b] \rightarrow \mathbb{C} \) is absolutely continuous if and only if it is of the form

\[ f(x) = f(a) + \int_a^x g(y) dy \]  \hspace{1cm} (11.71)

for some integrable function \( g \). Moreover, in this case \( f \) is differentiable a.e with respect to Lebesgue measure and \( f'(x) = g(x) \). In addition, \( f \) is of bounded variation and

\[ V_a^x(f) = \int_a^x |g(y)| dy. \]  \hspace{1cm} (11.72)

**Proof.** This is just a reformulation of the previous result. To see the last claim combine the last part of Theorem 11.47 with Lemma 11.17. \( \Box \)

In particular, since the fundamental theorem of calculus fails for the Cantor function, this function is an example of a continuous which is not absolutely continuous. Note that even if \( f \) is differentiable everywhere it might fail the fundamental theorem of calculus (Problem 11.46).
Finally, we note that in this case the integration by parts formula continues to hold.

**Lemma 11.50.** Let \( f, g \in BV[a,b] \), then
\[
\int_{[a,b]} f(x-)dg(x) = f(b-)g(b-) - f(a-)g(a-) - \int_{[a,b]} g(x+)df(x) \tag{11.73}
\]
as well as
\[
\int_{[a,b]} f(x+)dg(x) = f(b+)g(b+) - f(a+)g(a+) - \int_{[a,b]} g(x-)df(x). \tag{11.74}
\]

**Proof.** Since the formula is linear in \( f \) and holds if \( f \) is constant, we can assume \( f(a-) = 0 \) without loss of generality. Similarly, we can assume \( g(b-) = 0 \). Plugging \( f(x-) = \int_{[a,x]} df(y) \) into the left-hand side of the first formula we obtain from Fubini
\[
\int_{[a,b]} f(x-)dg(x) = \int_{[a,b]} \int_{[a,x]} df(y)dg(x)
\]
\[
= \int_{[a,b]} \int_{[a,b]} \chi_{\{(x,y)|y<x\}}(x,y)df(y)dg(x)
\]
\[
= \int_{[a,b]} \int_{[a,b]} \chi_{\{(x,y)|y<x\}}(x,y)dg(x)df(y)
\]
\[
= \int_{[a,b]} \int_{[y,b]} dg(x)df(y) = -\int_{[a,b]} g(y+)df(y).
\]
The second formula is shown analogously. \(\square\)

If both \( f, g \in AC[a,b] \) this takes the usual form
\[
\int_{a}^{b} f(x)g'(x)dx = f(b)g(b) - f(a)g(a) - \int_{a}^{b} g(x)f'(x)dx. \tag{11.75}
\]

**Problem 11.31.** Compute \( V_a^b(f) \) for \( f(x) = \text{sign}(x) \) on \([a, b] = [-1, 1] \).

**Problem 11.32.** Show (11.65).

**Problem 11.33.** Consider \( f_j(x) := x^j \cos(\pi/x) \) for \( j \in \mathbb{N} \). Show that \( f_j \in C[0,1] \) if we set \( f_j(0) = 0 \). Show that \( f_j \) is of bounded variation for \( j \geq 2 \) but not for \( j = 1 \).

**Problem 11.34.** Let \( \alpha \in (0,1) \) and \( \beta > 1 \) with \( \alpha \beta \leq 1 \). Set \( M := \sum_{k=1}^{\infty} k^{-\beta} \), \( x_0 := 0 \), and \( x_n := M^{-1} \sum_{k=1}^{n} k^{-\beta} \). Then we can define a function on \([0,1]\) as follows: Set \( g(0) := 0 \), \( g(x_n) := n^{-\beta} \), and
\[
g(x) := c_n |x - t_n x_n - (1 - t_n) x_{n+1}|, \quad x \in [x_n, x_{n+1}],
\]
where \( c_n \) and \( t_n \in [0,1] \) are chosen such that \( g \) is (Lipschitz) continuous. Show that \( f = g^\alpha \) is Hölder continuous of exponent \( \alpha \) but not of bounded variation. (Hint: What is the variation on each subinterval?)
Problem 11.35. Show that if \( f \in BV[a,b] \) then so is \( f^* \), \(|f|\) and
\[
V_a^b(f^*) = V_a^b(f), \quad V_a^b(|f|) \leq V_a^b(f).
\]
Moreover, show
\[
V_a^b(\text{Re}(f)) \leq V_a^b(f), \quad V_a^b(\text{Im}(f)) \leq V_a^b(f).
\]

Problem 11.36. Show that if \( f, g \in BV[a,b] \) then so is \( fg \) and
\[
V_a^b(fg) \leq V_a^b(f) \sup |g| + V_a^b(g) \sup |f|.
\]
Hence, together with the norm \( \|f\|_{BV} := \|f\|_\infty + V_a^b(f) \) the space \( BV[a,b] \) is a Banach algebra.

Problem 11.37. Suppose \( f \in AC[a,b] \). Show that \( f' \) vanishes a.e. on \( f^{-1}(c) \) for every \( c \). (Hint: Split the set \( f^{-1}(c) \) into isolated points and the rest.)

Problem 11.38. A function \( f : [a,b] \to \mathbb{R} \) is said to have the Luzin N property if it maps Lebesgue null sets to Lebesgue null sets. Show that absolutely continuous functions have the Luzin N property. Show that the Cantor function does not have the Luzin N property. (Hint: Use \((11.69)\) and recall that: A set \( A \subseteq \mathbb{R} \) is a null set if and only if for every \( \varepsilon > 0 \) there exists a countable set of intervals \( I_j \) which cover \( A \) and satisfy \( \sum |I_j| < \varepsilon \).)

Problem 11.39. The variation of a function \( f : [a,b] \to X \) with \( X \) a Banach space is defined as
\[
V_a^b(f) := \sup_{\text{partitions } P \text{ of } [a,b]} V(P, f), \quad V(P, f) := \sum_{k=1}^m \|f(x_k) - f(x_{k-1})\|.
\]
Show that \( f : [a,b] \to \mathbb{R}^n \) (with Euclidean norm) is of bounded variation if and only if every component is of bounded variation.

Recall that a curve \( \gamma : [a,b] \to \mathbb{R}^n \) is rectifiable if \( V_a^b(\gamma) < \infty \). In this case \( V_a^b(\gamma) \) is called the arc length of \( \gamma \). Conclude that \( \gamma \) is rectifiable if and only if each of its coordinate functions is of bounded variation. Show that if each coordinate function is absolutely continuous, then
\[
V_a^b(\gamma) = \int_a^b |\gamma'(t)| dt.
\]
(Hint: For the last part note that one inequality is easy. Then reduce it to the case when \( \gamma' \) is a step function.)

Problem 11.40. Show that if \( f, g \in AC[a,b] \) then so is \( fg \) and the product rule \((fg)' = f'g + fg'\) holds. Conclude that \( AC[a,b] \) is a closed subalgebra of the Banach algebra \( BV[a,b] \). (Hint: Integration by parts. For the second part use Problem 11.36 and \((11.72)\).)
Problem 11.41. Show that if \( f \in AC[a,b] \) is nondecreasing iff \( f' \geq 0 \) a.e. and prove the substitution rule
\[
\int_a^b g(f(x))f'(x)dx = \int_{f(a)}^{f(b)} g(y)dy
\]
in this case. Conclude that if \( h \) is absolutely continuous, then so is \( h \circ f \) and the chain rule \((h \circ f)' = (h' \circ f)f'\) holds.

Moreover, show that if \( f \in AC[a,b] \) is strictly increasing iff \( f' > 0 \) a.e. In this case \( f^{-1} \) is also absolutely continuous and the inverse function rule
\[
(f^{-1})' = \frac{1}{f'(f^{-1})}
\]
holds. (Hint: (9.54).)

Problem 11.42. Consider \( f(x) := x^2 \sin(\frac{\pi}{x}) \) on \([0,1]\) (here \( f(0) = 0 \)) and \( g(x) = \sqrt{|x|} \). Show that both functions are absolutely continuous, but \( g \circ f \) is not. Hence the monotonicity assumption in the previous problem is important. Show that if \( f \) is absolutely continuous and \( g \) Lipschitz, then \( g \circ f \) is absolutely continuous.

Problem 11.43 (Characterization of the exponential function). Show that every nontrivial locally integrable function \( f : \mathbb{R} \to \mathbb{C} \) satisfying
\[
f(x + y) = f(x)f(y), \quad x, y \in \mathbb{R},
\]
is of the form \( f(x) = e^{\alpha x} \) for some \( \alpha \in \mathbb{C} \). (Hint: Start with \( F(x) = \int_0^x f(t)dt \) and show \( F(x + y) - F(x) = F(y)f(x) \). Conclude that \( f \) is absolutely continuous.)

Problem 11.44. Let \( X \subseteq \mathbb{R} \) be an interval, \( Y \) some measure space, and \( f : X \times Y \to \mathbb{C} \) some measurable function. Suppose \( x \mapsto f(x,y) \) is absolutely continuous for a.e. \( y \) such that
\[
\int_a^b \int_Y \left| \frac{\partial}{\partial x} f(x,y) \right| d\mu(y)dx < \infty \quad (11.76)
\]
for every compact interval \([a,b] \subseteq X\) and \( \int_Y |f(c,y)|d\mu(y) < \infty \) for one \( c \in X \).

Show that
\[
F(x) := \int_Y f(x,y) d\mu(y) \quad (11.77)
\]
is absolutely continuous and
\[
F'(x) = \int_Y \frac{\partial}{\partial x} f(x,y) d\mu(y) \quad (11.78)
\]
in this case. (Hint: Fubini.)
Problem 11.45. Show that if \( f \in AC[a, b] \) and \( f' \in L^p(a, b) \), then \( f \) is Hölder continuous:
\[
|f(x) - f(y)| \leq \|f'\|_p |x - y|^{1 - \frac{1}{p}}.
\]
Show that the function \( f(x) = -\log(x)^{-1} \) is absolutely continuous but not Hölder continuous on \([0, \frac{1}{2}]\).

Problem 11.46. Consider \( f(x) := x^2 \sin\left(\frac{\pi}{x^2}\right) \) on \([0, 1]\) (here \( f(0) = 0 \)). Show that \( f \) is differentiable everywhere and compute its derivative. Show that its derivative is not integrable. In particular, this function is not absolutely continuous and the fundamental theorem of calculus does not hold for this function.

Problem 11.47. Show that the function from the previous problem is Hölder continuous of exponent \( \frac{1}{2} \). (Hint: Consider \( 0 < x < y \). There is an \( x' < y \) with \( f(x') = f(x) \) and \( (x')^{-2} - y^{-2} \leq 2\pi \). Hence \( (x')^{-1} - y^{-1} \leq \sqrt{2\pi} \). Now use the Cauchy–Schwarz inequality to estimate \( |f(y) - f(x)| = |f(y) - f(x')| \leq f_{x'} 1 \cdot f'(t)dt|\).
The dual of \( L^p \)

12.1. The dual of \( L^p \), \( p < \infty \)

By the Hölder inequality every \( g \in L^q(X, d\mu) \) gives rise to a linear functional on \( L^p(X, d\mu) \) and this clearly raises the question if every linear functional is of this form. For \( 1 \leq p < \infty \) this is indeed the case:

**Theorem 12.1.** Consider \( L^p(X, d\mu) \) with some \( \sigma \)-finite measure \( \mu \) and let \( q \) be the corresponding dual index, \( \frac{1}{p} + \frac{1}{q} = 1 \). Then the map \( g \in L^q \mapsto \ell_g \in (L^p)^* \) given by

\[
\ell_g(f) = \int_X g f \, d\mu \quad (12.1)
\]

is an isometric isomorphism for \( 1 \leq p < \infty \). If \( p = \infty \) it is at least isometric.

**Proof.** Given \( g \in L^q \) it follows from Hölder’s inequality that \( \ell_g \) is a bounded linear functional with \( \|\ell_g\| \leq \|g\|_q \). Moreover, \( \|\ell_g\| = \|g\|_q \) follows from Corollary 10.5.

To show that this map is surjective if \( 1 \leq p < \infty \), first suppose \( \mu(X) < \infty \), and choose some \( \ell \in (L^p)^* \). Since \( \|\chi_A\|_p = \mu(A)^{1/p} \), we have \( \chi_A \in L^p \) for every \( A \in \Sigma \) and we can define

\[
\nu(A) = \ell(\chi_A).
\]

Suppose \( A = \bigcup_{j=1}^{\infty} A_j \), where the \( A_j \)'s are disjoint. Then, by dominated convergence, \( \| \sum_{j=1}^{\infty} \chi_{A_j} - \chi_A \|_p \to 0 \) (this is false for \( p = \infty \)!) and hence

\[
\nu(A) = \ell(\sum_{j=1}^{\infty} \chi_{A_j}) = \sum_{j=1}^{\infty} \ell(\chi_{A_j}) = \sum_{j=1}^{\infty} \nu(A_j).
\]
Thus $\nu$ is a complex measure. Moreover, $\mu(A) = 0$ implies $\chi_A = 0$ in $L^p$ and hence $\nu(A) = \ell(\chi_A) = 0$. Thus $\nu$ is absolutely continuous with respect to $\mu$ and by the complex Radon–Nikodym theorem $d\nu = g\,d\mu$ for some $g \in L^1(X, d\mu)$. In particular, we have

$$\ell(f) = \int_X fg\,d\mu$$

for every simple function $f$. Next let $A_n = \{x||g| < n\}$, then $g_n = g\chi_{A_n} \in L^q$ and by Lemma 10.6 we conclude $\|g_n\|_q \leq \|\ell\|$. Letting $n \to \infty$ shows $g \in L^q$ and finishes the proof for finite $\mu$.

If $\mu$ is $\sigma$-finite, let $X_n \nearrow X$ with $\mu(X_n) < \infty$. Then for every $n$ there is some $g_n$ on $X_n$ and by uniqueness of $g_n$ we must have $g_n = g_m$ on $X_n \cap X_m$. Hence there is some $g$ and by $\|g_n\|_q \leq \|\ell\|$ independent of $n$, we have $g \in L^q$. By construction $\ell(f\chi_{X_n}) = \ell_g(f\chi_{X_n})$ for every $f \in L^p$ and letting $n \to \infty$ shows $\ell(f) = \ell_g(f)$.

**Corollary 12.2.** Let $\mu$ be some $\sigma$-finite measure. Then $L^p(X, d\mu)$ is reflexive for $1 < p < \infty$.

**Proof.** Identify $L^p(X, d\mu)^*$ with $L^q(X, d\mu)$ and choose $h \in L^p(X, d\mu)^*$. Then there is some $f \in L^p(X, d\mu)$ such that

$$h(g) = \int g(x)f(x)d\mu(x), \quad g \in L^q(X, d\mu) \cong L^p(X, d\mu)^*.$$ 

But this implies $h(g) = g(f)$, that is, $h = J(f)$, and thus $J$ is surjective. \qed

Note that in the case $0 < p < 1$, where $L^p$ fails to be a Banach space, the dual might even be empty (see Problem 12.1)!

**Problem 12.1.** Show that $L^p(0, 1)$ is a quasinormed space if $0 < p < 1$ (cf. Problem 1.14). Moreover, show that $L^p(0, 1)^* = \{0\}$ in this case. (Hint: Suppose there were a nontrivial $\ell \in L^p(0, 1)^*$. Start with $f_0 \in L^p$ such that $|\ell(f_0)| \geq 1$. Set $g_0 = \chi_{(0,s]}f$ and $h_0 = \chi_{(s,1]}f$, where $s \in (0,1)$ is chosen such that $\|g_0\|_p = \|h_0\|_p = 2^{-1/p}\|f_0\|_p$. Then $|\ell(g_0)| \geq \frac{1}{2}$ or $|\ell(h_0)| \geq \frac{1}{2}$ and we set $f_1 = 2g_0$ in the first case and $f_1 = 2h_0$ else. Iterating this procedure gives a sequence $f_n$ with $|\ell(f_n)| \geq 1$ and $\|f_n\|_p = 2^{-n(1/p-1)}\|f_0\|_p$.)

### 12.2. The dual of $L^\infty$ and the Riesz representation theorem

In the last section we have computed the dual space of $L^p$ for $p < \infty$. Now we want to investigate the case $p = \infty$. Recall that we already know that the dual of $L^\infty$ is much larger than $L^1$ since it cannot be separable in general.

**Example.** Let $\nu$ be a complex measure. Then

$$\ell_\nu(f) = \int_X fd\nu \quad (12.2)$$
12.2. The dual of $L^\infty$ and the Riesz representation theorem

is a bounded linear functional on $B(X)$ (the Banach space of bounded measurable functions) with norm

$$\|\ell_\nu\| = |\nu|(X)$$

by (11.24) and Corollary 11.18. If $\nu$ is absolutely continuous with respect to $\mu$, then it will even be a bounded linear functional on $L^\infty(X,d\mu)$ since the integral will be independent of the representative in this case. ∎

So the dual of $B(X)$ contains all complex measures. However, this is still not all of $B(X)^*$. In fact, it turns out that it suffices to require only finite additivity for $\nu$.

Let $(X,\Sigma)$ be a measurable space. A complex content $\nu$ is a map $\nu: \Sigma \to \mathbb{C}$ such that (finite additivity)

$$\nu\left(\bigcup_{k=1}^{n} A_k\right) = \sum_{k=1}^{n} \nu(A_k), \quad A_j \cap A_k = \emptyset, \quad j \neq k.$$  \hspace{1cm} (12.4)

A content is called positive if $\nu(A) \geq 0$ for all $A \in \Sigma$ and given $\nu$ we can define its total variation $|\nu|(A)$ as in (11.15). The same proof as in Theorem 11.12 shows that $|\nu|$ is a positive content. However, since we do not require $\sigma$-additivity, it is not clear that $|\nu|(X)$ is finite. Hence we will call $\nu$ finite if $|\nu|(X) < \infty$. As in (11.17), (11.18) we can split every content into a complex linear combination of four positive contents.

Given a content $\nu$ we can define the corresponding integral for simple functions $s(x) = \sum_{k=1}^{n} \alpha_k \chi_{A_k}$ as usual

$$\int_A s \, d\nu = \sum_{k=1}^{n} \alpha_k \nu(A_k \cap A).$$ \hspace{1cm} (12.5)

As in the proof of Lemma 9.1 one shows that the integral is linear. Moreover,

$$|\int_A s \, d\nu| \leq |\nu|(A) \|s\|_\infty$$ \hspace{1cm} (12.6)

and for a finite content this integral can be extended to all of $B(X)$ such that

$$|\int_X f \, d\nu| \leq |\nu|(X) \|f\|_\infty$$ \hspace{1cm} (12.7)

by Theorem 1.12 (compare Problem 9.7). However, note that our convergence theorems (monotone convergence, dominated convergence) will no longer hold in this case (unless $\nu$ happens to be a measure).

In particular, every complex content gives rise to a bounded linear functional on $B(X)$ and the converse also holds:
Theorem 12.3. Every bounded linear functional $\ell \in B(X)^*$ is of the form

$$\ell(f) = \int_X f \, d\nu$$

for some unique finite complex content $\nu$ and $\|\ell\| = |\nu|(X)$.

**Proof.** Let $\ell \in B(X)^*$ be given. If there is a content $\nu$ at all it is uniquely determined by $\nu(A) = \ell(\chi_A)$. Using this as definition for $\nu$, we see that finite additivity follows from linearity of $\ell$. Moreover, (12.8) holds for characteristic functions and by

$$\ell\left(\sum_{k=1}^n \alpha_k \chi_{A_k}\right) = \sum_{k=1}^n \alpha_k \nu(A_k) = \sum_{k=1}^n |\nu(A_k)|, \quad \alpha_k = \text{sign}(\nu(A_k)),$$

we see $|\nu|(X) \leq \|\ell\|$.

Since the characteristic functions are total, (12.8) holds everywhere by continuity and (12.7) shows $\|\ell\| \leq |\nu|(X)$. □

It is also easy to tell when $\nu$ is positive. To this end call $\ell$ a **positive functional** if $\ell(f) \geq 0$ whenever $f \geq 0$.

**Corollary 12.4.** Let $\ell \in B^*(X)$ be associated with the finite content $\nu$. Then $\nu$ will be a positive content if and only if $\ell$ is a positive functional. Moreover, every $\ell \in B^*(X)$ can be written as a complex linear combination of four positive functionals.

**Proof.** Clearly $\ell \geq 0$ implies $\nu(A) = \ell(\chi_A) \geq 0$. Conversely $\nu(A) \geq 0$ implies $\ell(s) \geq 0$ for every simple $s \geq 0$. Now for $f \geq 0$ we can find a sequence of simple functions $s_n$ such that $\|s_n - f\|_\infty \to 0$. Moreover, by $|s_n - f| \leq |s_n - f|$ we can assume $s_n$ to be nonnegative. But then $\ell(f) = \lim_{n \to \infty} \ell(s_n) \geq 0$ as required.

The last part follows by splitting the content $\nu$ into a linear combination of positive contents. □

Remark: To obtain the dual of $L^\infty(X,d\mu)$ from this you just need to restrict to those linear functionals which vanish on $N(X,d\mu)$ (cf. Problem 12.2), that is, those whose content is absolutely continuous with respect to $\mu$ (note that the Radon–Nikodym theorem does not hold unless the content is a measure).

**Example.** Consider $B(\mathbb{R})$ and define

$$\ell(f) = \lim_{\varepsilon \downarrow 0} (\lambda f(-\varepsilon) + (1 - \lambda)f(\varepsilon)), \quad \lambda \in [0,1],$$

for $f$ in the subspace of bounded measurable functions which have left and right limits at 0. Since $\|\ell\| = 1$ we can extend it to all of $B(\mathbb{R})$ using the
Hahn–Banach theorem. Then the corresponding content \( \nu \) is no measure:

\[
\lambda = \nu([-1,0)) = \nu\left(\bigcup_{n=1}^{\infty} \left[-\frac{1}{n}, -\frac{1}{n+1}\right]\right) \neq \sum_{n=1}^{\infty} \nu\left(\left[-\frac{1}{n}, -\frac{1}{n+1}\right]\right) = 0. \tag{12.10}
\]

Observe that the corresponding distribution function (defined as in (8.19)) is nondecreasing but not right continuous! If we render \( \nu \) right continuous, we get the the distribution function of the Dirac measure (centered at 0). In addition, the Dirac measure has the same integral at least for continuous functions!

Based on this observation we can give a simple proof of the Riesz representation for compact intervals. The general version will be shown in the next section.

**Theorem 12.5 (Riesz representation).** Let \( I = [a,b] \subseteq \mathbb{R} \) be a compact interval. Every bounded linear functional \( \ell \in C(I)^* \) is of the form

\[
\ell(f) = \int_I f \, d\nu \tag{12.11}
\]

for some unique complex Borel measure \( \nu \) and \( \|\ell\| = |\nu|(I) \).

**Proof.** By the Hahn–Banach theorem we can extend \( \ell \) to a bounded linear functional \( \overline{\ell} \in B(I)^* \) we have a corresponding content \( \overline{\nu} \). Splitting this content into positive parts it is no restriction to assume \( \overline{\nu} \) is positive.

Now the idea is as follows: Define a distribution function for \( \overline{\nu} \) as in (8.19). By finite additivity of \( \overline{\nu} \) it will be nondecreasing and we can use Theorem 8.13 to obtain an associated measure \( \nu \) whose distribution function coincides with \( \overline{\nu} \) except possibly at points where \( \nu \) is discontinuous. It remains to show that the corresponding integral coincides with \( \ell \) for continuous functions.

Let \( f \in C(I) \) be given. Fix points \( a < x_1^n < x_2^n < \ldots < x_k^n < b \) such that \( x_0^n \to a, x_k^n \to b \), and \( \sup_{k} |x_{k-1}^n - x_k^n| \to 0 \) as \( n \to \infty \). Then the sequence of simple functions

\[
f_n(x) = f(x_0^n)\chi_{[x_0^n,x_1^n]} + f(x_1^n)\chi_{[x_1^n,x_2^n]} + \cdots + f(x_{n-1}^n)\chi_{[x_{n-1}^n,x_n^n]}.
\]
converges uniformly to \( f \) by continuity of \( f \) (and the fact that \( f \) vanishes as \( x \to \pm \infty \) in the case \( I = \mathbb{R} \)). Moreover,

\[
\int_I f \, d\nu = \lim_{n \to \infty} \int_I f_n \, d\nu = \lim_{n \to \infty} \sum_{k=1}^n f(x^n_{k-1})(\nu(x^n_k) - \nu(x^n_{k-1}))
\]

\[
= \lim_{n \to \infty} \sum_{k=1}^n f(x^n_{k-1})(\tilde{\nu}(x^n_k) - \tilde{\nu}(x^n_{k-1})) = \lim_{n \to \infty} \int_I f_n \, d\tilde{\nu}
\]

\[
= \int_I f \, d\tilde{\nu} = \ell(f)
\]

provided the points \( x^n_k \) are chosen to stay away from all discontinuities of \( \nu(x) \) (recall that there are at most countably many).

To see \( \|\ell\| = |\nu|(I) \) recall \( d\nu = hd|\nu| \) where \( |h| = 1 \) (Corollary 11.18). Now choose continuous functions \( h_n(x) \to h(x) \) pointwise a.e. (Theorem 10.15). Using \( \tilde{h}_n = \frac{h_n}{\max(1,|h_n|)} \) we even get such a sequence with \( |\tilde{h}_n| \leq 1 \). Hence

\[
\ell(\tilde{h}_n) = \int \tilde{h}_n^* h \, d|\nu| \to \int |h|^2 \, d|\nu| = |\nu|(I) \text{ implying } \|\ell\| \geq |\nu|(I). \]

The converse follows from (12.7).

**Problem 12.2.** Let \( M \) be a closed subspace of a Banach space \( X \). Show that \( (X/M)^* \cong \{\ell \in X^* | M \subseteq \text{Ker}(\ell)\} \) (cf. Theorem 4.27).

### 12.3. The Riesz–Markov representation theorem

In this section we want to generalize Theorem 12.5. To this end \( X \) will be a metric space with the Borel \( \sigma \)-algebra. Given a Borel measure \( \mu \) the integral

\[
\ell(f) := \int_X f \, d\mu
\]  

(12.12)

will define a linear functional \( \ell \) on the set of continuous functions with compact support \( C_c(X) \). If \( \mu \) were bounded we could drop the requirement for \( f \) to have compact support, but we do not want to impose this restriction here. However, in an arbitrary metric space there might not be many continuous functions with compact support. In fact, if \( f \in C_c(X) \) and \( x \in X \) is such that \( f(x) \neq 0 \), then \( f^{-1}(B_r(f(x))) \) will be a relatively compact neighborhood of \( x \) whenever \( 0 < r < |f(x)| \). So in order to be able to see all of \( X \), we will assume that every point has a relatively compact neighborhood, that is, \( X \) is locally compact.

Moreover, note that positivity of \( \mu \) implies that \( \ell \) is **positive** in the sense that \( \ell(f) \geq 0 \) if \( f \geq 0 \). This raises the question if there are any other requirements for a linear functional to be of the form (12.12). The purpose of this section is to prove that there are none, that is, there is a one-to-one connection between positive linear functionals on \( C_c(X) \) and positive Borel measures on \( X \).
As a preparation let us reflect how \( \mu \) could be recovered from \( \ell \) as in \((12.12)\). Given a Borel set \( A \) it seems natural to try to approximate the characteristic function \( \chi_A \) by continuous functions form the inside or the outside. However, if you try this for the rational numbers in the case \( X = \mathbb{R} \), then this works neither from the inside nor the outside. So we have to be more modest. If \( K \) is a compact set, we can choose a sequence \( f_n \in C_c(X) \) with \( f_n \downarrow \chi_K \) (Problem 12.3). In particular,
\[
\mu(K) = \lim_{n \to \infty} \int_X f_n \, d\mu
\]
(12.13)
by dominated convergence. So we can recover the measure of compact sets from \( \ell \) and hence \( \mu \) if it is inner regular. In particular, for every positive linear functional there can be at most one inner regular measure. This shows how \( \mu \) should be defined given a linear functional \( \ell \). Nevertheless it will be more convenient for us to approximate characteristic functions of open sets from the inside since we want to define an outer measure and use the Carathéodory construction. Hence, given a positive linear functional \( \ell \) we define
\[
\rho(O) := \sup\{ \ell(f) \mid f \in C_c(X), \ f \prec O \}
\]
(12.14)
for any open set \( O \). Here \( f \prec O \) is short hand for \( f \leq \chi_O \) and \( \text{supp}(f) \subseteq O \). Since \( \ell \) is positive, so is \( \rho \). Note that it is not clear that this definition will indeed coincide with \( \mu(O) \) if \( \ell \) is given by \((12.12)\) unless \( O \) has a compact exhaustion. However, this is of no concern for us at this point.

**Lemma 12.6.** Given a positive linear functional \( \ell \) on \( C_c(X) \) the set function \( \rho \) defined in \((12.14)\) has the following properties:

(i) \( \rho(\emptyset) = 0 \),

(ii) monotonicity \( \rho(O_1) \leq \rho(O_2) \) if \( O_1 \subseteq O_2 \),

(iii) \( \rho \) is finite for relatively compact sets,

(iv) \( \rho(O) \leq \sum_n \rho(O_n) \) for every countable open cover \( \{O_n\} \) of \( O \), and

(v) additivity \( \rho(O_1 \cup O_2) = \rho(O_1) + \rho(O_2) \) if \( O_1 \cap O_2 = \emptyset \).

**Proof.** (i) and (ii) are clear. To see (iii) note that if \( \overline{O} \) is compact, then by Urysohn’s lemma there is a function \( f \in C_c(X) \) which is one on \( \overline{O} \) implying \( \rho(O) \leq \ell(f) \). To see (iv) let \( f \in C_c(X) \) with \( f \prec O \). Then finitely many of the sets \( O_1, \ldots, O_N \) will cover \( K := \text{supp}(f) \). Hence we can choose a partition of unity \( h_1, \ldots, h_{N+1} \) (Lemma 0.28) subordinate to the cover \( O_1, \ldots, O_N, X \setminus K \). Then \( \chi_K \leq h_1 + \cdots + h_N \) and hence
\[
\ell(f) = \sum_{j=1}^N \ell(h_j f) \leq \sum_{j=1}^N \rho(O_j) \leq \sum_n \rho(O_n).
\]
To see (v) note that \( f_1 \prec O_1 \) and \( f_2 \prec O_2 \) implies \( f_1 + f_2 \prec O_1 \cup O_2 \) and hence \( \ell(f_1) + \ell(f_2) = \ell(f_1 + f_2) \leq \rho(O_1 \cup O_2) \). Taking the supremum over \( f_1 \) and \( f_2 \) shows \( \rho(O_1) + \rho(O_2) \leq \rho(O_1 \cup O_2) \). The reverse inequality follows from the previous item.

**Lemma 12.7.** Let \( \ell \) be a positive linear functional on \( C_c(X) \) and let \( \rho \) be defined as in (12.14). Then

\[
\mu^*(A) := \inf \left\{ \rho(O) \mid A \subseteq O, \text{ O open} \right\},
\]

(12.15)
defines a metric outer measure on \( X \).

**Proof.** Consider the outer measure (Lemma 8.2)

\[
\nu^*(A) := \inf \left\{ \sum_{n=1}^{\infty} \rho(O_n) \mid A \subseteq \bigcup_{n=1}^{\infty} O_n, \text{ O_n open} \right\}.
\]

Then we clearly have \( \nu^*(A) \leq \mu^*(A) \). Moreover, if \( \nu^*(A) < \mu^*(A) \) we can find an open cover \( \{O_n\} \) such that \( \sum_{n} \rho(O_n) \leq \mu^*(A) - \epsilon \). But for \( O = \bigcup_n O_n \) we have \( \mu^*(A) \leq \rho(O) \leq \sum_n \rho(O_n) \), a contradiction. Hence \( \mu^* = \nu^* \) and we have an outer measure.

To see that \( \mu^* \) is a metric outer measure let \( A_1, A_2 \) with \( \text{dist}(A_1, A_2) > 0 \) be given. Then, there are disjoint open sets \( O_1 \supseteq A_1 \) and \( O_2 \supseteq A_2 \). Hence for \( O \supseteq A_1 \cup A_2 \subseteq O \) we have

\[
\mu^*(A_1) + \mu^*(A_2) \leq \rho(O_1 \cap O) + \rho(O_2 \cap O) \leq \rho(O)
\]

and taking the infimum over all \( O \) we have \( \mu^*(A_1) + \mu^*(A_2) \leq \mu^*(A_1 \cup A_2) \).

The converse follows from subadditivity and hence \( \mu^* \) is a metric outer measure.

So Theorem 8.9 gives us a corresponding measure \( \mu \) defined on the Borel \( \sigma \)-algebra by Lemma 8.11. By construction this Borel measure will be outer regular and it will also be inner regular as the next lemma shows. Note that if one is willing to make the extra assumption of separability for \( X \), this will come for free from Corollary 8.22.

**Lemma 12.8.** The Borel measure \( \mu \) associated with \( \mu^* \) from (12.15) is regular.

**Proof.** Since \( \mu \) is outer regular by construction it suffices to show

\[
\mu(O) = \sup_{K \subseteq O, \text{ K compact}} \mu(K)
\]

for every open set \( O \subseteq X \). Now denote the last supremum by \( \alpha \) and observe \( \alpha \leq \mu(O) \) by monotonicity. For the converse we can assume \( \alpha < \infty \) without loss of generality. Then, by the definition of \( \mu(O) = \rho(O) \) we can find some \( f \in C_c(X) \) with \( f \prec O \) such that \( \mu(O) \leq \ell(f) + \epsilon \). Since \( K := \text{supp}(f) \subseteq O \) is compact we have \( \mu(O) \leq \ell(f) + \epsilon \leq \mu(K) + \epsilon \leq \alpha + \epsilon \) and as \( \epsilon > 0 \) is arbitrary this establishes the claim.
Now we are ready to show

**Theorem 12.9** (Riesz–Markov representation). Let $X$ be a locally compact metric space. Then every positive linear functional $\ell : C_c(X) \to \mathbb{C}$ gives rise to a unique regular Borel measure $\mu$ such that (12.12) holds.

**Proof.** We have already constructed a corresponding Borel measure $\mu$ and it remains to show that $\ell$ is given by (12.12). To this end observe that if $f \in C_c(X)$ satisfies $\chi_O \leq f \leq \chi_C$, where $O$ is open and $C$ closed, then $\mu(O) \leq \ell(f) \leq \mu(C)$. In fact, every $g \prec O$ satisfies $\ell(g) \leq \ell(f)$ and hence $\mu(O) = \rho(O) \leq \ell(f)$. Similarly, for every $\tilde{O} \supseteq C$ we have $f \prec \tilde{O}$ and hence $\ell(f) \leq \rho(\tilde{O})$ implying $\ell(f) \leq \mu(C)$.

Now the next step is to split $f$ into smaller pieces for which this estimate can be applied. To this end suppose $0 \leq f \leq 1$ and define $g^n_k = \min(f, \frac{k}{n})$ for $0 \leq k \leq n$. Clearly $g^n_0 = 0$ and $g^n_n = f$. Setting $f^n_k = g^n_k - g^n_{k-1}$ for $1 \leq k \leq n$ we have $f = \sum_{k=1}^n f^n_k$ and $\frac{1}{n} \chi_{C^n_k} \leq f^n_k \leq \frac{1}{n} \chi_{O^n_k}$ where $O^n_k = \{x \in X | f(x) > \frac{k}{n}\}$ and $C^n_k = \overline{O^n_k} = \{x \in X | f(x) \geq \frac{k}{n}\}$. Summing over $k$ we have $\frac{1}{n} \sum_{k=1}^n \chi_{C^n_k} \leq f \leq \frac{1}{n} \sum_{k=0}^{n-1} \chi_{O^n_k}$ as well as

$$\frac{1}{n} \sum_{k=1}^n \mu(O^n_k) \leq \ell(f) \leq \frac{1}{n} \sum_{k=0}^{n-1} \mu(C^n_k)$$

Hence we obtain

$$\int f \, d\mu - \frac{\mu(O^n_0)}{n} \leq \frac{1}{n} \sum_{k=1}^n \mu(O^n_k) \leq \ell(f) \leq \frac{1}{n} \sum_{k=0}^{n-1} \mu(C^n_k) \leq \int f \, d\mu + \frac{\mu(C^n_0)}{n}$$

and letting $n \to \infty$ establishes the claim since $C^n_0 = \text{supp}(f)$ is compact and hence has finite measure. $\square$

Note that this might at first sight look like a contradiction since (12.12) gives a linear functional even if $\mu$ is not regular. However, in this case the Riesz–Markov theorem merely says that there will be a corresponding regular measure which gives rise to the same integral for continuous functions. Moreover, using (12.13) one even sees that both measures agree on compact sets.

As a consequence we can also identify the dual space of $C_0(X)$ (i.e. the closure of $C_c(X)$ as a subspace of $C_b(X)$). Note that $C_0(X)$ is separable if $X$ is locally compact and separable (Lemma 0.34). Also recall that a complex measure is regular if all four positive measures in the Jordan decomposition (11.19) are. By Lemma 11.20 this is equivalent to the total variation being regular.
Theorem 12.10 (Riesz–Markov representation). Let $X$ be a locally compact metric space. Every bounded linear functional $\ell \in C_0(X)^*$ is of the form

$$\ell(f) = \int_X f \, d\nu$$

(12.16)

for some unique regular complex Borel measure $\nu$ and $\|\ell\| = |\nu|(X)$. Moreover, $\ell$ will be positive if and only if $\nu$ is.

If $X$ is compact this holds for $C(X) = C_0(X)$.

**Proof.** First of all observe that (11.24) shows that for every regular complex measure $\nu$ equation (12.16) gives a linear functional $\ell$ with $\|\ell\| \leq |\nu|(X)$. This functional will be positive if $\nu$ is. Moreover, we have $d\nu = h \, d|\nu|$ (Corollary 11.18) and by Theorem 10.15 we can find a sequence $h_n \in C_c(X)$ with $h_n(x) \to h(x)$ pointwise a.e. Using $\tilde{h}_n = h_n \max(1, |h_n|)$ we even get such a sequence with $|\tilde{h}_n| \leq 1$. Hence $\ell(\tilde{h}_n) = \int \tilde{h}_n^* h \, d|\nu| \to \int |h|^2 d|\nu| = |\nu|(X)$ implying $\|\ell\| \geq |\nu|(X)$.

Conversely, let $\ell$ be given. By the Hahn–Banach theorem we can extend $\ell$ to a bounded linear functional $\ell \in B(X)^*$ which can be written as a linear combinations of positive functionals by Corollary 12.4. Hence it is no restriction to assume $\ell$ is positive. But for positive $\ell$ the previous theorem implies existence of a corresponding regular measure $\nu$ such that (12.16) holds for all $f \in C_c(X)$. Since $C_c(X) = C_0(X)$ (12.16) holds for all $f \in C_0(X)$ by continuity. \hfill \Box

**Example.** Note that the dual space of $C_b(X)$ will in general be larger. For example, consider $C_b(\mathbb{R})$ and define $\ell(f) = \lim_{x \to \infty} f(x)$ on the subspace of functions from $C_b(\mathbb{R})$ for which this limit exists. Extend $\ell$ to a bounded linear functional on all of $C_b(\mathbb{R})$ using Hahn–Banach. Then $\ell$ restricted to $C_0(\mathbb{R})$ is zero and hence there is no associated measure such that (12.16) holds. \hfill \Diamond

As a consequence we can extend Helly’s selection theorem. We call a sequence of complex measures $\nu_n$ vaguely convergent to a measure $\nu$ if

$$\int_X f \, d\nu_n \to \int_X f \, d\nu, \quad f \in C_c(X).$$

(12.17)

This generalizes our definition for positive measures from Section 11.7. Moreover, note that in the case that the sequence is bounded, $|\nu_n|(X) \leq M$, we get (12.17) for all $f \in C_0(X)$. Indeed, choose $g \in C_c(X)$ such that $\|f - g\|_{\infty} < \varepsilon$ and note that $\limsup_n |\int_X f \, d\nu_n - \int_X f \, d\nu| \leq |\nu|(X)$.

**Theorem 12.11.** Let $X$ be a locally compact metric space. Then every bounded sequence $\nu_n$ of regular complex measures, that is $|\nu_n|(X) \leq M$, has
a vaguely convergent subsequence whose limit is regular. If all \( \nu_n \) are positive every limit of a convergent subsequence is again positive.

**Proof.** Let \( Y = C_0(X) \). Then we can identify the space of regular complex measure \( M_{\text{reg}}(X) \) as the dual space \( Y^* \) by the Riesz–Markov theorem. Moreover, every bounded sequence has a weak-* convergent subsequence by the Banach–Alaoglu theorem (Theorem 5.10) and this subsequence converges in particular vaguely.

If the measures are positive, then \( \ell_n(f) = \int f \, d\nu_n \geq 0 \) for every \( f \geq 0 \) and hence \( \ell(f) = \int f \, d\nu \geq 0 \) for every \( f \geq 0 \), where \( \ell \in Y^* \) is the limit of some convergent subsequence. Hence \( \nu \) is positive by the Riesz–Markov representation theorem. \( \square \)

Recall once more that in the case where \( X \) is locally compact and separable, regularity will automatically hold for every Borel measure.

**Problem 12.3.** \( X \) be a locally compact metric space. Show that for every compact set \( K \) there is a sequence \( f_n \in C_c(X) \) with \( 0 \leq f_n \leq 1 \) and \( f_n \downarrow \chi_K \). (Hint: Urysohn’s lemma.)
Chapter 13

The Fourier transform

13.1. The Fourier transform on $L^1$ and $L^2$

For $f \in L^1(\mathbb{R}^n)$ we define its Fourier transform via

$$\hat{f}(p) \equiv \mathcal{F}(f)(p) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ipx} f(x) d^n x.$$  (13.1)

Here $px = p_1 x_1 + \cdots + p_n x_n$ is the usual scalar product in $\mathbb{R}^n$ and we will use $|x| = \sqrt{x_1^2 + \cdots + x_n^2}$ for the Euclidean norm.

Lemma 13.1. The Fourier transform is a bounded map from $L^1(\mathbb{R}^n)$ into $C_0(\mathbb{R}^n)$ satisfying

$$\|\hat{f}\|_{\infty} \leq (2\pi)^{-n/2} \|f\|_1.$$  (13.2)

Proof. Since $|e^{-ipx}| = 1$ the estimate (13.2) is immediate from

$$|\hat{f}(p)| \leq \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} |e^{-ipx} f(x)| d^n x = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} |f(x)| d^n x.$$  

Moreover, a straightforward application of the dominated convergence theorem shows that $\hat{f}$ is continuous. \hfill \Box

Note that if $f$ is nonnegative we have equality: $\|\hat{f}\|_{\infty} = (2\pi)^{-n/2} \|f\|_1 = \hat{f}(0)$.

The following simple properties are left as an exercise.
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Lemma 13.2. Let f ∈ L1 (Rn ). Then
(f (x + a))∧ (p) = eiap fˆ(p),

(13.3)

n

a∈R ,

(13.4)

λ > 0,

(13.5)

∧

f (x)) (p) = fˆ(p − a),
1
p
(f (λx))∧ (p) = n fˆ( ),
λ
λ
(f (−x))∧ (p) = (f )∧ (−p).

(e

ixa

a ∈ Rn ,

(13.6)

Next we look at the connection with differentiation.
Lemma 13.3. Suppose f ∈ C 1 (Rn ) such that lim|x|→∞ f (x) = 0 and
f, ∂j f ∈ L1 (Rn ) for some 1 ≤ j ≤ n. Then
(∂j f )∧ (p) = ipj fˆ(p).

(13.7)

Similarly, if f (x), xj f (x) ∈ L1 (Rn ) for some 1 ≤ j ≤ n, then fˆ(p) is differentiable with respect to pj and
(xj f (x))∧ (p) = i∂j fˆ(p).

(13.8)

Proof. First of all, by integration by parts, we see
Z
1
∂
(∂j f )∧ (p) =
f (x)dn x
e−ipx
∂xj
(2π)n/2 Rn

Z 
1
∂ −ipx
e
f (x)dn x
=
−
∂xj
(2π)n/2 Rn
Z
1
=
ipj e−ipx f (x)dn x = ipj fˆ(p).
(2π)n/2 Rn
Similarly, the second formula follows from
Z
1
∧
xj e−ipx f (x)dn x
(xj f (x)) (p) =
(2π)n/2 Rn

Z 
1
∂ −ipx
∂ ˆ
=
i
e
f (x)dn x = i
f (p),
n/2
∂pj
∂pj
(2π)
Rn
where interchanging the derivative and integral is permissible by Problem 9.14. In particular, fˆ(p) is differentiable.

This result immediately extends to higher derivatives. To this end let
C ∞ (Rn ) be the set of all complex-valued functions which have partial derivatives of arbitrary order. For f ∈ C ∞ (Rn ) and α ∈ Nn0 we set
∂α f =

∂ |α| f
,
· · · ∂xαnn

∂xα1 1

xα = xα1 1 · · · xαnn ,

|α| = α1 + · · · + αn .

(13.9)


13.1. The Fourier transform on $L^1$ and $L^2$

An element $\alpha \in \mathbb{N}_0^n$ is called a **multi-index** and $|\alpha|$ is called its **order**. We will also set $(\lambda x)^\alpha = \lambda^{|\alpha|} x^\alpha$ for $\lambda \in \mathbb{R}$. Recall the **Schwartz space**

$$S(\mathbb{R}^n) = \{ f \in C^\infty(\mathbb{R}^n) | \sup_x |x^\alpha(\partial_\beta f)(x)| < \infty, \forall \alpha, \beta \in \mathbb{N}_0^n \} \quad (13.10)$$

which is a subspace of $L^p(\mathbb{R}^n)$ and which is dense for $1 \leq p < \infty$ (since $C_c^\infty(\mathbb{R}^n) \subset S(\mathbb{R}^n)$). Together with the seminorms $\|x^\alpha(\partial_\beta f)(x)\|_\infty$ it is a Fréchet space. Note that if $f \in S(\mathbb{R}^n)$, then the same is true for $x^\alpha f(x)$ and $(\partial_\alpha f)(x)$ for every multi-index $\alpha$. Also, by Leibniz’ rule, the product of two Schwartz functions is again a Schwartz function.

**Lemma 13.4.** The Fourier transform satisfies $\mathcal{F} : S(\mathbb{R}^n) \to S(\mathbb{R}^n)$. Furthermore, for every multi-index $\alpha \in \mathbb{N}_0^n$ and every $f \in S(\mathbb{R}^n)$ we have

$$(\partial_\alpha f)^\wedge(p) = (ip)^\alpha \hat{f}(p), \quad (x^\alpha f(x))^\wedge(p) = i^{\alpha^0} |\alpha| \partial_\alpha \hat{f}(p). \quad (13.11)$$

**Proof.** The formulas are immediate from the previous lemma. To see that $\hat{f} \in S(\mathbb{R}^n)$ if $f \in S(\mathbb{R}^n)$, we begin with the observation that $\hat{f}$ is bounded by (13.2). But then $p^\alpha(\partial_\beta \hat{f})(p) = i^{-|\alpha|+|\beta|}(\partial_\alpha x^\beta f(x))^\wedge(p)$ is bounded since $\partial_\alpha x^\beta f(x) \in S(\mathbb{R}^n)$ if $f \in S(\mathbb{R}^n)$.

Hence we will sometimes write $pf(x)$ for $-i\partial f(x)$, where $\partial = (\partial_1, \ldots, \partial_n)$ is the **gradient**. Roughly speaking this lemma shows that the decay of a functions is related to the smoothness of its Fourier transform and the smoothness of a functions is related to the decay of its Fourier transform.

In particular, this allows us to conclude that the Fourier transform of an integrable function will vanish at $\infty$. Recall that we denote the space of all continuous functions $f : \mathbb{R}^n \to \mathbb{C}$ which vanish at $\infty$ by $C_0(\mathbb{R}^n)$.

**Corollary 13.5** (Riemann-Lebesgue). The Fourier transform maps $L^1(\mathbb{R}^n)$ into $C_0(\mathbb{R}^n)$.

**Proof.** First of all recall that $C_0(\mathbb{R}^n)$ equipped with the sup norm is a Banach space and that $S(\mathbb{R}^n)$ is dense (Problem 1.42). By the previous lemma we have $\hat{f} \in C_0(\mathbb{R}^n)$ if $f \in S(\mathbb{R}^n)$. Moreover, since $S(\mathbb{R}^n)$ is dense in $L^1(\mathbb{R}^n)$, the estimate (13.2) shows that the Fourier transform extends to a continuous map from $L^1(\mathbb{R}^n)$ into $C_0(\mathbb{R}^n)$.

Next we will turn to the inversion of the Fourier transform. As a preparation we will need the Fourier transform of a **Gaussian**.

**Lemma 13.6.** We have $e^{-z|x|^2/2} \in S(\mathbb{R}^n)$ for $\text{Re}(z) > 0$ and

$$\mathcal{F}(e^{-z|x|^2/2})(p) = \frac{1}{z^{n/2}} e^{-|p|^2/(2z)}. \quad (13.12)$$

Here $z^{n/2}$ is the standard branch with branch cut along the negative real axis.
Proof. Due to the product structure of the exponential, one can treat each coordinate separately, reducing the problem to the case \( n = 1 \) (Problem 13.3).

Let \( \phi_z(x) = \exp(-zx^2/2) \). Then \( \phi_z'(x) + z\phi_z(x) = 0 \) and hence \( i(p\hat{\phi}_z(p) + z\hat{\phi}'_z(p)) = 0 \). Thus \( \hat{\phi}_z(p) = c\phi_{1/\sqrt{z}}(p) \) and (Problem 9.22)

\[
c = \hat{\phi}_z(0) = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \exp(-zx^2/2)dx = \frac{1}{\sqrt{z}}
\]

at least for \( z > 0 \). However, since the integral is holomorphic for \( \text{Re}(z) > 0 \) by Problem 9.18, this holds for all \( z \) with \( \text{Re}(z) > 0 \) if we choose the branch cut of the root along the negative real axis. \( \Box \)

Now we can show

**Theorem 13.7.** The Fourier transform is a bounded injective map from \( L^1(\mathbb{R}^n) \) into \( C_0(\mathbb{R}^n) \). Its inverse is given by

\[
f(x) = \lim_{\varepsilon \downarrow 0} \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ipx-\varepsilon|p|^2/2} \hat{f}(p)d^n p,
\]

where the limit has to be understood in \( L^1 \). Moreover, (13.13) holds at every Lebesgue point (cf. Theorem 11.6) and hence in particular at every point of continuity.

Proof. Abbreviate \( \phi_\varepsilon(x) = (2\pi)^{-n/2} \exp(-\varepsilon|x|^2/2) \). Then the right-hand side is given by

\[
\int_{\mathbb{R}^n} \phi_\varepsilon(p)e^{ipx} \hat{f}(p)d^n p = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \phi_\varepsilon(p)e^{ipx} f(y)e^{-ipy}d^n yd^n p
\]

and, invoking Fubini and Lemma 13.2, we further see that this is equal to

\[
\int_{\mathbb{R}^n} \phi_\varepsilon(p)e^{ipx} \hat{f}(y)d^n y = \int_{\mathbb{R}^n} \frac{1}{\varepsilon^{n/2}} \phi_{1/\varepsilon}(y-x)f(y)d^n y.
\]

But the last integral converges to \( f \) in \( L^1(\mathbb{R}^n) \) by Lemma 10.18. Moreover, it is straightforward to see that it converges at every point of continuity. The case of Lebesgue points follows from Problem 14.9. \( \Box \)

Of course when \( \hat{f} \in L^1(\mathbb{R}^n) \), the limit is superfluous and we obtain

**Corollary 13.8.** Suppose \( f, \hat{f} \in L^1(\mathbb{R}^n) \). Then

\[
(\hat{f})^\vee = f,
\]

where

\[
\hat{f}(p) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ipx} f(x)d^n x = \hat{f}(-p).
\]

In particular, \( \mathcal{F} : F^1(\mathbb{R}^n) \rightarrow F^1(\mathbb{R}^n) \) is a bijection, where \( F^1(\mathbb{R}^n) = \{ f \in L^1(\mathbb{R}^n) | \hat{f} \in L^1(\mathbb{R}^n) \} \). Moreover, \( \mathcal{F} : S(\mathbb{R}^n) \rightarrow S(\mathbb{R}^n) \) is a bijection.
Observe that we have \( F^1(\mathbb{R}^n) \subset L^1(\mathbb{R}^n) \cap C_0(\mathbb{R}^n) \subset L^p(\mathbb{R}^n) \) for any \( p \in [1, \infty) \) (cf. also Problem 13.2) and choosing \( f \) continuous (13.14) will hold pointwise.

However, note that \( F : L^1(\mathbb{R}^n) \to C_0(\mathbb{R}^n) \) is not onto (cf. Problem 13.7). Nevertheless the inverse Fourier transform \( F^{-1} \) is a closed map from \( \text{Ran}(F) \to L^1(\mathbb{R}^n) \) by Lemma 4.8.

**Lemma 13.9.** Suppose \( f \in F^1(\mathbb{R}^n) \). Then \( f, \hat{f} \in L^2(\mathbb{R}^n) \) and
\[
\|f\|_2^2 = \|\hat{f}\|_2^2 \leq (2\pi)^{-n/2} \|f\|_1 \|\hat{f}\|_1
\]
(13.16) holds.

**Proof.** This follows from Fubini’s theorem since
\[
\int_{\mathbb{R}^n} |\hat{f}(p)|^2 d^n p = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} f(x) \ast \hat{f}(p) e^{i\langle x, p \rangle} d^n p d^n x
\]
\[
= \int_{\mathbb{R}^n} |f(x)|^2 d^n x
\]
for \( f, \hat{f} \in L^1(\mathbb{R}^n) \).

The identity \( \|f\|_2 = \|\hat{f}\|_2 \) is known as the **Plancherel identity**. Thus, by Theorem 1.12, we can extend \( F \) to all of \( L^2(\mathbb{R}^n) \) by setting \( F(f) = \lim_{m \to \infty} F(f_m) \), where \( f_m \) is an arbitrary sequence from, say, \( S(\mathbb{R}^n) \) converging to \( f \) in the \( L^2 \) norm.

**Theorem 13.10** (Plancherel). The Fourier transform \( F \) extends to a unitary operator \( F : L^2(\mathbb{R}^n) \to L^2(\mathbb{R}^n) \).

**Proof.** As already noted, \( F \) extends uniquely to a bounded operator on \( L^2(\mathbb{R}^n) \). Since Plancherel’s identity remains valid by continuity of the norm and since its range is dense, this extension is a unitary operator.

We also note that this extension is still given by (13.1) whenever the right-hand side is integrable.

**Lemma 13.11.** Let \( f \in L^1(\mathbb{R}^n) \cap L^2(\mathbb{R}^n) \), then (13.1) continues to hold, where \( F \) now denotes the extension of the Fourier transform from \( S(\mathbb{R}^n) \) to \( L^2(\mathbb{R}^n) \).

**Proof.** If \( f \) has compact support, then by Lemma 10.17 its mollification \( \phi_x * f \in C_c^\infty(\mathbb{R}^n) \) converges to \( f \) both in \( L^1 \) and \( L^2 \). Hence the claim holds for every \( f \) with compact support. Finally, for general \( f \in L^1(\mathbb{R}^n) \cap L^2(\mathbb{R}^n) \) consider \( f_m = f \chi_{B_m(0)} \). Then \( f_m \to f \) in both \( L^1(\mathbb{R}^n) \) and \( L^2(\mathbb{R}^n) \) and the claim follows.
In particular,
\[ \hat{f}(p) = \lim_{m \to \infty} \frac{1}{(2\pi)^{n/2}} \int_{|x| \leq m} e^{-ipx} f(x) d^n x, \tag{13.17} \]
where the limit has to be understood in \( L^2(\mathbb{R}^n) \) and can be omitted if \( f \in L^1(\mathbb{R}^n) \cap L^2(\mathbb{R}^n) \).

Another useful property is the convolution formula.

**Lemma 13.12.** The convolution
\[ (f * g)(x) = \int_{\mathbb{R}^n} f(y)g(x-y)d^n y = \int_{\mathbb{R}^n} f(x-y)g(y)d^n y \tag{13.18} \]
of two functions \( f, g \in L^1(\mathbb{R}^n) \) is again in \( L^1(\mathbb{R}^n) \) and we have Young’s inequality
\[ \|f * g\|_1 \leq \|f\|_1 \|g\|_1. \tag{13.19} \]
Moreover, its Fourier transform is given by
\[ (f * g)^\wedge = (2\pi)^{n/2} \hat{f} \hat{g}, \tag{13.20} \]

**Proof.** The fact that \( f * g \) is in \( L^1 \) together with Young’s inequality follows by applying Fubini’s theorem to \( h(x, y) = f(x - y)g(y) \) (in fact we have shown a more general version in Lemma 10.17). For the last claim we compute
\[
(f * g)^\wedge(p) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ipx} \int_{\mathbb{R}^n} f(y)g(x-y)d^n y d^n x
= \int_{\mathbb{R}^n} e^{-ipy} f(y) \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ip(x-y)} g(x-y)d^n x d^n y
= \int_{\mathbb{R}^n} e^{-ipy} f(y) \hat{g}(p)d^n y = (2\pi)^{n/2} \hat{f}(p) \hat{g}(p),
\]
where we have again used Fubini’s theorem. \( \Box \)

As a consequence we can also deal with the case of convolution on \( \mathcal{S}(\mathbb{R}^n) \) as well as on \( L^2(\mathbb{R}^n) \).

**Corollary 13.13.** The convolution of two \( \mathcal{S}(\mathbb{R}^n) \) functions as well as their product is in \( \mathcal{S}(\mathbb{R}^n) \) and
\[ (f * g)^\wedge = (2\pi)^{n/2} \hat{f} \hat{g}, \quad (fg)^\wedge = (2\pi)^{-n/2} \hat{f} \hat{g} \tag{13.21} \]
in this case.

**Proof.** Clearly the product of two functions in \( \mathcal{S}(\mathbb{R}^n) \) is again in \( \mathcal{S}(\mathbb{R}^n) \) (show this!). Since \( \mathcal{S}(\mathbb{R}^n) \subset L^1(\mathbb{R}^n) \) the previous lemma implies \( (f * g)^\wedge = (2\pi)^{n/2} \hat{f} \hat{g} \in \mathcal{S}(\mathbb{R}^n) \). Moreover, since the Fourier transform is injective on \( L^1(\mathbb{R}^n) \) we conclude \( f * g = (2\pi)^{n/2}(\hat{f} \hat{g})^\wedge \in \mathcal{S}(\mathbb{R}^n) \). Replacing \( f, g \) by \( \hat{f}, \hat{g} \)
Using integration by parts we have the
Replacing
Proof. in the last formula finally shows $\hat{f} \ast \hat{g} = (2\pi)^{n/2}(fg)^{\vee}$ and the claim follows by a simple change of variables using $\hat{f}(p) = \hat{f}(-p)$. □

**Corollary 13.14.** The convolution of two $L^2(\mathbb{R}^n)$ functions is in $\text{ Ran}(F) \subset C_0(\mathbb{R}^n)$ and we have $\|f \ast g\|_\infty \leq \|f\|_2\|g\|_2$ as well as

$$
(fg)^\wedge = (2\pi)^{-n/2}\hat{f} \ast \hat{g}, \quad (f \ast g)^\wedge = (2\pi)^{n/2}\hat{f}\hat{g}
$$

(13.22) in this case.

**Proof.** The inequality $\|f \ast g\|_\infty \leq \|f\|_2\|g\|_2$ is immediate from Cauchy–Schwarz and shows that the convolution is a continuous bilinear form from $L^2(\mathbb{R}^n)$ to $L^\infty(\mathbb{R}^n)$. Now take sequences $f_m, g_m \in \mathcal{S}(\mathbb{R}^n)$ converging to $f, g \in L^2(\mathbb{R}^n)$. Then using the previous corollary together with continuity of the Fourier transform from $L^1(\mathbb{R}^n)$ to $C_0(\mathbb{R}^n)$ and on $L^2(\mathbb{R}^n)$ we obtain

$$(fg)^\wedge = \lim_{m \to \infty} (f_m g_m)^\wedge = (2\pi)^{-n/2} \lim_{m \to \infty} \hat{f}_m \ast \hat{g}_m = (2\pi)^{-n/2}\hat{f} \ast \hat{g}.$$

Similarly,

$$(f \ast g)^\wedge = \lim_{m \to \infty} (f_m \ast g_m)^\wedge = (2\pi)^{n/2} \lim_{m \to \infty} \hat{f}_m \hat{g}_m = (2\pi)^{n/2}\hat{f}\hat{g}$$

from which that last claim follows since $F : \text{ Ran}(F) \to L^1(\mathbb{R}^n)$ is closed by Lemma 4.8. □

Finally, note that by looking at the Gaussian’s $\phi_\lambda(x) = \exp(-\lambda x^2/2)$ one observes that a well centered peak transforms into a broadly spread peak and vice versa. This turns out to be a general property of the Fourier transform known as **uncertainty principle.** One quantitative way of measuring this fact is to look at

$$
\|(x_j - x^0)f(x)\|_2^2 = \int_{\mathbb{R}^n} (x_j - x^0)^2 |f(x)|^2 d^n x
$$

(13.23) which will be small if $f$ is well concentrated around $x^0$ in the $j$’th coordinate direction.

**Theorem 13.15** (Heisenberg uncertainty principle). Suppose $f \in \mathcal{S}(\mathbb{R}^n)$. Then for any $x^0, p^0 \in \mathbb{R}$ we have

$$
\|(x_j - x^0)f(x)\|_2\|p_j - p^0\|_2 \geq \frac{\|f\|_2^2}{2}.
$$

(13.24)

**Proof.** Replacing $f(x)$ by $e^{ix_j p^0} f(x + x^0 e_j)$ (where $e_j$ is the unit vector into the $j$’th coordinate direction) we can assume $x^0 = p^0 = 0$ by Lemma 13.2. Using integration by parts we have

$$
\|f\|_2^2 = \int_{\mathbb{R}^n} |f(x)|^2 d^n x = -\int_{\mathbb{R}^n} x_j \partial_j |f(x)|^2 d^n x = -2\text{Re} \int_{\mathbb{R}^n} x_j f(x) \ast \partial_j f(x) d^n x.
$$
Hence, by Cauchy–Schwarz,
\[ \|f\|^2 \leq 2|\partial_x f(x)|\|\partial_x f(x)\| = 2|\partial_x f(x)|\|\hat{f}(p)\| \]
the claim follows. \(\Box\)

The name stems from quantum mechanics, where \(|f(x)|^2\) is interpreted as the probability distribution for the position of a particle and \(|\hat{f}(x)|^2\) is interpreted as the probability distribution for its momentum. Equation (13.24) says that the variance of both distributions cannot both be small and thus one cannot simultaneously measure position and momentum of a particle with arbitrary precision.

Another version states that \(f\) and \(\hat{f}\) cannot both have compact support.

**Theorem 13.16.** Suppose \(f \in L^2(\mathbb{R}^n)\). If both \(f\) and \(\hat{f}\) have compact support, then \(f = 0\).

**Proof.** Let \(A, B \subseteq \mathbb{R}^n\) be two compact sets and consider the subspace of all functions with \(\text{supp}(f) \subseteq A\) and \(\text{supp}(\hat{f}) \subseteq B\). Then
\[ f(x) = \int_{\mathbb{R}^n} K(x, y) f(y) d^n y, \]
where
\[ K(x, y) = \frac{1}{(2\pi)^n} \int_B e^{i(x-y) \cdot p} \chi_A(y) d^n p = \frac{1}{(2\pi)^n} \hat{\chi}_B(y-x) \chi_A(y). \]
Since \(K \in L^2(\mathbb{R}^n \times \mathbb{R}^n)\) the corresponding integral operator is Hilbert–Schmidt, and thus its eigenspace corresponding to the eigenvalue 1 can be at most finite dimensional.

Now if there is a nonzero \(f\), we can find a sequence of vectors \(x^n \to 0\) such the functions \(f_n(x) = f(x - x^n)\) are linearly independent (look at their supports) and satisfy \(\text{supp}(f_n) \subseteq 2A\), \(\text{supp}(\hat{f}_n) \subseteq B\). But this a contradiction by the first part applied to the sets \(2A\) and \(B\). \(\Box\)

**Problem 13.1.** Show that \(S(\mathbb{R}^n) \subset L^p(\mathbb{R}^n)\). (Hint: If \(f \in S(\mathbb{R}^n)\), then \(|f(x)| \leq C_m \prod_{j=1}^n (1 + x_j^2)^{-m}\) for every \(m\).)

**Problem 13.2.** Show that \(F^1(\mathbb{R}^n) \subset L^p(\mathbb{R}^n)\) with
\[ \|f\|_p \leq (2\pi)^{\frac{n}{2}} \|f\|_1 \|\hat{f}\|^{1-\frac{1}{p}}. \]
Moreover, show that \(S(\mathbb{R}^n) \subset F^1(\mathbb{R}^n)\) and conclude that \(F^1(\mathbb{R}^n)\) is dense in \(L^p(\mathbb{R}^n)\) for \(p \in [1, \infty)\). (Hint: Use \(x^p \leq x\) for \(0 \leq x \leq 1\) to show \(\|f\|_p \leq \|f\|_1^{1/p} \|f\|_1^{1/p}\).)

**Problem 13.3.** Suppose \(f_j \in L^1(\mathbb{R}), \ j = 1, \ldots, n\) and set \(f(x) = \prod_{j=1}^n f_j(x_j)\). Show that \(f \in L^1(\mathbb{R}^n)\) with \(\|f\|_1 = \prod_{j=1}^n \|f_j\|_1\) and \(\hat{f}(p) = \prod_{j=1}^n \hat{f}_j(p_j)\).
Problem 13.4. Compute the Fourier transform of the following functions \( f : \mathbb{R} \to \mathbb{C} \):

(i) \( f(x) = \chi_{(-1,1)}(x) \).

(ii) \( f(x) = \frac{1}{x^2 + k^2}, \quad \text{Re}(k) > 0 \).

Problem 13.5. A function \( f : \mathbb{R}^n \to \mathbb{C} \) is called spherically symmetric if it is invariant under rotations; that is, \( f(Ox) = f(x) \) for all \( O \in \text{SO}(\mathbb{R}^n) \) (equivalently, \( f \) depends only on the distance to the origin \(|x|\)). Show that the Fourier transform of a spherically symmetric function is again spherically symmetric.

Problem 13.6. Suppose \( f \in L^1(\mathbb{R}^n) \). If \( f \) is continuous at 0 and \( \hat{f}(p) \geq 0 \) then

\[
f(0) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} \hat{f}(p) d^n p.
\]

Use this to show the Plancherel identity for \( f \in L^1(\mathbb{R}^n) \cap L^2(\mathbb{R}^n) \) by applying it to \( F := f \ast \hat{f}, \) where \( \hat{f}(x) = f(-x^*) \).

Problem 13.7. Show that \( F : L^1(\mathbb{R}^n) \to C_0(\mathbb{R}^n) \) is not onto as follows:

(i) The range of \( F \) is dense.

(ii) \( F \) is onto if and only if it has a bounded inverse.

(iii) \( F \) has no bounded inverse.

(Hint for (iii): Consider \( \phi_z(x) = \exp(-z x^2/2) \) for \( z = \lambda + i\omega \) with \( \lambda > 0 \).)

Problem 13.8 (Wiener). Suppose \( f \in L^2(\mathbb{R}^n) \). Then the set \( \{ f(x+a) | a \in \mathbb{R}^n \} \) is total in \( L^2(\mathbb{R}^n) \) if and only if \( \hat{f}(p) \neq 0 \) a.e. (Hint: Use Lemma 13.2 and the fact that a subspace is total if and only if its orthogonal complement is zero.)

Problem 13.9. Suppose \( f(x)e^{k|x|} \in L^1(\mathbb{R}) \) for some \( k > 0 \). Then \( \hat{f}(p) \) has an analytic extension to the strip \( |\text{Im}(p)| < k \).

Problem 13.10. The Fourier transform of a complex measure \( \mu \) is defined by

\[
\hat{\mu}(p) := \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ipx} d\mu(x).
\]

Show that the Fourier transform is a bounded injective map from \( \mathcal{M}(\mathbb{R}^n) \to C_b(\mathbb{R}^n) \) satisfying \( \|\hat{\mu}\|_{\infty} \leq (2\pi)^{-n/2}\|\mu\|_{\mathbb{R}^n} \). Moreover, show \( (\mu * \nu)^w = (2\pi)^{n/2} \hat{\mu} \hat{\nu} \) (see Problem 10.25). (Hint: To see injectivity use Lemma 10.21 with \( f = 1 \) and a Schwartz function \( \varphi \).)

Problem 13.11. Consider the Fourier transform of a complex measure on \( \mathbb{R} \) as in the previous problem. Show that

\[
\lim_{r \to \infty} \frac{1}{\sqrt{2\pi}} \int_{-r}^{r} e^{ipr} - e^{ip} \frac{\hat{\mu}(p) dp}{ip} = \frac{\mu([a, b]) + \mu((a, b))}{2}.
\]
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(Hint: Insert the definition of \( \hat{\mu} \) and then use Fubini. To evaluate the limit you will need the Dirichlet integral from Problem 13.24).

Problem 13.12 (Lévy continuity theorem). Let \( \mu_m(\mathbb{R}^n), \mu(\mathbb{R}^n) \leq M \) be positive measures and suppose the Fourier transforms (cf. Problem 13.10) converge pointwise

\[ \hat{\mu}_m(p) \to \hat{\mu}(p) \]

for \( p \in \mathbb{R}^n \). Then \( \mu_n \to \mu \) vaguely. In fact, we have (11.58) for all \( f \in C_b(\mathbb{R}^n) \). (Hint: Show \( \int \hat{f} d\mu = \int f(p) \hat{\mu}(p) dp \) for \( f \in L^1(\mathbb{R}^n) \).)

13.2. Applications to linear partial differential equations

By virtue of Lemma 13.4 the Fourier transform can be used to map linear partial differential equations with constant coefficients to algebraic equations, thereby providing a mean of solving them. To illustrate this procedure we look at the famous Poisson equation, that is, given a function \( g \), find a function \( f \) satisfying

\[ -\Delta f = g. \tag{13.25} \]

For simplicity, let us start by investigating this problem in the space of Schwartz functions \( S(\mathbb{R}^n) \). Assuming there is a solution we can take the Fourier transform on both sides to obtain

\[ |p|^2\hat{f}(p) = \hat{g}(p) \Rightarrow \hat{f}(p) = |p|^{-2}\hat{g}(p). \tag{13.26} \]

Since the right-hand side is integrable for \( n \geq 3 \) we obtain that our solution is necessarily given by

\[ f(x) = (|p|^{-2}\hat{g}(p))^\vee(x). \tag{13.27} \]

In fact, this formula still works provided \( g(x), |p|^{-2}\hat{g}(p) \in L^1(\mathbb{R}^n) \). Moreover, if we additionally assume \( \hat{g} \in L^1(\mathbb{R}^n) \), then \( |p|^2\hat{f}(p) = \hat{g}(p) \in L^1(\mathbb{R}^n) \) and Lemma 13.3 implies that \( f \in C^2(\mathbb{R}^n) \) as well as that it is indeed a solution. Note that if \( n \geq 3 \), then \( |p|^{-2}\hat{g}(p) \in L^1(\mathbb{R}^n) \) follows automatically from \( g, \hat{g} \in L^1(\mathbb{R}^n) \) (show this!).

Moreover, we clearly expect that \( f \) should be given by a convolution. However, since \( |p|^{-2} \) is not in \( L^p(\mathbb{R}^n) \) for any \( p \), the formulas derived so far do not apply.

Lemma 13.17. Let \( 0 < \alpha < n \) and suppose \( g \in L^1(\mathbb{R}^n) \cap L^\infty(\mathbb{R}^n) \) as well as \( |p|^{-\alpha}\hat{g}(p) \in L^1(\mathbb{R}^n) \). Then

\[ (|p|^{-\alpha}\hat{g}(p))^\vee(x) = \int_{\mathbb{R}^n} I_\alpha(|x-y|)g(y) d^n y, \tag{13.28} \]

where the Riesz potential is given by

\[ I_\alpha(r) = \frac{\Gamma(\frac{n-\alpha}{2})}{2^\alpha \pi^{n/2} \Gamma(\frac{\alpha}{2})} \frac{1}{r^{n-\alpha}}. \tag{13.29} \]
Proof. Note that, while \(|.|^{-\alpha}\) is not in \(L^p(\mathbb{R}^n)\) for any \(p\), our assumption \(0 < \alpha < n\) ensures that the singularity at zero is integrable.

We set \(\phi_t(p) = \exp(-t|p|^2/2)\) and begin with the elementary formula

\[
|p|^{-\alpha} = c_\alpha \int_0^\infty \phi_t(p)t^\alpha/2-1 dt, \quad c_\alpha = \frac{1}{2^{\alpha/2}\Gamma(\alpha/2)},
\]

which follows from the definition of the gamma function (Problem 9.23) after a simple scaling. Since \(|p|^{-\alpha}\hat{g}(p)\) is integrable we can use Fubini and Lemma 13.6 to obtain

\[
(|p|^{-\alpha}\hat{g}(p))^\wedge(x) = \frac{c_\alpha}{(2\pi)^n/2} \int_{\mathbb{R}^n} e^{i xp} \left( \int_0^\infty \phi_t(p)t^\alpha/2-1 dt \right) \hat{g}(p)d^n p
\]

\[
= \frac{c_\alpha}{(2\pi)^n/2} \int_0^\infty \left( \int_{\mathbb{R}^n} e^{i xp} \phi_1/t(p) \hat{g}(p)d^n p \right) t^{(\alpha-n)/2-1} dt.
\]

Since \(\phi, g \in L^1\) we know by Lemma 13.12 that \(\hat{\phi} \hat{g} = (2\pi)^{-n/2}(\phi * g)^\wedge\). Moreover, since \(\hat{\phi} \hat{g} \in L^1\) Theorem 13.7 gives us \((\hat{\phi} \hat{g})^\vee = (2\pi)^{-n/2} \phi * g\). Thus, we can make a change of variables and use Fubini once again (since \(g \in L^\infty\))

\[
(|p|^{-\alpha}\hat{g}(p))^\wedge(x) = \frac{c_\alpha}{(2\pi)^n/2} \int_0^\infty \left( \int_{\mathbb{R}^n} \phi_1/t(x-y) g(y) dy \right) t^{(\alpha-n)/2-1} dt
\]

\[
= \frac{c_\alpha}{(2\pi)^n/2} \int_0^\infty \left( \int_{\mathbb{R}^n} \phi_t(x-y) g(y) dy \right) t^{(n-\alpha)/2-1} dt
\]

\[
= \frac{c_\alpha}{(2\pi)^n/2} \int_{\mathbb{R}^n} \left( \int_0^\infty \phi_t(x-y) t^{(n-\alpha)/2-1} dt \right) g(y)d^n y
\]

\[
= \frac{c_\alpha}{(2\pi)^n/2} \int_{\mathbb{R}^n} \frac{g(y)}{|x-y|^{n-\alpha}}d^n y
\]

to obtain the desired result. \(\square\)

Note that the conditions of the above theorem are, for example, satisfied if \(g, \hat{g} \in L^1(\mathbb{R}^n)\) which holds, for example, if \(g \in \mathcal{S}(\mathbb{R}^n)\). In summary, if \(g \in L^1(\mathbb{R}^n) \cap L^\infty(\mathbb{R}^n), \ |p|^{-2}\hat{g}(p) \in L^1(\mathbb{R}^n)\) and \(n \geq 3\), then

\[
f = \Phi \ast g \quad (13.30)
\]

is a classical solution of the Poisson equation, where

\[
\Phi(x) = \frac{\Gamma\left(\frac{n}{2} - 1\right)}{4\pi^{n/2}} \frac{1}{|x|^{n-2}}, \quad n \geq 3, \quad (13.31)
\]
is known as the fundamental solution of the Laplace equation.

A few words about this formula are in order. First of all, our original formula in Fourier space shows that the multiplication with \(|p|^{-2}\) improves the decay of \(\hat{g}\) and hence, by virtue of Lemma 13.4, \(f\) should have, roughly
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speaking, two derivatives more than \( g \). However, unless \( \hat{g}(0) \) vanishes, multiplication with \( |p|^{-2} \) will create a singularity at 0 and hence, again by Lemma 13.4, \( f \) will not inherit any decay properties from \( g \). In fact, evaluating the above formula with \( g = \chi_{B_1(0)} \) (Problem 13.13) shows that \( f \) might not decay better than \( \Phi \) even for \( g \) with compact support.

Moreover, our conditions on \( g \) might not be easy to check as it will not be possible to compute \( \hat{g} \) explicitly in general. So if one wants to deduce \( \hat{g} \in L^1(\mathbb{R}^n) \) from properties of \( g \), one could use Lemma 13.4 together with the Riemann–Lebesgue lemma to show that this condition holds if \( g \in C^k(\mathbb{R}^n) \), \( k > n - 2 \), such that all derivatives are integrable and all derivatives of order less than \( k \) vanish at \( \infty \) (Problem 13.14). This seems a rather strong requirement since our solution formula will already make sense under the sole assumption \( g \in L^1(\mathbb{R}^n) \). However, as the example \( g = \chi_{B_1(0)} \) shows, this solution might not be \( C^2 \) and hence one needs to weaken the notion of a solution if one wants to include such situations. This will lead us to the concepts of weak derivatives and Sobolev spaces. As a preparation we will develop some further tools which will allow us to investigate continuity properties of the operator \( \mathcal{L}_\alpha f = I_\alpha * f \) in the next section.

Before that, let us summarize the procedure in the general case. Suppose we have the following linear partial differential equations with constant coefficients:

\[
P(i\partial) f = g, \quad P(i\partial) = \sum_{\alpha \leq k} c_\alpha i^{|\alpha|} \partial_\alpha. \tag{13.32}
\]

Then the solution can be found via the procedure

\[
\begin{align*}
\hat{g} & \quad \xrightarrow{\mathcal{F}} \quad P^{-1} \hat{g} \\
g & \quad \xrightarrow{\mathcal{F}} \quad f
\end{align*}
\]

and is formally given by

\[
f(x) = (P(p)^{-1} \hat{g}(p))^\mathcal{V}(x). \tag{13.33}
\]

It remains to investigate the properties of the solution operator. In general, given a measurable function \( m \) one might try to define a corresponding operator via

\[
A_m f = (m\hat{g})^\mathcal{V}, \tag{13.34}
\]

in which case \( m \) is known as a **Fourier multiplier**. It is said to be an \( L^p \)-multiplier if \( A_m \) can be extended to a bounded operator in \( L^p(\mathbb{R}^n) \). For
example, it will be an $L^2$ multiplier if $m$ is bounded (in fact the converse is also true — Problem 13.15). As we have seen, in some cases $A_m$ can be expressed as a convolution, but this is not always the case as the trivial example $m = 1$ (corresponding to the identity operator) shows.

Another famous example which can be solved in this way is the Helmholtz equation

$$-\Delta f + f = g.$$  \hspace{1cm} (13.35)

As before we find that if $g(x), (1 + |p|^2)^{-1} \hat{g}(p) \in L^1(\mathbb{R}^n)$ then the solution is given by

$$f(x) = ((1 + |p|^2)^{-1} \hat{g}(p))^\vee(x).$$ \hspace{1cm} (13.36)

**Lemma 13.18.** Let $\alpha > 0$. Suppose $g \in L^1(\mathbb{R}^n) \cap L^\infty(\mathbb{R}^n)$ as well as $(1 + |p|^2)^{-\alpha/2} \hat{g}(p) \in L^1(\mathbb{R}^n)$. Then

$$((1 + |p|^2)^{-\alpha/2} \hat{g}(p))^\vee(x) = \int_{\mathbb{R}^n} J_{(n-\alpha)/2}(|x - y|)g(y)d^n y,$$ \hspace{1cm} (13.37)

where the Bessel potential is given by

$$J_\alpha(r) = \frac{2}{(4\pi)^{n/2} \Gamma(\frac{n}{2})} \left(\frac{r}{2}\right)^{(n-\alpha)/2} K_{(n-\alpha)/2}(r), \quad r > 0,$$ \hspace{1cm} (13.38)

with

$$K_\nu(r) = K_{-\nu}(r) = \frac{1}{2} \left(\frac{r}{2}\right)^\nu \int_0^\infty e^{-t-i^\nu r^2} \frac{dt}{t^{\nu+1}}, \quad r > 0, \nu \in \mathbb{R},$$ (13.39)

the modified Bessel function of the second kind of order $\nu$ ([25, (10.32.10)]).

**Proof.** We proceed as in the previous lemma. We set $\phi_t(p) = \exp(-t|p|^2/2)$ and begin with the elementary formula

$$\frac{\Gamma(\frac{n}{2})}{(1 + |p|^2)^{\alpha/2}} = \int_0^\infty t^{\alpha/2-1}e^{-t(1+|p|^2)}dt.$$

Since $g, \hat{g}(p)$ are integrable we can use Fubini and Lemma 13.6 to obtain

$$\left(\frac{\hat{g}(p)}{(1 + |p|^2)^{\alpha/2}}\right)^\vee(x) = \frac{\Gamma(\frac{n}{2})^{-1}}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{i\text{exp}\left(\int_0^\infty t^{\alpha/2-1}e^{-t(1+|p|^2)}dt\right)} \hat{g}(p)d^n p$$

$$= \frac{\Gamma(\frac{n}{2})^{-1}}{(4\pi)^{n/2}} \int_0^\infty \left(\int_{\mathbb{R}^n} e^{i\text{exp}\phi_{1/2t}(p)\hat{g}(p)d^n p} e^{-t t^{(\alpha-n)/2-1}dt}\right)$$

$$= \frac{\Gamma(\frac{n}{2})^{-1}}{(4\pi)^{n/2}} \int_0^\infty \left(\int_{\mathbb{R}^n} \phi_{1/2t}(x - y)g(y)d^n y\right) e^{-t t^{(\alpha-n)/2-1}dt}$$

$$= \frac{\Gamma(\frac{n}{2})^{-1}}{(4\pi)^{n/2}} \int_{\mathbb{R}^n} \phi_{1/2t}(x - y)e^{-t t^{(\alpha-n)/2-1}dt} g(y)d^n y$$

to obtain the desired result. Using Fubini in the last step is allowed since $g$ is bounded and $J_\alpha(|x|) \in L^1(\mathbb{R}^n)$ (Problem 13.16). \hfill \Box
Note that since the first condition \( g \in L^1(\mathbb{R}^n) \cap L^\infty(\mathbb{R}^n) \) implies \( g \in L^2(\mathbb{R}^n) \) and thus the second condition \((1 + |p|^2)^{-\alpha/2} \hat{g}(p) \in L^1(\mathbb{R}^n)\) will be satisfied if \( \frac{\alpha}{2} < \alpha \).

In particular, if \( g, \hat{g} \in L^1(\mathbb{R}^n) \), then
\[
f = J_1 * g
\]
is a solution of Helmholtz equation. Note that since our multiplier \((1 + |p|^2)^{-1}\) does not have a singularity near zero, the solution \( f \) will preserve (some) decay properties of \( g \). For example, it will map Schwartz functions to Schwartz functions and thus for every \( g \in \mathcal{S}(\mathbb{R}^n) \) there is a unique solution of the Helmholtz equation \( f \in \mathcal{S}(\mathbb{R}^n) \). This is also reflected by the fact that the Bessel potential decays much faster than the Riesz potential. Indeed, one can show that \([25, (10.25.3)]\)
\[
K_\nu(r) = \sqrt{\frac{\pi}{2r}} \exp\left(-r(1 + O(r^{-1}))\right)
\]
as \( r \to \infty \). The singularity near zero is of the same type as for \( I_\alpha \) since (see \([25, (10.30.2) \text{ and } (10.30.3)]\))
\[
K_\nu(r) = \begin{cases} 
\frac{\Gamma(\nu)}{2} \left(\frac{r}{2}\right)^{-\nu} + O(r^{-\nu+2}), & \nu > 0, \\
-\log(\frac{r}{2}) + O(1), & \nu = 0,
\end{cases}
\]
for \( r \to 0 \).

Problem 13.13. Show that for \( n = 3 \) we have
\[
(\Phi * \chi_{B_1(0)})(x) = \begin{cases} 
\frac{1}{4\pi|x|}, & |x| \geq 1, \\
\frac{1-|x|^2}{6}, & |x| \leq 1.
\end{cases}
\]
(Hint: Observe that the result depends only on \(|x|\). Then choose \( x = (0, 0, R) \) and evaluate the integral using spherical coordinates.)

Problem 13.14. Suppose \( g \in C^k(\mathbb{R}^n) \) and \( \partial^l_j g \in L^1(\mathbb{R}^n) \) for \( j = 1, \ldots, n \) and \( 0 \leq l \leq k \) as well as \( \lim_{|x| \to \infty} \partial^l_j g(x) = 0 \) for \( j = 1, \ldots, n \) and \( 0 \leq l < k \). Then
\[
|\hat{g}(p)| \leq \frac{C}{(1 + |p|^2)^{k/2}}.
\]

Problem 13.15. Show that \( m \) is an \( L^2 \) multiplier if and only if \( m \in L^\infty(\mathbb{R}^n) \).

Problem 13.16. Show
\[
\int_0^\infty J_\alpha(r)r^{n-1}dr = \frac{\Gamma(n/2)}{2\pi^{n/2}}, \quad \alpha > 0.
\]
Conclude that
\[
||J_\alpha * g||_p \leq ||g||_p.
\]
13.3. Sobolev spaces

We have already introduced Sobolev spaces in Section 10.6. In this section we present an alternate (and in particular independent) approach to Sobolev spaces of index two (the Hilbert space case) on all of $\mathbb{R}^n$.

We begin by introducing the **Sobolev space**

$$H^s(\mathbb{R}^n) = \{ f \in L^2(\mathbb{R}^n) \mid ||p|^r \hat{f}(p) \in L^2(\mathbb{R}^n) \}. \quad (13.43)$$

The most important case is when $r$ is an integer, however our definition makes sense for any $r \geq 0$. Moreover, note that $H^r(\mathbb{R}^n)$ becomes a Hilbert space if we introduce the scalar product

$$\langle f, g \rangle = \int_{\mathbb{R}^n} \hat{f}(p)^* \hat{g}(p)(1 + |p|^2)^r d^n p. \quad (13.44)$$

In particular, note that by construction $\mathcal{F}$ maps $H^r(\mathbb{R}^n)$ unitarily onto $L^2(\mathbb{R}^n), (1 + |p|^2)^r d^n p$. Clearly $H^{r+1}(\mathbb{R}^n) \subset H^r(\mathbb{R}^n)$ with the embedding being continuous. Moreover, $S(\mathbb{R}^n) \subset H^r(\mathbb{R}^n)$ and this subset is dense (since $S(\mathbb{R}^n)$ is dense in $L^2(\mathbb{R}^n), (1 + |p|^2)^r d^n p$).

The motivation for the definition (13.43) stems from Lemma 13.4 which allows us to extend differentiation to a larger class. In fact, every function in $H^r(\mathbb{R}^n)$ has partial derivatives up to order $[r]$, which are defined via

$$\partial_\alpha f = ((ip)^\alpha \hat{f}(p))^\vee, \quad f \in H^r(\mathbb{R}^n), |\alpha| \leq r. \quad (13.45)$$

By Lemma 13.4 this definition coincides with the usual one for every $f \in S(\mathbb{R}^n)$.

**Example.** Consider $f(x) = (1 - |x|)\chi_{[-1,1]}(x)$. Then $\hat{f}(p) = \sqrt{\frac{2 \cos(p)-1}{p^2}}$ and $f \in H^1(\mathbb{R})$. The weak derivative is $f'(x) = -\text{sign}(x)\chi_{[-1,1]}(x)$.

We also have

$$\int_{\mathbb{R}^n} g(x) (\partial_\alpha f)(x) d^n x = \langle g^*, (\partial_\alpha f) \rangle = \langle \hat{g}(p)^* (ip)^\alpha \hat{f}(p) \rangle$$

$$= (-1)^{|\alpha|} \langle (ip)^\alpha \hat{g}(p)^*, \hat{f}(p) \rangle = (-1)^{|\alpha|} \langle \partial_\alpha g^*, f \rangle$$

$$= (-1)^{|\alpha|} \int_{\mathbb{R}^n} (\partial_\alpha g)(x) f(x) d^n x, \quad (13.46)$$

for $f, g \in H^r(\mathbb{R}^n)$. Furthermore, recall that a function $h \in L^1_{loc}(\mathbb{R}^n)$ satisfying

$$\int_{\mathbb{R}^n} \varphi(x) h(x) d^n x = (-1)^{|\alpha|} \int_{\mathbb{R}^n} (\partial_\alpha \varphi)(x) f(x) d^n x, \quad \forall \varphi \in C_c^\infty(\mathbb{R}^n), \quad (13.47)$$

(Hint: Fubini.)
is called the weak derivative or the derivative in the sense of distributions of \( f \) (by Lemma 10.21 such a function is unique if it exists). Hence, choosing \( g = \varphi \) in (13.46), we see that functions in \( H^r(\mathbb{R}^n) \) have weak derivatives up to order \( r \), which are in \( L^2(\mathbb{R}^n) \). Moreover, the weak derivatives coincide with the derivatives defined in (13.45). Conversely, given (13.47) with \( f, h \in L^2(\mathbb{R}^n) \) we can use that \( \mathcal{F} \) is unitary to conclude \( \int_{\mathbb{R}^n} \varphi(p)^* \hat{h}(p) d^n p = \int_{\mathbb{R}^n} p^\alpha \varphi(p) \hat{f}(p) d^n p \) for all \( \varphi \in C_0^\infty(\mathbb{R}^n) \). By approximation this follows for \( \varphi \in H^r(\mathbb{R}^n) \) with \( r \geq |\alpha| \) and hence in particular for \( \varphi \in C_0^\infty(\mathbb{R}^n) \). Consequently \( p^\alpha f(p) = \hat{h}(p) \) a.e. implying that \( f \in H^r(\mathbb{R}^n) \) if all weak derivatives exist up to order \( r \) and are in \( L^2(\mathbb{R}^n) \).

In this connection the following norm for \( H^m(\mathbb{R}^n) \) with \( m \in \mathbb{N}_0 \) is more common:

\[
\|f\|_{2,m}^2 = \sum_{|\alpha| \leq m} \|\partial_\alpha f\|_2^2. \tag{13.48}
\]

By \( |p^\alpha| \leq |p|^{|\alpha|} \leq (1 + |p|^2)^{m/2} \) it follows that this norm is equivalent to (13.44).

**Example.** This definition of a weak derivative is tailored for the method of solving linear constant coefficient partial differential equations as outlined in Section 13.2. While Lemma 13.3 only gives us a sufficient condition on \( \hat{f} \) for \( f \) to be differentiable, the weak derivatives gives us necessary and sufficient conditions. For example, we see that the Poisson equation (13.25) will have a (unique) solution \( f \in H^2(\mathbb{R}^n) \) if and if \( |p|^{-2} \hat{g} \in L^2(\mathbb{R}^n) \). That this is not true for all \( g \in L^2(\mathbb{R}^n) \) is connected with the fact that \( |p|^{-2} \) is unbounded and hence no \( L^2 \) multiplier (cf. Problem 13.15). Consequently the range of \( \Delta \) when defined on \( H^2(\mathbb{R}^n) \) will not be all of \( L^2(\mathbb{R}^n) \) and hence the Poisson equation is not solvable within the class \( H^2(\mathbb{R}^n) \) for all \( g \in L^2(\mathbb{R}^n) \). Nevertheless, we get a unique weak solution under some conditions. Under which conditions this weak solution is also a classical solution can then be investigated separately.

Note that the situation is even simpler for the Helmholtz equation (13.35) since the corresponding multiplier \( (1 + |p|^2)^{-1} \) does map \( L^2 \) to \( L^2 \). Hence we get that the Helmholtz equation has a unique solution \( f \in H^2(\mathbb{R}^n) \) if and only if \( g \in L^2(\mathbb{R}^n) \). Moreover, \( f \in H^{r+2}(\mathbb{R}^n) \) if and only if \( g \in H^r(\mathbb{R}^n) \). \( \diamond \)

Of course a natural question to ask is when the weak derivatives are in fact classical derivatives. To this end observe that the Riemann–Lebesgue lemma implies that \( \partial_\alpha f(x) \in C_0(\mathbb{R}^n) \) provided \( p^\alpha \hat{f}(p) \in L^1(\mathbb{R}^n) \). Moreover, in this situation the derivatives will exist as classical derivatives:

**Lemma 13.19.** Suppose \( f \in L^1(\mathbb{R}^n) \) or \( f \in L^2(\mathbb{R}^n) \) with \( (1 + |p|^k) \hat{f}(p) \in L^1(\mathbb{R}^n) \) for some \( k \in \mathbb{N}_0 \). Then \( f \in C^k_0(\mathbb{R}^n) \), the set of functions with
continuous partial derivatives of order \( k \) all of which vanish at \( \infty \). Moreover,

\[
(\partial_\alpha f)(p) = (ip)^\alpha \hat{f}(p), \quad |\alpha| \leq k, \tag{13.49}
\]
in this case.

**Proof.** We begin by observing that by Theorem 13.7

\[
f(x) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ipx} \hat{f}(p)d^np.
\]

Now the claim follows as in the proof of Lemma 13.4 by differentiating the integral using Problem 9.14. \( \square \)

Now we are able to prove the following embedding theorem.

**Theorem 13.20** (Sobolev embedding). Suppose \( r > k + \frac{n}{2} \) for some \( k \in \mathbb{N}_0 \). Then \( H^r(\mathbb{R}^n) \) is continuously embedded into \( C^k_\infty(\mathbb{R}^n) \) with

\[
\|\partial_\alpha f\|_\infty \leq C_{n,r} \|f\|_{2,r}, \quad |\alpha| \leq k. \tag{13.50}
\]

**Proof.** Abbreviate \( \langle p \rangle = (1 + |p|^2)^{1/2} \). Now use \( |(ip)^\alpha \hat{f}(p)| \leq \langle p \rangle^{|\alpha|} |\hat{f}(p)| = \langle p \rangle^{-s} \cdot \langle p \rangle^{s+|\alpha|} |\hat{f}(p)| \). Now \( \langle p \rangle^{-s} \in L^2(\mathbb{R}^n) \) if \( s > \frac{n}{2} \) (use polar coordinates to compute the norm) and \( \langle p \rangle^{s+|\alpha|} |\hat{f}(p)| \in L^2(\mathbb{R}^n) \) if \( s + |\alpha| \leq r \). Hence \( \langle p \rangle^{s+|\alpha|} |\hat{f}(p)| \in L^1(\mathbb{R}^n) \) and the claim follows from the previous lemma. \( \square \)

In fact, we can even do a bit better.

**Lemma 13.21** (Morrey inequality). Suppose \( f \in H^{n/2+\gamma}(\mathbb{R}^n) \) for some \( \gamma \in (0, 1) \). Then \( f \in C^{0,\gamma}_0(\mathbb{R}^n) \), the set of functions which are Hölder continuous with exponent \( \gamma \) and vanish at \( \infty \). Moreover,

\[
|f(x) - f(y)| \leq C_{n,\gamma} \|\hat{f}(p)\|_{2,n/2+\gamma} |x - y|^{\gamma} \tag{13.51}
\]
in this case.

**Proof.** We begin with

\[
f(x + y) - f(x) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ipx}(e^{ipy} - 1)\hat{f}(p)d^np
\]

implying

\[
|f(x + y) - f(x)| \leq \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} \left|\frac{e^{ipy} - 1}{\langle p \rangle^{n/2+\gamma}}\right| \langle p \rangle^{n/2+\gamma} |\hat{f}(p)|d^np,
\]
where again $\langle p \rangle = (1 + |p|^2)^{1/2}$. Hence, after applying Cauchy–Schwarz, it remains to estimate (recall (9.42))

$$
\int_{\mathbb{R}^n} \left| e^{ipy} - 1 \right|^2 \frac{d^n p}{\langle p \rangle^{n+2\gamma}} \leq S_n \int_0^{1/|y|} \frac{\langle |y|r \rangle^2}{\langle r \rangle^{n+2\gamma}} r^{n-1} dr + S_n \int_{1/|y|}^{\infty} \frac{4}{\langle r \rangle^{n+2\gamma}} r^{n-1} dr
$$

\[
\leq \frac{S_n}{2(1-\gamma)} |y|^{2\gamma} + \frac{S_n}{2\gamma} |y|^{2\gamma} = \frac{S_n}{2\gamma(1-\gamma)} |y|^{2\gamma},
\]

where $S_n = nV_n$ is the surface area of the unit sphere in $\mathbb{R}^n$. \hfill \Box

Using this lemma we immediately obtain:

**Corollary 13.22.** Suppose $r \geq k + \gamma + \frac{n}{2}$ for some $k \in \mathbb{N}_0$ and $\gamma \in (0,1)$. Then $H^r(\mathbb{R}^n)$ is continuously embedded into $C^{k,\gamma}(\mathbb{R}^n)$, the set of functions in $C^k(\mathbb{R}^n)$ whose highest derivatives are Hölder continuous of exponent $\gamma$.

**Example.** The function $f(x) = \log(|x|)$ is in $H^1(\mathbb{R}^n)$ for $n \geq 3$. In fact, the weak derivatives are given by

$$
\partial_j f(x) = \frac{x_j}{|x|^2}, \quad (13.52)
$$

However, observe that $f$ is not continuous. \hfill \diamond

The last example shows that in the case $r < \frac{n}{2}$ functions in $H^r$ are no longer necessarily continuous. In this case we at least get an embedding into some better $L^p$ space:

**Theorem 13.23 (Sobolev inequality).** Suppose $0 < r < \frac{n}{2}$. Then $H^r(\mathbb{R}^n)$ is continuously embedded into $L^p(\mathbb{R}^n)$ with $p = \frac{2n}{n-2r}$, that is,

$$
\|f\|_p \leq \tilde{C}_{n,r} \|\rho^r \hat{f}(p)\|_2 \leq C_{n,r} \|f\|_2, \quad (13.53)
$$

**Proof.** We will give a prove based on the Hardy–Littlewood–Sobolev inequality to be proven in Theorem 14.10 below.

It suffices to prove the first inequality. Set $|\rho|^{r} \hat{f}(p) = \hat{g}(p) \in L^2$. Moreover, choose some sequence $f_m \in \mathcal{S} \to f \in H^r$. Then, by Lemma 13.17 $f_m = \mathcal{I}_r g_m$, and since the Hardy–Littlewood–Sobolev inequality implies that the map $\mathcal{I}_r : L^2 \to L^p$ is continuous, we have $\|f_m\|_p = \|\mathcal{I}_r g_m\|_p \leq \tilde{C} \|g_m\|_2 = \tilde{C} \|\rho^r \hat{f}_m(p)\|_2$ and the claim follows after taking limits. \hfill \Box

**Problem 13.17.** Use dilations $f(x) \mapsto f(\lambda x)$, $\lambda > 0$, to show that $p = \frac{2n}{n-2r}$ is the only index for which the Sobolev inequality $\|f\|_p \leq \tilde{C}_{n,r} \|\rho^r \hat{f}(p)\|_2$ can hold.
Problem 13.18. Suppose $f \in L^2(\mathbb{R}^n)$ show that $\varepsilon^{-1}(f(x + e_j \varepsilon) - f(x)) \to g_j(x)$ in $L^2$ if and only if $p_j \hat{f}(p) \in L^2$, where $e_j$ is the unit vector into the $j$'th coordinate direction. Moreover, show $g_j = \partial_j f$ if $f \in H^1(\mathbb{R}^n)$.

Problem 13.19. Show that $u$ is weakly differentiable in the interval $(0,1)$ if and only if $u$ is absolutely continuous and $u' = v$ in this case. (Hint: You will need that $\int_0^1 u(t)\varphi'(t)\,dt = 0$ for all $\varphi \in C_c^\infty(0,1)$ if and only if $u$ is constant, which is known as du Bois-Reymond lemma (Problem 10.26). To see this, choose some $\varphi_0 \in C_c^\infty(0,1)$ with $I(\varphi_0) = \int_0^1 \varphi_0(t)\,dt = 1$. Then invoke Lemma 10.21 and use that every $\varphi \in C_c^\infty(0,1)$ can be written as $\varphi(t) = \Phi'(t) + I(\varphi)\varphi_0(t)$ with $\Phi(t) = \int_0^t \varphi(s)\,ds - I(\varphi)\int_0^t \varphi_0(s)\,ds$.)

13.4. Applications to evolution equations

In this section we want to show how to apply these considerations to evolution equations. As a prototypical example we start with the Cauchy problem for the heat equation

$$u_t - \Delta u = 0, \quad u(0) = g. \quad (13.54)$$

It turns out useful to view $u(t, x)$ as a function of $t$ with values in a Banach space $X$. To this end we let $I \subseteq \mathbb{R}$ be some interval and denote by $C(I, X)$ the set of continuous functions from $I$ to $X$. Given $t \in I$ we call $u : I \to X$ differentiable at $t$ if the limit

$$\dot{u}(t) = \lim_{\varepsilon \to 0} \frac{u(t + \varepsilon) - u(t)}{\varepsilon} \quad (13.55)$$

exists. The set of functions $u : I \to X$ which are differentiable at all $t \in I$ and for which $\dot{u} \in C(I, X)$ is denoted by $C^1(I, X)$. As usual we set $C^{k+1}(I, X) = \{ u \in C^1(I, X) | \dot{u} \in C^k(I, X) \}$. Note that if $U \in \mathcal{L}(X, Y)$ and $u \in C^1(I, X)$, then $Uu \in C^1(I, Y)$ and $\frac{d}{dt}Uu = U\dot{u}$.

A strongly continuous operator semigroup (also $C_0$-semigroup) is a family of operators $T(t) \in \mathcal{L}(X)$, $t \geq 0$, such that

1. $T(t)g \in C([0, \infty), X)$ for every $g \in X$ (strong continuity) and
2. $T(0) = I$, $T(t + s) = T(t)T(s)$ for every $t, s \geq 0$ (semigroup property).

Given a strongly continuous semigroup we can define its generator $A$ as the linear operator

$$Af = \lim_{t \downarrow 0} \frac{1}{t} (T(t)f - f) \quad (13.56)$$

where the domain $\mathcal{D}(A)$ is precisely the set of all $f \in X$ for which the above limit exists. The key result is that if $A$ generates a $C_0$-semigroup $T(t)$,
then $u(t) := T(t)g$ will be the unique solution of the corresponding abstract Cauchy problem. More precisely we have (see Lemma 7.7):

**Lemma 13.24.** Let $T(t)$ be a $C_0$-semigroup with generator $A$. If $g \in X$ with $u(t) = T(t)g \in \mathcal{D}(A)$ for $t > 0$ then $u(t) \in C^1((0, \infty), X) \cap C([0, \infty), X)$ and $u(t)$ is the unique solution of the abstract Cauchy problem

$$
\dot{u}(t) = Au(t), \quad u(0) = g.
$$

(13.57)

This is, for example, the case if $g \in \mathcal{D}(A)$ in which case we even have $u(t) \in C^1([0, \infty), X)$.

After these preparations we are ready to return to our original problem (13.54). Let $g \in L^2(\mathbb{R}^n)$ and let $u \in C^1((0, \infty), L^2(\mathbb{R}^n))$ be a solution such that $u(t) \in H^2(\mathbb{R}^n)$ for $t > 0$. Then we can take the Fourier transform to obtain

$$
\hat{\dot{u}} + |p|^2 \hat{u} = 0, \quad \hat{u}(0) = \hat{g}.
$$

(13.58)

Next, one verifies (Problem 13.20) that the solution (in the sense defined above) of this differential equation is given by

$$
\hat{u}(t)(p) = \hat{g}(p)e^{-|p|^2t}.
$$

(13.59)

Accordingly, the solution of our original problem is

$$
u(t) = T_H(t)g, \quad T_H(t) = \mathcal{F}^{-1}e^{-|p|^2t} \mathcal{F}.
$$

(13.60)

Note that $T_H(t) : L^2(\mathbb{R}^n) \to L^2(\mathbb{R}^n)$ is a bounded linear operator with $\|T_H(t)\| \leq 1$ (since $|e^{-|p|^2t}| \leq 1$). In fact, for $t > 0$ we even have $T_H(t)g \in H^r(\mathbb{R}^n)$ for any $r \geq 0$ showing that $u(t)$ is smooth even for *rough* initial functions $g$. In summary,

**Theorem 13.25.** The family $T_H(t)$ is a $C_0$-semigroup whose generator is $\Delta$, $\mathcal{D}(\Delta) = H^2(\mathbb{R}^n)$.

**Proof.** That $H^2(\mathbb{R}^n) \subseteq \mathcal{D}(A)$ follows from Problem 13.20. Conversely, let $g \notin H^2(\mathbb{R}^n)$. Then $t^{-1}(e^{-|p|^2t} - 1) \to -|p|^2$ uniformly on every compact subset $K \subset \mathbb{R}^n$. Hence $\int_K |p|^2|\hat{g}(p)|^2\,d^np = \int_K |Ag(x)|^2\,d^nx$ which gives a contradiction as $K$ increases. \qed

Next we want to derive a more explicit formula for our solution. To this end we assume $g \in L^1(\mathbb{R}^n)$ and introduce

$$
\Phi_t(x) = \frac{1}{(4\pi t)^{n/2}}e^{-\frac{|x|^2}{4t}},
$$

(13.61)

known as the **fundamental solution** of the heat equation, such that

$$
\hat{\dot{u}}(t) = (2\pi)^{n/2}\hat{g}\hat{\Phi}_t = (\hat{\Phi}_t * \hat{g})^\wedge
$$

(13.62)
by Lemma 13.6 and Lemma 13.12. Finally, by injectivity of the Fourier transform (Theorem 13.7) we conclude

\[ u(t) = \Phi_t * g. \]  \hspace{1cm} (13.63)

Moreover, one can check directly that (13.63) defines a solution for arbitrary \( g \in L^p(\mathbb{R}^n) \).

**Theorem 13.26.** Suppose \( g \in L^p(\mathbb{R}^n), 1 \leq p \leq \infty \). Then (13.63) defines a solution for the heat equation which satisfies \( u \in C^\infty((0, \infty) \times \mathbb{R}^n) \). The solution has the following properties:

(i) If \( 1 \leq p < \infty \), then \( \lim_{t \downarrow 0} u(t) = g \) in \( L^p \). If \( p = \infty \) this holds for \( g \in C_0(\mathbb{R}^n) \).

(ii) If \( p = \infty \), then

\[ \|u(t)\|_\infty \leq \|g\|_\infty. \]  \hspace{1cm} (13.64)

If \( g \) is real-valued then so is \( u \) and

\[ \inf g \leq u(t) \leq \sup g. \]  \hspace{1cm} (13.65)

(iii) (Mass conservation) If \( p = 1 \), then

\[ \int_{\mathbb{R}^n} u(t, x) dx = \int_{\mathbb{R}^n} g(x) dx \]  \hspace{1cm} (13.66)

and

\[ \|u(t)\|_\infty \leq \frac{1}{(4\pi t)^{n/2}} \|g\|_1. \]  \hspace{1cm} (13.67)

**Proof.** That \( u \in C^\infty \) follows since \( \Phi \in C^\infty \) from Problem 9.14. To see the remaining claims we begin by noting (by Problem 9.22)

\[ \int_{\mathbb{R}^n} \Phi_t(x) dx = 1. \]  \hspace{1cm} (13.68)

Now (i) follows from Lemma 10.18, (ii) is immediate, and (iii) follows from Fubini. \( \square \)

Note that using Young’s inequality (14.10) (to be established below) we even have

\[ \|u(t)\|_\infty \leq \|\Phi_t\|_q \|g\|_p = \frac{1}{q^{\frac{p}{n}}} \frac{1}{(4\pi t)^{n/2}} \|g\|_p, \quad \frac{1}{p} + \frac{1}{q} = 1. \]  \hspace{1cm} (13.69)

Another closely related equation is the **Schrödinger equation**

\[-iu_t - \Delta u = 0, \quad u(0) = g. \]  \hspace{1cm} (13.70)

As before we obtain that the solution for \( g \in H^2(\mathbb{R}^n) \) is given by

\[ u(t) = T_S(t)g, \quad T_S(t) = F^{-1}e^{-|p|^2 t}F. \]  \hspace{1cm} (13.71)
Note that $T_S(t) : L^2(\mathbb{R}^n) \rightarrow L^2(\mathbb{R}^n)$ is a unitary operator (since $|e^{-|p|^2t}| = 1$):
\begin{equation}
\|u(t)\|_2 = \|g\|_2. \tag{13.72}
\end{equation}
However, while we have $T_H(t)g \in H^r(\mathbb{R}^n)$ whenever $g \in H^r(\mathbb{R}^n)$, unlike the heat equation, the Schrödinger equation does only preserve but not improve the regularity of the initial condition.

**Theorem 13.27.** The family $T_S(t)$ is a $C_0$-group whose generator is $i\Delta$, $\mathcal{D}(i\Delta) = H^2(\mathbb{R}^n)$.

As in the case of the heat equation, we would like to express our solution as a convolution with the initial condition. However, now we run into the problem that $e^{-|p|^2t}$ is not integrable. To overcome this problem we consider
\begin{equation}
f_\varepsilon(p) = e^{-(it+\varepsilon)p^2}, \quad \varepsilon > 0. \tag{13.73}
\end{equation}
Then, as before we have
\begin{equation}
(f_\varepsilon \hat{g}^\vee)(x) = \frac{1}{(4\pi(it+\varepsilon))^{n/2}} \int_{\mathbb{R}^n} e^{-\frac{|x-y|^2}{4(it+\varepsilon)}} g(y) d^n y \tag{13.74}
\end{equation}
and hence
\begin{equation}
T_S(t)g(x) = \frac{1}{(4\pi it)^{n/2}} \int_{\mathbb{R}^n} e^{\frac{|x-y|^2}{4it}} g(y) d^n y \tag{13.75}
\end{equation}
for $t \neq 0$ and $g \in L^2(\mathbb{R}^n) \cap L^1(\mathbb{R}^n)$. In fact, letting $\varepsilon \downarrow 0$ the left-hand side converges to $T_S(t)g$ in $L^2$ and the limit of the right-hand side exists pointwise by dominated convergence and its pointwise limit must thus be equal to its $L^2$ limit.

Using this explicit form, we can again draw some further consequences. For example, if $g \in L^2(\mathbb{R}^n) \cap L^1(\mathbb{R}^n)$, then $g(t) \in C(\mathbb{R}^n)$ for $t \neq 0$ (use dominated convergence and continuity of the exponential) and satisfies
\begin{equation}
\|u(t)\|_\infty \leq \frac{1}{|4\pi t|^{n/2}} \|g\|_1. \tag{13.76}
\end{equation}
Thus we have again spreading of wave functions in this case.

Finally we turn to the **wave equation**
\begin{equation}
\ddot{u} - \Delta u = 0, \quad u(0) = g, \quad u_t(0) = f. \tag{13.77}
\end{equation}
This equation will fit into our framework once we transform it to a first order system with respect to time:
\begin{equation}
u_t = v, \quad v_t = \Delta u, \quad u(0) = g, \quad v(0) = f. \tag{13.78}
\end{equation}
After applying the Fourier transform this system reads
\begin{equation}
\hat{u}_t = \hat{v}, \quad \hat{v}_t = -|p|^2 \hat{u}, \quad \hat{u}(0) = \hat{g}, \quad \hat{v}(0) = \hat{f}, \tag{13.79}
\end{equation}
and the solution is given by
\[ \hat{u}(t, p) = \cos(t|p|)\hat{g}(p) + \frac{\sin(t|p|)}{|p|} \hat{f}(p), \]
\[ \hat{v}(t, p) = -\sin(t|p|)|p|\hat{g}(p) + \cos(t|p|)\hat{f}(p). \] (13.80)

Hence for \((g, f) \in H^2(\mathbb{R}^n) \oplus H^1(\mathbb{R}^n)\) our solution is given by
\[ \begin{pmatrix} u(t) \\ v(t) \end{pmatrix} = T_W(t) \begin{pmatrix} g \\ f \end{pmatrix}, \quad T_W(t) = \mathcal{F}^{-1} \begin{pmatrix} \cos(t|p|) & \frac{\sin(t|p|)}{|p|} \\ -\sin(t|p|) & \cos(t|p|) \end{pmatrix} \mathcal{F}. \] (13.81)

**Theorem 13.28.** The family \(T_W(t)\) is a \(C_0\)-semigroup whose generator is 
\[ A = \begin{pmatrix} 0 & 1 \\ \Delta & 0 \end{pmatrix}, \mathcal{D}(A) = H^2(\mathbb{R}^n) \oplus H^1(\mathbb{R}^n). \]

Note that if we use \(w\) defined via \(\hat{w}(p) = |p|\hat{v}(p)\) instead of \(v\), then
\[ \begin{pmatrix} u(t) \\ w(t) \end{pmatrix} = \hat{T}_W(t) \begin{pmatrix} g \\ h \end{pmatrix}, \quad \hat{T}_W(t) = \mathcal{F}^{-1} \begin{pmatrix} \cos(t|p|) & \sin(t|p|) \\ -\sin(t|p|) & \cos(t|p|) \end{pmatrix} \mathcal{F}, \] (13.82)
where \(h\) is defined via \(\hat{h} = |p|\hat{f}\). In this case \(\hat{T}_W\) is unitary and thus
\[ \|u(t)\|^2 + \|w(t)\|^2 = \|g\|^2 + \|h\|^2. \] (13.83)

Note that \(\|w\|^2 = \langle w, w \rangle = \langle \hat{w}, \hat{w} \rangle = \langle \hat{v}, |p|^2\hat{v} \rangle = -\langle v, \Delta v \rangle\).

If \(n = 1\) we have \(\frac{\sin(t|p|)}{|p|} \in L^2(\mathbb{R})\) and hence we can get an expression in terms of convolutions. In fact, since the inverse Fourier transform of \(\frac{\sin(t|p|)}{|p|}\) is \(\sqrt{\frac{2}{\pi}}\chi_{[-1,1]}(p/t)\), we obtain
\[ u(t, x) = \int_{\mathbb{R}} \frac{1}{2} \chi_{[-t,t]}(x-y)f(y)dy = \frac{1}{2} \int_{x-t}^{x+t} f(y)dy \]
in the case \(g = 0\). But the corresponding expression for \(f = 0\) is just the time derivative of this expression and thus
\[ u(t, x) = \frac{1}{2} \frac{\partial}{\partial t} \int_{x-t}^{x+t} g(y)dy + \frac{1}{2} \int_{x-t}^{x+t} f(y)dy \]
\[ = \frac{g(x+t) + g(x-t)}{2} + \frac{1}{2} \int_{x-t}^{x+t} f(y)dy, \] (13.84)
which is known as d’Alembert’s formula.

To obtain the corresponding formula in \(n = 3\) dimensions we use the following observation
\[ \frac{\partial}{\partial t} \hat{\varphi}_t(p) = \frac{\sin(t|p|)}{|p|}, \quad \hat{\varphi}_t(p) = \frac{1 - \cos(t|p|)}{|p|^2}. \] (13.85)
where $\hat{\varphi}_t \in L^2(\mathbb{R}^3)$. Hence we can compute its inverse Fourier transform using
\[
\varphi_t(x) = \lim_{R \to \infty} \frac{1}{(2\pi)^{3/2}} \int_{B_R(0)} \hat{\varphi}_t(p) e^{ipx} d^3p \tag{13.86}
\]
using spherical coordinates (without loss of generality we can rotate our coordinate system, such that the third coordinate direction is parallel to $x$)
\[
\varphi_t(x) = \lim_{R \to \infty} \frac{1}{(2\pi)^{3/2}} \int_0^R \int_0^\pi \int_0^{2\pi} \frac{1 - \cos(tr)}{r^2} e^{ir|x|\cos(\theta)} r^2 \sin(\theta) d\varphi d\theta dr.
\tag{13.87}
\]
Evaluating the integrals we obtain
\[
\varphi_t(x) = \lim_{R \to \infty} \frac{1}{\sqrt{2\pi}} \int_0^R (1 - \cos(tr)) \int_0^\pi e^{ir|x|\cos(\theta)} \sin(\theta) d\theta dr
\]
\[
= \lim_{R \to \infty} \frac{\sqrt{2}}{\pi} \int_0^R (1 - \cos(tr)) \frac{\sin(r|x|)}{|x|r} dr
\]
\[
= \lim_{R \to \infty} \frac{1}{\sqrt{2\pi}} \int_0^R \left( \frac{2 \sin(r|x|)}{|x|r} + \frac{\sin(r(t - |x|))}{|x|r} - \frac{\sin(r(t + |x|))}{|x|r} \right) dr,
\]
\[
= \lim_{R \to \infty} \frac{1}{\sqrt{2\pi}|x|} (2 \text{Si}(R|x|) + \text{Si}(R(t - |x|)) - \text{Si}(R(t + |x|)),
\tag{13.88}
\]
where
\[
\text{Si}(z) = \int_0^z \frac{\sin(x)}{x} dx
\tag{13.89}
\]
is the sine integral. Using $\text{Si}(-x) = -\text{Si}(x)$ for $x \in \mathbb{R}$ and (Problem 13.24)
\[
\lim_{x \to \infty} \text{Si}(x) = \frac{\pi}{2}
\tag{13.90}
\]
we finally obtain (since the pointwise limit must equal the $L^2$ limit)
\[
\varphi_t(x) = \sqrt{\frac{\pi}{2}} \chi_{[0,t]}(|x|)
\tag{13.91}
\]
For the wave equation this implies (using Lemma 9.17)
\[
u(t, x) = \frac{1}{4\pi} \frac{\partial}{\partial t} \int_{B_{|y|}(|x|)} \frac{1}{|x - y|} f(y) d^3y
\]
\[
= \frac{1}{4\pi} \frac{\partial}{\partial t} \int_0^{|t|} \int_{S^2} \frac{1}{r} f(x - r\omega) r^2 d\sigma^2(\omega) dr
\]
\[
= \frac{t}{4\pi} \int_{S^2} f(x - t\theta) d\sigma^2(\omega)
\tag{13.92}
\]
and thus finally
\[
u(t, x) = \frac{\partial}{\partial t} \frac{t}{4\pi} \int_{S^2} g(x - t\omega) d\sigma^2(\omega) + \frac{t}{4\pi} \int_{S^2} f(x - t\omega) d\sigma^2(\omega),
\tag{13.93}
which is known as Kirchhoff’s formula.

Finally, to obtain a formula in \( n = 2 \) dimensions we use the method of descent: That is we use the fact, that our solution in two dimensions is also a solution in three dimensions which happens to be independent of the third coordinate direction. Unfortunately this does not fit within our current framework since such functions are not square integrable (unless they vanish identically). However, ignoring this fact and assuming our solution is given by Kirchhoff’s formula we can simplify the integral using the fact that \( f \) does not depend on \( x_3 \). Using spherical coordinates we obtain

\[
\frac{t}{4\pi} \int_{S^2} f(x - t\omega) d\sigma^2(\omega) =
\]

\[
= \frac{t}{2\pi} \int_0^{\pi/2} \int_0^{2\pi} f(x_1 - t\sin(\theta)\cos(\varphi), x_2 - t\sin(\theta)\sin(\varphi)) \sin(\theta) d\theta d\varphi
\]

\[
= \frac{t}{2\pi} \int_0^1 \int_0^{2\pi} \frac{f(x_1 - t\rho\cos(\varphi), x_2 - t\rho\sin(\varphi))}{\sqrt{1 - \rho^2}} \rho d\rho d\varphi
\]

\[
= \frac{t}{2\pi} \int_{B_1(0)} \frac{f(x - ty)}{\sqrt{1 - |y|^2}} \, d^2y
\]

which gives Poisson’s formula

\[
u(t, x) = \frac{\partial}{\partial t} \frac{t}{2\pi} \int_{B_1(0)} \frac{g(x - ty)}{\sqrt{1 - |y|^2}} \, d^2y + \frac{t}{2\pi} \int_{B_1(0)} \frac{f(x - ty)}{\sqrt{1 - |y|^2}} \, d^2y. \quad (13.94)
\]

**Problem 13.20.** Show that \( u(t) \) defined in \((13.59)\) is in \( C^1((0, \infty), L^2(\mathbb{R}^n)) \) and solves \((13.58)\). (Hint: \( |e^{-t|p|^2} - 1| \leq t|p|^2 \) for \( t \geq 0 \).)

**Problem 13.21.** Suppose \( u(t) \in C^1(I, X) \). Show that for \( s, t \in I \)

\[
\|u(t) - u(s)\| \leq M|t - s|, \quad M = \sup_{\tau \in [s, t]} \|\frac{du}{dt}(\tau)\|.
\]

(Hint: Consider \( d(\tau) = \|u(\tau) - u(s)\| - \tilde{M}(\tau - s) \) for \( \tau \in [s, t] \). Suppose \( \tau_0 \) is the largest \( \tau \) for which the claim holds with \( \tilde{M} > M \) and find a contradiction if \( \tau_0 < t \).)

**Problem 13.22.** Solve the transport equation

\[
u_t + a \partial_x u = 0, \quad u(0) = g,
\]

using the Fourier transform.

**Problem 13.23.** Suppose \( A \in \mathcal{L}(X) \). Show that

\[
T(t) = \exp(tA) = \sum_{j=0}^{\infty} \frac{t^j}{j!} A^j
\]

defines a \( C_0 \) (semi)group with generator \( A \). Show that it is fact uniformly continuous: \( T(t) \in C([0, \infty), \mathcal{L}(X)) \).
Problem 13.24. Show the Dirichlet integral
\[
\lim_{R \to \infty} \int_0^R \frac{\sin(x)}{x} dx = \frac{\pi}{2}.
\]
Show also that the sine integral is bounded
\[
|\text{Si}(x)| \leq \min(x, \pi(1 + \frac{1}{2ex})), \quad x > 0.
\]
(Hint: Write \(\text{Si}(R) = \int_0^R \int_0^\infty \sin(x)e^{-xt} dt \, dx\) and use Fubini.)

13.5. Tempered distributions

In many situations, in particular when dealing with partial differential equations, it turns out convenient to look at generalized functions, also known as distributions.

To begin with we take a closer look at the Schwartz space \(S(\mathbb{R}^m)\), defined in (13.10), which already turned out to be a convenient class for the Fourier transform. For our purpose it will be crucial to have a notion of convergence in \(S(\mathbb{R}^m)\) and the natural choice is the topology generated by the seminorms
\[
q_n(f) = \sum_{|\alpha|,|\beta| \leq n} \|x^\alpha(\partial^\beta f)(x)\|_{\infty},
\]
where the sum runs over all multi indices \(\alpha, \beta \in \mathbb{N}_0^m\) of order less than \(n\). Unfortunately these seminorms cannot be replaced by a single norm (and hence we do not have a Banach space) but there is at least a metric
\[
d(f, g) = \sum_{n=1}^\infty \frac{1}{2^n} \frac{q_n(f - g)}{1 + q_n(f - g)}
\]
and \(S(\mathbb{R}^m)\) is complete with respect to this metric and hence a Fréchet space:

Lemma 13.29. The Schwartz space \(S(\mathbb{R}^m)\) together with the family of seminorms \(q_n\) is a Fréchet space.

Proof. It suffices to show completeness. Since a Cauchy sequence \(f_n\) in particular a Cauchy sequence in \(C^\infty(\mathbb{R}^m)\) there is a limit \(f \in C^\infty(\mathbb{R}^m)\) such that all derivatives converge uniformly. Moreover, since Cauchy sequences are bounded \(\|x^\alpha(\partial^\beta f_n)(x)\|_{\infty} \leq C_{\alpha, \beta}\) we obtain \(\|x^\alpha(\partial^\beta f)(x)\|_{\infty} \leq C_{\alpha, \beta}\) and thus \(f \in S(\mathbb{R}^m)\). \(\square\)

We refer to Section 5.4 for further background on Fréchet space. However, for our present purpose it is sufficient to observe that \(f_n \rightarrow f\) if and only if \(q_k(f_n - f) \rightarrow 0\) for every \(k \in \mathbb{N}_0\). Moreover, (cf. Corollary 5.16) a linear map \(A : S(\mathbb{R}^m) \rightarrow S(\mathbb{R}^m)\) is continuous if and only if for every \(j \in \mathbb{N}_0\) there is some \(k \in \mathbb{N}_0\) and a corresponding constant \(C_k\) such that
$q_j(Af) \leq C_k q_k(f)$ and a linear functional $\ell : S(\mathbb{R}^m) \to \mathbb{C}$ is continuous if and only if there is some $k \in \mathbb{N}_0$ and a corresponding constant $C_k$ such that $|\ell(f)| \leq C_k q_k(f)$.

Now the set of all continuous linear functionals, that is the dual space $S^*(\mathbb{R}^m)$, is known as the space of tempered distributions. To understand why this generalizes the concept of a function we begin by observing that any locally integrable function which does not grow too fast gives rise to a distribution.

**Example.** Let $g$ be a locally integrable function of at most polynomial growth, that is, there is some $k \in \mathbb{N}_0$ such that $C_k = \int_{\mathbb{R}^m} |g(x)|(1 + |x|)^{-k}dm \ x < \infty$. Then $T_g(f) = \int_{\mathbb{R}^m} g(x)f(x)dm \ x$ is a distribution. To see that $T_g$ is continuous observe that $|T_g(f)| \leq C_k q_k(f)$. Moreover, note that by Lemma 10.21 the distribution $T_g$ and the function $g$ uniquely determine each other.

The next question is if there are distributions which are not functions.

**Example.** Let $x_0 \in \mathbb{R}^m$ then $\delta_{x_0}(f) = f(x_0)$ is a distribution, the Dirac delta distribution centered at $x_0$. Continuity follows from $|\delta_{x_0}(f)| \leq q_0(f)$ Formally $\delta_{x_0}$ can be written as $T_{\delta_{x_0}}$ as in the previous example where $\delta_{x_0}$ is the Dirac $\delta$-function which satisfies $\delta_{x_0}(x) = 0$ for $x \neq x_0$ and $\delta_{x_0}(x) = \infty$ such that $\int_{\mathbb{R}^m} \delta_{x_0}(x)f(x)dm \ x = f(x_0)$. This is of course nonsense as one can easily see that $\delta_{x_0}$ cannot be expressed as $T_g$ with at locally integrable function of at most polynomial growth (show this). However, giving a precise mathematical meaning to the Dirac $\delta$-function was one of the main motivations to develop distribution theory.

**Example.** This example can be easily generalized: Let $\mu$ be a Borel measure on $\mathbb{R}^m$ such that $C_k = \int_{\mathbb{R}^m} (1 + |x|)^{-k}d\mu(x) < \infty$ for some $k$, then $T_{\mu}(f) = \int_{\mathbb{R}^m} f(x)d\mu(x)$ is a distribution since $|T_{\mu}(f)| \leq C_k q_k(f)$.

**Example.** Another interesting distribution in $S^*(\mathbb{R})$ is given by

$$(p.v. \frac{1}{x})(f) = \lim_{\varepsilon \downarrow 0} \int_{|x| > \varepsilon} \frac{f(x)}{x} dx.$$
To see that this is a distribution note that by the mean value theorem
\[ |(p.v. \frac{1}{x})(f)| = \int_{\varepsilon < |x| < 1} \frac{f(x) - f(0)}{x} \, dx + \int_{1 < |x|} \frac{f(x)}{x} \, dx \leq \int_{\varepsilon < |x| < 1} \left| \frac{f(x) - f(0)}{x} \right| \, dx + \int_{1 < |x|} \frac{|x f(x)|}{x^2} \, dx \leq 2 \sup_{|x| \leq 1} |f'(x)| + 2 \sup_{|x| \geq 1} |x f(x)|. \]

This shows \( |(p.v. \frac{1}{x})(f)| \leq 2 q_1(f) \). \( \diamond \)

Of course, to fill distribution theory with life, we need to extend the classical operations for functions to distributions. First of all, addition and multiplication by scalars comes for free, but we can easily do more. The general principle is always the same: For any continuous linear operator \( A : \mathcal{S}(\mathbb{R}^m) \to \mathcal{S}(\mathbb{R}^m) \) there is a corresponding adjoint operator \( A' : \mathcal{S}^*(\mathbb{R}^m) \to \mathcal{S}^*(\mathbb{R}^m) \) which extends the effect on functions (regarded as distributions of type \( T_g \)) to all distributions. We start with a simple example illustrating this procedure.

Let \( h \in \mathcal{S}(\mathbb{R}^m) \), then the map \( A : \mathcal{S}(\mathbb{R}^m) \to \mathcal{S}(\mathbb{R}^m), f \mapsto h \cdot f \) is continuous. In fact, continuity follows from the Leibniz rule
\[
\partial_{\alpha}(h \cdot f) = \sum_{\beta \leq \alpha} \binom{\alpha}{\beta} (\partial_{\beta} h)(\partial_{\alpha-\beta} f),
\]
where \( \binom{\alpha}{\beta} = \frac{\alpha!}{\beta! (\alpha-\beta)!} \), \( \alpha! = \prod_{j=1}^m (\alpha_j)! \), and \( \beta \leq \alpha \) means \( \beta_j \leq \alpha_j \) for \( 1 \leq j \leq m \). In particular, \( q_j(h \cdot f) \leq C_j q_j(h) q_j(f) \) which shows that \( A \) is continuous and hence the adjoint is well defined via
\[
(A'T)(f) = T(Af). \tag{13.97}
\]

Now what is the effect on functions? For a distribution \( T_g \) given by an integrable function as above we clearly have
\[
(A'T_g)(f) = T_g(h f) = \int_{\mathbb{R}^m} g(x)(h(x)f(x))d^m x = \int_{\mathbb{R}^m} (g(x)h(x))f(x)d^m x = T_{gh}(f). \tag{13.98}
\]

So the effect of \( A' \) on functions is multiplication by \( h \) and hence \( A' \) generalizes this operation to arbitrary distributions. We will write \( A'T = h \cdot T \) for notational simplicity. Note that since \( f \) can even compensate a polynomial growth, \( h \) could even be a smooth functions all whose derivatives grow at most polynomially (e.g. a polynomial):
\[
C^\infty_{pg}(\mathbb{R}^m) = \{ h \in C^\infty(\mathbb{R}^m) \mid \forall \alpha \in \mathbb{N}_0^m \exists C, n : |\partial_{\alpha} h(x)| \leq C(1 + |x|)^n \}. \tag{13.99}
\]
In summary we can define

\[(h \cdot T)(f) = T(h \cdot f), \quad h \in C^\infty_{pq}(\mathbb{R}^m).\]  

(13.100)

**Example.** Let \( h \) be as above and \( \delta_{x_0}(f) = f(x_0) \). Then

\[ h \cdot \delta_{x_0}(f) = \delta_{x_0}(h \cdot f) = h(x_0)f(x_0) \]

and hence \( h \cdot \delta_{x_0} = h(x_0)\delta_{x_0} \). \( \diamond \)

Moreover, since Schwartz functions have derivatives of all orders, the same is true for tempered distributions! To this end let \( \alpha \) be a multi-index and consider \( D_\alpha : S(\mathbb{R}^m) \rightarrow S(\mathbb{R}^m) \), \( f \mapsto (-1)^{|\alpha|}\partial_\alpha f \) (the reason for the extra \((-1)^{|\alpha|}\) will become clear in a moment) which is continuous since \( q_n(D_\alpha f) \leq q_n(|\alpha|)(f) \). Again we let \( D'_\alpha \) be the corresponding adjoint operator and compute its effect on distributions given by functions \( g \):

\[ (D'_\alpha T_g)(f) = T_g((-1)^{|\alpha|}\partial_\alpha f) = (-1)^{|\alpha|}\int_{\mathbb{R}^m} g(x)(\partial_\alpha f(x))d^m x \]

\[ = \int_{\mathbb{R}^m} (\partial_\alpha g(x))f(x)d^m x = T_{\partial_\alpha g}(f), \]  

(13.101)

where we have used integration by parts in the last step which is (e.g.) permissible for \( g \in C^{|\alpha|}_{pq}(\mathbb{R}^m) \) with \( C^k_{pq}(\mathbb{R}^m) = \{ h \in C^k(\mathbb{R}^m)|\forall |\alpha| \leq k \exists C, n : |\partial_\alpha h(x)| \leq C(1 + |x|^n) \} \).

Hence for every multi-index \( \alpha \) we define

\[ (\partial_\alpha T)(f) = (-1)^{|\alpha|}T(\partial_\alpha f). \]  

(13.102)

**Example.** Let \( \alpha \) be a multi-index and \( \delta_{x_0}(f) = f(x_0) \). Then

\[ \partial_\alpha \delta_{x_0}(f) = (-1)^{|\alpha|}\delta_{x_0}(\partial_\alpha f) = (-1)^{|\alpha|}(\partial_\alpha f)(x_0). \]  

(13.103)

Finally we use the same approach for the Fourier transform \( \mathcal{F} : S(\mathbb{R}^m) \rightarrow S(\mathbb{R}^m) \), which is also continuous since \( q_n(\hat{f}) \leq C_nq_n(f) \) by Lemma 13.4. Since Fubini implies

\[ \int_{\mathbb{R}^m} g(x)\hat{f}(x)d^m x = \int_{\mathbb{R}^m} \hat{g}(x)f(x)d^m x \]

(13.104)

for \( g \in L^1(\mathbb{R}^m) \) (or \( g \in L^2(\mathbb{R}^m) \)) and \( f \in S(\mathbb{R}^m) \) we define the Fourier transform of a distribution to be

\[ (\mathcal{FT})(f) \equiv \hat{T}(f) = T(\hat{f}) \]

such that \( \mathcal{F}T_g = T_g \) for \( g \in L^1(\mathbb{R}^m) \) (or \( g \in L^2(\mathbb{R}^m) \)).

**Example.** Let us compute the Fourier transform of \( \delta_{x_0}(f) = f(x_0) \):

\[ \hat{\delta}_{x_0}(f) = \delta_{x_0}(\hat{f}) = \hat{f}(x_0) = \frac{1}{(2\pi)^{m/2}} \int_{\mathbb{R}^m} e^{-ix_0 x} f(x)d^m x = T_g(f), \]
where \( g(x) = (2\pi)^{-m/2}e^{-ix_0x} \).

**Example.** A slightly more involved example is the Fourier transform of \( p.v.\frac{1}{x} \):

\[
\left(\left(p.v.\frac{1}{x}\right)\right)^\wedge(f) = \lim_{\varepsilon \downarrow 0} \int \frac{\hat{f}(x)}{x} \, dx = \lim_{\varepsilon \downarrow 0} \int_{|x| < |1/\varepsilon|} e^{-iyx/x} f(y) \, dy \, dx
\]

where we have used the sine integral (13.89). Moreover, Problem 13.24 shows that we can use dominated convergence to get:

\[
\left(\left(p.v.\frac{1}{x}\right)\right)^\wedge(f) = -i\pi \int_{\mathbb{R}} \text{sign}(y) f(y) \, dy,
\]

that is, \( \left(\left(p.v.\frac{1}{x}\right)\right)^\wedge = -i\pi \text{sign}(y) \).

Note that since \( \mathcal{F} : \mathcal{S}(\mathbb{R}^m) \to \mathcal{S}(\mathbb{R}^m) \) is a homeomorphism, so is its adjoint \( \mathcal{F}^\prime : \mathcal{S}^*(\mathbb{R}^m) \to \mathcal{S}^*(\mathbb{R}^m) \). In particular, its inverse is given by

\[
\hat{T}(f) = T(h).
\]

Moreover, all the operations for \( \mathcal{F} \) carry over to \( \mathcal{F}^\prime \). For example, from Lemma 13.4 we immediately obtain

\[
(\partial^\alpha T)^\wedge = (i\nu)^\alpha \hat{T}, \quad (x^\alpha T)^\wedge = i^{\alpha j} \partial_j \hat{T}.
\]

Similarly one can extend Lemma 13.2 to distributions.

Next we turn to convolutions. Since (Fubini)

\[
\int_{\mathbb{R}^m} (h * g)(x) f(x) \, dx = \int_{\mathbb{R}^m} g(x)(\tilde{h} * f)(x) \, dx,
\]

for integrable functions \( f, g, h \) we define

\[
(h * T)(f) = T(\tilde{h} * f), \quad h \in \mathcal{S}(\mathbb{R}^m),
\]

which is well defined by Corollary 13.13. Moreover, Corollary 13.13 immediately implies

\[
(h * T)^\wedge = (2\pi)^{n/2} \hat{h} \hat{T}, \quad (hT)^\wedge = (2\pi)^{-n/2} \hat{h} \hat{T}, \quad h \in \mathcal{S}(\mathbb{R}^m).
\]
13.5. Tempered distributions

**Example.** Note that the Dirac delta distribution acts like an identity for convolutions since

\[(h * \delta_0)(f) = \delta_0(\tilde{h} * f) = (\tilde{h} * f)(0) = \int_{\mathbb{R}^m} h(y)f(y) = T_h(f)\].

In the last example the convolution is associated with a function. This turns out always to be the case.

**Theorem 13.30.** For every \(T \in \mathcal{S}^*(\mathbb{R}^m)\) and \(h \in \mathcal{S}(\mathbb{R}^m)\) we have that \(h * T\) is associated with the function

\[h * T = T_g, \quad g(x) = T(h(x - .)) \in C^\infty_p(\mathbb{R}^m).\]  (13.110)

**Proof.** By definition \((h * T)(f) = T(\tilde{h} * f)\) and since \((\tilde{h} * f)(x) = \int h(y - x)f(y)d^m y\) the distribution \(T\) acts on \(h(y - .)\) and we should be able to pull out the integral by linearity. To make this idea work let us replace the integral by a Riemann sum

\[(\tilde{h} * f)(x) = \lim_{n \to \infty} \sum_{j=1}^{n^{2m}} h(y^n_j - x)f(y^n_j)|Q^n_j|,\]

where \(Q^n_j\) is a partition of \([-\frac{n}{2}, \frac{n}{2}]^m\) into \(n^{2m}\) cubes of side length \(\frac{1}{n}\) and \(y^n_j\) is the midpoint of \(Q^n_j\). Then, if this Riemann sum converges to \(h * f\) in \(\mathcal{S}(\mathbb{R}^m)\), we have

\[(h * T)(f) = \lim_{n \to \infty} \sum_{j=1}^{n^{2m}} g(y^n_j)f(y^n_j)|Q^n_j|\]

and of course we expect this last limit to converge to the corresponding integral. To be able to see this we need some properties of \(g\). Since

\[|h(z - x) - h(z - y)| \leq q_1(h)|x - y|\]

by the mean value theorem and similarly

\[q_n(h(z - x) - h(z - y)) \leq C_n g_{n+1}(h)|x - y|\]

we see that \(x \mapsto h(. - x)\) is continuous in \(\mathcal{S}(\mathbb{R}^m)\). Consequently \(g\) is continuous. Similarly, if \(x = x_0 + \varepsilon e_j\) with \(e_j\) the unit vector in the \(j\)th coordinate direction,

\[q_n \left( \frac{1}{\varepsilon} (h(. - x) - h(. - x_0)) - \partial_j h(. - x_0) \right) \leq C_n g_{n+2}(h)\varepsilon\]

which shows \(\partial_j g(x) = T((\partial_j h)(x - .))\). Applying this formula iteratively gives

\[\partial_\alpha g(x) = T((\partial_\alpha h)(x - .))\]  (13.111)
and hence \( g \in C^\infty(\mathbb{R}^m) \). Furthermore, \( g \) has at most polynomial growth since \(|T(f)| \leq Cq_n(f)\) implies
\[
|g(x)| = |T(h(-x))| \leq Cq_n(h(-x)) \leq \tilde{C}(1 + |x|^n)q(h).
\]
Combining this estimate with (13.111) even gives \( g \in C^q_{pg}(\mathbb{R}^m) \).

In particular, since \( g \cdot f \in \mathcal{S}(\mathbb{R}^m) \) the corresponding Riemann sum converges and we have \( h * T = T_g \).

It remains to show that our first Riemann sum for the convolution converges in \( \mathcal{S}(\mathbb{R}^m) \). It suffices to show
\[
\sup_x |x|^N \left| \int_{|y| > n/2} h(y - x)f(y)d^m y \right| \leq Cq_N(h) \int_{|y| > n/2} (1 + |y|^N)|f(y)|d^m y \to 0.
\]
The second one is
\[
\sup_x |x|^N \left| \sum_{j=1}^{n^2m} \int_{Q_j^n} (h(y^n_j - x)f(y^n_j) - h(y - x)f(y))d^m y \right|
\]
and the integrand can be estimated by
\[
|y|^N|\hbar(y^n_j - x)f(y^n_j) - h(y - x)f(y)|
\leq |x|^N|\hbar(y^n_j - x) - h(y - x)||f(y^n_j)| + |x|^N|h(y - x)||f(y^n_j) - f(y)|
\leq \left(q_{N+1}(h)(1 + |y^n_j|^N)|f(y^n_j)| + q_N(h)(1 + |y|^N)|\partial f(y^n_j)| \right)|y - y^n_j|
\]
and the claim follows since \(|f(y)| + |\partial f(y)| \leq C(1 + |y|)^{-N-m-1}. \quad \square

**Example.** If we take \( T = \frac{1}{\pi}(p.v.\frac{1}{x}) \), then
\[
(T * h)(x) = \lim_{\varepsilon \to 0} \frac{1}{\pi} \int_{|y| > \varepsilon} \frac{h(x - y)}{y}dy
= \lim_{\varepsilon \to 0} \frac{1}{\pi} \int_{|x-y| > \varepsilon} \frac{h(y)}{x-y}dy
\]
which is known as the **Hilbert transform** of \( h \). Moreover,
\[
(T * h)^\wedge(p) = \sqrt{2\pi}i\text{sign}(p)h(p)
\]
as distributions and hence the Hilbert transform extends to a bounded operator on \( L^2(\mathbb{R}) \).

As a consequence we get that distributions can be approximated by functions.
13.5. Tempered distributions

**Theorem 13.31.** Let $\phi_\varepsilon$ be the standard mollifier. Then $\phi_\varepsilon * T \to T$ in $S^*(\mathbb{R}^m)$.

**Proof.** We need to show $\phi_\varepsilon * T(f) = T(\phi_\varepsilon * f)$ for any $f \in S(\mathbb{R}^m)$. This follows from continuity since $\phi_\varepsilon * f \to f$ in $S(\mathbb{R}^m)$ as can be easily seen (the derivatives follow from Lemma 10.17 (ii)). □

Note that Lemma 10.17 (i) and (ii) implies

$$\partial_\alpha (h * T) = (\partial_\alpha h) * T = h * (\partial_\alpha T).$$

(13.112)

When working with distributions it is also important to observe that, in contradistinction to smooth functions, they can be supported at a single point. Here the **support** $\text{supp}(T)$ of a distribution is the smallest closed set for which

$$\text{supp}(f) \subseteq \mathbb{R}^m \setminus V \implies T(f) = 0.$$

An example of a distribution supported at 0 is the Dirac delta distribution $\delta_0$ as well as all of its derivatives. It turns out that these are in fact the only examples.

**Lemma 13.32.** Suppose $T$ is a distribution supported at $x_0$. Then

$$T = \sum_{|\alpha| \leq n} c_\alpha \partial_\alpha \delta_{x_0}.$$  

(13.113)

**Proof.** For simplicity of notation we suppose $x_0 = 0$. First of all there is some $n$ such that $|T(f)| \leq C q_n(f)$. Write

$$T = \sum_{|\alpha| \leq n} c_\alpha \partial_\alpha \delta_0 + \tilde{T},$$

where $c_\alpha = \frac{T(x^\alpha)}{\alpha!}$. Then $\tilde{T}$ vanishes on every polynomial of degree at most $n$, has support at 0, and still satisfies $|\tilde{T}(f)| \leq \tilde{C} q_n(f)$. Now let $\phi_\varepsilon(x) = \phi(\frac{x}{\varepsilon})$, where $\phi$ has support in $B_1(0)$ and equals 1 in a neighborhood of 0. Then $\tilde{T}(f) = \tilde{T}(g) = \tilde{T}(\phi_\varepsilon g)$, where $g(x) = f(x) - \sum_{|\alpha| \leq n} \frac{f^{(\alpha)(0)}}{\alpha!} x^\alpha$. Since $|\partial_\beta g(x)| \leq C \varepsilon^{n+1-|\beta|}$ for $x \in B_\varepsilon(0)$ Leibniz’ rule implies $q_n(\phi_\varepsilon g) \leq C \varepsilon$. Hence $|\tilde{T}(f)| = |\tilde{T}(\phi_\varepsilon g)| \leq \tilde{C} q_n(\phi_\varepsilon g) \leq \tilde{C} \varepsilon$ and since $\varepsilon > 0$ is arbitrary we have $|\tilde{T}(f)| = 0$, that is, $\tilde{T} = 0$. □

**Example.** Let us try to solve the Poisson equation in the sense of distributions. We begin with solving

$$-\Delta T = \delta_0.$$

Taking the Fourier transform we obtain

$$|p|^2 \tilde{T} = (2\pi)^{-m/2}.$$
and since $|p|^{-1}$ is a bounded locally integrable function in $\mathbb{R}^m$ for $m \geq 2$ the above equation will be solved by

$$
\Phi = (2\pi)^{-m/2} \left( \frac{1}{|p|^2} \right)^\vee.
$$

Explicitly, $\Phi$ must be determined from

$$
\Phi(f) = (2\pi)^{-m/2} \int_{\mathbb{R}^m} \hat{f}(p) \frac{1}{|p|^2} d^m p = (2\pi)^{-m/2} \int_{\mathbb{R}^m} \hat{f}(p) |p|^2 d^m p
$$

and evaluating Lemma 13.17 at $x = 0$ we obtain that $\Phi = (2\pi)^{-m/2} T_{I_2}$ where $I_2$ is the Riesz potential. Note, that $\Phi$ is not unique since we could add a polynomial of degree two corresponding to a solution of the homogeneous equation $(|p|^2 T = 0$ implies that supp$(T) = 0$ and comparing with Lemma 13.32 shows $T = \sum_{|\alpha| \leq 2} c_\alpha \partial_\alpha \delta_0$ and hence $T = \sum_{|\alpha| \leq 2} \tilde{c}_\alpha x^\alpha$).

Given $h \in S(\mathbb{R}^m)$ we can now consider $h * \Phi$ which solves

$$
-\Delta (h * \Phi) = h * (-\Delta \Phi) = h * \delta_0 = h,
$$

where in the last equality we have identified $h$ with $T_h$. Note that since $h * \Phi$ is associated with a function in $C^\infty(\mathbb{R}^m)$ our distributional solution is also a classical solution. This gives the formal calculations with the Dirac delta function found in many physics textbooks a solid mathematical meaning. 

Note that while we have been quite successful in generalizing many basic operations to distributions, our approach is limited to linear operations! In particular, it is not possible to define nonlinear operations, for example the product of two distributions within this framework. In fact, there is no associative product of two distributions extending the product of a distribution by a function from above.

**Example.** Consider the distributions $\delta_0$, $x$, and $p.v. \frac{1}{x}$ in $S^*(\mathbb{R})$. Then

$$
x \cdot \delta_0 = 0, \quad x \cdot p.v. \frac{1}{x} = 1.
$$

Hence if there would be an associative product of distributions we would get

$$
0 = (x \cdot \delta_0) \cdot p.v. \frac{1}{x} = \delta_0 \cdot (x \cdot p.v. \frac{1}{x}) = \delta_0.
$$

This is known as Schwartz’ impossibility result. However, if one is content with preserving the product of functions, Colombeau algebras will do the trick.

**Problem 13.25.** Compute the derivative of $g(x) = \text{sign}(x)$ in $S^*(\mathbb{R})$.

**Problem 13.26.** Let $h \in C^\infty_{pg}(\mathbb{R}^m)$ and $T \in S^*(\mathbb{R}^m)$. Show

$$
\partial_\alpha (h \cdot T) = \sum_{\beta \leq \alpha} \binom{\alpha}{\beta} \partial_\beta h (\partial_{\alpha-\beta} T).
$$
Problem 13.27. Show that $\text{supp}(T_g) = \text{supp}(g)$ for locally integrable functions.
Chapter 14

Interpolation

14.1. Interpolation and the Fourier transform on $L^p$

We will fix some measure space $(X, \mu)$ and abbreviate $L^p = L^p(X, d\mu)$ for notational simplicity. If $f \in L^{p_0} \cap L^{p_1}$ for some $p_0 < p_1$ then it is not hard to see that $f \in L^p$ for every $p \in [p_0, p_1]$ and we have (Problem 14.1) the Lyapunov inequality

$$\|f\|_p \leq \|f\|_{p_0}^{1-\theta} \|f\|_{p_1}^\theta,$$

where $\frac{1}{p} = \frac{1-\theta}{p_0} + \frac{\theta}{p_1}$, $\theta \in (0, 1)$. Note that $L^{p_0} \cap L^{p_1}$ contains all integrable simple functions. Moreover, the latter functions are dense in $L^p$ for $1 \leq p < \infty$ (for $p = \infty$ this is only true if the measure is finite — cf. Problem 10.15).

This is a first occurrence of an interpolation technique. Next we want to turn to operators. For example, we have defined the Fourier transform as an operator from $L^1 \to L^\infty$ as well as from $L^2 \to L^2$ and the question is if this can be used to extend the Fourier transform to the spaces in between.

Denote by $L^{p_0} + L^{p_1}$ the space of (equivalence classes) of measurable functions $f$ which can be written as a sum $f = f_0 + f_1$ with $f_0 \in L^{p_0}$ and $f_1 \in L^{p_1}$ (clearly such a decomposition is not unique and different decompositions will differ by elements from $L^{p_0} \cap L^{p_1}$). Then we have

$$L^p \subseteq L^{p_0} + L^{p_1}, \quad p_0 < p < p_1,$$

since we can always decompose a function $f \in L^p$, $1 \leq p < \infty$, as $f = f \chi_{\{|f(x)| \leq 1\}} + f \chi_{\{|f(x)| > 1\}}$ with $f \chi_{\{|f(x)| \leq 1\}} \in L^p \cap L^\infty$ and $f \chi_{\{|f(x)| > 1\}} \in L^1 \cap L^p$. Hence, if we have two operators $A_0 : L^{p_0} \to L^{q_0}$ and $A_1 : L^{p_1} \to L^{q_1}$ which coincide on the intersection, $A_0|_{L^{p_0} \cap L^{p_1}} = A_1|_{L^{p_0} \cap L^{p_1}}$, we can extend
them by virtue of

\[ A : L^{p_0} + L^{p_1} \to L^{q_0} + L^{q_1}, \quad f_0 + f_1 \mapsto A_0 f_0 + A_1 f_1 \quad (14.3) \]

(check that \( A \) is indeed well-defined, i.e., independent of the decomposition of \( f \) into \( f_0 + f_1 \)). In particular, this defines \( A \) on \( L^p \) for every \( p \in (p_0, p_1) \) and the question is if \( A \) restricted to \( L^p \) will be a bounded operator into some \( L^q \) provided \( A_0 \) and \( A_1 \) are bounded.

To answer this question we begin with a result from complex analysis.

**Theorem 14.1** (Hadamard three-lines theorem). Let \( S \) be the open strip \( \{ z \in \mathbb{C} | 0 < \text{Re}(z) < 1 \} \) and let \( F : S \to \mathbb{C} \) be continuous and bounded on \( S \) and holomorphic in \( S \). If

\[
|F(z)| \leq \begin{cases} 
M_0, & \text{Re}(z) = 0, \\
M_1, & \text{Re}(z) = 1,
\end{cases} \quad (14.4)
\]

then

\[
|F(z)| \leq M_0^{1-\text{Re}(z)} M_1^{\text{Re}(z)} \quad (14.5)
\]

for every \( z \in S \).

**Proof.** Without loss of generality we can assume \( M_0, M_1 > 0 \) and after the transformation \( F(z) \to \frac{M_0 - z}{M_1 - z} F(z) \) even \( M_0 = M_1 = 1 \). Now we consider the auxiliary function

\[ F_n(z) = e^{(z^2 - 1)/n} F(z) \]

which still satisfies \( |F_n(z)| \leq 1 \) for \( \text{Re}(z) = 0 \) and \( \text{Re}(z) = 1 \) since \( \text{Re}(z^2 - 1) \leq -\text{Im}(z)^2 \leq 0 \) for \( z \in S \). Moreover, by assumption \( |F(z)| \leq M \) implying \( |F_n(z)| \leq 1 \) for \( |\text{Im}(z)| \geq \sqrt{\log(M)n} \). Since we also have \( |F_n(z)| \leq 1 \) for \( |\text{Im}(z)| \leq \sqrt{\log(M)n} \) by the maximum modulus principle we see \( |F_n(z)| \leq 1 \) for all \( z \in S \). Finally, letting \( n \to \infty \) the claim follows. \( \square \)

Now we are able to show the **Riesz–Thorin interpolation theorem**

**Theorem 14.2** (Riesz–Thorin). Let \( (X, d\mu) \) and \( (Y, d\nu) \) be \( \sigma \)-finite measure spaces and \( 1 \leq p_0, p_1, q_0, q_1 \leq \infty \). If \( A \) is a linear operator on

\[ A : L^{p_0}(X, d\mu) + L^{p_1}(X, d\mu) \to L^{q_0}(Y, d\nu) + L^{q_1}(Y, d\nu) \quad (14.6) \]

satisfying

\[ \|Af\|_{q_0} \leq M_0 \|f\|_{p_0}, \quad \|Af\|_{q_1} \leq M_1 \|f\|_{p_1}, \quad (14.7) \]

then \( A \) has continuous restrictions

\[ A_\theta : L^{p_0}(X, d\mu) \to L^{q_0}(Y, d\nu), \quad \frac{1}{p_0} = \frac{1 - \theta}{p_0} + \frac{\theta}{p_1}, \quad \frac{1}{q_0} = \frac{1 - \theta}{q_0} + \frac{\theta}{q_1} \quad (14.8) \]

satisfying \( \|A_\theta\| \leq M_0^{1-\theta} M_1^{\theta} \) for every \( \theta \in (0, 1) \).
14.1. Interpolation and the Fourier transform on $L^p$

**Proof.** In the case $p_0 = p_1 = \infty$ the claim is immediate from (14.1) and hence we can assume $p_0 < \infty$ in which case the space of integrable simple functions is dense in $L^{p_0}$. We will also temporarily assume $q_0 < \infty$. Then, by Lemma 10.6 it suffices to show

$$\left| \int (Af)(y)g(y)\,dv(y) \right| \leq M_0^{1-\theta} M_1^\theta,$$

where $f, g$ are simple functions with $\|f\|_{p_0} = \|g\|_{q_0'} = 1$ and $\frac{1}{q_0} + \frac{1}{q_0'} = 1$.

Now choose simple functions $f(x) = \sum_j \alpha_j \chi_{A_j}(x)$, $g(x) = \sum_k \beta_k \chi_{B_k}(x)$ with $\|f\|_1 = \|g\|_1 = 1$ and set $f_z(x) = \sum_j |\alpha_j|^{1/p^z} \text{sign}(\alpha_j) \chi_{A_j}(x)$, $g_z(y) = \sum_k |\beta_k|^{1-1/q^z} \text{sign}(\beta_k) \chi_{B_k}(y)$ such that $\|f_z\|_{p_0} = \|g_z\|_{q_0'} = 1$ for $\theta = \text{Re}(z) \in [0, 1]$. Moreover, note that both functions are entire and thus the function

$$F(z) = \int (Af_z)(y)g_z\,dv(y)$$

satisfies the assumptions of the three-lines theorem. Hence we have the required estimate for integrable simple functions. Now let $f \in L^{p_0}$ and split it according to $f = f_0 + f_1$ with $f_0 \in L^{p_0} \cap L^{p_1}$ and $f_1 \in L^{p_1} \cap L^{p_0}$ and approximate both by integrable simple functions (cf. Problem 10.15).

It remains to consider the case $p_0 < p_1$ and $q_0 = q_1 = \infty$. In this case we can proceed as before using again Lemma 10.6 and a simple function for $g = g_z$.

Note that the proof shows even a bit more

**Corollary 14.3.** Let $A$ be an operator defined on the space of integrable simple functions satisfying (14.7). Then $A$ has continuous extensions $A_{p_0}$ as in the Riesz–Thorin theorem which will agree on $L^{p_0}(X,d\mu) \cap L^{p_1}(X,d\mu)$.

As a consequence we get two important inequalities:

**Corollary 14.4** (Hausdorff–Young inequality). The Fourier transform extends to a continuous map $F : L^p(\mathbb{R}^n) \to L^q(\mathbb{R}^n)$, for $1 \leq p \leq 2$, $\frac{1}{p} + \frac{1}{q} = 1$, satisfying

$$2\pi^{-n/(2q)}\|\hat{f}\|_q \leq (2\pi)^{-n/(2p)}\|f\|_p. \quad (14.9)$$

We remark that the Fourier transform does not extend to a continuous map $F : L^p(\mathbb{R}^n) \to L^q(\mathbb{R}^n)$, for $p > 2$ (Problem 14.2). Moreover, its range is dense for $1 < p \leq 2$ but not all of $L^q(\mathbb{R}^n)$ unless $p = q = 2$.

**Corollary 14.5** (Young inequality). Let $f \in L^p(\mathbb{R}^n)$ and $g \in L^q(\mathbb{R}^n)$ with $\frac{1}{p} + \frac{1}{q} \geq 1$. Then $f(y)g(x-y)$ is integrable with respect to $y$ for a.e. $x$ and the convolution satisfies $f \ast g \in L^r(\mathbb{R}^n)$ with

$$\|f \ast g\|_r \leq \|f\|_p \|g\|_q, \quad (14.10)$$
where $\frac{1}{r} = \frac{1}{p} + \frac{1}{q} - 1$.

**Proof.** We consider the operator $A_g f = f \ast g$ which satisfies $\|A_g f\|_q \leq \|g\|_q \|f\|_p$ for every $f \in L^1$ by Lemma 10.17. Similarly, Hölder’s inequality implies $\|A_g f\|_q \leq \|g\|_q \|f\|_q'$ for every $f \in L^{q'}$, where $\frac{1}{q} + \frac{1}{q'} = 1$. Hence the Riesz–Thorin theorem implies that $A_g$ extends to an operator $A_g : L^p \to L^{q'}$, where $\frac{1}{p} = \frac{1}{q} - \frac{1}{q'} = 1 - \frac{1}{q}$ and $\frac{1}{q'} = \frac{1}{q} + \frac{1}{\infty} = \frac{1}{p} + \frac{1}{q} - 1$. To see that $f(y)g(x - y)$ is integrable a.e. consider $f_n(x) = \chi_{|x| \leq n}(x) \max(n, |f(x)|)$. Then the convolution $(f_n \ast |g|)(x)$ is finite and converges for every $x$ by monotone convergence. Moreover, since $f_n \to |f|$ in $L^p$ we have $f_n \ast |g| \to A_g f$ in $L^{q'}$, which finishes the proof.

Combining the last two corollaries we obtain:

**Corollary 14.6.** Let $f \in L^p(\mathbb{R}^n)$ and $g \in L^q(\mathbb{R}^n)$ with $\frac{1}{r} = \frac{1}{p} + \frac{1}{q} - 1 \geq 0$ and $1 \leq r, p, q \leq 2$. Then

$$
(f \ast g)^\wedge = (2\pi)^{n/2} \hat{f} \hat{g}.
$$

**Proof.** By Corollary 13.13 the claim holds for $f, g \in \mathcal{S}(\mathbb{R}^n)$. Now take a sequence of Schwartz functions $f_m \to f$ in $L^p$ and a sequence of Schwartz functions $g_m \to g$ in $L^q$. Then the left-hand side converges in $L^{q'}$, where $\frac{1}{q'} = 2 - \frac{1}{r} - \frac{1}{q}$, by the Young and Hausdorff-Young inequalities. Similarly, the right-hand side converges in $L^{q'}$ by the generalized Hölder (Problem 10.8) and Hausdorff-Young inequalities.

**Problem 14.1.** Show (14.1). (Hint: Generalized Hölder inequality from Problem 10.8.)

**Problem 14.2.** Show that the Fourier transform does not extend to a continuous map $\mathcal{F} : L^p(\mathbb{R}^n) \to L^q(\mathbb{R}^n)$, for $p > 2$. Use the closed graph theorem to conclude that $\mathcal{F}$ is not onto for $1 \leq p < 2$. (Hint for the case $n = 1$: Consider $\phi_z(x) = \exp(-xz^2/2)$ for $z = \lambda + i\omega$ with $\lambda > 0$.)

**Problem 14.3** (Young inequality). Let $K(x, y)$ be measurable and suppose

$$
\sup_x \|K(x, \cdot)\|_{L^r(Y, dv)} \leq C, \quad \sup_y \|K(\cdot, y)\|_{L^r(X, d\mu)} \leq C.
$$

where $\frac{1}{r} = \frac{1}{q} - \frac{1}{p} \geq 1$ for some $1 \leq p \leq q \leq \infty$. Then the operator $K : L^p(Y, dv) \to L^q(X, d\mu)$, defined by

$$(Kf)(x) = \int_Y K(x, y)f(y)dv(y),$$

for $\mu$-almost every $x$, is bounded with $\|K\| \leq C$. (Hint: Show $\|Kf\|_\infty \leq C\|f\|_{r'}$, $\|Kf\|_r \leq C\|f\|_1$ and use interpolation.)
14.2. The Marcinkiewicz interpolation theorem

In this section we are going to look at another interpolation theorem which might be helpful in situations where the Riesz–Thorin interpolation theorem does not apply. In this respect recall, that \( f(x) = \frac{1}{x} \) just fails to be integrable over \( \mathbb{R} \). To include such functions we begin by slightly weakening the \( L^p \) norms. To this end we consider the distribution function

\[
E_f(r) = \mu(\{ x \in X \mid |f(x)| > r \})
\]

of a measurable function \( f : X \to \mathbb{C} \) with respect to \( \mu \). Note that \( E_f \) is decreasing and right continuous. Given, the distribution function we can compute the \( L^p \) norm via (Problem 9.19)

\[
\|f\|_p^p = \int_0^\infty r^{p-1} E_f(r) dr, \quad 1 \leq p < \infty.
\]

In the case \( p = \infty \) we have

\[
\|f\|_\infty = \inf \{ r \geq 0 \mid E_f(r) = 0 \}.
\]

Another relationship follows from the observation

\[
\|f\|_p^p = \int_X |f|^p d\mu \geq \int_{|f| > r} r^p d\mu = r^p E_f(r)
\]

which yields Markov’s inequality

\[
E_f(r) \leq r^{-p} \|f\|_p^p.
\]

Motivated by this we define the **weak \( L^p \) norm**

\[
\|f\|_{p,w} = \sup_{r > 0} r E_f(r)^{1/p}, \quad 1 \leq p < \infty,
\]

and the corresponding spaces \( L^{p,w}(X, d\mu) \) consist of all equivalence classes of functions which are equal a.e. for which the above norm is finite. Clearly the distribution function and hence the weak \( L^p \) norm depend only on the equivalence class. Despite its name the weak \( L^p \) norm turns out to be only a quasinorm (Problem 14.4). By construction we have

\[
\|f\|_{p,w} \leq \|f\|_p
\]

and thus \( L^p(X, d\mu) \subseteq L^{p,w}(X, d\mu) \). In the case \( p = \infty \) we set \( \|\cdot\|_{\infty,w} = \|\cdot\|_{\infty} \).

**Example.** Consider \( f(x) = \frac{1}{x} \) in \( \mathbb{R} \). Then clearly \( f \notin L^1(\mathbb{R}) \) but

\[
E_f(r) = \{|x| \frac{1}{x} > r\} = \{|x| < r^{-1}\} = \frac{2}{r}
\]

shows that \( f \in L^{1,w}(\mathbb{R}) \) with \( \|f\|_{1,w} = 2 \). Slightly more general the function \( f(x) = |x|^{-n/p} \notin L^p(\mathbb{R}^n) \) but \( f \in L^{p,w}(\mathbb{R}^n) \). Hence \( L^{p,w}(\mathbb{R}^n) \) is strictly larger than \( L^p(\mathbb{R}^n) \).
Now we are ready for our interpolation result. We call an operator $T : L^p(X, d\mu) \to L^q(X, d\nu)$ subadditive if it satisfies

$$\|T(f + g)\|_q \leq \|T(f)\|_q + \|T(g)\|_q.$$  \hfill (14.18)

It is said to be of strong type $(p, q)$ if

$$\|T(f)\|_q \leq C_{p,q}\|f\|_p$$  \hfill (14.19)

and of weak type $(p, q)$ if

$$\|T(f)\|_{q,w} \leq C_{p,q,w}\|f\|_p.$$  \hfill (14.20)

By (14.17) strong type $(p, q)$ is indeed stronger than weak type $(p, q)$ and we have $C_{p,q,w} \leq C_{p,q}$.

**Theorem 14.7 (Marcinkiewicz).** Let $(X, d\mu)$ and $(Y, d\nu)$ be measure spaces and $1 \leq p_0 < p_1 \leq \infty$. Let $T$ be a subadditive operator defined for all $f \in L^p(X, d\mu)$, $p \in [p_0, p_1]$. If $T$ is of weak type $(p_0, p_0)$ and $(p_1, p_1)$ then it is also of strong type $(p, p)$ for every $p_0 < p < p_1$.

**Proof.** We begin by assuming $p_1 < \infty$. Fix $f \in L^p$ as well as some number $s > 0$ and decompose $f = f_0 + f_1$ according to

$$f_0 = f \chi_{\{|f| > s\}} \in L^{p_0} \cap L^p, \quad f_1 = f \chi_{\{|f| \leq s\}} \in L^p \cap L^{p_1}.$$  

Next we use (14.12),

$$\|T(f)\|_p = p \int_0^\infty r^{p-1} E_{T(f)}(r) dr = p2^p \int_0^\infty r^{p-1} E_{T(f)}(2r) dr$$

and observe

$$E_{T(f)}(2r) \leq E_{T(f_0)}(r) + E_{T(f_1)}(r)$$

since $|T(f)| \leq |T(f_0)| + |T(f_1)|$ implies $|T(f)| > 2r$ only if $|T(f_0)| > r$ or $|T(f_1)| > r$. Now using (14.15) our assumption implies

$$E_{T(f_0)}(r) \leq \left( \frac{C_0 \|f_0\|_{p_0}}{r} \right)^{p_0}, \quad E_{T(f_1)}(r) \leq \left( \frac{C_1 \|f_1\|_{p_1}}{r} \right)^{p_1}$$

and choosing $s = r$ we obtain

$$E_{T(f)}(2r) \leq \frac{C_0^{p_0}}{r^{p_0}} \int_{\{|f| > r\}} |f|^{p_0} d\mu + \frac{C_1^{p_1}}{r^{p_1}} \int_{\{|f| \leq r\}} |f|^{p_1} d\mu.$$  

In summary we have $\|T(f)\|_p \leq p2^p(C_0^{p_0} I_1 + C_1^{p_1} I_2)$ with

$$I_0 = \int_0^\infty \int_X r^{p-p_0-1} \chi_{\{|f| > r\}} |f(x)|^{p_0} d\mu(x) dr$$

$$= \int_X |f(x)|^{p_0} \int_0^\infty r^{p-p_0-1} dr d\mu(x) = \frac{1}{p - p_0} \|f\|_p^p$$

and $I_1$ similarly.
14.2. The Marcinkiewicz interpolation theorem

and

\[
I_1 = \int_0^\infty \int_X r^{p-p_1-1} \chi\{r(x) \leq r \} |f(x)|^{p_1} d\mu(x) \, dr
= \int_X |f(x)|^{p_1} \int_{|f(x)|}^\infty r^{p-p_1-1} \, dr \, d\mu(x) = \frac{1}{p_1 - p} \|f\|_p^{p_1}.
\]

This is the desired estimate

\[
\|T(f)\|_p \leq 2 \left( \frac{p}{p - p_0} \right) C_0^{p_0} + \frac{p}{p_1 - p} C_1^{p_1} \|f\|_p.
\]

The case \( p_1 = \infty \) is similar: Split \( f \in L^{p_0} \) according to

\( f_0 = f \chi\{|f| > s/C_1\} \in L^{p_0} \cap L^p \), \( f_1 = f \chi\{|f| \leq s/C_1\} \in L^p \cap L^\infty \)

(if \( C_1 = 0 \) there is nothing to prove). Then \( \|T(f_1)\|_\infty \leq s/C_1 \) and hence \( E_{T(f_1)}(s) = 0 \). Thus

\[
E_{T(f)}(2r) \leq \frac{C_0^{p_0}}{r^{p_0}} \int_{\{x: |f| > r/C_1\}} |f|^{p_0} \, d\mu
\]

and we can proceed as before to obtain

\[
\|T(f)\|_p \leq 2 \left( \frac{p}{p - p_0} \right)^{1/p} C_0^{p_0/p} C_1^{1-p_0/p} \|f\|_p,
\]

which is again the desired estimate. \( \square \)

As with the Riesz–Thorin theorem there is also a version for operators which are of weak type \((p_0, q_0)\) and \((p_1, q_1)\) but the proof is slightly more involved and the above diagonal version will be sufficient for our purpose.

As a first application we will use it to investigate the **Hardy–Littlewood maximal function** defined for any locally integrable function in \( \mathbb{R}^n \) via

\[
\mathcal{M}(f)(x) = \sup_{r>0} \frac{1}{|B_r(x)|} \int_{B_r(x)} |f(y)| \, dy.
\]  

(14.21)

By the dominated convergence theorem, the integral is continuous with respect to \( x \) and consequently (Problem 8.18) \( \mathcal{M}(f) \) is lower semicontinuous (and hence measurable). Moreover, its value is unchanged if we change \( f \) on sets of measure zero, so \( \mathcal{M} \) is well defined for functions in \( L^p(\mathbb{R}^n) \). However, it is unclear if \( \mathcal{M}(f)(x) \) is finite a.e. at this point. If \( f \) is bounded we of course have the trivial estimate

\[
\|\mathcal{M}(f)\|_\infty \leq \|f\|_\infty.
\]  

(14.22)

**Theorem 14.8** (Hardy–Littlewood maximal inequality). The maximal function is of weak type \((1,1)\),

\[
E_{\mathcal{M}(f)}(r) \leq \frac{3^n}{r} \|f\|_1,
\]  

(14.23)
and of strong type \((p, p)\),
\[
\|\mathcal{M}(f)\|_p \leq 2 \left( \frac{3^np}{p-1} \right)^{1/p} \|f\|_p,
\]
(14.24)
for every \(1 < p \leq \infty\).

**Proof.** The first estimate follows literally as in the proof of Lemma 11.5 and combining this estimate with the trivial one (14.22) the Marcinkiewicz interpolation theorem yields the second. \(\square\)

Using this fact, our next aim is to prove the Hardy–Littlewood–Sobolev inequality. As a preparation we show

**Lemma 14.9.** Let \(\phi \in L^1(\mathbb{R}^n)\) be a radial, \(\phi(x) = \phi_0(|x|)\) with \(\phi_0\) positive and nonincreasing. Then we have the following estimate for convolutions with integrable functions:
\[
|\phi \ast f(x)| \leq \|\phi\|_1 \mathcal{M}(f)(x).
\]
(14.25)

**Proof.** By approximating \(\phi_0\) with simple functions of the same type, it suffices to prove that case where \(\phi_0 = \sum_{j=1}^p \alpha_j \chi_{[0, r_j]}\) with \(\alpha_j > 0\). Then
\[
(\phi \ast f)(x) = \sum_j \alpha_j |B_{r_j}(0)| \int_{B_{r_j}(x)} \frac{f(y)}{|B_{r_j}(x)|} d^n y
\]
and the estimate follows upon taking absolute values and observing \(\|\phi\|_1 = \sum_j \alpha_j |B_{r_j}(0)|\). \(\square\)

Now we will apply this to the Riesz potential (13.29) of order \(\alpha\):
\[
\mathcal{I}_\alpha f = I_\alpha \ast f.
\]
(14.26)

**Theorem 14.10** (Hardy–Littlewood–Sobolev inequality). Let \(0 < \alpha < n\), \(p \in (1, \frac{n}{\alpha})\), and \(q = \frac{pn}{n-\alpha} \in (\frac{n}{n-\alpha}, \infty)\) (i.e., \(\frac{\alpha}{n} = \frac{1}{p} - \frac{1}{q}\)). Then \(\mathcal{I}_\alpha\) is of strong type \((p, q)\),
\[
\|\mathcal{I}_\alpha f\|_q \leq C_{p, \alpha, n}\|f\|_p.
\]
(14.27)

**Proof.** We split the Riesz potential into two parts
\[
I_\alpha = I^0_\alpha + I^\infty_\alpha, \quad I^0_\alpha = I_\alpha \chi(0, \varepsilon), \quad I^\infty_\alpha = I_\alpha \chi(\varepsilon, \infty),
\]
where \(\varepsilon > 0\) will be determined later. Note that \(I^0_\alpha(\cdot) \in L^1(\mathbb{R}^n)\) and \(I^\infty_\alpha(\cdot) \in L^r(\mathbb{R}^n)\) for every \(r \in (\frac{n}{n-\alpha}, \infty)\). In particular, since \(p' = \frac{p}{p-1} \in (\frac{n}{n-\alpha}, \infty)\), both integrals converge absolutely by the Young inequality (14.10). Next we will estimate both parts individually. Using Lemma 14.9 we obtain
\[
|I^0_\alpha f(x)| \leq \int_{|y| < \varepsilon} \frac{d^n y}{|y|^{n-\alpha}} \mathcal{M}(f)(x) = \frac{(n-1)V_n}{\alpha - 1} \varepsilon^n \mathcal{M}(f)(x).
\]
On the other hand, using Hölder’s inequality we infer

\[ |\mathcal{I}_\alpha f(x)| \leq \left( \int_{|y| \geq \varepsilon} \frac{d^n y}{|y|^{(n-\alpha)p'}} \right)^{1/p'} \|f\|_p = \left( \frac{(n-1)V_n}{p'(n-\alpha) - n} \right)^{1/p'} \varepsilon^{\alpha-n/p} \|f\|_p. \]

Now we choose \( \varepsilon = \left( \frac{\|f\|_p}{\mathcal{M}_\alpha(x)} \right)^{p/n} \) such that

\[ |\mathcal{I}_\alpha f(x)| \leq \tilde{C} \|f\|^\theta \mathcal{M}(f)(x)^{1-\theta}, \quad \theta = \frac{\alpha p}{n} \in \left( \frac{\alpha}{n}, 1 \right), \]

where \( \tilde{C}/2 \) is the larger of the two constants in the estimates for \( \mathcal{I}_\alpha^0 f \) and \( \mathcal{I}_\alpha^\infty f \). Taking the \( L^q \) norm in the above expression gives

\[ \|\mathcal{I}_\alpha f\|_q \leq \tilde{C} \|f\|_p \mathcal{M}(f)^{1-\theta} \|f\|_p = \tilde{C} \|f\|_p \mathcal{M}(f)^{1-\theta} \|f\|_p = \tilde{C} \|f\|_p \mathcal{M}(f)^{1-\theta} \]

and the claim follows from the Hardy–Littlewood maximal inequality. \( \square \)

**Problem 14.4.** Show that \( E_f = 0 \) if and only if \( f = 0 \). Moreover, show \( E_{f+g}(r + s) \leq E_f(r) + E_g(s) \) and \( E_{\alpha f}(r) = E_f(r/|\alpha|) \) for \( \alpha \neq 0 \). Conclude that \( L^{p,w}(X, d\mu) \) is a quasinormed space with

\[ \|f + g\|_{p,w} \leq 2(\|f\|_{p,w} + \|g\|_{p,w}), \quad \|\alpha f\|_{p,w} = |\alpha|\|f\|_{p,w}. \]

**Problem 14.5.** Show \( f(x) = |x|^{-n/p} \in L^{p,w}(\mathbb{R}^n) \). Compute \( \|f\|_{p,w} \).

**Problem 14.6.** Show that if \( f_n \to f \) in weak-\( L^p \), then \( f_n \to f \) in measure. (Hint: Problem 8.22.)

**Problem 14.7.** The right continuous generalized inverse of the distribution function is known as the **decreasing rearrangement** of \( f \):

\[ f^*(t) = \inf\{r \geq 0 | E_f(r) \leq t\} \]

(see Section 9.4 for basic properties of the generalized inverse). Note that \( f^* \) is decreasing with \( f^*(0) = \|f\|_\infty \). Show that

\[ E_{f^*} = E_f \]

and in particular \( \|f^*\|_p = \|f\|_p \).

**Problem 14.8.** Show that the maximal function of an integrable function is finite at every Lebesgue point.

**Problem 14.9.** Let \( \phi \) be a nonnegative nonincreasing radial function with \( \|\phi\|_1 = 1 \). Set \( \phi_\varepsilon(x) = \varepsilon^{-n} \phi(x/\varepsilon) \). Show that for integrable \( f \) we have \( (\phi_\varepsilon * f)(x) \to f(x) \) at every Lebesgue point. (Hint: Split \( \phi = \phi^\delta + \phi^\delta \) into a part with compact support \( \phi^\delta \) and a rest by setting \( \phi^\delta(x) = \min(\delta, \phi(x)) \). To handle the compact part use Problem 10.23. To control the contribution of the rest use Lemma 14.9.)
Problem 14.10. For $f \in L^1(0, 1)$ define

$$T(f)(x) = e^{i\arg(\int_0^1 f(y)dy)} f(x).$$

Show that $T$ is subadditive and norm preserving. Show that $T$ is not continuous.
Part 3

Nonlinear Functional Analysis
Chapter 15

Analysis in Banach spaces

15.1. Differentiation and integration in Banach spaces

We first review some basic facts from calculus in Banach spaces. Most facts will be similar to the situation of multivariable calculus from functions from $\mathbb{R}^n$ to $\mathbb{R}^m$. To emphasize this we will use $|.|$ for the norm in this section.

Let $X$ and $Y$ be two Banach spaces and let $U$ be an open subset of $X$. Denote by $C(U,Y)$ the set of continuous functions from $U \subseteq X$ to $Y$ and by $\mathcal{L}(X,Y) \subset C(X,Y)$ the Banach space of (bounded) linear functions equipped with the operator norm

$$
\|L\| := \sup_{|u|=1} |Lu|.
$$

(15.1)

Then a function $F : U \to Y$ is called differentiable at $x \in U$ if there exists a linear function $dF(x) \in \mathcal{L}(X,Y)$ such that

$$
F(x + u) = F(x) + dF(x) u + o(u),
$$

(15.2)

where $o, O$ are the Landau symbols. Explicitly

$$
\lim_{u \to 0} \frac{|F(x + u) - F(x) - dF(x) u|}{|u|} = 0.
$$

(15.3)

The linear map $dF(x)$ is called the Fréchet derivative of $F$ at $x$. It is uniquely defined since if $dG(x)$ were another derivative we had $(dF(x) - dG(x))u = o(u)$ implying that for every $\varepsilon > 0$ we can find a $\delta > 0$ such that $|(dF(x) - dG(x))u| \leq \varepsilon |u|$ whenever $|u| \leq \delta$. By homogeneity of the norm we conclude $\|dF(x) - dG(x)\| \leq \varepsilon$ and since $\varepsilon > 0$ is arbitrary $dF(x) = dG(x)$. 
Note that for this argument to work it is crucial that we can approach \(x\) from arbitrary directions \(u\) which explains our requirement that \(U\) should be open.

**Example.** Let \(X\) be a Hilbert space and consider \(F : X \to \mathbb{R}\) given by \(F(x) := |x|^2\). Then
\[
F(x + u) = \langle x + u, x + u \rangle = |x|^2 + 2\Re \langle x, u \rangle + |u|^2 = F(x) + 2\Re \langle x, u \rangle + o(u).
\]
Hence if \(X\) is a real Hilbert space, then \(F\) is differentiable with \(dF(x)u = 2\Re \langle x, u \rangle\). However, if \(X\) is a complex Hilbert space, then \(F\) is not differentiable. In fact, in case of a complex Hilbert (or Banach) space, we obtain a version of complex differentiability which of course is much stronger than real differentiability.

**Example.** Suppose \(f \in C^1(\mathbb{R})\) with \(f(0) = 0\). Let \(X := \ell^p(\mathbb{N})\), then
\[
F : X \to X, \quad (x_n)_{n \in \mathbb{N}} \mapsto (f(x_n))_{n \in \mathbb{N}}
\]
is differentiable for every \(x \in X\) with derivative given by the multiplication operator
\[
(dF(x))_n = f'(x_n)u_n.
\]
First of all note that the mean value theorem implies \(|f(t)| \leq M_R|t|\) for \(|t| \leq R\) with \(M_R := \sup_{|t| \leq R}|f'(t)|\). Hence, since \(||x||_\infty \leq ||x||_p\), we have \(||F(x)||_p \leq M_{||x||_\infty}||x||_p\) and \(F\) is well defined. This also shows that multiplication by \(f'(x_n)\) is a bounded linear map. To establish differentiability we use
\[
f(t + s) - f(t) - f'(t)s = s \int_0^1 (f'(t + s\tau) - f'(t))d\tau
\]
and since \(f'\) is uniformly continuous on every compact interval, we can find a \(\delta > 0\) for every given \(R > 0\) and \(\varepsilon > 0\) such that
\[
|f'(t + s) - f'(t)| < \varepsilon \quad \text{if} \quad |s| < \delta, \ |t| < R.
\]
Now for \(x, u \in X\) with \(||x||_\infty < R\) and \(||u||_\infty < \delta\) we have \(|f(x_n + u_n) - f(x_n) - f'(x_n)u_n| < \varepsilon|u_n|\) and hence
\[
||F(x + u) - F(x) - dF(x)u||_p < \varepsilon||u||_p
\]
which establishes differentiability. Moreover, using uniform continuity of \(f\) on compact sets a similar argument shows that \(dF\) is continuous (observe that the operator norm of a multiplication operator by a sequence is the sup norm of the sequence) and hence we even have \(F \in C^1(X, X)\).

Differentiability implies existence of directional derivatives
\[
\delta F(x, u) := \lim_{\varepsilon \to 0} \frac{F(x + \varepsilon u) - F(x)}{\varepsilon}, \quad \varepsilon \in \mathbb{R} \setminus \{0\}, \quad (15.4)
\]
which are also known as Gâteaux derivative or variational derivative. Indeed, if $F$ is differentiable at $x$, then (15.2) implies
\[
\delta F(x, u) = dF(x)u. \tag{15.5}
\]
However, note that existence of the Gâteaux derivative (i.e. the limit on the right-hand side in (15.4)) for all $u \in X$ does not imply differentiability. In fact, the Gâteaux derivative might be unbounded or it might even fail to be linear in $u$. Some authors require the Gâteaux derivative to be a bounded linear operator and in this case we will write $\delta F(x, u) = \delta F(x)u$ but even this additional requirement does not imply differentiability in general. Note that in any case the Gâteaux derivative is homogenous, that is, if $\delta F(x, u)$ exists, then $\delta F(x, \lambda u)$ exists for every $\lambda \in \mathbb{R}$ and
\[
\delta F(x, \lambda u) = \lambda \delta F(x, u), \quad \lambda \in \mathbb{R}. \tag{15.6}
\]

**Example.** The function $F : \mathbb{R}^2 \to \mathbb{R}$ given by $F(x, y) = \frac{x^3}{x^2+y^2}$ for $(x, y) \neq 0$ and $F(0, 0) = 0$ is Gâteaux differentiable at 0 with Gâteaux derivative
\[
\delta F(0, (u, v)) = \lim_{\varepsilon \to 0} \frac{F(\varepsilon u, \varepsilon v)}{\varepsilon} = F(u, v),
\]
which is clearly nonlinear.

The function $F : \mathbb{R}^2 \to \mathbb{R}$ given by $F(x, y) = x$ for $y = x^2$ and $F(x, 0) = 0$ else is Gâteaux differentiable at 0 with Gâteaux derivative $\delta F(0) = 0$, which is clearly linear. However, $F$ is not differentiable.

If you take a linear function $L : X \to Y$ which is unbounded, then $L$ is everywhere Gâteaux differentiable with derivative equal to $Lu$, which is linear but, by construction, not bounded.

**Example.** Let $X := L^2(0, 1)$ and consider
\[
F : X \to X, \quad x \mapsto \sin(x).
\]
First of all note that by $|\sin(t)| \leq |t|$ our map is indeed from $X$ to $X$ and since sine is Lipschitz continuous we get the same for $F : \|F(x) - F(y)\|_2 \leq \|x - y\|_2$. Moreover, $F$ is Gâteaux differentiable at $x = 0$ with derivative given by
\[
\delta F(0) = I,
\]
but it is not differentiable at $x = 0$.

To see that the Gâteaux derivative is the identity note that
\[
\lim_{\varepsilon \to 0} \frac{\sin(\varepsilon u(t))}{\varepsilon} = u(t)
\]
pointwise and hence
\[
\lim_{\varepsilon \to 0} \left\| \frac{\sin(\varepsilon u(\cdot))}{\varepsilon} - u(\cdot) \right\|_2 = 0
\]
by dominated convergence since \( \left| \frac{\sin(\varepsilon u(t))}{\varepsilon} \right| \leq |u(t)| \).

To see that \( F \) is not differentiable let
\[
u_n = \pi \chi_{[0,1/n]}, \quad \|\nu_n\|_2 = \frac{\pi}{\sqrt{n}}
\]
and observe that \( F(\nu_n) = 0 \), implying that
\[
\frac{\|F(\nu_n) - \nu_n\|_2}{\|\nu_n\|_2} = 1
\]
does not converge to 0. Note that this problem does not occur in \( X := C[0,1] \) (Problem 15.2).

We will mainly consider Fréchet derivatives in the remainder of this chapter as it will allow a theory quite close to the usual one for multivariable functions.

**Lemma 15.1.** Suppose \( F : U \to Y \) is differentiable at \( x \in U \). Then \( F \) is continuous at \( x \). Moreover, we can find constants \( M, \delta > 0 \) such that
\[
|F(x + u) - F(x)| \leq M|u|, \quad |u| \leq \delta.
\]

**Proof.** For every \( \varepsilon > 0 \) we can find a \( \delta > 0 \) such that \( |F(x + u) - F(x) - dF(x)u| \leq \varepsilon|u| \) for \( |u| \leq \delta \). Now chose \( M = \|dF(x)\| + \varepsilon \). \( \square \)

**Example.** Note that this lemma fails for the Gâteaux derivative as the example of an unbounded linear function shows. In fact, it already fails in \( \mathbb{R}^2 \) as the function \( F : \mathbb{R}^2 \to \mathbb{R} \) given by \( F(x,y) = 1 \) for \( y = x^2 \neq 0 \) and \( F(x,0) = 0 \) else shows: It is Gâteaux differentiable at 0 with \( \delta F(0) = 0 \) but it is not continuous since \( \lim_{\varepsilon \to 0} F(\varepsilon^2, \varepsilon) = 1 \neq 0 = F(0,0) \).

Note that this in particular implies that at every point of differentiability there is a neighborhood where \( F \) is Lipschitz continuous. However, of course we have linearity (which is easy to check):

**Lemma 15.2.** Suppose \( F, G : U \to Y \) are differentiable at \( x \in U \) and \( \alpha, \beta \in \mathbb{C} \). Then \( \alpha F + \beta G \) is differentiable at \( x \) with \( d(\alpha F + \beta G)(x) = \alpha dF(x) + \beta dG(x) \). Similarly, if the Gâteaux derivatives \( \delta F(x,u) \) and \( \delta G(x,u) \) exist, then so does \( \delta(F + G)(x,u) = \delta F(x,u) + \delta G(x,u) \).

If \( F \) is differentiable for all \( x \in U \) we call \( F \) differentiable. In this case we get a map
\[
dF : \quad U \to \mathcal{L}(X,Y) \quad x \mapsto dF(x). \tag{15.8}
\]
If \( dF : U \to \mathcal{L}(X,Y) \) is continuous, we call \( F \) continuously differentiable and write \( F \in C^1(U,Y) \).
If $X$ or $Y$ has a (finite) product structure, then the computation of the partial derivatives can be reduced as usual. The following facts are simple and can be shown as in the case of $X = \mathbb{R}^n$ and $Y = \mathbb{R}^m$.

Let $Y := \bigtimes_{j=1}^{m} Y_j$ and let $F : X \to Y$ be given by $F = (F_1, \ldots, F_m)$ with $F_j : X \to Y_j$. Then $F \in C^1(X,Y)$ if and only if $F_j \in C^1(X,Y_j)$, $1 \leq j \leq m$, and in this case $dF = (dF_1, \ldots, dF_m)$. Similarly, if $X = \bigtimes_{i=1}^{n} X_i$, then one can define the partial derivative $\partial_i F \in \mathcal{L}(X_i, Y)$, which is the derivative of $F$ considered as a function of the $i$-th variable alone (the other variables being fixed). We have $dF u = \sum_{i=1}^{n} \partial_i F u_i$, $u = (u_1, \ldots, u_n) \in X$, and $F \in C^1(X,Y)$ if and only if all partial derivatives exist and are continuous.

**Example.** In the case of $X = \mathbb{R}^n$ and $Y = \mathbb{R}^m$, the matrix representation of $dF$ with respect to the canonical basis in $\mathbb{R}^n$ and $\mathbb{R}^m$ is given by the partial derivatives $\partial_i F_j(x)$ and is called Jacobi matrix of $F$ at $x$.

Given $F \in C^1(U, Y)$ we have $dF \in C(U, \mathcal{L}(X,Y))$ and we can define the second derivative (provided it exists) via

$$dF(x + v) = dF(x) + d^2 F(x)v + o(v). \quad (15.9)$$

In this case $d^2 F : U \to \mathcal{L}(X, \mathcal{L}(X,Y))$ which maps $x$ to the linear map $v \mapsto d^2 F(x)v$ which for fixed $v$ is a linear map $u \mapsto (d^2 F(x)v)u$. Equivalently, we could regard $d^2 F(x)$ as a map $d^2 F(x) : X^2 \to Y$; $(u, v) \mapsto (d^2 F(x)v)u$ which is linear in both arguments. That is, $d^2 F(x)$ is a bilinear map $X^2 \to Y$.

The corresponding norm on $\mathcal{L}(X, \mathcal{L}(X,Y))$ explicitly spelled out reads

$$\|d^2 F(x)\| = \sup_{|v|=1} \|d^2 F(x)v\| = \sup_{|u|=|v|=1} \|(d^2 F(x)v)u\|. \quad (15.10)$$

**Example.** Note that if $F \in \mathcal{L}(X,Y)$, then $dF(x) = F$ (independent of $x$) and $d^2 F(x) = 0$.

**Example.** Let $X$ be a real Hilbert space and $F(x) = |x|^2$. Then we have already seen $dF(x)u = \langle x, u \rangle$ and hence

$$dF(x + v)u = \langle x + v, u \rangle = \langle x, u \rangle + \langle v, u \rangle = dF(x)u + \langle v, u \rangle$$

which shows $(d^2 F(x)v)u = \langle v, u \rangle$.

**Example.** Suppose $f \in C^2(\mathbb{R})$ with $f(0) = 0$ and continue the example from page 436. Then we have $F \in C^2(X,Y)$ with $d^2 F(x)v$ the multiplication operator by the sequence $f''(x_n) v_n$, that is,

$$((d^2 F(x)v)u)_n = f''(x_n) v_n u_n.$$ 

Indeed, arguing in a similar fashion we can find a $\delta_1$ such that $|f'(x_n + v_n) - \cdot f'(x_n) - f''(x_n) v_n| \leq \varepsilon |v_n|$ whenever $|x| \to R$ and $\|v\|_\infty < \delta_1$. Hence

$$\|dF(x + v) - dF(x) - d^2 F(x)v\| < \varepsilon \|v\|_p.$$
which shows differentiability. Moreover, since \( \|d^2 F(x)\| = \|f''(x)\|_\infty \) one also easily verifies that \( F \in C^2(X,X) \) using uniform continuity of \( f'' \) on compact sets.

We can iterate the procedure of differentiation and write \( F \in C^r(U,Y) \), \( r \geq 1 \), if the \( r \)-th derivative of \( F \), \( d^r F \) (i.e., the derivative of the \((r-1)\)-th derivative of \( F \)), exists and is continuous. Note that \( d^r F(x) \) will be a multilinear map in \( r \) arguments as we will show below. Finally, we set \( C^\infty(U,Y) = \bigcap_{r \in \mathbb{N}} C^r(U,Y) \) and, for notational convenience, \( C^0(U,Y) = C(U,Y) \) and \( d^0 F = F \).

**Example.** Let \( X \) be a Banach algebra. Consider the multiplication \( M : X \times X \to X \). Then

\[
\partial_1 M(x,y)u = uy, \quad \partial_2 M(x,y)u = xu
\]

and hence

\[
dM(x,y)(u_1, u_2) = u_1 y + xu_2.
\]

Consequently \( dM \) is linear in \((x,y)\) and hence

\[
(d^2 M(x,y)(v_1, v_2))(u_1, u_2) = u_1 v_2 + v_1 u_2
\]

Consequently all differentials of order higher than two will vanish and in particular \( M \in C^\infty(X \times X, X) \).

If \( F \) is bijective and \( F, F^{-1} \) are both of class \( C^r \), \( r \geq 1 \), then \( F \) is called a **diffeomorphism** of class \( C^r \).

For the composition of mappings we have the usual **chain rule**.

**Lemma 15.3** (Chain rule). Let \( U \subseteq X, V \subseteq Y \) and \( F \in C^r(U,V) \) and \( G \in C^r(V,Z) \), \( r \geq 1 \). Then \( G \circ F \in C^r(U,Z) \) and

\[
d(G \circ F)(x) = dG(F(x)) \circ dF(x), \quad x \in X. \quad (15.11)
\]

**Proof.** Fix \( x \in U, y = F(x) \in V \) and let \( u \in X \) such that \( v = dF(x)u \) with \( x+u \in U \) and \( y+v \in V \) for \( |u| \) sufficiently small. Then \( F(x+u) = y+v+o(u) \) and, with \( \tilde{v} = v+o(u) \),

\[
G(F(x+u)) = G(y + \tilde{v}) = G(y) + dG(y)\tilde{v} + o(\tilde{v}).
\]

Using \( |\tilde{v}| \leq \|dF(x)||u| + |o(u)| \) we see that \( o(\tilde{v}) = o(u) \) and hence

\[
G(F(x+u)) = G(y) + dG(y)v + o(u) = G(F(x)) + dG(F(x)) \circ dF(x)u + o(u)
\]

as required. This establishes the case \( r = 1 \). The general case follows from induction.

In particular, if \( \lambda \in Y^* \) is a bounded linear functional, then \( d(\lambda \circ F) = d\lambda \circ dF = \lambda \circ dF \). As an application of this result we obtain
15.1. Differentiation and integration in Banach spaces

**Theorem 15.4** (Schwarz). Suppose \( F \in C^2(\mathbb{R}^n, Y) \). Then

\[
\partial_i \partial_j F = \partial_j \partial_i F
\]

for any \( 1 \leq i, j \leq n \).

**Proof.** First of all note that \( \partial_j F(x) \in \mathcal{L}(\mathbb{R}, Y) \) and thus it can be regarded as an element of \( Y \). Clearly the same applies to \( \partial_i \partial_j F(x) \). Let \( \lambda \in Y^* \) be a bounded linear functional, then \( \lambda \circ F \in C^2(\mathbb{R}^2, \mathbb{R}) \) and hence \( \partial_i \partial_j (\lambda \circ F) = \partial_j \partial_i (\lambda \circ F) \) by the classical theorem of Schwarz. Moreover, by our remark preceding this lemma \( \partial_i \partial_j (\lambda \circ F) = \partial_i \lambda (\partial_j F) = \lambda (\partial_i \partial_j F) \) and hence \( \lambda (\partial_i \partial_j F) = \lambda (\partial_j \partial_i F) \) for every \( \lambda \in Y^* \) implying the claim. □

Now we let \( F \in C^2(X, Y) \) and look at the function \( G : \mathbb{R}^2 \to Y \), \((t, s) \mapsto G(t, s) = F(x + tu + sv)\). Then one computes

\[
\partial_t G(t, s) \bigg|_{t=0} = dF(x + sv)u
\]

and hence

\[
\partial_s \partial_t G(t, s) \bigg|_{(s, t)=0} = \partial_s dF(x + sv)u \bigg|_{s=0} = (d^2 F(x)u)v.
\]

Since by the previous lemma the order of the derivatives is irrelevant, we obtain

\[
d^2 F(u, v) = d^2 F(v, u),
\]

that is, \( d^2 F \) is a symmetric bilinear form. This result easily generalizes to higher derivatives. To this end we introduce some notation first.

A function \( L : \times_{j=1}^n X_j \to Y \) is called **multilinear** if it is linear with respect to each argument. It is not hard to see that \( L \) is continuous if and only if

\[
\|L\| = \sup_{x:|x_1| = \ldots = |x_n| = 1} |L(x_1, \ldots, x_n)| < \infty. \tag{15.13}
\]

If we take \( n \) copies of the same space, the set of multilinear functions \( L : X^n \to Y \) will be denoted by \( \mathcal{L}^n(X, Y) \). A multilinear function is called **symmetric** provided its value remains unchanged if any two arguments are switched. With the norm from above it is a Banach space and in fact there is a canonical isometric isomorphism between \( \mathcal{L}^n(X, Y) \) and \( \mathcal{L}(X, \mathcal{L}^{n-1}(X, Y)) \) given by \( L : (x_1, \ldots, x_n) \mapsto L(x_1, \ldots, x_n) \) maps to \( x_1 \mapsto L(x_1, \cdot) \).

**Lemma 15.5.** Suppose \( F \in C^r(X, Y) \). Then for every \( x \in X \) we have that

\[
d^r F(x)(u_1, \ldots, u_r) = \partial_{t_1} \cdots \partial_{t_r} F(x + \sum_{i=1}^r t_i u_i) \bigg|_{t_1=\ldots=t_r=0}. \tag{15.14}
\]

Moreover, \( d^r F(x) \in \mathcal{L}^r(X, Y) \) is a bounded symmetric multilinear form.
Proof. The representation (15.14) follows using induction as before. Symmetry follows since the order of the partial derivatives can be interchanged by Lemma 15.4.

Finally, note that to each \( L \in \mathcal{L}^n(X,Y) \) we can assign its polar form \( L \in C(X,Y) \) using \( L(x) = L(x, \ldots, x), x \in X \). If \( L \) is symmetric it can be reconstructed using polarization (Problem 15.4):

\[
L(u_1, \ldots, u_n) = \frac{1}{n!} \partial_{t_1} \cdots \partial_{t_n} L(\sum_{i=1}^{n} t_i u_i). \tag{15.15}
\]

We also have the following version of the product rule: Suppose \( L \in \mathcal{L}^2(X,Y) \), then \( L \in C^1(X^2,Y) \) with

\[
dL(x)u = L(u_1, x_2) + L(x_1, u_2) \tag{15.16}
\]

since

\[
L(x_1 + u_1, x_2 + u_2) - L(x_1, x_2) = L(u_1, x_2) + L(x_1, u_2) + L(u_1, u_2)
= L(u_1, x_2) + L(x_1, u_2) + O(|u|^2) \tag{15.17}
\]

as \( |L(u_1, u_2)| \leq \|L\| |u_1||u_2| = O(|u|^2) \). If \( X \) is a Banach algebra and \( L(x_1, x_2) = x_1 x_2 \) we obtain the usual form of the product rule.

Next we have the following mean value theorem.

**Theorem 15.6 (Mean value).** Suppose \( U \subseteq X \) and \( F: U \to Y \) is Gâteaux differentiable at every \( x \in U \). If \( U \) is convex, then

\[
|F(x) - F(y)| \leq M|x - y|, \quad M := \sup_{0 \leq t \leq 1} |\delta F((1 - t)x + ty, x - y)|. \tag{15.18}
\]

Conversely, (for any open \( U \)) if

\[
|F(x) - F(y)| \leq M|x - y|, \quad x, y \in U, \tag{15.19}
\]

then

\[
\sup_{x \in U, |e| = 1} |\delta F(x, e)| \leq M. \tag{15.20}
\]

**Proof.** Abbreviate \( f(t) = F((1 - t)x + ty), 0 \leq t \leq 1, \) and hence \( df(t) = \delta F((1 - t)x + ty, y - x) \) implying \( |df(t)| \leq \tilde{M} := M|x - y| \) by (15.6). For the first part it suffices to show

\[
\phi(t) = |f(t) - f(0)| - (\tilde{M} + \delta)t \leq 0
\]
for any $\delta > 0$. Let $t_0 = \max\{t \in [0, 1] | \phi(t) \leq 0\}$. If $t_0 < 1$ then
\[
\phi(t_0 + \varepsilon) = |f(t_0 + \varepsilon) - f(t_0) + f(t_0) - f(0)| - (\tilde{M} + \delta)(t_0 + \varepsilon)
\leq |f(t_0 + \varepsilon) - f(t_0)| - (\tilde{M} + \delta)\varepsilon + \phi(t_0)
\leq |df(t_0)\varepsilon + o(\varepsilon)| - (\tilde{M} + \delta)\varepsilon
\leq (\tilde{M} + o(1) - \tilde{M} - \delta)\varepsilon = (-\delta + o(1))\varepsilon \leq 0,
\]
for $\varepsilon \geq 0$, small enough. Thus $t_0 = 1$.

To prove the second claim suppose we can find an $e \in X$, $|e| = 1$ such that $|\delta F(x_0, e)| = M + \delta$ for some $\delta > 0$ and hence
\[
M\varepsilon \geq |F(x_0 + \varepsilon e) - F(x_0)| = |\delta F(x_0, e)\varepsilon + o(\varepsilon)|
\geq (M + \delta)\varepsilon - |o(\varepsilon)| > M\varepsilon
\]
since we can assume $|o(\varepsilon)| < \varepsilon\delta$ for $\varepsilon > 0$ small enough, a contradiction. $\square$

Note that in the infinite dimensional case continuity of $dF$ does not suffice to conclude boundedness on bounded closed sets.

**Example.** Let $X$ be an infinite Hilbert space and $\{u_n\}_{n \in \mathbb{N}}$ some orthonormal set. Then the family of functions $F_n(x) = \max(0, 1 - 2||x - u_n||)$ is continuous with disjoint supports. Hence $F(x) = \sum_{n \in \mathbb{N}} nF_n(x)$ is also continuous (show this). But $F$ is not bounded on the unit ball since $F(u_n) = n$. $\diamond$

As an immediate consequence we obtain

**Corollary 15.7.** Suppose $U$ is a connected subset of a Banach space $X$. A Gâteaux differentiable mapping $F : U \to Y$ is constant if and only if $\delta F = 0$. In addition, if $F_{1,2} : U \to Y$ and $\delta F_1 = \delta F_2$, then $F_1$ and $F_2$ differ only by a constant.

Now we turn to integration. We will only consider the case of mappings $f : I \to X$ where $I = [a, b] \subset \mathbb{R}$ is a compact interval and $X$ is a Banach space. A function $f : I \to X$ is called **simple** if the image of $f$ is finite, $f(I) = \{x_i\}_{i=1}^n$, and if each inverse image $f^{-1}(x_i)$, $1 \leq i \leq n$ is a Borel set. The set of simple functions $S(I, X)$ forms a linear space and can be equipped with the sup norm. The corresponding Banach space obtained after completion is called the set of **regulated functions** $R(I, X)$.

Observe that $C(I, X) \subset R(I, X)$. In fact, consider the functions $f_n = \sum_{i=0}^{n-1} f(t_i)\chi_{[t_i, t_{i+1})} \in S(I, X)$, where $t_i = a + i\frac{b-a}{n}$ and $\chi$ is the characteristic function. Since $f \in C(I, X)$ is uniformly continuous, we infer that $f_n$ converges uniformly to $f$. 
For \( f \in S(I, X) \) we can define a linear map \( \int : S(I, X) \to X \) by
\[
\int_a^b f(t)dt = \sum_{i=1}^n x_i \mu(f^{-1}(x_i)),
\]
where \( \mu \) denotes the Lebesgue measure on \( I \). This map satisfies
\[
\left| \int_a^b f(t)dt \right| \leq |f|(b-a).
\]
and hence it can be extended uniquely to a linear map \( \int : R(I, X) \to X \) with the same norm \((b-a)\). We even have
\[
\left| \int_a^b f(t)dt \right| \leq \int_a^b |f(t)|dt
\]
since this holds for simple functions by the triangle inequality and hence for all functions by approximation.

In addition, if \( \lambda \in X^* \) is a continuous linear functional, then
\[
\lambda(\int_a^b f(t)dt) = \int_a^b \lambda(f(t))dt, \quad f \in R(I, X).
\]

We will use the usual conventions \( \int_{t_1}^{t_2} f(s)ds = \int_a^b \chi_{(t_1,t_2)}(s)f(s)ds \) and \( \int_{t_1}^{t_2} f(s)ds = -\int_{t_2}^{t_1} f(s)ds \).

If \( I \subseteq \mathbb{R} \), we have an isomorphism \( \mathcal{L}(I, X) \equiv X \) and if \( F : I \to X \) we will write \( \dot{F}(t) \) instead of \( dF(t) \) if we regard \( dF(t) \) as an element of \( X \). Note that in this case the Gâteaux and Fréchet derivatives coincide as we have
\[
\dot{F}(t) = \lim_{\varepsilon \to 0} \frac{F(t+\varepsilon) - F(t)}{\varepsilon}.
\]

By \( C^1(I, X) \) we will denote the set of functions from \( C(I, X) \) which are differentiable in the interior of \( I \) and for which the derivative has a continuous extension to \( C(I, X) \).

**Theorem 15.8** (Fundamental theorem of calculus). *Suppose \( F \in C^1(I, X) \), then*
\[
F(t) = F(a) + \int_a^t \dot{F}(s)ds.
\]

*Conversely, if \( f \in C(I, X) \), then \( F(t) = \int_a^t f(s)ds \in C^1(I, X) \) and \( \dot{F}(t) = f(t) \).*

**Proof.** Let \( f \in C(I, X) \) and set \( G(t) = \int_a^t f(s)ds \in C^1(I, X) \). Then \( \dot{F}(t) = f(t) \) as can be seen from
\[
|\int_a^{t+\varepsilon} f(s)ds - \int_a^t f(s)ds - f(t)\varepsilon| = |\int_t^{t+\varepsilon} (f(s) - f(t))ds| \leq |\varepsilon| \sup_{s\in[t,t+\varepsilon]} |f(s) - f(t)|.
\]
Hence if \( F \in C^1(I, X) \) then \( G(t) = \int_0^t (\dot{F}(s))ds \) satisfied \( \dot{G} = \dot{F} \) and hence \( F(t) = C + G(t) \) by Corollary 15.7. Choosing \( t = a \) finally shows \( F(a) = C \). \( \square \)

As a simple application we obtain a generalization of the well-known fact that continuity of the directional derivatives implies continuous differentiability.

**Lemma 15.9.** Suppose \( F : U \subseteq X \rightarrow Y \) is Gâteaux differentiable such that the Gâteaux derivative is linear and continuous, \( \delta F \in C(U, \mathcal{L}(X, Y)) \). Then \( F \in C^1(U, Y) \) and \( dF = \delta F \).

**Proof.** By assumption \( f(t) = F(x + tu) \) is in \( C^1([0, 1], Y) \) for \( u \) with sufficiently small norm. Moreover, by definition we have \( \dot{f} = \delta F(x + tu)u \) and using the fundamental theorem of calculus we obtain

\[
F(x + u) - F(x) = f(1) - f(0) = \int_0^1 \dot{f}(t)dt = \int_0^1 \delta F(x + tu)u dt
\]

where the last equality follows from continuity of the integral since it clearly holds for simple functions. Consequently

\[
|F(x + u) - F(x) - \delta F(x)u| = \left| \int_0^1 ((\delta F(x + tu) - \delta F(x))dt)u \right|
\]

\[
\leq \int_0^1 (\|\delta F(x + tu) - \delta F(x)\| dt) |u|
\]

\[
\leq \max_{t \in [0,1]} \|\delta F(x + tu) - \delta F(x)\| |u|.
\]

By the continuity assumption on \( \delta F \), the right-hand side is \( o(u) \) as required. \( \square \)

As another consequence we obtain **Taylor’s theorem**.

**Theorem 15.10 (Taylor).** Suppose \( U \subseteq X \) and \( F \in C^{r+1}(U, Y) \). Then

\[
F(x + u) = F(x) + dF(x)u + \frac{1}{2} d^2 F(x)u^2 + \cdots + \frac{1}{r!} d^r F(x)u^r
\]

\[
+ \left( \frac{1}{r!} \int_0^1 \left( 1 - t \right)^r d^{r+1} F(x + tu)dt \right) u^{r+1},
\]

(15.27)

where \( u^k = (u, \ldots, u) \in X^k \).

**Proof.** As in the proof of the previous lemma, the case \( r = 0 \) is just the fundamental theorem of calculus applied to \( f(t) = F(x + tu) \). For the induction step we use integration by parts. To this end let \( f_j \in C^1([0, 1], X_j) \),
$L \in \mathcal{L}^2(X_1 \times X_2, Y)$ bilinear. Then the product rule (15.16) and the fundamental theorem of calculus imply

$$\int_0^1 L(\dot{f}_1(t), f_2(t))dt = L(f_1(1), f_2(1)) - L(f_1(0), f_2(0)) - \int_0^1 L(f_1(t), \dot{f}_2(t))dt.$$  

Hence applying integration by parts with $L(y, t) = ty$, $f_1(t) = d^r F(x + ut)$, and $f_2(t) = \frac{(1-t)^{r+1}}{(r+1)!}$ establishes the induction step. □

Of course this also gives the Peano form for the remainder:

**Corollary 15.11.** Suppose $U \subseteq X$ and $F \in C^r(U, Y)$. Then

$$F(x+u) = F(x) + dF(x)u + \frac{1}{2} d^2 F(x)u^2 + \cdots + \frac{1}{r!} d^r F(x)u^r + o(|u|^r).$$ (15.28)

**Proof.** Just estimate

$$\left| \left( \frac{1}{(r-1)!} \int_0^1 (1-t)^{r-1} d^r F(x + tu)dt - \frac{1}{r!} d^r F(x) \right) u^r \right|$$

$$\leq \frac{|u|^r}{(r-1)!} \int_0^1 (1-t)^{r-1} \|d^r F(x + tu) - d^r F(x)\| dt$$

$$\leq \frac{|u|^r}{r!} \sup_{0 \leq t \leq 1} \|d^r F(x + tu) - d^r F(x)\|.$$ □

Finally we remark that it is often necessary to equip $C^r(U, Y)$ with a norm. A suitable choice is

$$\|F\| = \sum_{0 \leq j \leq r} \sup_{x \in U} \|d^j F(x)\|. \quad (15.29)$$

The set of all $r$ times continuously differentiable functions for which this norm is finite forms a Banach space which is denoted by $C_b^r(U, Y)$.

In the definition of differentiability we have required $U$ to be open. Of course there is no stringent reason for this and (15.3) could simply be required for all sequences from $U \setminus \{x\}$ converging to $x$. However, note that the derivative might not be unique in case you miss some directions (the ultimate problem occurring at an isolated point). Our requirement avoids all these issues. Moreover, there is usually another way of defining differentiability at a boundary point: By $C^r(\overline{U}, Y)$ we denote the set of all functions in $C^r(U, Y)$ all whose derivatives of order up to $r$ have a continuous extension to $\overline{U}$. Note that if you can approach a boundary point along a half-line then the fundamental theorem of calculus shows that the extension coincides with the Gâteaux derivative.

**Problem 15.1.** Let $X$ be a Hilbert space, $A \in \mathcal{L}(X)$ and $F(x) := \langle x, Ax \rangle$. Compute $d^n F$.  

---
Problem 15.2. Let $X := C[0, 1]$ and suppose $f \in C^1(\mathbb{R})$. Show that

$$F : C[0, 1] \to C[0, 1], \quad x \mapsto f \circ x$$

is differentiable for every $x \in \ell^p(\mathbb{N})$ with derivative given by

$$(dF(x)y)(t) = f'(x(t))y(t).$$

Problem 15.3. Let $X := \ell^2(\mathbb{N})$, $Y := \ell^1(\mathbb{N})$ and $F : X \to Y$ given by $F(x)_j := x_j^2$. Show $F \in C^\infty(X,Y)$ and compute all derivatives.

Problem 15.4. Show (15.15).

Problem 15.5. Let $X$ be a Banach algebra, $I \subseteq \mathbb{R}$ an open interval, and $x, y \in C^1(I, X)$. Show that $xy \in C^1(I, X)$ with

$$(xy)\dot[] = \dot{x}y + xy.$$

Problem 15.6. Let $X$ be a Banach algebra and $G(X)$ the group of invertible elements. Show that $I : G(X) \to G(X)$, $x \mapsto x^{-1}$ is differentiable with

$$dI(x)y = -x^{-1}yx.$$

(Hint: (6.9))

15.2. Minimizing functionals

Many problems in applications leading to finding the minimum (or maximum) of a given functional $F : X \to \mathbb{R}$. Since the minima of $-F$ are the maxima of $F$ and vice versa we will restrict our attention to minima only. Of course if $X = \mathbb{R}$ (or $\mathbb{R}^n$) we can find the local extrema by searching for the zeros of the derivative and then checking the second derivative to determine if it is a minim or maximum. In fact, by virtue of our version of Taylor’s theorem (15.28) we see that $F$ will take values above and below $F(x)$ in a vicinity of $x$ if we can find some $u$ such that $dF(x)u \neq 0$. Hence $dF(x) = 0$ is clearly a necessary condition for a local extremum. Moreover, if $dF(x) = 0$ we can go one step further and conclude that all values in a vicinity of $x$ will lie above $F(x)$ provided the second derivative $d^2F(x)$ is positive in the sense that there is some $c > 0$ such that $d^2F(x)u^2 > c$ for all directions $u \in B_1(0)$. While this gives a viable solution to the problem of finding local extrema, we can easily do a bit better. To this end we look at the variations of $f$ along lines trough $x$, that is, we look at the behavior of the function

$$f(t) := F(x + tu) \quad (15.30)$$

for a fixed direction $u \in B_1(0)$. Then, if $F$ has a local extremum at $x$ the same will be true for $f$ and hence a necessary condition for an extremum is that the Gâteaux derivative vanishes in every direction: $\delta F(x, u) = 0$ for all unit vectors $u$. Similarly, a necessary condition for a local minimum
at $x$ is that $f$ has a local minimum at 0 for all unit vectors $u$. For example $\delta^2 F(x,u) > 0$ for all unit vectors $u$. Here the higher order Gâteaux derivatives are defined as

$$\delta^n F(x,u) := \left( \frac{d}{dt} \right)^n F(x + tu) \bigg|_{t=0} \quad (15.31)$$

with the derivative defined as a limit as in (15.4). That is we have the recursive definition $\delta^n F(x,u) = \lim_{\varepsilon \to 0} \frac{1}{\varepsilon^n} (\delta^{n-1} F(x+\varepsilon u,u) - \delta^{n-1} F(x,u))$. Note that if $\delta^n F(x,u)$ exists, then $\delta^n F(x,\lambda u)$ exists for every $\lambda \in \mathbb{R}$ and

$$\delta^n F(x,\lambda u) = \lambda^n \delta^n F(x,u), \quad \lambda \in \mathbb{R}. \quad (15.32)$$

However, the condition $\delta^2 F(x,u) > 0$ for all unit vectors $u$ is not sufficient as the following example shows.

**Example.** Let $X = \mathbb{R}^2$ and set $F(x,y) = xy\sin(\frac{1}{x})$ if $y = x^2$ and $F(x,y) = 0$ else. Then $F(\alpha t, \beta t)$ has a local minimum at $t = 0$ for every $(\alpha, \beta) \in \mathbb{R}^2$ but $F$ has no local minimum at $(0,0)$. □

**Lemma 15.12.** Suppose $F : U \to \mathbb{R}$ has Gâteaux derivatives up to the order of two. A necessary condition for $x \in U$ to be a local extremum is that $\delta F(x,u) = 0$ and $\delta^2 F(x,u) \geq 0$ for all $u \in X$. A sufficient condition for a strict local minimum is if addition $\delta^2 F(x,u) \geq c > 0$ for all $u \in \partial B_1(0)$ and $\delta^2 F$ is continuous at $x$ uniformly with respect to $u \in \partial B_1(0)$.

**Proof.** The necessary conditions have already been established. To see the sufficient conditions note that the assumptions on $\delta^2 F$ imply that there is some $\varepsilon > 0$ such that $\delta^2 F(y,u) \geq \frac{c}{2}$ for all $y \in B_\varepsilon(x)$ and all $u \in \partial B_1(0)$. Equivalently, $\delta^2 F(y,u) \geq \frac{c}{2} |u|^2$ for all $y \in B_\varepsilon(x)$ and all $u \in X$. Hence applying Taylor’s theorem to $f(t)$ using $\tilde{f}(t) = \delta^2 F(x + tu,u)$ gives

$$F(x + u) = f(1) = f(0) + \int_0^1 (1 - s)\tilde{f}(s)ds \geq F(x) + \frac{c}{4} |u|^2$$

for $u \in B_\varepsilon(0)$. □

Note that if $F \in C^2(U,\mathbb{R})$ then $\delta^2 F(x,u) = d^2 F(x)u^2$ and we obtain

**Corollary 15.13.** Suppose $F \in C^2(U,\mathbb{R})$. A sufficient condition for $x \in U$ to be a strict local minimum is $dF(x) = 0$ and $d^2 F(x)u^2 \geq c|u|^2$ for all $u \in X$.

**Proof.** Observe that by $|\delta^2 F(x,u) - \delta^2 F(y,u)| \leq \|d^2 F(x) - d^2 F(y)\||u|^2$ the continuity requirement from the previous lemma is satisfied. □

**Example.** If $X$ is a Hilbert space, then the symmetric bilinear form $d^2 F$ has a corresponding self-adjoint operator $A \in \mathcal{L}(X)$ such that $d^2 F(u,v) = \langle u, Av \rangle$ and the condition $d^2 F(x)u^2 \geq c|u|^2$ is equivalent to the spectral...
15.2. Minimizing functionals

condition $\sigma(A) \subset [c, \infty)$. In the finite dimensional case $A$ is of course the Jacobi matrix and the spectral conditions says that all eigenvalues must be positive.

\begin{itemize}
\item \textbf{Example.} Let $X = \ell^p(\mathbb{N})$ and consider

$$F(x) := \sum_{n \in \mathbb{N}} \left( \frac{x_n^2}{2n^2} - x_n^4 \right).$$

Then $F \in C^2(X, \mathbb{R})$ with $F(0) = 0$, $dF(0) = 0$ and $d^2F(x)u^2 = \sum_n n^{-2}u_n^2 > 0$ for $u \neq 0$. However, $F(\delta^{mn}/m) < 0$ shows that $0$ is no local minimum. So the condition $d^2F(x)u^2 > 0$ is not sufficient in infinite dimensions. It is however, sufficient in finite dimensions since compactness of the unit ball leads to the stronger condition $d^2F(x,u) \geq c > 0$ for all $u \in \partial B_1(0)$.

\item \textbf{Example.} Suppose we have a particle given whose location at time $t$ is given by $q(t)$. Then the \textbf{least action principle} principle states that if the particle moves from $q(a)$ to $q(b)$ the path of the particle will make the action functional

$$S(q) := \int_a^b L(t, q(t), \dot{q}(t))dt$$

stationary, that is

$$\delta S(q) = 0.$$ 

Here $L : \mathbb{R} \times \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$ is the \textbf{Lagrangian} of the system. The name suggests that the action should attain a minimum, but this is not always the case and hence it is also referred to as \textbf{stationary action principle}.

More precisely, let $L \in C^2(\mathbb{R}^{2n+1}, \mathbb{R})$ and in order to incorporate the requirement that the initial and end points are fixed, we take $X = \{ x \in C^2([a,b], \mathbb{R}^n) | x(a) = x(b) = 0 \}$ and consider

$$q(t) := q(a) + \frac{t-a}{b-a}q(b) + x(t), \quad x \in X.$$ 

Hence we want to compute the Gâteaux derivative of $F(x) := S(q)$, where $x$ and $q$ are related as above with $q(a), q(b)$ fixed. Then

$$\delta F(x,u) = \frac{d}{dt} \bigg|_{t=0} \int_a^b L(s, q(s) + t u(s), \dot{q}(s) + t \dot{u}(s)) ds$$

$$= \int_a^b \left( L_q(s, q(s), \dot{q}(s))u(s) + L_{\dot{q}}(s, q(s), \dot{q}(s))\dot{u}(s) \right) ds$$

$$= \int_a^b \left( L_q(s, q(s), \dot{q}(s))u(s) - \partial_q L_q(s, q(s), \dot{q}(s))u(s) \right) ds,$$
where we have used integration by parts (including the boundary conditions) to obtain the last equality. Here $L_q, L_{\dot{q}}$ are the gradients with respect to $q, \dot{q}$, respectively, and products are understood as scalar products in $\mathbb{R}^n$.

If we want this to vanish for all $u \in X$ we obtain the corresponding **Euler–Lagrange equation**

$$\partial_s L_{\dot{q}}(s, q(s), \dot{q}(s)) = L_q(s, q(s), \dot{q}(s)),$$

For example, for a classical particle of mass $m > 0$ moving in a conservative force field described by a potential $V \in C^1(\mathbb{R}^n, \mathbb{R})$ the Lagrangian is given by the difference between kinetic and potential energy

$$L(t, q, \dot{q}) := \frac{m}{2} \dot{q}^2 - V(q)$$

and the Euler–Lagrange equations read

$$m\ddot{q} = -V_q(q),$$

which are just Newton’s equation of motion.

Finally we note that the situation simplifies a lot when $F$ is convex. Our first observation is that a local minimum is automatically a global one.

**Lemma 15.14.** Suppose $C \subseteq X$ is convex and $F : C \to \mathbb{R}$ is convex. Every local minimum is a global minimum. Moreover, if $F$ is strictly convex then the minimum is unique.

**Proof.** Suppose $x$ is a local minimum and $F(y) < F(x)$. Then $F(\lambda y + (1 - \lambda)x) \leq \lambda F(y) + (1 - \lambda)F(x) < F(x)$ for $\lambda \in (0, 1)$ contradicts the fact that $x$ is a local minimum. If $x, y$ are two global minima, then $F(\lambda y + (1 - \lambda)x) < F(y) = F(x)$ yielding a contradiction unless $x = y$. \hfill \diamond

Moreover, to find the global minimum it suffices to find a point where the Gâteaux derivative vanishes.

**Lemma 15.15.** Suppose $C \subseteq X$ is convex and $F : C \to \mathbb{R}$ is convex. If the Gâteaux derivative exists at an interior point $x \in C$ and satisfies $\delta F(x, u) = 0$ for all $u \in X$, then $x$ is a global minimum.

**Proof.** By assumption $f(t) := F(x + tu)$ is a convex function defined on an interval containing 0 with $f'(0) = 0$. If $y$ is another point we can choose $u = y - x$ and Lemma 10.2 (iii) implies $F(y) = f(1) \geq f(0) = F(x)$. \hfill \square

As in the one-dimensional case, convexity can be read off from the second derivative.

**Lemma 15.16.** Suppose $C \subseteq X$ is open and convex and $F : C \to \mathbb{R}$ has Gâteaux derivatives up to order two. Then $F$ is convex if and only if
\[ \delta^2 F(x, u) \geq 0 \text{ for all } x \in C \text{ and } u \in X. \] Moreover, \( F \) is strictly convex if \( \delta^2 F(x, u) > 0 \) for all \( x \in C \) and \( u \in X \setminus \{0\} \).

**Proof.** We consider \( f(t) := F(x + tu) \) as before such that \( f'(t) = \delta F(x + tu, u) \), \( f''(t) = \delta^2 F(x + tu, u) \). Moreover, note that \( f \) is (strictly) convex for all \( x \in C \) and \( u \in X \setminus \{0\} \) if and only if \( F \) is (strictly) convex. Indeed, if \( F \) is (strictly) convex so is \( f \) as is easy to check. To see the converse note

\[
F(\lambda y + (1 - \lambda)x) = f(\lambda) \leq \lambda f(1) - (1 - \lambda)f(0) = \lambda F(y) - (1 - \lambda)F(x)
\]
with strict inequality if \( f \) is strictly convex. The rest follows from Problem 10.5. \( \square \)

There is also a version using only first derivatives plus the concept of a monotone operator. A map \( F : U \subseteq X \to X' \) is **monotone** if

\[
(F(x) - F(y))(x - y) \geq 0, \quad x, y \in U.
\]
It is called **strictly monotone** if we have strict inequality for \( x \neq y \). Monotone operators will be the topic of Chapter 19.

**Lemma 15.17.** Suppose \( C \subseteq X \) is open and convex and \( F : C \to \mathbb{R} \) has Gâteaux derivatives \( \delta F(x) \) for every \( x \in C \). Then \( F \) is (strictly) convex if and only if \( \delta F \) is (strictly) monotone.

**Proof.** Note that by assumption \( \delta F : C \to X' \) and the claim follows as in the previous lemma from Problem 10.5 since \( f'(t) = \delta F(x + tu)u \) which shows that \( \delta F \) is (strictly) monotone if and only if \( f' \) is (strictly) increasing. \( \square \)

**Example.** The length of a curve \( q : [a, b] \to \mathbb{R}^n \) is given by (cf. Problem 11.39)

\[
\int_a^b |q'(s)| \, ds.
\]
Of course we know that the shortest curve between two given points \( q_0 \) and \( q_1 \) is a straight line. Notwithstanding that this is evident defining the length as the total variation (cf. again Problem 11.39), let us show this by seeking the minimum of the following functional

\[
F(x) := \int_a^b |q'(s)| \, ds, \quad q(t) = x(t) + q_0 + \frac{t - a}{b - a}(q_1 - q_0)
\]
for \( x \in X := \{ x \in C^1([a, b], \mathbb{R}^n) | x(a) = x(b) = 0 \} \). Unfortunately our integrand will not be differentiable unless \( \dot{q} \geq c \). However, since the absolute value is convex, so is \( F \) and it will suffice to search for a local minimum within the convex open set \( C := \{ x \in X | \dot{x} < \frac{|q_1 - q_0|}{2(b - a)} \} \). We compute

\[
\delta F(x, u) = \int_a^b \frac{q'(s)u'(s)}{|q'(s)|} \, ds
\]
which shows (Problem 10.26) that \( q' \) must be constant. Hence the local minimum in \( C \) is indeed a straight line and this must also be a global minimum in \( X \). However, since the length of a curve is independent of its parametrization, this minimum is not unique!

\[ \diamond \]

**Example.** The time for a particle to slide along a curve \( y(x) \) from \( y(0) \) to \( y(x_0) \) is given by

\[
T(y(x)) = \frac{1}{\sqrt{2g}} \int_{0}^{x_0} \sqrt{\frac{1 + y'(x)^2}{x}} \, dx.
\]

The **Brachistochrone problem**, as posed by Johann Bernoulli, asks for the curve which minimizes the time given \( y(0) \) and \( y(x_0) \).

Note that since the function \( t \mapsto \sqrt{1 + t^2} \) is convex, we obtain that \( T \) is convex. Hence it suffices to find a zero of

\[
\delta T(y, u) = \frac{1}{\sqrt{2g}} \int_{0}^{x_0} \frac{y'(x)u'(x)}{\sqrt{x(1 + y'(x)^2)}} \, dx
\]

which shows (Problem 10.26) that \( \frac{y'}{\sqrt{x(1 + y'^2)}} = C^{-1/2} \) is constant or equivalently

\[
y'(x) = \sqrt{\frac{x}{C - x}}
\]

and hence

\[
y(x) = y(0) + C \arctan \left( \sqrt{\frac{x}{C - x}} \right) - \sqrt{x(C - x)}.
\]

The constant \( C \) has to be chosen such that \( y(x_0) \) matches the given value. \( \diamond \)

**Problem 15.7.** Consider the least action principle for a classical one-dimensional particle. Show that

\[
\delta^2 F(x, u) = \int_{a}^{b} \left( m \ddot{u}(s)^2 - V''(q(s))u(s)^2 \right) ds.
\]

Moreover, show that we have indeed a minimum if \( V'' \leq 0 \).

### 15.3. Contraction principles

Let \( X \) be a Banach space. A fixed point of a mapping \( F : C \subseteq X \rightarrow C \) is an element \( x \in C \) such that \( F(x) = x \). Moreover, \( F \) is called a contraction if there is a contraction constant \( \theta \in [0, 1) \) such that

\[
|F(x) - F(\tilde{x})| \leq \theta |x - \tilde{x}|, \quad x, \tilde{x} \in C.
\]

(15.33)

Note that a contraction is continuous. We also recall the notation \( F^n(x) = F(F^{n-1}(x)), F^0(x) = x \).
Theorem 15.18 (Contraction principle). Let $C$ be a nonempty closed subset of a Banach space $X$ and let $F : C \to C$ be a contraction, then $F$ has a unique fixed point $\bar{x} \in C$ such that
\[ |F^n(x) - \bar{x}| \leq \frac{\theta^n}{1 - \theta} |F(x) - x|, \quad x \in C. \quad (15.34) \]

Proof. If $x = F(x)$ and $\tilde{x} = F(\tilde{x})$, then $|x - \tilde{x}| = |F(x) - F(\tilde{x})| \leq \theta |x - \tilde{x}|$ shows that there can be at most one fixed point.

Concerning existence, fix $x_0 \in C$ and consider the sequence $x_n = F^n(x_0)$. We have
\[ |x_{n+1} - x_n| \leq \theta |x_n - x_{n-1}| \leq \cdots \leq \theta^n |x_1 - x_0| \]
and hence by the triangle inequality (for $n > m$)
\[
|x_n - x_m| \leq \sum_{j=m+1}^{n} |x_j - x_{j-1}| \leq \theta^m \sum_{j=0}^{n-m-1} \theta^j |x_1 - x_0| \\
\leq \frac{\theta^m}{1 - \theta} |x_1 - x_0|. \quad (15.35)
\]

Thus $x_n$ is Cauchy and tends to a limit $\bar{x}$. Moreover,
\[ |F(\bar{x}) - \bar{x}| = \lim_{n \to \infty} |x_{n+1} - x_n| = 0 \]
shows that $\bar{x}$ is a fixed point and the estimate (15.34) follows after taking the limit $m \to \infty$ in (15.35). \qed

Note that we can replace $\theta^n$ by any other summable sequence $\theta_n$ (Problem 15.9):

Theorem 15.19 (Weissinger). Let $C$ be a nonempty closed subset of a Banach space $X$. Suppose $F : C \to C$ satisfies
\[ |F^n(x) - F^n(y)| \leq \theta_n |x - y|, \quad x, y \in C, \quad (15.36) \]
with $\sum_{n=1}^{\infty} \theta_n < \infty$. Then $F$ has a unique fixed point $\bar{x}$ such that
\[ |F^n(x) - \bar{x}| \leq \left( \sum_{j=n}^{\infty} \theta_j \right) |F(x) - x|, \quad x \in C. \quad (15.37) \]

Next, we want to investigate how fixed points of contractions vary with respect to a parameter. Let $X, Y$ be Banach spaces, $U \subseteq X$, $V \subseteq Y$ be open and consider $F : \overline{U} \times V \to U$. The mapping $F$ is called a uniform contraction if there is a $\theta \in [0, 1)$ such that
\[ |F(x, y) - F(\tilde{x}, y)| \leq \theta |x - \tilde{x}|, \quad x, \tilde{x} \in \overline{U}, \ y \in V, \quad (15.38) \]
that is, the contraction constant $\theta$ is independent of $y$. 
**Theorem 15.20** (Uniform contraction principle). Let $U, V$ be nonempty open subsets of Banach spaces $X, Y$, respectively. Let $F : U \times V \to U$ be a uniform contraction and denote by $\overline{y}(y) \in U$ the unique fixed point of $F(\cdot, y)$. If $F \in C^r(U \times V, U)$, $r \geq 0$, then $\overline{y}(\cdot) \in C^r(V, U)$.

**Proof.** Let us first show that $\overline{y}(y)$ is continuous. From

$$|\overline{y}(y + v) - \overline{y}(y)| = |F(\overline{y}(y + v), y + v) - F(\overline{y}(y), y + v) + F(\overline{y}(y), y + v) - F(\overline{y}(y), y)|$$

$$\leq \theta|\overline{y}(y + v) - \overline{y}(y)| + |F(\overline{y}(y), y + v) - F(\overline{y}(y), y)|$$

(15.39)

we infer

$$|\overline{y}(y + v) - \overline{y}(y)| \leq \frac{1}{1 - \theta}|F(\overline{y}(y), y + v) - F(\overline{y}(y), y)|$$

(15.40)

and hence $\overline{y}(y) \in C(V, U)$. Now let $r = 1$ and let us formally differentiate $\overline{y}(y) = F(\overline{y}(y), y)$ with respect to $y$,

$$d\overline{y}(y) = \partial_x F(\overline{y}(y), y) d\overline{y}(y) + \partial_y F(\overline{y}(y), y).$$

(15.41)

Considering this as a fixed point equation $T(x', y) = x'$, where $T(\cdot, y) : \mathcal{L}(Y, X) \to \mathcal{L}(Y, X), x' \mapsto \partial_x F(\overline{y}(y), y)x' + \partial_y F(\overline{y}(y), y)$ is a uniform contraction since we have $\|\partial_x F(\overline{y}(y), y)\| \leq \theta$ by Theorem 15.6. Hence we get a unique continuous solution $\overline{y}'(y)$. It remains to show

$$\overline{y}(y + v) - \overline{y}(y) - \overline{y}'(y)v = o(v).$$

(15.42)

Let us abbreviate $u = \overline{y}(y + v) - \overline{y}(y)$, then using (15.41) and the fixed point property of $\overline{y}(y)$ we see

$$(1 - \partial_x F(\overline{y}(y), y))(u - \overline{y}'(y)v) =$$

$$= F(\overline{y}(y) + u, y + v) - F(\overline{y}(y), y) - \partial_x F(\overline{y}(y), y)u - \partial_y F(\overline{y}(y), y)v$$

$$= o(u) + o(v)$$

(15.43)

since $F \in C^4(U \times V, U)$ by assumption. Moreover, $\|(1 - \partial_x F(\overline{y}(y), y))^{-1}\| \leq (1 - \theta)^{-1}$ and $u = O(v)$ (by (15.40)) implying $u - \overline{y}'(y)v = o(v)$ as desired. Finally, suppose that the result holds for some $r - 1 \geq 1$. Thus, if $F$ is $C^r$, then $\overline{y}(y)$ is at least $C^{r-1}$ and the fact that $d\overline{y}(y)$ satisfies (15.41) implies $\overline{y}(y) \in C^r(V, U)$.

As an important consequence we obtain the implicit function theorem.

**Theorem 15.21** (Implicit function). Let $X, Y$, and $Z$ be Banach spaces and let $U, V$ be open subsets of $X, Y$, respectively. Let $F \in C^r(U \times V, Z)$, $r \geq 0$, and fix $(x_0, y_0) \in U \times V$. Suppose $\partial_x F \in C(U \times V, Z)$ exists (if $r = 0$) and $\partial_x F(x_0, y_0) \in \mathcal{L}(X, Z)$ is an isomorphism. Then there exists an open neighborhood $U_1 \times V_1 \subseteq U \times V$ of $(x_0, y_0)$ such that for each $y \in V_1$ there
exists a unique point \((\xi(y), y) \in U_1 \times V_1\) satisfying \(F(\xi(y), y) = F(x_0, y_0)\). Moreover, \(\xi\) is in \(C^r(V_1, Z)\) and fulfills (for \(r \geq 1\))

\[
d\xi(y) = -(\partial_x F(\xi(y), y))^{-1} \circ \partial_y F(\xi(y), y).
\] (15.44)

**Proof.** Using the shift \(F \to F - F(x_0, y_0)\) we can assume \(F(x_0, y_0) = 0\). Next, the fixed points of \(G(x, y) = x - (\partial_x F(x_0, y_0))^{-1} F(x, y)\) are the solutions of \(F(x, y) = 0\). The function \(G\) has the same smoothness properties as \(F\) and since \(\partial_x G(x_0, y_0) = 0\), we can find balls \(U_1\) and \(V_1\) around \(x_0\) and \(y_0\) such that \(\|\partial_x G(x, y)\| \leq \theta < 1\) for \((x, y) \in U_1 \times V_1\). Thus by the mean value theorem (Theorem 15.6) \(G(., y)\) is a uniform contraction on \(U_1\) for \(y \in V_1\). Moreover, choosing the radius of \(V_1\) to be less then \(\frac{1-\theta}{\theta} r\) where \(r\) is the radius of \(U_1\), the mean value theorem also shows

\[
|G(x, y) - x_0| = |G(x, y) - G(x_0, y_0)| \leq \theta(|x - x_0| + |y - y_0|) < r
\]
for \((x, y) \in U_1 \times V_1\), that is, \(G : \overline{U_1} \times V_1 \to U_1\). The rest follows from the uniform contraction principle. Formula (15.44) follows from differentiating \(F(\xi(y), y) = 0\) using the chain rule. \(\square\)

Note that our proof is constructive, since it shows that the solution \(\xi(y)\) can be obtained by iterating \(x - (\partial_x F(x_0, y_0))^{-1}(F(x, y) - F(x_0, y_0))\).

Moreover, as a corollary of the implicit function theorem we also obtain the inverse function theorem.

**Theorem 15.22** (Inverse function). Suppose \(F \in C^r(U, Y)\), \(r \geq 1\), \(U \subseteq X\), and let \(dF(x_0)\) be an isomorphism for some \(x_0 \in U\). Then there are neighborhoods \(U_1, V_1\) of \(x_0, F(x_0)\), respectively, such that \(F \in C^r(U_1, V_1)\) is a diffeomorphism.

**Proof.** Apply the implicit function theorem to \(G(x, y) = y - F(x)\). \(\square\)

**Example.** Let \(X\) be a Banach algebra and \(G(X)\) the group of invertible elements. We have seen that multiplication is \(C^\infty(X \times X, X)\) and hence taking the inverse is also \(C^\infty(G(X), G(X))\). Consequently, \(G(X)\) is an (in general infinite-dimensional) **Lie group.** \(\diamondsuit\)

Further applications will be given in the next section.

**Problem 15.8.** Derive Newton’s method for finding the zeros of a twice continuously differentiable function \(f(x)\),

\[
x_{n+1} = F(x_n), \quad F(x) = x - \frac{f(x)}{f'(x)},
\]
from the contraction principle by showing that if \(x\) is a zero with \(f'(x) \neq 0\), then there is a corresponding closed interval \(C\) around \(x\) such that the assumptions of Theorem 15.18 are satisfied.
Problem 15.9. Prove Theorem 15.19. Moreover, suppose \( F : C \to C \) and that \( F^n \) is a contraction. Show that the fixed point of \( F^n \) is also one of \( F \). Hence Theorem 15.19 (except for the estimate) can also be considered as a special case of Theorem 15.18 since the assumption implies that \( F^n \) is a contraction for \( n \) sufficiently large.

15.4. Ordinary differential equations

As a first application of the implicit function theorem, we prove (local) existence and uniqueness for solutions of ordinary differential equations in Banach spaces. Let \( X \) be a Banach space and \( U \subseteq X \). Denote by \( C_b(I,U) \) the Banach space of bounded continuous functions equipped with the sup norm.

The following lemma, known as omega lemma, will be needed in the proof of the next theorem.

Lemma 15.23. Suppose \( I \subseteq \mathbb{R} \) is a compact interval and \( f \in C^r(U,Y) \). Then \( f_* \in C^r(C_b(I,U),C_b(I,Y)) \), where

\[
(f_*x)(t) = f(x(t)).
\] (15.45)

Proof. Fix \( x_0 \in C_b(I,U) \) and \( \varepsilon > 0 \). For each \( t \in I \) we have a \( \delta(t) > 0 \) such that \( \overline{B}_{2\delta(t)}(x_0(t)) \subseteq U \) and \( |f(x) - f(x_0(t))| \leq \varepsilon/2 \) for all \( x \) with \( |x - x_0(t)| \leq 2\delta(t) \). The balls \( B_{\delta(t)}(x_0(t)), t \in I \), cover the set \( \{x_0(t)\}_{t \in I} \) and since \( I \) is compact, there is a finite subcover \( B_{\delta(t_j)}(x_0(t_j)), 1 \leq j \leq n \).

Let \( |x - x_0| \leq \delta := \min_{1 \leq j \leq n} \delta(t_j) \). Then for each \( t \in I \) there is \( t_i \) such that \( |x_0(t) - x_0(t_j)| \leq \delta(t_j) \) and hence \( |f(x(t)) - f(x_0(t_j))| \leq |f(x(t)) - f(x_0(t_j))| + |f(x_0(t_j)) - f(x_0(t))| \leq \varepsilon \) since \( |x(t) - x_0(t_j)| \leq |x(t) - x_0(t)| + |x_0(t) - x_0(t_j)| \leq 2\delta(t_j) \). This settles the case \( r = 0 \).

Next let us turn to \( r = 1 \). We claim that \( df_* \) is given by \( (df_*x_0)x(t) = df(x_0(t))x(t) \). Hence we need to show that for each \( \varepsilon > 0 \) we can find a \( \delta > 0 \) such that

\[
\sup_{t \in I} |f(x_0(t) + x(t)) - f(x_0(t)) - df(x_0(t))x(t)| \leq \varepsilon \sup_{t \in I} |x(t)|
\] (15.46)

whenever \( |x| = \sup_{t \in I} |x(t)| \leq \delta \). By assumption we have

\[
|f(x_0(t) + x(t)) - f(x_0(t)) - df(x_0(t))x(t)| \leq \varepsilon |x(t)|
\] (15.47)

whenever \( |x(t)| \leq \delta(t) \). Now argue as before to show that \( \delta(t) \) can be chosen independent of \( t \). It remains to show that \( df_* \) is continuous. To see this we use the linear map

\[
\lambda : C_b(I,L(X,Y)) \to L(C_b(I,X),C_b(I,Y)), \quad T \mapsto T_*
\] (15.48)
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where \((T_x)(t) = T(t)x(t)\). Since we have
\[
|T_x| = \sup_{t \in I} |T(t)x(t)| \leq \sup_{t \in I} \|T(t)\| \|x(t)| \leq |T| \|x|,
\]
we infer \(|\lambda| \leq 1\) and hence \(\lambda\) is continuous. Now observe \(df_\ast = \lambda \circ (df)_\ast\).

The general case \(r > 1\) follows from induction. \(\square\)

Now we come to our existence and uniqueness result for the initial value problem in Banach spaces.

**Theorem 15.24.** Let I be an open interval, \(U\) an open subset of a Banach space \(X\) and \(\Lambda\) an open subset of another Banach space. Suppose \(F \in C^r(I \times U \times \Lambda, X), r \geq 1\), then the initial value problem
\[
\dot{x} = F(t, x, \lambda), \quad x(t_0) = x_0, \quad (t_0, x_0, \lambda) \in I \times U \times \Lambda,
\]
has a unique solution \(x(t, t_0, x_0, \lambda) \in C^r(I_1 \times I_2 \times U_1 \times \Lambda_1, X)\), where \(I_{1,2}, U_1, \) and \(\Lambda_1\) are open subsets of I, U, and \(\Lambda\), respectively. The sets \(I_2, U_1,\) and \(\Lambda_1\) can be chosen to contain any point \((t_0, x_0, \lambda)\) of \(I \times U \times \Lambda\), respectively.

**Proof.** Adding \(t\) and \(\lambda\) to the dependent variables \(x\), that is considering \((\tau, x, \lambda) \in \mathbb{R} \times X \times \Lambda\) and augmenting the differential equation according to \((\hat{\tau}, \hat{x}, \hat{\lambda}) = (1, F(\tau, x, \lambda), 0)\), we can assume that \(F\) is independent of \(t\) and \(\lambda\). Moreover, by a translation we can even assume \(t_0 = 0\).

Our goal is to invoke the implicit function theorem. In order to do this we introduce an additional parameter \(\varepsilon \in \mathbb{R}\) and consider
\[
\dot{x} = \varepsilon F(x_0 + x), \quad x \in D^1 = \{x \in C^1([-1, 1], B_\delta(0))| x(0) = 0\},
\]
such that we know the solution for \(\varepsilon = 0\). The implicit function theorem will show that solutions still exist as long as \(\varepsilon\) remains small. At first sight this doesn’t seem to be good enough for us since our original problem corresponds to \(\varepsilon = 1\). But since \(\varepsilon\) corresponds to a scaling \(t \to \varepsilon t\), the solution for one \(\varepsilon > 0\) suffices. Now let us turn to the details.

Our problem (15.51) is equivalent to looking for zeros of the function
\[
G : \quad D^1 \times U_0 \times \mathbb{R} \quad \to \quad C^1([-1, 1], X),
\]
\[
(x, x_0, \varepsilon) \quad \mapsto \quad \dot{x} - \varepsilon F(x_0 + x),
\]
where \(U_0\) is a neighborhood of \(x_0\) and \(\delta\) sufficiently small such that \(U_0 + B_\delta(0) \subseteq U\). Lemma 15.23 ensures that this function is \(C^1\). Now fix \(x_0\), then \(G(0, x_0, 0) = 0\) and \(\partial_x G(0, x_0, 0) = T\), where \(T x = \dot{x}\). Since \((T^{-1} x)(t) = \int_0^t x(s) \, ds\) we can apply the implicit function theorem to conclude that there is a unique solution \(x(x_0, \varepsilon) \in C^1(U_1 \times (-\varepsilon_0, \varepsilon_0), D^1) \subseteq C^1([-1, 1] \times U_1 \times (-\varepsilon_0, \varepsilon_0), X)\). In particular, the map \((t, x_0) \mapsto x_0 + x(x_0, \varepsilon)(t/\varepsilon)\) is in
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C^1((−ε,ε) × U_1, X). Hence it is the desired solution of our original problem. This settles the case r = 1.

For r > 1 we use induction. Suppose F ∈ C^{r+1} and let x(t, x_0) be the solution which is at least C^r. Moreover, y(t, x_0) := ∂_{x_0}x(t, x_0) satisfies
\[ \dot{y} = \partial_x F(x(t, x_0))y, \quad y(0) = I, \]
and hence y(t, x_0) ∈ C^r. Moreover, the differential equation shows ∂_t x(t, x_0) = F(x(t, x_0)) ∈ C^r which shows x(t, x_0) ∈ C^{r+1}. □

Example. The simplest example is a linear equation
\[ \dot{x} = Ax, \quad x(0) = x_0, \]
where A ∈ L(X). Then it is easy to verify that the solution is given by
\[ x(t) = \exp(tA)x_0, \]
where
\[ \exp(tA) = \sum_{k=0}^{\infty} \frac{t^k}{k!} A^k. \]
It is easy to check that the last series converges absolutely (cf. also Problem 1.32) and solves the differential equation (Problem 15.10). ⊳

Example. The classical example is a linear equation
\[ \dot{x} = x^2, \quad x(0) = x_0, \]
in X = R with solution
\[ x(t) = \frac{x_0}{1 - x_0 t}, \quad t \in \begin{cases} \ (-\infty, \frac{1}{x_0}), & x_0 > 0, \\ \ R, & x_0 = 0, \\ \ (\frac{1}{x_0}, \infty), & x_0 < 0. \end{cases} \]
shows that solutions might not exist for all t ∈ R even though the differential equation is defined for all t ∈ R. ⊳

This raises the question about the maximal interval on which a solution of the initial value problem (15.50) can be defined.

Suppose that solutions of the initial value problem (15.50) exist locally and are unique (as guaranteed by Theorem 15.24). Let φ_1, φ_2 be two solutions of (15.50) defined on the open intervals I_1, I_2, respectively. Let I = I_1 ∩ I_2 = (t_−, t_+) and let (t_−, t_+) be the maximal open interval on which both solutions coincide. I claim that (t_−, t_+) = (T_−, T_+). In fact, if t_+ < T_+, both solutions would also coincide at t_+ by continuity. Next, considering the initial value problem with initial condition x(t_+) = φ_1(t_+) = φ_2(t_+) shows that both solutions coincide in a neighborhood of t_+ by local uniqueness. This contradicts maximality of t_+ and hence t_+ = T_+. Similarly, t_− = T_−.
Moreover, we get a solution

\[ \phi(t) = \begin{cases} \phi_1(t), & t \in I_1, \\ \phi_2(t), & t \in I_2, \end{cases} \]

defined on \( I_1 \cup I_2 \). In fact, this even extends to an arbitrary number of solutions and in this way we get a (unique) solution defined on some maximal interval.

**Theorem 15.25.** Suppose the initial value problem (15.50) has a unique local solution (e.g. the conditions of Theorem 15.24 are satisfied). Then there exists a unique maximal solution defined on some maximal interval \( I(t_0, x_0) = (T_-(t_0, x_0), T_+(t_0, x_0)) \).

**Proof.** Let \( S \) be the set of all solutions \( \phi \) of (15.50) which are defined on an open interval \( I_\phi \). Let \( I = \bigcup_{\phi \in S} I_\phi \), which is again open. Moreover, if \( t_1 > t_0 \in I \), then \( t_1 \in I_\phi \) for some \( \phi \) and thus \( [t_0, t_1] \subseteq I_\phi \subseteq I \). Similarly for \( t_1 < t_0 \) and thus \( I \) is an open interval containing \( t_0 \). In particular, it is of the form \( I = (T_-, T_+) \). Now define \( \phi_{\max}(t) \) on \( I \) by \( \phi_{\max}(t) = \phi(t) \) for some \( \phi \in S \) with \( t \in I_\phi \). By our above considerations any two \( \phi \) will give the same value, and thus \( \phi_{\max}(t) \) is well-defined. Moreover, for every \( t_1 > t_0 \) there is some \( \phi \in S \) such that \( t_1 \in I_\phi \) and \( \phi_{\max}(t) = \phi(t) \) for \( t \in (t_0 - \varepsilon, t_1 + \varepsilon) \) which shows that \( \phi_{\max} \) is a solution. By construction there cannot be a solution defined on a larger interval. \( \square \)

The solution found in the previous theorem is called the maximal solution. A solution defined for all \( t \in \mathbb{R} \) is called a global solution. Clearly every global solution is maximal.

The next result gives a simple criterion for a solution to be global.

**Lemma 15.26.** Suppose \( F \in C^1(\mathbb{R} \times X, X) \) and let \( x(t) \) be a maximal solution of the initial value problem (15.50). Suppose \( |F(t, x(t))| \) is bounded on finite \( t \)-intervals. Then \( x(t) \) is a global solution.

**Proof.** Let \( (T_-, T_+) \) be the domain of \( x(t) \) and suppose \( T_+ < \infty \). Then \( |F(t, x(t))| \leq C \) for \( t \in (t_0, T_+) \) and for \( t_0 s < t < T_+ \) we have

\[
|x(t) - x(s)| \leq \int_s^t |x(\tau)|d\tau = \int_s^t |F(\tau, x(\tau))|d\tau \leq C|t - s|.
\]

Thus \( x(t_n) \) is Cauchy whenever \( t_n \) is and hence \( \lim_{t \to T_+} x(t) = x_+ \) exists. Now let \( y(t) \) be the solution satisfying the initial condition \( y(T_+) = x_+ \). Then

\[ \tilde{x}(t) = \begin{cases} x(t), & t < T_+, \\ y(t), & t \geq T_+, \end{cases} \]

is a larger solution contradicting maximality of \( T_+ \). \( \square \)
Example. Finally, we want to apply this to a famous example, the so-called FPU lattices (after Enrico Fermi, John Pasta, and Stanislaw Ulam who investigated such systems numerically). This is a simple model of a linear chain of particles coupled via nearest neighbor interactions. Let us assume for simplicity that all particles are identical and that the interaction is described by a potential $V \in C^2(\mathbb{R})$. Then the equation of motions are given by

$$\ddot{q}_n(t) = V'(q_{n+1} - q_n) - V'(q_n - q_{n-1}), \quad n \in \mathbb{Z},$$

where $q_n(t) \in \mathbb{R}$ denotes the position of the $n$'th particle at time $t \in \mathbb{R}$ and the particle index $n$ runs through all integers. If the potential is quadratic, $V(r) = k \frac{1}{2} r^2$, then we get the discrete linear wave equation

$$\ddot{q}_n(t) = k(q_{n+1}(t) - 2q_n(t) + q_{n-1}(t)).$$

If we use the fact that the Jacobi operator $Aq_n = q_{n+1} - 2q_n + q_{n-1}$ is a bounded operator in $X = l^p(\mathbb{Z})$ we can easily solve this system as in the case of ordinary differential equations. In fact, if $q^0 = q(0)$ and $p^0 = \dot{q}(0)$ are the initial conditions then one can easily check (cf. Problem 15.10) that the solution is given by

$$q(t) = \cos(tA^{1/2})q^0 + \frac{\sin(tA^{1/2})}{A^{1/2}}p^0.$$

In the Hilbert space case $p = 2$ these functions of our operator $A$ could be defined via the spectral theorem but here we just use the more direct definition

$$\cos(tA^{1/2}) = \sum_{k=0}^{\infty} \frac{t^{2k}}{(2k)!} A^k, \quad \frac{\sin(tA^{1/2})}{A^{1/2}} = \sum_{k=0}^{\infty} \frac{t^{2k}}{(2k+1)!} A^k.$$

In the general case an explicit solution is no longer possible but we are still able to show global existence under appropriate conditions. To this end we will assume that $V$ has a global minimum at 0 and hence looks like $V(r) = V(0) + \frac{k}{2} r^2 + o(r^2)$. As $V(0)$ does not enter our differential equation we will assume $V(0) = 0$ without loss of generality. Moreover, we will also introduce $p_n = \dot{q}_n$ to have a first order system

$$\dot{q}_n = p_n, \quad \dot{p}_n = V'(q_{n+1} - q_n) - V'(q_n - q_{n-1}).$$

Since $V' \in C^1(\mathbb{R})$ with $V'(0) = 0$ it gives rise to a $C^1$ map on $l^p(\mathbb{N})$ (see the example on page 436). Since the same is true for shifts, the chain rule implies that the right-hand side of our system is a $C^1$ map and hence Theorem 15.24 gives us existence of a local solution. To get global solutions we will need a bound on solutions. This will follow from the fact that the energy of the
15.4. Ordinary differential equations

system

\[ H(p, q) = \sum_{n \in \mathbb{Z}} \left( \frac{p_n^2}{2} + V(q_{n+1} - q_n) \right) \]

is conserved. To ensure that the above sum is finite we will choose \( X = \ell^2(\mathbb{Z}) \oplus \ell^2(\mathbb{Z}) \) as our underlying Banach (in this case even Hilbert) space. Recall that since we assume \( V \) to have a minimum at 0 we have \( |V(r)| \leq CRr^2 \) for \( |r| < R \) and hence \( H(p, q) < \infty \) for \((p, q) \in X\). Under these assumptions it is easy to check that \( H \in C^1(X, \mathbb{R}) \) and that

\[
\frac{d}{dt} H(p(t), q(t)) = \sum_{n \in \mathbb{Z}} \left( \dot{p}_n(t)p_n(t) + V'(q_{n+1}(t) - q_n(t))(q_{n+1}(t) - \dot{q}_n(t)) \right)
\]

\[
= \sum_{n \in \mathbb{Z}} \left( (V'(q_{n+1} - q_n) - V'(q_n - q_{n-1}))p_n(t) \\
+ V'(q_{n+1}(t) - q_n(t))(p_{n+1}(t) - p_n(t)) \right)
\]

\[
= \sum_{n \in \mathbb{Z}} \left( -V'(q_n - q_{n-1})p_n(t) + V'(q_{n+1}(t) - q_n(t))p_{n+1}(t) \right)
\]

= 0

provided \((p(t), q(t))\) solves our equation. Consequently, since \( V \geq 0 \),

\[ \|p(t)\|_2 \leq 2H(p(t), q(t)) = 2H(p(0), q(0)). \]

Moreover, \( q_n(t) = q_n(0) + \int_0^t p_n(s)ds \) (note that since the \( \ell^2 \) norm is stronger than the \( \ell^\infty \) norm, \( q_n(t) \) is differentiable for fixed \( n \)) implies

\[ \|q(t)\|_2 \leq \|q(0)\|_2 + \int_0^t \|p_n(s)\|_2 ds \leq \|q(0)\|_2 + 2H(p(0), q(0))t. \]

So Lemma 15.26 ensures that solutions are global in \( X \). Of course every solution from \( X \) is also a solutions from \( Y = \ell^p(\mathbb{Z}) \oplus \ell^p(\mathbb{Z}) \) for all \( p \geq 2 \) (since the \( \|\cdot\|_2 \) norm is stronger than the \( \|\cdot\|_p \) norm for \( p \geq 2 \)).

Examples include the original FPU \( \beta \)-model

\[ V_\beta(r) = \frac{1}{2}r^2 + \frac{\beta}{4}r^4, \beta > 0, \]

and the famous Toda lattice \( V(r) = e^{-r} + r - 1 \).

It should be mentioned that the above theory does not suffice to cover partial differential equations. In fact, if we replace the difference operator by a differential operator we run into the problem that differentiation is not a continuous process!

**Problem 15.10.** Let

\[ f(z) = \sum_{j=0}^\infty f_j z^j, \quad |z| < R, \]
be a convergent power series with convergence radius $R > 0$. Suppose $X$ is a Banach space and $A \in \mathcal{L}(X)$ is a bounded operator with $\|A\| < R$. Show that

$$f(tA) = \sum_{j=0}^{\infty} f_j t^j A^j$$

is in $C^\infty(I, \mathcal{L}(X))$, $I = (-R\|A\|^{-1}, R\|A\|^{-1})$ and

$$\frac{d^n}{dt^n} f(tA) = A^n f^{(n)}(tA), \quad n \in \mathbb{N}_0.$$

(Compare also Problem 1.32.)

**Problem 15.11.** Consider the FPU $\alpha$-model $V_\alpha(r) = \frac{1}{2}r^2 + \frac{\alpha}{3}r^3$. Show that solutions satisfying $\|q_{n+1}(0) - q_n(0)\|_\infty < \frac{1}{|\alpha|}$ and $H(p(0), q(0)) < \frac{1}{6\alpha^2}$ are global in $X = \ell^2(\mathbb{Z}) \oplus \ell^2(\mathbb{Z})$. (Hint: Of course local solutions follow from our considerations above. Moreover, note that $V(r)$ has a maximum at $r = -\frac{1}{3\alpha}$. Now use conservation of energy to conclude that the solution cannot escape the region $|r| < \frac{1}{|\alpha|}$.)
Chapter 16

The Brouwer mapping degree

16.1. Introduction

Many applications lead to the problem of finding all zeros of a mapping \( f : U \subseteq X \rightarrow X \), where \( X \) is some (real) Banach space. That is, we are interested in the solutions of

\[
f(x) = 0, \quad x \in U.
\]  
(16.1)

In most cases it turns out that this is too much to ask for, since determining the zeros analytically is in general impossible.

Hence one has to ask some weaker questions and hope to find answers for them. One such question would be "Are there any solutions, respectively, how many are there?". Luckily, these questions allow some progress.

To see how, let's consider the case \( f \in \mathcal{H}(\mathbb{C}) \), where \( \mathcal{H}(U) \) denotes the set of holomorphic functions on a domain \( U \subset \mathbb{C} \). Recall the concept of the winding number from complex analysis. The winding number of a path \( \gamma : [0,1] \rightarrow \mathbb{C} \setminus \{z_0\} \) around a point \( z_0 \in \mathbb{C} \) is defined by

\[
n(\gamma, z_0) := \frac{1}{2\pi i} \int_{\gamma} \frac{dz}{z - z_0} \in \mathbb{Z}.
\]  
(16.2)

It gives the number of times \( \gamma \) encircles \( z_0 \) taking orientation into account. That is, encirclements in opposite directions are counted with opposite signs.

In particular, if we pick \( f \in \mathcal{H}(\mathbb{C}) \) one computes (assuming \( 0 \not\in f(\gamma) \))

\[
n(f(\gamma), 0) = \frac{1}{2\pi i} \int_{\gamma} \frac{f'(z)}{f(z)} dz = \sum_k n(\gamma, z_k) \alpha_k,
\]  
(16.3)
where $z_k$ denote the zeros of $f$ and $\alpha_k$ their respective multiplicity. Moreover, if $\gamma$ is a Jordan curve encircling a simply connected domain $U \subset \mathbb{C}$, then $n(\gamma, z_k) = 0$ if $z_k \not\in U$ and $n(\gamma, z_k) = 1$ if $z_k \in U$. Hence $n(f(\gamma), 0)$ counts the number of zeros inside $U$.

However, this result is useless unless we have an efficient way of computing $n(f(\gamma), 0)$ (which does not involve the knowledge of the zeros $z_k$). This is our next task.

Now, let’s recall how one would compute complex integrals along complicated paths. Clearly, one would use homotopy invariance and look for a simpler path along which the integral can be computed and which is homotopic to the original one. In particular, if $f : \gamma \to \mathbb{C} \setminus \{0\}$ and $g : \gamma \to \mathbb{C} \setminus \{0\}$ are homotopic, we have $n(f(\gamma), 0) = n(g(\gamma), 0)$ (which is known as Rouché’s theorem).

More explicitly, we need to find a mapping $g$ for which $n(g(\gamma), 0)$ can be computed and a homotopy $H : [0, 1] \times \gamma \to \mathbb{C} \setminus \{0\}$ such that $H(0, z) = f(z)$ and $H(1, z) = g(z)$ for $z \in \gamma$. For example, how many zeros of $f(z) = \frac{1}{2}z^6 + z - \frac{1}{3}$ lie inside the unit circle? Consider $g(z) = z$, then $H(t, z) = (1 - t)f(z) + tg(z)$ is the required homotopy since $|f(z) - g(z)| < |g(z)|$, $|z| = 1$, implying $H(t, z) \neq 0$ on $[0, 1] \times \gamma$. Hence $f(z)$ has one zero inside the unit circle.

Summarizing, given a (sufficiently smooth) domain $U$ with enclosing Jordan curve $\partial U$, we have defined a degree $\deg(f, U, z_0) = n(f(\partial U), z_0) = n(f(\partial U) - z_0, 0) \in \mathbb{Z}$ which counts the number of solutions of $f(z) = z_0$ inside $U$. The invariance of this degree with respect to certain deformations of $f$ allowed us to explicitly compute $\deg(f, U, z_0)$ even in nontrivial cases.

Our ultimate goal is to extend this approach to continuous functions $f : \mathbb{R}^n \to \mathbb{R}^n$. However, such a generalization runs into several problems. First of all, it is unclear how one should define the multiplicity of a zero. But even more severe is the fact, that the number of zeros is unstable with respect to small perturbations. For example, consider $f_\varepsilon : [-1, 2] \to \mathbb{R}$, $x \mapsto x^2 - \varepsilon$. Then $f_\varepsilon$ has no zeros for $\varepsilon < 0$, one zero for $\varepsilon = 0$, two zeros for $0 < \varepsilon \leq 1$, one for $1 < \varepsilon \leq \sqrt{2}$, and none for $\varepsilon > \sqrt{2}$. This shows the following facts.

(i) Zeros with $f' \neq 0$ are stable under small perturbations.

(ii) The number of zeros can change if two zeros with opposite sign change (i.e., opposite signs of $f'$) run into each other.

(iii) The number of zeros can change if a zero drops over the boundary.

Hence we see that we cannot expect too much from our degree. In addition, since it is unclear how it should be defined, we will first require some basic
16.2. Definition of the mapping degree and the determinant formula

To begin with, let us introduce some useful notation. Throughout this section $U$ will be a bounded open subset of $\mathbb{R}^n$. For $f \in C^1(U, \mathbb{R}^n)$ the Jacobi matrix of $f$ at $x \in U$ is $df(x) = (\partial_x f_j(x))_{1 \leq i, j \leq n}$ and the Jacobi determinant of $f$ at $x \in U$ is

$$J_f(x) := \det df(x).$$  \hfill (16.4)

The set of regular values is

$$\text{RV}(f) := \{ y \in \mathbb{R}^n | \forall x \in f^{-1}(y) : J_f(x) \neq 0 \}. \hfill (16.5)$$

Its complement $\text{CV}(f) := \mathbb{R}^n \setminus \text{RV}(f)$ is called the set of critical values. We set $C^r(U, \mathbb{R}^n) := \{ f \in C^r(U, \mathbb{R}^n) | df \in C(U, \mathbb{R}^n), 0 \leq j \leq r \}$ and

$$D^r_y(U, \mathbb{R}^n) := \{ f \in C^r(U, \mathbb{R}^n) | y \notin f(\partial U) \}, \quad D_y(U, \mathbb{R}^n) := D^0_y(U, \mathbb{R}^n)$$

for $y \in \mathbb{R}^n$. We will use the topology induced by the sup norm for $C^r(U, \mathbb{R}^n)$ such that it becomes a Banach space (cf. Section 15.1).

Note that, since $U$ is bounded, $\partial U$ is compact and so is $f(\partial U)$ if $f \in C(U, \mathbb{R}^n)$. In particular,

$$\text{dist}(y, f(\partial U)) = \inf_{x \in \partial U} |y - f(x)|$$

is positive for $f \in D^1_y(U, \mathbb{R}^n)$ and thus $D^1_y(U, \mathbb{R}^n)$ is an open subset of $C^r(U, \mathbb{R}^n)$.

Now that these things are out of the way, we come to the formulation of the requirements for our degree.

A function deg which assigns each $f \in D_y(U, \mathbb{R}^n)$, $y \in \mathbb{R}^n$, a real number $\deg(f, U, y)$ will be called degree if it satisfies the following conditions.

\begin{enumerate}
  \item [(D1).] $\deg(f, U, y) = \deg(f - y, U, 0)$ \textit{(translation invariance)}.
  \item [(D2).] $\deg(1, U, y) = 1$ if $y \in U$ \textit{(normalization)}.
  \item [(D3).] If $U_1, U_2$ are open, disjoint subsets of $U$ such that $y \notin f(U \setminus (U_1 \cup U_2))$, then
     \begin{align*}
     \deg(f, U, y) &= \deg(f, U_1, y) + \deg(f, U_2, y) \quad \text{\textit{(additivity)}}.
     \end{align*}
  \item [(D4).] If $H(t) = (1 - t)f + tg \in D_y(U, \mathbb{R}^n)$, $t \in [0, 1]$, then
     \begin{align*}
     \deg(f, U, y) &= \deg(g, U, y) \quad \text{\textit{(homotopy invariance)}}.
     \end{align*}
\end{enumerate}

Before we draw some first conclusions from this definition, let us discuss the properties (D1)–(D4) first. (D1) is natural since $\deg(f, U, y)$ should have something to do with the solutions of $f(x) = y$, $x \in U$, which is the same
as the solutions of \( f(x) - y = 0, \ x \in U \). (D2) is a normalization since any multiple of deg would also satisfy the other requirements. (D3) is also quite natural since it requires deg to be additive with respect to components. In addition, it implies that sets where \( f \neq y \) do not contribute. (D4) is not that natural since it already rules out the case where deg is the cardinality of \( f^{-1}(U) \). On the other hand it will give us the ability to compute \( \text{deg}(f, U, y) \) in several cases.

**Theorem 16.1.** Suppose \( \text{deg} \) satisfies (D1)–(D4) and let \( f, g \in D_y(\overline{U}, \mathbb{R}^n) \), then the following statements hold.

(i). We have \( \text{deg}(f, \emptyset, y) = 0 \). Moreover, if \( U_i, \ 1 \leq i \leq N, \) are disjoint open subsets of \( U \) such that \( y \notin f(\overline{U} \setminus \bigcup_{i=1}^{N} U_i) \), then \( \text{deg}(f, U, y) = \sum_{i=1}^{N} \text{deg}(f, U_i, y) \).

(ii). If \( y \notin f(U) \), then \( \text{deg}(f, U, y) = 0 \) (but not the other way round). Equivalently, if \( \text{deg}(f, U, y) \neq 0 \), then \( y \in f(U) \).

(iii). If \( |f(x) - g(x)| < \text{dist}(y, f(\partial U)) \), \( x \in \partial U \), then \( \text{deg}(f, U, y) = \text{deg}(g, U, y) \). In particular, this is true if \( f(x) = g(x) \) for \( x \in \partial U \).

**Proof.** For the first part of (i) use (D3) with \( U_1 = U \) and \( U_2 = \emptyset \). For the second part use \( U_2 = \emptyset \) in (D3) if \( i = 1 \) and the rest follows from induction. For (ii) use \( i = 1 \) and \( U_1 = \emptyset \) in (ii). For (iii) note that \( H(t, x) = (1 - t)f(x) + tg(x) \) satisfies \( |H(t, x) - y| \geq \text{dist}(y, f(\partial U)) - |f(x) - g(x)| \) for \( x \) on the boundary. \( \square \)

Next we show that (D4) implies several at first sight much stronger looking facts.

**Theorem 16.2.** We have that \( \text{deg}(., U, y) \) and \( \text{deg}(f, ., .) \) are both continuous. In fact, we even have

(i). \( \text{deg}(., U, y) \) is constant on each component of \( D_y(\overline{U}, \mathbb{R}^n) \).

(ii). \( \text{deg}(f, ., .) \) is constant on each component of \( \mathbb{R}^n \setminus f(\partial U) \).

Moreover, if \( H : [0, 1] \times \overline{U} \to \mathbb{R}^n \) and \( y : [0, 1] \to \mathbb{R}^n \) are both continuous such that \( H(t) \in D_y(t)(U, \mathbb{R}^n), \ t \in [0, 1] \), then \( \text{deg}(H(0), U, y(0)) = \text{deg}(H(1), U, y(1)) \).

**Proof.** For (i) let \( C \) be a component of \( D_y(\overline{U}, \mathbb{R}^n) \) and let \( d_0 \in \text{deg}(C, U, y) \). It suffices to show that \( \text{deg}(., U, y) \) is locally constant. But if \( |g - f| < \text{dist}(y, f(\partial U)) \), then \( \text{deg}(f, U, y) = \text{deg}(g, U, y) \) by (D4) since \( |H(t) - y| \geq |f - y| - |g - f| > 0 \), \( H(t) = (1 - t)f + tg \). The proof of (ii) is similar. For the remaining part observe, that if \( H : [0, 1] \times \overline{U} \to \mathbb{R}^n, \ (t, x) \mapsto H(t, x), \) is continuous, then so is \( H : [0, 1] \to C(\overline{U}, \mathbb{R}^n), \ t \mapsto H(t), \) since \( \overline{U} \) is compact.

Hence, if in addition \( H(t) \in D_y(\overline{U}, \mathbb{R}^n) \), then \( \text{deg}(H(t), U, y) \) is independent
of $t$ and if $y = y(t)$ we can use \( \deg(H(0), U, y(0)) = \deg(H(t) - y(t), U, 0) = \deg(H(1), U, y(1)). \) \hfill \Box

Note that this result also shows why \( \deg(f, U, y) \) cannot be defined meaningful for $y \in f(\partial D)$. Indeed, approaching $y$ from within different components of $\mathbb{R}^n \setminus f(\partial U)$ will result in different limits in general!

In addition, note that if $Q$ is a closed subset of a locally pathwise connected space $X$, then the components of $X \setminus Q$ are open (in the topology of $X$) and pathwise connected (the set of points for which a path to a fixed point $x_0$ exists is both open and closed).

Now let us try to compute \( \deg \) using its properties. Let start with a simple case and suppose $f \in C^1(U, \mathbb{R}^n)$ and $y \notin CV(f) \cup f(\partial U)$. Without restriction we consider $y = 0$. In addition, we avoid the trivial case $f^{-1}(y) = \emptyset$. Since the points of $f^{-1}(0)$ inside $U$ are isolated (use $J_f(x) \neq 0$ and the inverse function theorem) they can only cluster at the boundary $\partial U$. But this is also impossible since $f$ would equal $y$ at the limit point on the boundary by continuity. Hence $f^{-1}(0) = \{ x^i \}_{i=1}^N$. Picking sufficiently small neighborhoods $U(x^i)$ around $x^i$ we consequently get

\[
\deg(f, U, 0) = \sum_{i=1}^N \deg(f, U(x^i), 0). \tag{16.8}
\]

It suffices to consider one of the zeros, say $x^1$. Moreover, we can even assume $x^1 = 0$ and $U(x^1) = B_\delta(0)$. Next we replace $f$ by its linear approximation around 0. By the definition of the derivative we have

\[
f(x) = df(0)x + |x|r(x), \quad r \in C(B_\delta(0), \mathbb{R}^n), \quad r(0) = 0. \tag{16.9}
\]

Now consider the homotopy $H(t, x) = df(0)x + (1 - t)|x|r(x)$. In order to conclude $\deg(f, B_\delta(0), 0) = \deg(df(0), B_\delta(0), 0)$ we need to show $0 \notin H(t, \partial B_\delta(0))$. Since $J_f(0) \neq 0$ we can find a constant $\lambda$ such that $|df(0)x| \geq \lambda|x|$ and since $r(0) = 0$ we can decrease $\delta$ such that $|r| < \lambda$. This implies $|H(t, x)| \geq ||df(0)x| - (1 - t)|x||r(x)|| \geq \lambda \delta - \delta|r| > 0$ for $x \in \partial B_\delta(0)$ as desired.

In summary we have

\[
\deg(f, U, 0) = \sum_{i=1}^N \deg(df(x^i), U(x^i), 0) \tag{16.10}
\]

and it remains to compute the degree of a nonsingular matrix. To this end we need the following lemma.

\textbf{Lemma 16.3.} Two nonsingular matrices $M_{1,2} \in \text{GL}(n)$ are homotopic in $\text{GL}(n)$ if and only if $\text{sign det } M_1 = \text{sign det } M_2$. 

Proof. We will show that any given nonsingular matrix $M$ is homotopic to 
$\text{diag}(\text{sign det } M, 1, \ldots, 1)$, where $\text{diag}(m_1, \ldots, m_n)$ denotes a diagonal matrix with diagonal entries $m_i$.

In fact, note that adding one row to another and multiplying a row by a positive constant can be realized by continuous deformations such that all intermediate matrices are nonsingular. Hence we can reduce $M$ to a diagonal matrix $\text{diag}(m_1, \ldots, m_n)$ with $(m_i)^2 = 1$. Next,

$$
\begin{pmatrix}
\pm \cos(\pi t) & \mp \sin(\pi t) \\
\sin(\pi t) & \cos(\pi t)
\end{pmatrix},
$$

shows that $\text{diag}(\pm 1, 1)$ and $\text{diag}(\mp 1, -1)$ are homotopic. Now we apply this result to all two by two subblocks as follows. For each $i$ starting from $n$ and going down to 2 transform the subblock $\text{diag}(m_{i-1}, m_i)$ into $\text{diag}(1, 1)$ respectively $\text{diag}(-1, 1)$. The result is the desired form for $M$.

To conclude the proof note that a continuous deformation within $\text{GL}(n)$ cannot change the sign of the determinant since otherwise the determinant would have to vanish somewhere in between (i.e., we would leave $\text{GL}(n)$). □

Using this lemma we can now show the main result of this section.

Theorem 16.4. Suppose $f \in D^1_y(U, \mathbb{R}^n)$ and $y \notin \text{CV}(f)$, then a degree satisfying (D1)–(D4) satisfies

$$
\deg(f, U, y) = \sum_{x \in f^{-1}(y)} \text{sign } J_f(x), \quad (16.11)
$$

where the sum is finite and we agree to set $\sum_{x \in \emptyset} = 0$.

Proof. By the previous lemma we obtain

$$
\deg(df(0), B_0(0), 0) = \deg(\text{diag}(\text{sign } J_f(0), 1, \ldots, 1), B_0(0), 0)
$$

since $\det M \neq 0$ is equivalent to $Mx \neq 0$ for $x \in \partial B_0(0)$. Hence it remains to show $\deg(df(0), B_0(0), 0) = \text{sign } J_f(0)$.

If $\text{sign } J_f(0) = 1$ this is true by (D2). Otherwise we can replace $df(0)$ by $M_\varepsilon = \text{diag}(-1, 1, \ldots, 1)$ and it remains to show $\deg(M_\varepsilon, B_1(0), 0) = -1$.

Abbreviate $U_1 = B_1(0) = \{x \in \mathbb{R}^n | x_i < 1, 1 \leq i \leq n\}$, $U_2 = \{x \in \mathbb{R}^n | 1 < x_1 < 3, |x_i| < 1, 2 \leq i \leq n\}$, $U = \{x \in \mathbb{R}^n | -1 < x_1 < 3, |x_i| < 1, 2 \leq i \leq n\}$, and $g(r) = 2 - |r - 1|, h(r) = 1 - r^2$. Now consider the two functions $f_1(x) = (1 - g(x_1)h(x_2) \cdots h(x_n), x_2, \ldots, x_n)$ and $f_2(x) = (1, x_2, \ldots, x_n)$. Clearly $f_1^{-1}(0) = \{x^1, x^2\}$ with $x^1 = 0$, $x^2 = (2, \ldots, 0)$ and $f_2^{-1}(0) = \emptyset$. Since $f_1(x) = f_2(x)$ for $x \in \partial U$ we infer $\deg(f_1, U, 0) = \deg(f_2, U, 0) = 0$. Moreover, we have $\deg(f_1, U_1, 0) = \deg(f_1, U_1, 0) + \deg(f_1, U_2, 0)$ and hence $\deg(M_\varepsilon, U_1, 0) = \deg(df_1(x^1)) = \deg(f_1, U_1, 0) = -\deg(f_1, U_2, 0) = -\deg(df_1(x^2)) = -\deg(\mathbb{I}, U_2, 0) = -1$ as claimed. □
Up to this point we have only shown that a degree (provided there is one at all) necessarily satisfies (16.11). Once we have shown that regular values are dense, it will follow that the degree is uniquely determined by (16.11) since the remaining values follow from point (iii) of Theorem 16.1. On the other hand, we don’t even know whether a degree exists. Hence we need to show that (16.11) can be extended to $f \in D_y(\overline{U}, \mathbb{R}^n)$ and that this extension satisfies our requirements (D1)–(D4).

16.3. Extension of the determinant formula

Our present objective is to show that the determinant formula (16.11) can be extended to all $f \in D_y(U, \mathbb{R}^n)$. This will be done in two steps, where we will show that $\deg(f, U, y)$ as defined in (16.11) is locally constant with respect to both $y$ (step one) and $f$ (step two).

Before we work out the technical details for these two steps, we prove that the set of regular values is dense as a warm up. This is a consequence of a special case of Sard’s theorem which says that $CV(f)$ has zero measure.

**Lemma 16.5 (Sard).** Suppose $f \in C^1(U, \mathbb{R}^n)$, then the Lebesgue measure of $CV(f)$ is zero.

**Proof.** Since the claim is easy for linear mappings our strategy is as follows. We divide $U$ into sufficiently small subsets. Then we replace $f$ by its linear approximation in each subset and estimate the error.

Let $CP(f) = \{x \in U | J_f(x) = 0\}$ be the set of critical points of $f$. We first pass to cubes which are easier to divide. Let $\{Q_i\}_{i \in \mathbb{N}}$ be a countable cover for $U$ consisting of open cubes such that $Q_i \subset U$. Then it suffices to prove that $f(CP(f) \cap Q_i)$ has zero measure since $CV(f) = f(CP(f)) = \bigcup_i f(CP(f) \cap Q_i)$ (the $Q_i$’s are a cover).

Let $Q$ be any of these cubes and denote by $\rho$ the length of its edges. Fix $\varepsilon > 0$ and divide $Q$ into $N^n$ cubes $Q_i$ of length $\rho/N$. These cubes don’t have to be open and hence we can assume that they cover $Q$. Since $df(x)$ is uniformly continuous on $Q$ we can find an $N$ (independent of $i$) such that

\[
|f(x) - f(\bar{x}) - df(\bar{x})(x - \bar{x})| \leq \int_0^1 |df(\bar{x} + t(x - \bar{x})) - df(\bar{x})||x - \bar{x}|dt \leq \frac{\varepsilon \rho}{N}
\]

for $\bar{x}, x \in Q_i$. Now pick a $Q_i$ which contains a critical point $\bar{x}_i \in CP(f)$. Without restriction we assume $\bar{x}_i = 0$, $f(\bar{x}_i) = 0$ and set $M = df(\bar{x}_i)$. By $\det M = 0$ there is an orthonormal basis $\{b_i\}_{1 \leq i \leq n}$ of $\mathbb{R}^n$ such that $b_n$ is
orthogonal to the image of $M$. In addition,
\[ Q_i \subseteq \left\{ \sum_{i=1}^{n} \lambda_i b_i \right\} \subseteq \left\{ \sum_{i=1}^{n} \lambda_i \leq \sqrt{n} \frac{\rho}{N} \right\} \]
and hence there is a constant (again independent of $i$) such that
\[ MQ_i \subseteq \left\{ \sum_{i=1}^{n} \lambda_i b_i \right\} \leq C \frac{\rho}{N} \]
(e.g., $C = \sqrt{n} \max_{x \in \mathbb{Q}} |df(x)|$). Next, by our estimate (16.12) we even have
\[ f(Q_i) \subseteq \left\{ \sum_{i=1}^{n} \lambda_i b_i \right\} \leq (C + \varepsilon) \frac{\rho}{N}, \quad |\lambda_n| \leq \varepsilon \frac{\rho}{N} \]
and hence the measure of $f(Q_i)$ is smaller than $\tilde{C}_\varepsilon$. Since there are at most $N^n$ such $Q_i$’s, we see that the measure of $f(CP(f) \cap Q)$ is smaller than $\tilde{C}_\varepsilon$.

Having this result out of the way we can come to step one and two from above.

**Step 1: Admitting critical values**

By (ii) of Theorem 16.2, $\deg(f, U, y)$ should be constant on each component of $\mathbb{R}^n \setminus f(\partial U)$. Unfortunately, if we connect $y$ and a nearby regular value $\tilde{y}$ by a path, then there might be some critical values in between. To overcome this problem we need a definition for $\deg$ which works for critical values as well. Let us try to look for an integral representation. Formally (16.11) can be written as $\deg(f, U, y) = \int_U \delta_y(f(x)) J_f(x) dx$, where $\delta_y(.)$ is the Dirac distribution at $y$. But since we don’t want to mess with distributions, we replace $\delta_y(.)$ by $\phi_\varepsilon(\cdot - y)$, where $\{\phi_\varepsilon\}_{\varepsilon > 0}$ is a family of functions such that $\phi_\varepsilon$ is supported on the ball $B_\varepsilon(0)$ of radius $\varepsilon$ around 0 and satisfies $\int_{\mathbb{R}^n} \phi_\varepsilon(x) dx = 1$.

**Lemma 16.6.** Let $f \in D_1^1(\overline{U}, \mathbb{R}^n)$, $y \notin CV(f)$. Then
\[ \deg(f, U, y) = \int_U \phi_\varepsilon(f(x) - y) J_f(x) dx \] (16.13)
for all positive $\varepsilon$ smaller than a certain $\varepsilon_0$ depending on $f$ and $y$. Moreover, $\text{supp}(\phi_\varepsilon(f(\cdot) - y)) \subset U$ for $\varepsilon < \text{dist}(y, f(\partial U))$.

**Proof.** If $f^{-1}(y) = \emptyset$, we can set $\varepsilon_0 = \text{dist}(y, f(\overline{U}))$, implying $\phi_\varepsilon(f(x) - y) = 0$ for $x \in \overline{U}$.
## 16.3. Extension of the determinant formula

If \( f^{-1}(y) = \{x^i\}_{1 \leq i \leq N} \), we can find an \( \varepsilon_0 > 0 \) such that \( f^{-1}(B_{\varepsilon_0}(y)) \) is a union of disjoint neighborhoods \( U(x^i) \) of \( x^i \) by the inverse function theorem. Moreover, after possibly decreasing \( \varepsilon_0 \) we can assume that \( f|_{U(x^i)} \) is a bijection and that \( J_f(x) \) is nonzero on \( U(x^i) \). Again \( \phi_\varepsilon(f(x) - y) = 0 \) for \( x \in \overline{U} \setminus \bigcup_{i=1}^N U(x^i) \) and hence

\[
\int_U \phi_\varepsilon(f(x) - y)J_f(x)dx = \sum_{i=1}^N \int_{U(x^i)} \phi_\varepsilon(f(x) - y)J_f(x)dx
\]

\[
= \sum_{i=1}^N \text{sign}(J_f(x)) \int_{B_{\varepsilon_0}(y)} \phi_\varepsilon(\tilde{x})d\tilde{x} = \deg(f, U, y),
\]

where we have used the change of variables \( \tilde{x} = f(x) - y \) in the second step. \( \square \)

Our new integral representation makes sense even for critical values. But since \( \varepsilon \) depends on \( y \), continuity with respect to \( y \) is not clear. This will be shown next at the expense of requiring \( f \in C^2 \) rather than \( f \in C^1 \).

The key idea is to rewrite \( \deg(f, U, y^2) - \deg(f, U, y^1) \) as an integral over a divergence (here we will need \( f \in C^2 \)) supported in \( U \) and then apply Stokes theorem. For this purpose the following result will be used.

### Lemma 16.7

Suppose \( f \in C^2(U, \mathbb{R}^n) \) and \( u \in C^1(\mathbb{R}^n, \mathbb{R}^n) \), then

\[
(\text{div } u)(f)J_f = \text{div } D_f(u),
\]

where \( D_f(u)_{j,j} \) is the determinant of the matrix obtained from \( df \) by replacing the \( j \)-th column by \( u(f) \).

**Proof.** We compute

\[
\text{div } D_f(u) = \sum_{j=1}^n \partial_{x_j} D_f(u)_{j,k} = \sum_{j,k=1}^n D_f(u)_{j,k},
\]

where \( D_f(u)_{j,k} \) is the determinant of the matrix obtained from the matrix associated with \( D_f(u)_{j,j} \) by applying \( \partial_{x_j} \) to the \( k \)-th column. Since \( \partial_{x_j} \partial_{x_k} f = \partial_{x_k} \partial_{x_j} f \) we infer \( D_f(u)_{j,k} = -D_f(u)_{k,j}, j \neq k \), by exchanging the \( k \)-th and the \( j \)-th column. Hence

\[
\text{div } D_f(u) = \sum_{i=1}^n D_f(u)_{i,i}.
\]

Now let \( J_f^{(i,j)}(x) \) denote the \( (i, j) \) minor of \( df(x) \) and recall \( \sum_{i=1}^n J_f^{(i,j)} \partial_{x_i} f_k = \delta_{j,k} J_f \). Using this to expand the determinant \( D_f(u)_{i,i} \) along the \( i \)-th column
The Brouwer mapping degree shows
\[
\text{div } D_f(u) = \sum_{i,j=1}^{n} J_f^{(i,j)} \partial_{x_i} u_j(f) = \sum_{i,j=1}^{n} J_f^{(i,j)} \sum_{k=1}^{n} (\partial_{x_k} u_j)(f) \partial_{x_k} f_k
\]
\[
= \sum_{j,k=1}^{n} (\partial_{x_k} u_j)(f) \sum_{i=1}^{n} J_f^{(i,j)} \partial_{x_j} f_k = \sum_{j=1}^{n} (\partial_{x_j} u_j)(f) J_f
\]
as required. $\square$

Now we can prove Lemma 16.8. Suppose $f \in C^2(U, \mathbb{R}^n)$. Then $\deg(f, U, .)$ is constant in each ball contained in $\mathbb{R}^n \setminus f(\partial U)$, whenever defined.

**Proof.** Fix $\tilde{y} \in \mathbb{R}^n \setminus f(\partial U)$ and consider the largest ball $B_\rho(\tilde{y})$, $\rho = \text{dist}(\tilde{y}, f(\partial U))$ around $\tilde{y}$ contained in $\mathbb{R}^n \setminus f(\partial U)$. Pick $y^1 \in B_\rho(\tilde{y}) \cap \text{RV}(f)$ and consider
\[
\deg(f, U, y^2) - \deg(f, U, y^1) = \int_{U} (\phi_\varepsilon(f(x) - y^2) - \phi_\varepsilon(f(x) - y^1)) J_f(x) dx
\]
for suitable $\phi_\varepsilon \in C^2(\mathbb{R}^n, \mathbb{R})$ and suitable $\varepsilon > 0$. Now observe
\[
(div u)(y) = \int_{0}^{1} z_j \partial_{y_j} \phi(y + tz) dt
\]
\[
= \int_{0}^{1} \left( \frac{d}{dt} \phi(y + tz) \right) dt = \phi_\varepsilon(y - y^2) - \phi_\varepsilon(y - y^1),
\]
where
\[
u(y) = z \int_{0}^{1} \phi(y + tz) dt, \quad \phi(y) = \phi_\varepsilon(y - y^1), \quad z = y^1 - y^2,
\]
and apply the previous lemma to rewrite the integral as $\int_{U} \text{div } D_f(u) dx$. Since the integrand vanishes in a neighborhood of $\partial U$ we can extend it to all of $\mathbb{R}^n$ by setting it zero outside $U$ and choose a cube $Q \supset U$. Then elementary coordinatewise integration (or Stokes theorem) gives $\int_{U} \text{div } D_f(u) dx = \int_{Q} \text{div } D_f(u) dx = \int_{\partial Q} D_f(u) dF = 0$ since $u$ is supported inside $B_\rho(\tilde{y})$ provided $\varepsilon$ is small enough (e.g., $\varepsilon < \rho - \max\{|y^i - \tilde{y}^i|\}_{i=1,2}$). $\square$

As a consequence we can define
\[
\deg(f, U, y) = \deg(f, U, \tilde{y}), \quad y \notin f(\partial U), \quad f \in C^2(\overline{U}, \mathbb{R}^n), \quad (16.15)
\]
where $\tilde{y}$ is a regular value of $f$ with $|\tilde{y} - y| < \text{dist}(y, f(\partial U))$. 
Remark 16.9. Let me remark a different approach due to Kronecker. For $U$ with sufficiently smooth boundary we have

$$\deg(f,U,0) = \frac{1}{|S^{n-1}|} \int_{\partial U} D_f(x) dF = \frac{1}{|S^n|} \int_{\partial U} \frac{1}{|f^n|} D_f(x) dF, \quad \tilde{f} = \frac{f}{|f|},$$

(16.16)

for $f \in C^2_y(U, \mathbb{R}^n)$. Explicitly we have

$$\deg(f,U,0) = \frac{1}{|S^{n-1}|} \int_{\partial U} \sum_{j=1}^n (-1)^{j-1} \frac{f_j}{|f^n|} df_1 \wedge \cdots \wedge df_{j-1} \wedge df_{j+1} \wedge \cdots \wedge df_n.$$  

(16.17)

Since $\tilde{f} : \partial U \to S^{n-1}$ the integrand can also be written as the pull back $\tilde{f}^* dS$ of the canonical surface element $dS$ on $S^{n-1}$.

This coincides with the boundary value approach for complex functions (note that holomorphic functions are orientation preserving).

Step 2: Admitting continuous functions

Our final step is to remove the condition $f \in C^2$. As before we want the degree to be constant in each ball contained in $D_y(U, \mathbb{R}^n)$. For example, fix $f \in D_y(U, \mathbb{R}^n)$ and set $\rho = \text{dist}(y, f(\partial U)) > 0$. Choose $f^i \in C^2(U, \mathbb{R}^n)$ such that $|f^i - f| < \rho$, implying $f^i \in D_y(U, \mathbb{R}^n)$. Then $H(t, x) = (1-t)f^1(x) + tf^2(x) \in D_y(U, \mathbb{R}^n) \cap C^2(U, \mathbb{R}^n)$, $t \in [0,1]$, and $|H(t) - f| < \rho$. If we can show that $\deg(H(t), U, y)$ is locally constant with respect to $t$, then it is continuous with respect to $t$ and hence constant (since $[0,1]$ is connected).

Consequently we can define

$$\deg(f, U, y) = \deg(\tilde{f}, U, y), \quad f \in D_y(U, \mathbb{R}^n),$$

(16.18)

where $\tilde{f} \in C^2(U, \mathbb{R}^n)$ with $|\tilde{f} - f| < \text{dist}(y, f(\partial U))$.

It remains to show that $t \mapsto \deg(H(t), U, y)$ is locally constant.

Lemma 16.10. Suppose $f \in C^2_y(U, \mathbb{R}^n)$. Then for each $\tilde{f} \in C^2(U, \mathbb{R}^n)$ there is an $\varepsilon > 0$ such that $\deg(f + t \tilde{f}, U, y) = \deg(f, U, y)$ for all $t \in (-\varepsilon, \varepsilon)$.

Proof. If $f^{-1}(y) = \emptyset$ the same is true for $f + t g$ if $|t| < \text{dist}(y, f(U))/|g|$. Hence we can exclude this case. For the remaining case we use our usual strategy of considering $y \in \text{RV}(f)$ first and then approximating general $y$ by regular ones.

Suppose $y \in \text{RV}(f)$ and let $f^{-1}(y) = \{x^i\}_{j=1}^N$. By the implicit function theorem we can find disjoint neighborhoods $U(x^i)$ such that there exists a unique solution $x^i(t) \in U(x^i)$ of $(f + t g)(x) = y$ for $|t| < \varepsilon_1$. By reducing $U(x^i)$ if necessary, we can even assume that the sign of $J_{f + t g}$ is constant on
$U(x^i)$. Finally, let $\varepsilon_2 = \text{dist}(y, f(U \cup \bigcup_{i=1}^{N} U(x^i)))/|g|$. Then $|f + tg - y| > 0$ for $|t| < \varepsilon_2$ and $\varepsilon = \min(\varepsilon_1, \varepsilon_2)$ is the quantity we are looking for.

It remains to consider the case $y \in CV(f)$. Pick a regular value $\tilde{y} \in B_{\rho/3}(y)$, where $\rho = \text{dist}(y, f(\partial U))$, implying $\deg(f, U, y) = \deg(f, U, \tilde{y})$. Then we can find an $\tilde{\varepsilon} > 0$ such that $\deg(f, U, \tilde{y}) = \deg(f + tg, U, \tilde{y})$ for $|t| < \tilde{\varepsilon}$. Setting $\varepsilon = \min(\tilde{\varepsilon}, \rho/(3|g|))$ we infer $\tilde{y} - (f + tg)(x) \geq \rho/3$ for $x \in \partial U$, that is $|\tilde{y} - y| < \text{dist}(\tilde{y}, (f + tg)(\partial U))$, and thus $\deg(f + tg, U, \tilde{y}) = \deg(f + tg, U, y)$. Putting it all together implies $\deg(f, U, y) = \deg(f + tg, U, y)$ for $|t| < \varepsilon$ as required. \hfill \square

Now we can finally prove our main theorem.

**Theorem 16.11.** There is a unique degree $\deg$ satisfying (D1)-(D4). Moreover, $\deg(., U, y) : D_y(\overline{U}, \mathbb{R}^n) \to \mathbb{Z}$ is constant on each component and given $f \in D_y(\overline{U}, \mathbb{R}^n)$ we have

$$\deg(f, U, y) = \sum_{x \in \hat{f}^{-1}(y)} \text{sign } J_{\hat{f}}(x) \quad (16.19)$$

where $\hat{f} \in D_y^2(\overline{U}, \mathbb{R}^n)$ is in the same component of $D_y(\overline{U}, \mathbb{R}^n)$, say $|f - \hat{f}| < \text{dist}(y, f(\partial U))$, such that $y \in RV(\hat{f})$.

**Proof.** Our previous considerations show that $\deg$ is well-defined and locally constant with respect to the first argument by construction. Hence $\deg(., U, y) : D_y(\overline{U}, \mathbb{R}^n) \to \mathbb{Z}$ is continuous and thus necessarily constant on components since $\mathbb{Z}$ is discrete. (D2) is clear and (D1) is satisfied since it holds for $\hat{f}$ by construction. Similarly, taking $U_{1,2}$ as in (D3) we can require $|f - \hat{f}| < \text{dist}(y, f(\partial U \setminus (U_1 \cup U_2)))$. Then (D3) is satisfied since it also holds for $\hat{f}$ by construction. Finally, (D4) is a consequence of continuity. \hfill \square

To conclude this section, let us give a few simple examples illustrating the use of the Brouwer degree.

**Example.** First, let’s investigate the zeros of

$$f(x_1, x_2) := (x_1 - 2x_2 + \cos(x_1 + x_2), x_2 + 2x_1 + \sin(x_1 + x_2)).$$

Denote the linear part by

$$g(x_1, x_2) := (x_1 - 2x_2, x_2 + 2x_1).$$

Then we have $|g(x)| = \sqrt{5}|x|$ and $|f(x) - g(x)| = 1$ and hence $h(t) = (1-t)g + t f = g + t(f-g)$ satisfies $|h(t)| \geq |g| - t|f-g| > 0$ for $|x| > 1/\sqrt{5}$ implying

$$\deg(f, B_r(0), 0) = \deg(g, B_r(0), 0) = 1, \quad r > 1/\sqrt{5}.$$ 

Moreover, since $J_f(x) = 5 + 3\cos(x_1 + x_2) + \sin(x_1 + x_2) > 1$ the determinant formula (16.11) for the degree implies that $f(x) = 0$ has a unique solution
in \( \mathbb{R}^2 \). This solution even has to lie on the circle \( |x| = 1/\sqrt{5} \) since \( f(x) = 0 \)
implies \( 1 = |f(x) - g(x)| = |g(x)| = \sqrt{5}|x| \).

Next let us prove the following result which implies the hairy ball (or hedgehog) theorem.

**Theorem 16.12.** Suppose \( U \) contains the origin and let \( f : \partial U \to \mathbb{R}^n \setminus \{0\} \)
be continuous. If \( n \) is odd, then there exists a \( x \in \partial U \) and a \( \lambda \neq 0 \) such that \( f(x) = \lambda x \).

**Proof.** By Theorem 16.15 we can assume \( f \in C(U, \mathbb{R}^n) \) and since \( n \) is odd we have \( \deg(-I, U, 0) = -1 \). Now if \( \deg(f, U, 0) \neq -1 \), then \( H(t, x) = (1-t)f(x) - tx \) must have a zero \( (t_0, x_0) \in (0,1) \times \partial U \) and hence \( f(x_0) = \frac{t_0}{1-t_0}x_0 \). Otherwise, if \( \deg(f, U, 0) = -1 \) we can apply the same argument to
\( H(t, x) = (1-t)f(x) + tx \). \( \square \)

In particular, this result implies that a continuous tangent vector field on the unit sphere \( f : S^{n-1} \to \mathbb{R}^n \) (with \( f(x)x = 0 \) for all \( x \in S^n \)) must vanish somewhere if \( n \) is odd. Or, for \( n = 3 \), you cannot smoothly comb a hedgehog without leaving a bald spot or making a parting. It is however possible to comb the hair smoothly on a torus and that is why the magnetic containers in nuclear fusion are toroidal.

Another simple consequence is the fact that a vector field on \( \mathbb{R}^n \), which points outwards (or inwards) on a sphere, must vanish somewhere inside the sphere.

**Theorem 16.13.** Suppose \( f : \overline{B_R(0)} \to \mathbb{R}^n \) is continuous and satisfies
\[
f(x)x > 0, \quad |x| = R.
\]
Then \( f(x) \) vanishes somewhere inside \( B_R(0) \).

**Proof.** If \( f \) does not vanish, then \( H(t, x) = (1-t)x + tf(x) \) must vanish at some point \( (t_0, x_0) \in (0,1) \times \partial B_R(0) \) and thus
\[
0 = H(t_0, x_0)x_0 = (1-t_0)R^2 + t_0 f(x_0)x_0.
\]
But the last part is positive by assumption, a contradiction. \( \square \)

### 16.4. The Brouwer fixed-point theorem

Now we can show that the famous Brouwer fixed-point theorem is a simple consequence of the properties of our degree.

**Theorem 16.14 (Brouwer fixed point).** Let \( K \) be a topological space homeomorphic to a compact, convex subset of \( \mathbb{R}^n \) and let \( f \in C(K, K) \), then \( f \) has at least one fixed point.
Proof. Clearly we can assume $K \subset \mathbb{R}^n$ since homeomorphisms preserve fixed points. Now let $K = B_r(0)$. If there is a fixed-point on the boundary $\partial B_r(0)$ we are done. Otherwise $H(t, x) = x - tf(x)$ satisfies $0 \notin H(t, \partial B_r(0))$ since $|H(t, x)| \geq |x - tf(x)| \geq (1 - t)r > 0$, $0 \leq t < 1$. And the claim follows from $\deg(x - f(x), B_r(0), 0) = \deg(x, B_r(0), 0) = 1$.

Now let $K$ be convex. Then $K \subseteq B_r(0)$ and, by Theorem 16.15 below, we can find a continuous retraction $R : \mathbb{R}^n \rightarrow K$ (i.e., $R(x) = x$ for $x \in K$) and consider $\tilde{f} = f \circ R \in \mathcal{C}(\overline{B_r(0)}, \overline{B_r(0)})$. By our previous analysis, there is a fixed point $x = \tilde{f}(x) \in \text{hull}(f(K)) \subseteq K$.

Note that any compact, convex subset of a finite dimensional Banach space (complex or real) is isomorphic to a compact, convex subset of $\mathbb{R}^n$ since linear transformations preserve both properties. In addition, observe that all assumptions are needed. For example, the map $f : \mathbb{R} \rightarrow \mathbb{R}, x \mapsto x + 1$, has no fixed point ($\mathbb{R}$ is homeomorphic to a bounded set but not to a compact one). The same is true for the map $f : \partial B_1(0) \rightarrow \partial B_1(0), x \mapsto -x$ ($\partial B_1(0) \subset \mathbb{R}^n$ is simply connected for $n \geq 3$ but not homeomorphic to a convex set).

It remains to prove the result from topology needed in the proof of the Brouwer fixed-point theorem. It is a variant of the Tietze extension theorem.

Theorem 16.15. Let $X$ be a metric space, $Y$ a Banach space and let $K$ be a closed subset of $X$. Then $F \in \mathcal{C}(K, Y)$ has a continuous extension $F \in \mathcal{C}(X, Y)$ such that $F(X) \subseteq \text{hull}(F(K))$.

Proof. Consider the open cover $\{B_{\rho(x)}(x)\}_{x \in X \setminus K}$ for $X \setminus K$, where $\rho(x) = \text{dist}(x, K)/2$. Choose a (locally finite) partition of unity $\{\phi_\lambda\}_{\lambda \in \Lambda}$ subordinate to this cover and set

$$F(x) := \sum_{\lambda \in \Lambda} \phi_\lambda(x)F(x_\lambda) \text{ for } x \in X \setminus K,$$

where $x_\lambda \in K$ satisfies $\text{dist}(x_\lambda, \text{supp } \phi_\lambda) \leq 2\text{dist}(K, \text{supp } \phi_\lambda)$. By construction, $F$ is continuous except for possibly at the boundary of $K$. Fix $x_0 \in \partial K$, $\varepsilon > 0$ and choose $\delta > 0$ such that $|F(x) - F(x_0)| \leq \varepsilon$ for all $x \in K$ with $|x - x_0| < 4\delta$. We will show that $|F(x) - F(x_0)| \leq \varepsilon$ for all $x \in X$ with $|x - x_0| < \delta$. Suppose $x \notin K$, then $|F(x) - F(x_0)| \leq \sum_{\lambda \in \Lambda} |\phi_\lambda(x)||F(x_\lambda) - F(x_0)|$. By our construction, $x_\lambda$ should be close to $x$ for all $\lambda$ with $x \in \text{supp } \phi_\lambda$ since $x$ is close to $K$. In fact, if $x \in \text{supp } \phi_\lambda$ we have

$$|x - x_\lambda| \leq \text{dist}(x_\lambda, \text{supp } \phi_\lambda) + d(\text{supp } \phi_\lambda) \leq 2\text{dist}(K, \text{supp } \phi_\lambda) + d(\text{supp } \phi_\lambda),$$

where $d(\text{supp } \phi_\lambda) = \sup_{x, y \in \text{supp } \phi_\lambda} |x - y|$. Since our partition of unity is subordinate to the cover $\{B_{\rho(x)}(x)\}_{x \in X \setminus K}$ we can find a $\tilde{x} \in X \setminus K$ such
that \( \text{supp} \phi_\lambda \subset B_{\rho(x)}(\tilde{x}) \) and hence \( d(\text{supp} \phi_\lambda) \leq \rho(\tilde{x}) \leq \text{dist}(K, B_{\rho(x)}(\tilde{x})) \leq \text{dist}(K, \text{supp} \phi_\lambda) \). Putting it all together implies that we have \( |x - x_\lambda| \leq 3 \text{dist}(K, \text{supp} \phi_\lambda) \leq 3|x_0 - x| \) whenever \( x \in \text{supp} \phi_\lambda \) and thus

\[
|x_0 - x_\lambda| \leq |x_0 - x| + |x - x_\lambda| \leq 4|x_0 - x| \leq 4 \delta
\]

as expected. By our choice of \( \delta \) we have \( |F(x_\lambda) - F(x_0)| \leq \varepsilon \) for all \( \lambda \) with \( \phi_\lambda(x) \neq 0 \). Hence \( |F(x) - F(x_0)| \leq \varepsilon \) whenever \( |x - x_0| \leq \delta \) and we are done.

As an easy example of how to use the Brouwer fixed point theorem we show the famous **Perron–Frobenius theorem**.

**Theorem 16.16** (Perron–Frobenius). Let \( A \) be an \( n \times n \) matrix all whose entries are nonnegative and there is an \( m \) such the entries of \( A^m \) are all positive. Then \( A \) has a positive eigenvalue and the corresponding eigenvector can be chosen to have strictly positive components.

**Proof.** We equip \( \mathbb{R}^n \) with the norm \( |x|_1 = \sum_{j=1}^{n} |x_j| \) and set \( \Delta := \{ x \in \mathbb{R}^n | x_j \geq 0, |x|_1 = 1 \} \). For \( x \in \Delta \) we have \( Ax \neq 0 \) (since \( A^m x \neq 0 \)) and hence

\[
f : \Delta \rightarrow \Delta, \quad x \mapsto \frac{Ax}{|Ax|_1}
\]

has a fixed point \( x_0 \) by the Brouwer fixed point theorem. Then \( Ax_0 = |Ax_0|_1 x_0 \) and \( x_0 \) has strictly positive components since \( A^m x_0 = |Ax_0|^m x_0 \) has.

Let me remark that the Brouwer fixed point theorem is equivalent to the fact that there is no continuous retraction \( R : \overline{B}_1(0) \rightarrow \partial B_1(0) \) (with \( R(x) = x \) for \( x \in \partial B_1(0) \)) from the unit ball to the unit sphere in \( \mathbb{R}^n \).

In fact, if \( R \) would be such a retraction, \( -R \) would have a fixed point \( x_0 \in \partial B_1(0) \) by Brouwer’s theorem. But then \( x_0 = -R(x_0) = -x_0 \) which is impossible. Conversely, if a continuous function \( f : \overline{B}_1(0) \rightarrow \overline{B}_1(0) \) has no fixed point we can define a retraction \( R(x) = f(x) + t(x)(x - f(x)) \), where \( t(x) \geq 0 \) is chosen such that \( |R(x)|^2 = 1 \) (i.e., \( R(x) \) lies on the intersection of the line spanned by \( x, f(x) \) with the unit sphere).

Using this equivalence the Brouwer fixed point theorem can also be derived easily by showing that the homology groups of the unit ball \( \overline{B}_1(0) \) and its boundary (the unit sphere) differ (see, e.g., [29] for details).

Finally, we also derive the following important consequence known as **invariance of domain theorem**.

**Theorem 16.17** (Brouwer). Let \( U \subseteq \mathbb{R}^n \) be open and let \( f : U \rightarrow \mathbb{R}^n \) be continuous and injective. Then \( f(U) \) is also open.
**Proof.** By scaling and translation it suffices to show that if \( f : B_1(0) \to \mathbb{R}^n \) is injective, then \( f(0) \) is an inner point for \( f(B_1(0)) \). Abbreviate \( C = B_1(0) \).

Since \( C \) is compact so is \( f(C) \) and thus \( f : C \to f(C) \) is a homeomorphism. In particular, \( f^{-1} : f(C) \to C \) is continuous and can be extended to a continuous left inverse \( g : \mathbb{R}^n \to \mathbb{R}^n \) (i.e., \( g(f(x)) = x \) for all \( x \in C \).

Note that \( g \) has a zero in \( f(C) \), namely \( f(0) \), which is stable in the sense that any perturbation \( \tilde{g} : f(C) \to \mathbb{R}^n \) satisfying \( |\tilde{g}(y) - g(y)| \leq 1 \) for all \( y \in f(C) \) also has a zero. To see this apply the Brower fixed point theorem to the function \( F(x) = x - \tilde{g}(f(x)) = g(f(x)) - \tilde{g}(f(x)) \) which maps \( C \) to \( C \) by assumption.

Our strategy is to find a contradiction to this fact. Since \( g(f(0)) = 0 \) vanishes there is some \( \varepsilon \) such that \( |g(y)| \leq \frac{1}{3} \) for \( y \in B_2(f(0)) \). If \( f(0) \) were not in the interior of \( f(C) \) we can find some \( z \in B_2(f(0)) \) which is not in \( f(C) \). After a translation we can assume \( z = 0 \) without loss of generality, that is, \( 0 \notin f(C) \) and \( |f(0)| < \varepsilon \). In particular, we also have \( |g(y)| \leq \frac{1}{3} \) for \( y \in B_{\varepsilon}(0) \).

Next consider the map \( \varphi : f(C) \to \mathbb{R}^n \) given by

\[
\varphi(y) := \begin{cases} 
  y, & |y| > \varepsilon, \\
  \varepsilon \frac{y}{|y|}, & |y| \leq \varepsilon.
\end{cases}
\]

It is continuous away from 0 and its range is contained in \( \Sigma_1 \cup \Sigma_2 \), where \( \Sigma_1 = \{ y \in f(C) | |y| \geq \varepsilon \} \) and \( \Sigma_2 = \{ y \in \mathbb{R}^n | |y| = \varepsilon \} \).

Since \( f \) is injective, \( g \) does not vanish on \( \Sigma_1 \) and since \( \Sigma_1 \) is compact there is a \( \delta \) such that \( |g(y)| \geq \delta \) for \( y \in \Sigma_1 \). We may even assume \( \delta < \frac{1}{3} \).

Next, by the Stone–Weierstraß theorem we can find a polynomial \( P \) such that

\[ |P(y) - g(y)| < \delta \]

for all \( y \in \Sigma \). In particular, \( P \) does not vanish on \( \Sigma_1 \). However, it could vanish on \( \Sigma_2 \). But since \( \Sigma_2 \) has measure zero, so has \( P(\Sigma_2) \) and we can find an arbitrarily small value which is not in \( P(\Sigma_2) \). Shifting \( P \) by such a value we can assume that \( P \) does not vanish on \( \Sigma_1 \cup \Sigma_2 \).

Now chose \( \tilde{g} : f(C) \to \mathbb{R}^n \) according to

\[ \tilde{g}(y) = P(\varphi(y)). \]

Then \( \tilde{g} \) is a continuous function which does not vanish. Moreover, if \( |y| \geq \varepsilon \) we have

\[ |g(y) - \tilde{g}(y)| = |g(y) - P(y)| < \delta < \frac{1}{3}. \]

And if \( |y| < \varepsilon \) we have \( |g(y)| \leq \frac{1}{3} \) and \( |g(\varphi(y))| \leq \frac{1}{3} \) implying

\[ |g(y) - \tilde{g}(y)| \leq |g(y) - g(\varphi(y))| + |g(\varphi(y)) - P(\varphi(y))| \leq \frac{2}{3} + \delta \leq 1. \]
Thus $\tilde{g}$ contradicts our above observation. \qed

An easy consequence worth while noting is the topological invariance of dimension:

**Corollary 16.18.** If $m < n$ and $U$ is a nonempty open subset of $\mathbb{R}^n$, then there is no continuous injective mapping from $U$ to $\mathbb{R}^m$.

**Proof.** Suppose there were such a map and extend it to a map from $U$ to $\mathbb{R}^n$ by setting the additional coordinates equal to zero. The resulting map contradicts the invariance of domain theorem. \qed

In particular, $\mathbb{R}^m$ and $\mathbb{R}^n$ are not homeomorphic for $m \neq n$.

### 16.5. Kakutani’s fixed-point theorem and applications to game theory

In this section we want to apply Brouwer’s fixed-point theorem to show the existence of Nash equilibria for $n$-person games. As a preparation we extend Brouwer’s fixed-point theorem to set valued functions. This generalization will be more suitable for our purpose.

Denote by $\text{CS}(K)$ the set of all nonempty convex subsets of $K$.

**Theorem 16.19** (Kakutani). Suppose $K$ is a compact convex subset of $\mathbb{R}^n$ and $f : K \to \text{CS}(K)$. If the set

$$\Gamma = \{(x,y)|y \in f(x)\} \subseteq K^2 \quad (16.21)$$

is closed, then there is a point $x \in K$ such that $x \in f(x)$.

**Proof.** Our strategy is to apply Brouwer’s theorem, hence we need a function related to $f$. For this purpose it is convenient to assume that $K$ is a simplex

$$K = (v_1, \ldots, v_m), \quad m \leq n,$$

where $v_i$ are the vertices. If we pick $y_i \in f(v_i)$ we could set

$$f^1(x) = \sum_{i=1}^{m} \lambda_i y_i,$$

where $\lambda_i$ are the barycentric coordinates of $x$ (i.e., $\lambda_i \geq 0$, $\sum_{i=1}^{m} \lambda_i = 1$ and $x = \sum_{i=1}^{m} \lambda_i v_i$). By construction, $f^1 \in C(K,K)$ and there is a fixed point $x^1$. But unless $x^1$ is one of the vertices, this doesn’t help us too much. So let’s choose a better function as follows. Consider the $k$-th barycentric subdivision and for each vertex $v_i$ in this subdivision pick an element $y_i \in f(v_i)$. The resulting map

$$f^{k+1}(x) = \sum_{i=1}^{m} \lambda_i y^{k+1}_i,$$

where $y^{k+1}_i$ are the barycentric coordinates of $x$ (i.e., $y^{k+1}_i \geq 0$, $\sum_{i=1}^{m} y^{k+1}_i = 1$ and $x = \sum_{i=1}^{m} y^{k+1}_i v_i$). By construction, $f^{k+1} \in C(K,K)$ and there is a fixed point $x^{k+1}$. But unless $x^{k+1}$ is one of the vertices, this doesn’t help us too much. So let’s choose a better function as follows. Consider the $k$-th barycentric subdivision and for each vertex $v_i$ in this subdivision pick an element $y_i \in f(v_i)$.
Now define $f^k(v_i) = y_i$ and extend $f^k$ to the interior of each subsimplex as before. Hence $f^k \in C(K, K)$ and there is a fixed point

$$x^k = \sum_{i=1}^{m} \lambda^k_i v^k_i = \sum_{i=1}^{m} \lambda^k_i y^k_i, \quad y^k_i = f^k(v^k_i),$$  \hspace{1cm} (16.22)

in the subsimplex $\langle v^k_1, \ldots, v^k_m \rangle$. Since $(x^k, \lambda^0_1, \ldots, \lambda^k_m, y^k_1, \ldots, y^k_m) \in K \times [0,1]^m \times K^m$ we can assume that this sequence converges to some limit $(x^0, \lambda^0_1, \ldots, \lambda^0_m, y^0_1, \ldots, y^0_m)$ after passing to a subsequence. Since the subsimplices shrink to a point, this implies $v^k_i \to x^0$ and hence $y^k_i \in f(x^0)$ since $(v^k_i, y^k_i) \in \Gamma \to (v^0_i, y^0_i) \in \Gamma$ by the closedness assumption. Now (16.22) tells us

$$x^0 = \sum_{i=1}^{m} \lambda^0_i y^0_i \in f(x^0)$$

since $f(x^0)$ is convex and the claim holds if $K$ is a simplex.

If $K$ is not a simplex, we can pick a simplex $S$ containing $K$ and proceed as in the proof of the Brouwer theorem. \hfill \Box

If $f(x)$ contains precisely one point for all $x$, then Kakutani’s theorem reduces to the Brouwer’s theorem.

Now we want to see how this applies to game theory.

An $n$-person game consists of $n$ players who have $m_i$ possible actions to choose from. The set of all possible actions for the $i$-th player will be denoted by $\Phi_i = \{1, \ldots, m_i\}$. An element $\varphi_i \in \Phi_i$ is also called a pure strategy for reasons to become clear in a moment. Once all players have chosen their move $\varphi_i$, the payoff for each player is given by the payoff function

$$R_i(\varphi), \quad \varphi = (\varphi_1, \ldots, \varphi_n) \in \Phi = \prod_{i=1}^{n} \Phi_i$$  \hspace{1cm} (16.23)

of the $i$-th player. We will consider the case where the game is repeated a large number of times and where in each step the players choose their action according to a fixed strategy. Here a strategy $s_i$ for the $i$-th player is a probability distribution on $\Phi_i$, that is, $s_i = (s^1_i, \ldots, s_{m_i}^i)$ such that $s^k_i \geq 0$ and $\sum_{k=1}^{m_i} s^k_i = 1$. The set of all possible strategies for the $i$-th player is denoted by $S_i$. The number $s^k_i$ is the probability for the $k$-th pure strategy to be chosen. Consequently, if $s = (s_1, \ldots, s_n) \in S = \prod_{i=1}^{n} S_i$ is a collection of strategies, then the probability that a given collection of pure strategies gets chosen is

$$s(\varphi) = \prod_{i=1}^{n} s_i(\varphi), \quad s_i(\varphi) = s^k_i, \quad \varphi = (k_1, \ldots, k_n) \in \Phi$$  \hspace{1cm} (16.24)
(assuming all players make their choice independently) and the expected payoff for player \(i\) is

\[
R_i(s) = \sum_{\varphi \in \Phi} s(\varphi) R_i(\varphi).
\] (16.25)

By construction, \(R_i(s)\) is continuous.

The question is of course, what is an optimal strategy for a player? If the other strategies are known, a best reply of player \(i\) against \(s\) would be a strategy \(\tilde{s}_i\) satisfying

\[
R_i(s \setminus \tilde{s}_i) = \max_{\tilde{s}_i \in \tilde{S}_i} R_i(s \setminus \tilde{s}_i)
\] (16.26)

Here \(s \setminus \tilde{s}_i\) denotes the strategy combination obtained from \(s\) by replacing \(s_i\) by \(\tilde{s}_i\). The set of all best replies against \(s\) for the \(i\)-th player is denoted by \(B_i(s)\). Explicitly, \(\tilde{s}_i \in B(s)\) if and only if \(\tilde{s}_i^k = 0\) whenever \(R_i(s \setminus k) < \max_{1 \leq l \leq m_i} R_i(s \setminus l)\) (in particular \(B_i(s) \neq \emptyset\)).

Let \(s, \bar{s} \in S\), we call \(\bar{s}\) a best reply against \(s\) if \(\bar{s}_i\) is a best reply against \(s\) for all \(i\). The set of all best replies against \(s\) is \(B(s) = \prod_{i=1}^n B_i(s)\).

A strategy combination \(\bar{s} \in S\) is a Nash equilibrium for the game if it is a best reply against itself, that is,

\[
\bar{s} \in B(\bar{s}).
\] (16.27)

Or, put differently, \(\bar{s}\) is a Nash equilibrium if no player can increase his payoff by changing his strategy as long as all others stick to their respective strategies. In addition, if a player sticks to his equilibrium strategy, he is assured that his payoff will not decrease no matter what the others do.

To illustrate these concepts, let us consider the famous prisoners dilemma. Here we have two players which can choose to defect or cooperate. The payoff is symmetric for both players and given by the following diagram

\[
\begin{array}{c|cc}
R_1 & d_2 & c_2 \\
\hline
d_1 & 0 & 2 \\
c_1 & -1 & 1 \\
\end{array}
\begin{array}{c|cc}
R_2 & d_2 & c_2 \\
\hline
d_1 & 0 & -1 \\
c_1 & 2 & 1 \\
\end{array}
\] (16.28)

where \(c_i\) or \(d_i\) means that player \(i\) cooperates or defects, respectively. It is easy to see that the (pure) strategy pair \((d_1, d_2)\) is the only Nash equilibrium for this game and that the expected payoff is 0 for both players. Of course, both players could get the payoff 1 if they both agree to cooperate. But if one would break this agreement in order to increase his payoff, the other one would get less. Hence it might be safer to defect.

Now that we have seen that Nash equilibria are a useful concept, we want to know when such an equilibrium exists. Luckily we have the following result.
Theorem 16.20 (Nash). Every n-person game has at least one Nash equilibrium.

Proof. The definition of a Nash equilibrium begs us to apply Kakutani’s theorem to the set valued function \( s \mapsto B(s) \). First of all, \( S \) is compact and convex and so are the sets \( B(s) \). Next, observe that the closedness condition of Kakutani’s theorem is satisfied since if \( s^m \in S \) and \( \bar{s}^m \in B(s^n) \) both converge to \( s \) and \( \bar{s} \), respectively, then (16.26) for \( s^m, \bar{s}^m \)

\[
R_i(s^m \setminus \tilde{s}_i) \leq R_i(s^m \setminus \bar{s}_i^m), \quad \tilde{s}_i \in S_i, \ 1 \leq i \leq n,
\]

implies (16.26) for the limits \( s, \bar{s} \)

\[
R_i(s \setminus \tilde{s}_i) \leq R_i(s \setminus \bar{s}_i), \quad \tilde{s}_i \in S_i, \ 1 \leq i \leq n,
\]

by continuity of \( R_i(s) \). \( \square \)

16.6. Further properties of the degree

We now prove some additional properties of the mapping degree. The first one will relate the degree in \( \mathbb{R}^n \) with the degree in \( \mathbb{R}^m \). It will be needed later on to extend the definition of degree to infinite dimensional spaces. By virtue of the canonical embedding \( \mathbb{R}^m \hookrightarrow \mathbb{R}^m \times \{0\} \subset \mathbb{R}^n \) we can consider \( \mathbb{R}^m \) as a subspace of \( \mathbb{R}^n \).

Theorem 16.21 (Reduction property). Let \( f \in C(\overline{U}, \mathbb{R}^m) \) and \( y \in \mathbb{R}^m \setminus (\mathbb{I} + f)(\partial U) \), then

\[
\deg(\mathbb{I} + f, U, y) = \deg(\mathbb{I} + f_m, U_m, y), \tag{16.29}
\]

where \( f_m = f|_{U_m} \), where \( U_m \) is the projection of \( U \) to \( \mathbb{R}^m \).

Proof. Choose a \( \tilde{f} \in C^2(U, \mathbb{R}^m) \) sufficiently close to \( f \) such that \( y \in RV(\tilde{f}) \). Let \( x \in (\mathbb{I} + \tilde{f})^{-1}(y) \), then \( x = y - f(x) \in \mathbb{R}^m \) implies \( (\mathbb{I} + \tilde{f})^{-1}(y) = (\mathbb{I} + \tilde{f}_m)^{-1}(y) \). Moreover,

\[
J_{\mathbb{I} + \tilde{f}}(x) = \det(\mathbb{I} + \tilde{f})(x) = \det\left( \begin{array}{cc} \delta_{ij} + \partial_j \tilde{f}_i(x) & \partial_j \tilde{f}_j(x) \\ 0 & \delta_{ij} \end{array} \right) = \det(\delta_{ij} + \partial_j \tilde{f}_i) = J_{\mathbb{I} + \tilde{f}_m}(x)
\]

shows \( \deg(\mathbb{I} + f, U, y) = \deg(\mathbb{I} + \tilde{f}, U, y) = \deg(\mathbb{I} + \tilde{f}_m, U_m, y) = \deg(\mathbb{I} + f_m, U_m, y) \) as desired. \( \square \)

Let \( U \subseteq \mathbb{R}^n \) and \( f \in C(\overline{U}, \mathbb{R}^n) \) be as usual. By Theorem 16.2 we know that \( \deg(f, U, y) \) is the same for every \( y \) in a connected component of \( \mathbb{R}^n \setminus f(\partial U) \). We will denote these components by \( K_j \) and write \( \deg(f, U, y) = \deg(f, U, K_j) \) if \( y \in K_j \).
16.6. Further properties of the degree

**Theorem 16.22** (Product formula). Let $U \subseteq \mathbb{R}^n$ be a bounded and open set and denote by $G_j$ the connected components of $\mathbb{R}^n \setminus f(\partial U)$. If $g \circ f \in D_y(U, \mathbb{R}^n)$, then

$$\deg(g \circ f, U, y) = \sum_j \deg(f, U, G_j) \deg(g, G_j, y),$$  \hfill (16.30)

where only finitely many terms in the sum are nonzero.

**Proof.** Since $f(\overline{U})$ is is compact, we can find an $r > 0$ such that $f(\overline{U}) \subseteq B_r(0)$. Moreover, since $g^{-1}(y)$ is closed, $g^{-1}(y) \cap B_r(0)$ is compact and hence can be covered by finitely many components $\{G_j\}_{j=1}^m$. In particular, the others will have $\deg(f, G_k, y) = 0$ and hence only finitely many terms in the above sum are nonzero.

We begin by computing $\deg(g \circ f, U, y)$ in the case where $f, g \in C^1$ and $y \not\in CV(g \circ f)$. Since $d(g \circ f)(x) = g'(f(x))df(x)$ the claim is a straightforward calculation

$$\deg(g \circ f, U, y) = \sum_{x \in (g \circ f)^{-1}(y)} \text{sign}(J_{g \circ f}(x))$$

$$= \sum_{x \in (g \circ f)^{-1}(y)} \text{sign}(J_g(f(x))) \text{sign}(J_f(x))$$

$$= \sum_{z \in g^{-1}(y)} \text{sign}(J_g(z)) \sum_{x \in f^{-1}(z)} \text{sign}(J_f(x))$$

$$= \sum_{z \in g^{-1}(y)} \text{sign}(J_g(z)) \deg(f, U, z)$$

and, using our cover $\{G_j\}_{j=1}^m$,

$$\deg(g \circ f, U, y) = \sum_{j=1}^m \sum_{z \in g^{-1}(y) \cap G_j} \text{sign}(J_g(z)) \deg(f, U, z)$$

$$= \sum_{j=1}^m \deg(f, U, G_j) \sum_{z \in g^{-1}(y) \cap G_j} \text{sign}(J_g(z))$$

$$= \sum_{j=1}^m \deg(f, U, G_j) \deg(g, G_j, y).$$

Moreover, this formula still holds for $y \in CV(g \circ f)$ and for $g \in C$ by construction of the Brouwer degree. However, the case $f \in C$ will need a closer investigation since the sets $G_j$ depend on $f$. To overcome this problem we will introduce the sets

$$L_l = \{z \in \mathbb{R}^n \setminus f(\partial U) | \deg(f, U, z) = l\}.$$
Observe that $L_l, l > 0$, must be a union of some sets of $\{G_j\}_{j=1}^m$.

Now choose $\tilde{f} \in C^1$ such that $|f(x) - \tilde{f}(x)| < 2^{-1} \text{dist}(g^{-1}(y), f(\partial U))$ for $x \in \overline{U}$ and define $\tilde{K}_j, \tilde{L}_l$ accordingly. Then we have $U_l \cap g^{-1}(y) = \tilde{U}_l \cap g^{-1}(y)$ by Theorem 16.1 (iii). Moreover,

$$\deg(g \circ f, U, y) = \deg(g \circ \tilde{f}, U, y) = \sum_j \deg(\tilde{f}, U, \tilde{K}_j) \deg(g, \tilde{K}_j, y)$$

$$= \sum_{l>0} l \deg(g, \tilde{U}_l, y) = \sum_{l>0} l \deg(g, U_l, y)$$

$$= \sum_j \deg(f, U, G_j) \deg(g, G_j, y)$$

which proves the claim. \qed

16.7. The Jordan curve theorem

In this section we want to show how the product formula (16.30) for the Brouwer degree can be used to prove the famous Jordan curve theorem which states that a homeomorphic image of the circle dissects $\mathbb{R}^2$ into two components (which necessarily have the image of the circle as common boundary). In fact, we will even prove a slightly more general result.

**Theorem 16.23.** Let $C_j \subset \mathbb{R}^n, j = 1, 2$, be homeomorphic compact sets. Then $\mathbb{R}^n \setminus C_1$ and $\mathbb{R}^n \setminus C_2$ have the same number of connected components.

**Proof.** Denote the components of $\mathbb{R}^n \setminus C_1$ by $H_j$ and those of $\mathbb{R}^n \setminus C_2$ by $K_j$. Let $h : C_1 \to C_2$ be a homeomorphism with inverse $k : C_2 \to C_1$. By Theorem 16.15 we can extend both to $\mathbb{R}^n$. Then Theorem 16.1 (iii) and the product formula imply

$$1 = \deg(k \circ h, H_j, y) = \sum_l \deg(h, H_j, G_l) \deg(k, G_l, y)$$

for any $y \in H_j$. Now we have

$$\bigcup_i K_i = \mathbb{R}^n \setminus C_1 \subset \mathbb{R}^n \setminus h(\partial H_j) \subset \bigcup_l G_l$$

and hence for every $i$ we have $K_i \subseteq G_l$ for some $l$ since components are maximal connected sets. Let $N_l = \{i | K_i \subseteq G_l\}$ and observe that we have

$$\deg(k, G_l, y) = \sum_{i \in N_l} \deg(k, K_i, y) \quad \text{and} \quad \deg(h, H_j, G_l) = \deg(h, H_j, K_i)$$

for every $i \in N_l$. Therefore,

$$1 = \sum_l \sum_{i \in N_l} \deg(h, H_j, K_i) \deg(k, K_i, y) = \sum_i \deg(h, H_j, K_i) \deg(k, K_i, H_j)$$

By reversing the role of $C_1$ and $C_2$, the same formula holds with $H_j$ and $K_i$ interchanged.
Hence
\[ \sum_i 1 = \sum_i \sum_j \deg(h, H_j, K_i) \deg(k, K_i, H_j) = \sum_j 1 \]
shows that if either the number of components of \( \mathbb{R}^n \setminus C_1 \) or the number of components of \( \mathbb{R}^n \setminus C_2 \) is finite, then so is the other and both are equal. Otherwise there is nothing to prove. \( \square \)
Chapter 17

The Leray–Schauder mapping degree

17.1. The mapping degree on finite dimensional Banach spaces

The objective of this section is to extend the mapping degree from \( \mathbb{R}^n \) to general Banach spaces. Naturally, we will first consider the finite dimensional case.

Let \( X \) be a (real) Banach space of dimension \( n \) and let \( \phi \) be any isomorphism between \( X \) and \( \mathbb{R}^n \). Then, for \( f \in D_y(U, X) \), \( U \subset X \) open, \( y \in X \), we can define

\[
\deg(f, U, y) = \deg(\phi \circ f \circ \phi^{-1}, \phi(U), \phi(y)) \tag{17.1}
\]

provided this definition is independent of the isomorphism chosen. To see this let \( \psi \) be a second isomorphism. Then \( A = \psi \circ \phi^{-1} \in \text{GL}(n) \). Abbreviate \( f^* = \phi \circ f \circ \phi^{-1}, \ y^* = \phi(y) \) and pick \( \tilde{f}^* \in C^1_y(\phi(U), \mathbb{R}^n) \) in the same component of \( D_y(\phi(U), \mathbb{R}^n) \) as \( f^* \) such that \( y^* \in RV(f^*) \). Then \( A \circ \tilde{f}^* \circ A^{-1} \in C^1_y(\psi(U), \mathbb{R}^n) \) is the same component of \( D_y(\psi(U), \mathbb{R}^n) \) as \( A \circ f^* \circ A^{-1} = \psi \circ f \circ \psi^{-1} \) (since \( A \) is also a homeomorphism) and

\[
J_{A \circ \tilde{f}^* \circ A^{-1}}(A y^*) = \det(A)J_{\tilde{f}^*}(y^*) \det(A^{-1}) = J_{\tilde{f}^*}(y^*) \tag{17.2}
\]

by the chain rule. Thus we have \( \deg(\psi \circ f \circ \psi^{-1}, \psi(U), \psi(y)) = \deg(\phi \circ f \circ \phi^{-1}, \phi(U), \phi(y)) \) and our definition is independent of the basis chosen. In addition, it inherits all properties from the mapping degree in \( \mathbb{R}^n \). Note also that the reduction property holds if \( \mathbb{R}^m \) is replaced by an arbitrary subspace \( X_1 \) since we can always choose \( \phi : X \to \mathbb{R}^n \) such that \( \phi(X_1) = \mathbb{R}^m \).
Our next aim is to tackle the infinite dimensional case. The general idea is to approximate \( F \) by finite dimensional maps (in the same spirit as we approximated continuous \( f \) by smooth functions). To do this we need to know which maps can be approximated by finite dimensional operators. Hence we have to recall some basic facts first.

### 17.2. Compact maps

Let \( X, Y \) be Banach spaces and \( U \subset X \). A map \( F : U \subset X \to Y \) is called **finite dimensional** if its range is finite dimensional. In addition, it is called **compact** if it is continuous and maps bounded sets into relatively compact ones. The set of all compact maps is denoted by \( \mathcal{C}(U, Y) \) and the set of all compact, finite dimensional maps is denoted by \( \mathcal{F}(U, Y) \). Both sets are normed linear spaces and we have \( \mathcal{F}(U, Y) \subseteq \mathcal{C}(U, Y) \subseteq \mathcal{C}_b(U, Y) \) (recall that compact sets are automatically bounded).

If \( U \) is compact, then \( \mathcal{C}(U, Y) = \mathcal{C}(U, Y) \) (since the continuous image of a compact set is compact) and if \( \dim(Y) < \infty \), then \( \mathcal{F}(U, Y) = \mathcal{C}(U, Y) \). In particular, if \( U \subset \mathbb{R}^n \) is bounded, then \( \mathcal{F}(\overline{U}, \mathbb{R}^n) = \mathcal{C}(\overline{U}, \mathbb{R}^n) = \mathcal{C}(U, \mathbb{R}^n) \).

Now let us collect some results needed in the sequel.

**Lemma 17.1.** If \( K \subset X \) is compact, then for every \( \varepsilon > 0 \) there is a finite dimensional subspace \( X_\varepsilon \subset X \) and a continuous map \( P_\varepsilon : K \to X_\varepsilon \) such that \( |P_\varepsilon(x) - x| \leq \varepsilon \) for all \( x \in K \).

**Proof.** Pick \( \{x_i\}_{i=1}^n \subset K \) such that \( \bigcup_{i=1}^n B_\varepsilon(x_i) \) covers \( K \). Let \( \{\phi_i\}_{i=1}^n \) be a partition of unity (restricted to \( K \)) subordinate to \( \{B_\varepsilon(x_i)\}_{i=1}^n \), that is, \( \phi_i \in C(K, [0, 1]) \) with \( \text{supp}(\phi_i) \subset B_\varepsilon(x_i) \) and \( \sum_{i=1}^n \phi_i(x) = 1, \ x \in K \). Set

\[
P_\varepsilon(x) = \sum_{i=1}^n \phi_i(x)x_i,
\]

then

\[
|P_\varepsilon(x) - x| = \left| \sum_{i=1}^n \phi_i(x)x - \sum_{i=1}^n \phi_i(x)x_i \right| \leq \sum_{i=1}^n \phi_i(x)|x - x_i| \leq \varepsilon. \tag*{□}
\]

This lemma enables us to prove the following important result.

**Theorem 17.2.** Let \( U \) be bounded, then the closure of \( \mathcal{F}(U, Y) \) in \( \mathcal{C}(U, Y) \) is \( \mathcal{C}(U, Y) \).

**Proof.** Suppose \( F_N \in \mathcal{C}(U, Y) \) converges to \( F \). If \( F \notin \mathcal{C}(U, Y) \) then we can find a sequence \( x_n \in U \) such that \( |F(x_n) - F(x_m)| \geq \rho > 0 \) for \( n \neq m \). If \( N \)
17.3. The Leray–Schauder mapping degree

is so large that $|F - F_N| \leq \rho/4$, then

$$|F_N(x_n) - F_N(x_m)| \geq |F(x_n) - F(x_m)| - |F_N(x_n) - F(x_n)|$$

$$- |F_N(x_m) - F(x_m)|$$

$$\geq \rho - 2\frac{\rho}{4} = \frac{\rho}{2}$$

This contradiction shows $\overline{F(U, y)} \subseteq C(U, Y)$. Conversely, let $K = \overline{F(U)}$ and choose $P_\varepsilon$ according to Lemma 17.1, then $F_\varepsilon = P_\varepsilon \circ F \in F(U, Y)$ converges to $F$. Hence $C(U, Y) \subseteq \overline{F(U, y)}$ and we are done. \qed

Finally, let us show some interesting properties of mappings $I + F$, where $F \in C(U, Y)$.

**Lemma 17.3.** Let $U$ be bounded and closed. Suppose $F \in C(U, X)$, then $I + F$ is proper (i.e., inverse images of compact sets are compact) and maps closed subsets to closed subsets.

**Proof.** Let $A \subseteq U$ be closed and $y_n = (I + F)(x_n) \in (I + F)(A)$ converges to some point $y$. Since $y_n - x_n = F(x_n) \in F(U)$ we can assume that $y_n - x_n \to z$ after passing to a subsequence and hence $x_n \to x = y - z \in A$. Since $y = x + F(x) \in (I + F)(A)$, $(I + F)(A)$ is closed.

Next, let $U$ be closed and $K \subseteq Y$ be compact. Let $\{x_n\} \subseteq (I + F)^{-1}(K)$. Then we can pass to a subsequence $y_{n_m} = x_{n_m} + F(x_{n_m})$ such that $y_{n_m} \to y$. As before this implies $x_{n_m} \to x$ and thus $(I + F)^{-1}(K)$ is compact. \qed

Now we are all set for the definition of the Leray–Schauder degree, that is, for the extension of our degree to infinite dimensional Banach spaces.

### 17.3. The Leray–Schauder mapping degree

For $U \subseteq X$ we set

$$\mathcal{D}_y(\overline{U}, X) = \{F \in C(\overline{U}, X) | y \notin (I + F)(\partial U)\} \tag{17.3}$$

and $\mathcal{F}_y(\overline{U}, X) = \{F \in \mathcal{F}(\overline{U}, X) | y \notin (I + F)(\partial U)\}$. Note that for $F \in \mathcal{D}_y(\overline{U}, X)$ we have $\text{dist}(y, (I + F)(\partial U)) > 0$ since $I + F$ maps closed sets to closed sets.

Abbreviate $\rho = \text{dist}(y, (I + F)(\partial U))$ and pick $F_1 \in \mathcal{F}(\overline{U}, X)$ such that $|F - F_1| < \rho$ implying $F_1 \in \mathcal{F}_y(\overline{U}, X)$. Next, let $X_1$ be a finite dimensional subspace of $X$ such that $F_1(U) \subseteq X_1$, $y \in X_1$ and set $U_1 = U \cap X_1$. Then we have $F_1 \in \mathcal{F}_y(\overline{U_1}, X_1)$ and might define

$$\deg(I + F, U, y) = \deg(I + F_1, U_1, y) \tag{17.4}$$

provided we show that this definition is independent of $F_1$ and $X_1$ (as above). Pick another map $F_2 \in \mathcal{F}(\overline{U}, X)$ such that $|F - F_2| < \rho$ and let $X_2$ be a
corresponding finite dimensional subspace as above. Consider \( X_0 = X_1 + X_2, \) \( U_0 = U \cap X_0, \) then \( F_i \in \mathcal{F}_\gamma(U_0, X_0), i = 1, 2, \) and
\[
\deg(\mathbb{I} + F_i, U_0, y) = \deg(\mathbb{I} + F_i, U_i, y), \quad i = 1, 2,
\]
(17.5) by the reduction property. Moreover, set \( H(t) = \mathbb{I} + (1-t)F_1 + tF_2 \) implying \( H(t) \subset t \in [0, 1], \) since \( |H(t) - (\mathbb{I} + F)| < \rho \) for \( t \in [0, 1]. \) Hence homotopy invariance
\[
\deg(\mathbb{I} + F_1, U_0, y) = \deg(\mathbb{I} + F_2, U_0, y)
\]
(17.6) shows that (17.4) is independent of \( F_1, X_1. \)

**Theorem 17.4.** Let \( U \) be a bounded open subset of a (real) Banach space \( X \) and let \( F \in \mathcal{D}_\gamma(\overline{U}, X), y \in X. \) Then the following hold true.

(i). \( \deg(\mathbb{I} + F, U, y) = \deg(\mathbb{I} + F - y, U, 0). \)

(ii). \( \deg(\mathbb{I}, U, y) = 1 \) if \( y \in U. \)

(iii). If \( U_{1,2} \) are open, disjoint subsets of \( U \) such that \( y \notin f(\overline{U}\setminus(U_1 \cup U_2)), \)
then \( \deg(\mathbb{I} + F, U, y) = \deg(\mathbb{I} + F, U_1, y) + \deg(\mathbb{I} + F, U_2, y). \)

(iv). If \( H : [0,1] \times \overline{U} \to X \) and \( y : [0,1] \to X \) are both continuous such that \( H(t) \in \mathcal{D}_\gamma(t)(U, \mathbb{R}^n), t \in [0,1], \) then \( \deg(\mathbb{I} + H(0), U, y(0)) = \deg(\mathbb{I} + H(1), U, y(1)). \)

**Proof.** Except for (iv) all statements follow easily from the definition of the degree and the corresponding property for the degree in finite dimensional spaces. Considering \( H(t, x) - y(t), \) we can assume \( y(t) = 0 \) by (i). Since \( H([0,1], \partial U) \) is compact, we have \( \rho = \text{dist}(y, H([0,1], \partial U)) > 0. \) By Theorem 17.2 we can pick \( H_1 \in \mathcal{F}([0,1] \times U, X) \) such that \( |H(t) - H_1(t)| < \rho, \)
\( t \in [0,1]. \) this implies \( \deg(\mathbb{I} + H(t), U, 0) = \deg(\mathbb{I} + H_1(t), U, 0) \) and the rest follows from Theorem 16.2.

In addition, Theorem 16.1 and Theorem 16.2 hold for the new situation as well (no changes are needed in the proofs).

**Theorem 17.5.** Let \( F, G \in \mathcal{D}_\gamma(U, X), \) then the following statements hold.

(i). We have \( \deg(\mathbb{I} + F, 0, y) = 0. \) Moreover, if \( U_i, \) \( 1 \leq i \leq N, \) are disjoint open subsets of \( U \) such that \( y \notin (\mathbb{I} + F)(U \setminus \bigcup_{i=1}^{N} U_i), \)
then \( \deg(\mathbb{I} + F, U, y) = \sum_{i=1}^{N} \deg(\mathbb{I} + F, U_i, y). \)

(ii). If \( y \notin (\mathbb{I} + F)(U), \) then \( \deg(\mathbb{I} + F, U, y) = 0 \) (but not the other way round). Equivalently, if \( \deg(\mathbb{I} + F, U, y) \neq 0, \) then \( y \in (\mathbb{I} + F)(U). \)

(iii). If \( |f(x) - g(x)| < \text{dist}(y, f(\partial U)), \) \( x \in \partial U, \) then \( \deg(f, U, y) = \deg(g, U, y). \) In particular, this is true if \( f(x) = g(x) \) for \( x \in \partial U. \)

(iv). \( \deg(\mathbb{I} + ., U, y) \) is constant on each component of \( \mathcal{D}_\gamma(\overline{U}, X). \)

(v). \( \deg(\mathbb{I} + F, U, .) \) is constant on each component of \( X \setminus f(\partial U). \)
17.4. The Leray–Schauder principle and the Schauder fixed-point theorem

As a first consequence we note the Leray–Schauder principle which says that a priori estimates yield existence.

**Theorem 17.6** (Leray–Schauder principle). Suppose $F \in \mathcal{C}(X, X)$ and any solution $x$ of $x = tF(x), t \in [0, 1]$ satisfies the a priori bound $|x| \leq M$ for some $M > 0$, then $F$ has a fixed point.

**Proof.** Pick $\rho > M$ and observe $\deg(\mathbb{I} - F, B_\rho(0), 0) = \deg(\mathbb{I}, B_\rho(0), 0) = 1$ using the compact homotopy $H(t, x) := -tF(x)$. Here $H(t) \in \mathcal{D}_0(B_\rho(0), X)$ due to the a priori bound.

Now we can extend the Brouwer fixed-point theorem to infinite dimensional spaces as well.

**Theorem 17.7** (Schauder fixed point). Let $K$ be a closed, convex, and bounded subset of a Banach space $X$. If $F \in \mathcal{C}(K, K)$, then $F$ has at least one fixed point. The result remains valid if $K$ is only homeomorphic to a closed, convex, and bounded subset.

**Proof.** Since $K$ is bounded, there is a $\rho > 0$ such that $K \subseteq B_\rho(0)$. By Theorem 16.15 we can find a continuous retraction $R : X \to K$ (i.e., $R(x) = x$ for $x \in K$) and consider $\tilde{F} = F \circ R \in \mathcal{C}(\overline{B}_\rho(0), \overline{B}_\rho(0))$. The compact homotopy $H(t, x) := -t\tilde{F}(x)$ shows that $\deg(\mathbb{I} - \tilde{F}, B_\rho(0), 0) = \deg(\mathbb{I}, B_\rho(0), 0) = 1$. Hence there is a point $x_0 = \tilde{F}(x_0) \in K$. Since $\tilde{F}(x_0) = F(x_0)$ for $x_0 \in K$ we are done.

**Example.** Consider the nonlinear integral equation

$$x = F(x), \quad F(x)(t) := \int_0^1 e^{-ts} \cos(\lambda x(s))ds$$

in $X := C[0, 1]$ with $\lambda > 0$. Then one checks that $F \in C(X, X)$ since

$$|F(x)(t) - F(y)(t)| \leq \int_0^1 e^{-ts} |\cos(\lambda x(s)) - \cos(\lambda y(s))|ds$$

$$\leq \int_0^1 e^{-ts}\lambda|x(s) - y(s)|ds \leq \lambda\|x - y\|_\infty.$$

In particular, for $\lambda < 1$ we have a contraction and the contraction principle gives us existence of a unique fixed point. Moreover, proceeding similarly, one obtains estimates for the norm of $F(x)$ and its derivative:

$$\|F(x)\|_\infty \leq 1, \quad \|F(x)\|_\infty \leq 1.$$
Hence the Arzelà–Ascoli theorem (Theorem 0.37) implies that the image of $F$ is a compact subset of the unit ball and hence $F \in \mathcal{C}(B_1(0), B_1(0))$. Thus the Schauder fixed point theorem guarantees a fixed point for all $\lambda > 0$.

Finally, let us prove another fixed-point theorem which covers several others as special cases.

**Theorem 17.8.** Let $U \subset X$ be open and bounded and let $F \in \mathcal{C}(\overline{U}, X)$. Suppose there is an $x_0 \in U$ such that

$$F(x) - x_0 \neq \alpha(x - x_0), \quad x \in \partial U, \alpha \in (1, \infty). \tag{17.7}$$

Then $F$ has a fixed point.

**Proof.** Consider $H(t, x) := x - x_0 - t(F(x) - x_0)$, then we have $H(t, x) \neq 0$ for $x \in \partial U$ and $t \in [0, 1]$ by assumption. If $H(1, x) = 0$ for some $x \in \partial U$, then $x$ is a fixed point and we are done. Otherwise we have $\deg(I - F, U, 0) = \deg(I - x_0, U, 0) = \deg(I, U, x_0) = 1$ and hence $F$ has a fixed point.

Now we come to the anticipated corollaries.

**Corollary 17.9.** Let $U \subset X$ be open and bounded and let $F \in \mathcal{C}(\overline{U}, X)$. Then $F$ has a fixed point if one of the following conditions holds.

(i) $U = B_\rho(0)$ and $F(\partial U) \subseteq \overline{U}$ (Rothe).

(ii) $U = B_\rho(0)$ and $|F(x) - x|^2 \geq |F(x)|^2 - |x|^2$ for $x \in \partial U$ (Altman).

(iii) $X$ is a Hilbert space, $U = B_\rho(0)$ and $\langle F(x), x \rangle \leq |x|^2$ for $x \in \partial U$ (Krasnosel’ski).

**Proof.** (1). $F(\partial U) \subseteq \overline{U}$ and $F(x) = \alpha x$ for $|x| = \rho$ implies $|\alpha|\rho \leq \rho$ and hence (17.7) holds. (2). $F(x) = \alpha x$ for $|x| = \rho$ implies $(\alpha - 1)^2 \rho^2 \geq (\alpha^2 - 1)\rho^2$ and hence $\alpha \leq 0$. (3). Special case of (2) since $|F(x) - x|^2 = |F(x)|^2 - 2\langle F(x), x \rangle + |x|^2$. □

### 17.5. Applications to integral and differential equations

In this section we want to show how our results can be applied to integral and differential equations. To be able to apply our results we will need to know that certain integral operators are compact.

**Lemma 17.10.** Suppose $I = [a, b] \subset \mathbb{R}$ and $f \in \mathcal{C}(I \times I \times \mathbb{R}^n, \mathbb{R}^n)$, $\tau \in \mathcal{C}(I, I)$, then

$$F : \mathcal{C}(I, \mathbb{R}^n) \rightarrow \mathcal{C}(I, \mathbb{R}^n)$$

$$x(t) \mapsto F(x)(t) = \int_a^{\tau(t)} f(t, s, x(s)) ds \tag{17.8}$$

is compact.
Proof. We first need to prove that $F$ is continuous. Fix $x_0 \in C(I, \mathbb{R}^n)$ and $\varepsilon > 0$. Set $\rho = |x_0| + 1$ and abbreviate $\overline{B} = \overline{B}_\rho(0) \subset \mathbb{R}^n$. The function $f$ is uniformly continuous on $Q = I \times I \times \overline{B}$ since $Q$ is compact. Hence for $\varepsilon_1 = \varepsilon/(b-a)$ we can find a $\delta \in (0,1]$ such that $|f(t, s, x) - f(t, s, y)| \leq \varepsilon_1$ for $|x - y| < \delta$. But this implies

$$|F(x) - F(x_0)| = \sup_{t \in I} \left| \int_a^{\tau(t)} f(t, s, x(s)) - f(t, s, x_0(s))ds \right|$$

$$\leq \sup_{t \in I} \int_a^{\tau(t)} \left| f(t, s, x(s)) - f(t, s, x_0(s)) \right| ds$$

$$\leq \sup_{t \in I} (b-a)\varepsilon_1 = \varepsilon,$$

for $|x - x_0| < \delta$. In other words, $F$ is continuous. Next we note that if $U \subset C(I, \mathbb{R}^n)$ is bounded, say $|U| < \rho$, then

$$|F(U)| \leq \sup_{x \in U} \left| \int_a^{\tau(t)} f(t, s, x(s))ds \right| \leq (b-a)M,$$

where $M = \max |f(I, I, \overline{B}_\rho(0))|$. Moreover, the family $F(U)$ is equicontinuous. Fix $\varepsilon$ and $\varepsilon_1 = \varepsilon/(2(b-a))$, $\varepsilon_2 = \varepsilon/(2M)$. Since $f$ and $\tau$ are uniformly continuous on $I \times I \times \overline{B}_\rho(0)$ and $I$, respectively, we can find a $\delta > 0$ such that $|f(t, s, x) - f(t_0, s, x)| \leq \varepsilon_1$ and $|\tau(t) - \tau(t_0)| \leq \varepsilon_2$ for $|t - t_0| < \delta$. Hence we infer for $|t - t_0| < \delta$

$$|F(x)(t) - F(x)(t_0)| = \left| \int_a^{\tau(t)} f(t, s, x(s))ds - \int_a^{\tau(t_0)} f(t_0, s, x(s))ds \right|$$

$$\leq \int_a^{\tau(t_0)} \left| f(t, s, x(s)) - f(t_0, s, x(s)) \right| ds + \int_{\tau(t_0)}^{\tau(t)} \left| f(t, s, x(s)) \right| ds$$

$$\leq (b-a)\varepsilon_1 + \varepsilon_2 M = \varepsilon.$$

This implies that $F(U)$ is relatively compact by the Arzelà–Ascoli theorem (Theorem 0.37). Thus $F$ is compact.

As a first application we use this result to show existence of solutions to integral equations.

**Theorem 17.11.** Let $F$ be as in the previous lemma. Then the integral equation

$$x - \lambda F(x) = y, \quad \lambda \in \mathbb{R}, y \in C(I, \mathbb{R}^n) \quad (17.9)$$

has at least one solution $x \in C(I, \mathbb{R}^n)$ if $|\lambda| \leq \rho/M(\rho)$, where $M(\rho) = (b-a) \max_{(s,t,x) \in I \times I \times \overline{B}_\rho(0)} |f(s, t, x - y(s))|$ and $\rho > 0$ is arbitrary.
Proof. Note that, by our assumption on $\lambda$, $\lambda F + y$ maps $B_\rho(y)$ into itself. Now apply the Schauder fixed-point theorem.

This result immediately gives the Peano theorem for ordinary differential equations.

**Theorem 17.12 (Peano).** Consider the initial value problem

$$\dot{x} = f(t, x), \quad x(t_0) = x_0,$$

where $f \in C(I \times U, \mathbb{R}^n)$ and $I \subset \mathbb{R}$ is an interval containing $t_0$. Then (17.10) has at least one local solution $x \in C^1([t_0 - \varepsilon, t_0 + \varepsilon], \mathbb{R}^n)$, $\varepsilon > 0$. For example, any $\varepsilon$ satisfying $\varepsilon M(\varepsilon, \rho) \leq \rho$, $\rho > 0$ with $M(\varepsilon, \rho) = \max |f([t_0 - \varepsilon, t_0 + \varepsilon] \times B_\rho(x_0))|$ works. In addition, if $M(\varepsilon, \rho) \leq \tilde{M}(\varepsilon)(1 + \rho)$, then there exists a global solution.

Proof. For notational simplicity we make the shift $t \to t - t_0$, $x \to x - x_0$, $f(t, x) \to f(t + t_0, x + t_0)$ and assume $t_0 = 0$, $x_0 = 0$. In addition, it suffices to consider $t \geq 0$ since $t \to -t$ amounts to $f \to -f$.

Now observe, that (17.10) is equivalent to

$$x(t) - \int_0^t f(s, x(s))ds = 0, \quad x \in C([-\varepsilon, \varepsilon], \mathbb{R}^n)$$

and the first part follows from our previous theorem. To show the second, fix $\varepsilon > 0$ and assume $M(\varepsilon, \rho) \leq \tilde{M}(\varepsilon)(1 + \rho)$. Then

$$|x(t)| \leq \int_0^t |f(s, x(s))|ds \leq \tilde{M}(\varepsilon) \int_0^t (1 + |x(s)|)ds$$

implies $|x(t)| \leq \exp(\tilde{M}(\varepsilon)\varepsilon)$ by Gronwall’s inequality. Hence we have an a priori bound which implies existence by the Leray–Schauder principle. Since $\varepsilon$ was arbitrary we are done. □
Chapter 18

The stationary Navier–Stokes equation

18.1. Introduction and motivation

In this chapter we turn to partial differential equations. In fact, we will only consider one example, namely the stationary Navier–Stokes equation. Our goal is to use the Leray–Schauder principle to prove an existence and uniqueness result for solutions.

Let \( U (\neq \emptyset) \) be an open, bounded, and connected subset of \( \mathbb{R}^3 \). We assume that \( U \) is filled with an incompressible fluid described by its velocity field \( v_j(t,x) \) and its pressure \( p(t,x) \), \( (t,x) \in \mathbb{R} \times U \). The requirement that our fluid is incompressible implies \( \partial_j v_j = 0 \) (we sum over two equal indices from 1 to 3), which follows from the Gauss theorem since the flux through any closed surface must be zero.

Rather than just writing down the equation, let me give a short physical motivation. To obtain the equation which governs such a fluid we consider the forces acting on a small cube spanned by the points \((x_1, x_2, x_3)\) and \((x_1 + \Delta x_1, x_2 + \Delta x_2, x_3 + \Delta x_3)\). We have three contributions from outer forces, pressure differences, and viscosity.

The outer force density (force per volume) will be denoted by \( K_j \) and we assume that it is known (e.g. gravity).

The force from pressure acting on the surface through \((x_1, x_2, x_3)\) normal to the \( x_1 \)-direction is \( p \Delta x_2 \Delta x_3 \delta_{1j} \). The force from pressure acting on the opposite surface is \(-(p + \partial_1 p \Delta x_1) \Delta x_2 \Delta x_3 \delta_{1j} \). In summary, we obtain

\[-(\partial_j p) \Delta V, \quad (18.1)\]
where $\Delta V = \Delta x_1 \Delta x_2 \Delta x_3$.

The viscosity acting on the surface through $(x_1, x_2, x_3)$ normal to the $x_1$-direction is $-\eta \Delta x_2 \Delta x_3 \partial_1 v_j$ by some physical law. Here $\eta > 0$ is the viscosity constant of the fluid. On the opposite surface we have $\eta \Delta x_2 \Delta x_3 \partial_1 (v_j + \partial_1 v_j \Delta x_1)$. Adding up the contributions of all surface we end up with

$$\eta \Delta V \partial_i \partial_i v_j.$$  \hfill (18.2)

Putting it all together we obtain from Newton’s law

$$\rho \Delta V \frac{d}{dt}v_j(t, x(t)) = \eta \Delta V \partial_i \partial_i v_j(t, x(t)) - (\partial_j p(t, x(t)) + \Delta V K_j(t, x(t))),$$  \hfill (18.3)

where $\rho > 0$ is the density of the fluid. Dividing by $\Delta V$ and using the chain rule yields the **Navier–Stokes equation**

$$\rho \partial_t v_j = \eta \partial_i \partial_i v_j - \rho (v_i \partial_i) v_j - \partial_j p + K_j.$$  \hfill (18.4)

Note that it is no restriction to assume $\rho = 1$.

In what follows we will only consider the stationary Navier–Stokes equation

$$0 = \eta \partial_i \partial_i v_j - (v_i \partial_i) v_j - \partial_j p + K_j.$$  \hfill (18.5)

In addition to the incompressibility condition $\partial_j v_j = 0$ we also require the boundary condition $v|_{\partial U} = 0$, which follows from experimental observations.

In summary, we consider the problem (18.5) for $v$ in (e.g.) $X = \{v \in C^2(\overline{U}, \mathbb{R}^3) | \partial_j v_j = 0 \text{ and } v|_{\partial U} = 0\}$.

Our strategy is to rewrite the stationary Navier–Stokes equation in integral form, which is more suitable for our further analysis. For this purpose we need to introduce some function spaces first.

### 18.2. An insert on Sobolev spaces

Let $U$ be a bounded open subset of $\mathbb{R}^n$ and let $L^p(U, \mathbb{R})$ denote the Lebesgue spaces of $p$ integrable functions with norm

$$\|u\|_p = \left(\int_U |u(x)|^p dx\right)^{1/p}.  \hfill (18.6)$$

In the case $p = 2$ we even have a scalar product

$$\langle u, v \rangle_2 = \int_U u(x)v(x)dx$$  \hfill (18.7)

and our aim is to extend this case to include derivatives.

Given the set $C^1(U, \mathbb{R})$ we can consider the scalar product

$$\langle u, v \rangle_{2,1} = \int_U u(x)v(x)dx + \int_U (\partial_j u)(x)(\partial_j v)(x)dx.$$  \hfill (18.8)
Taking the completion with respect to the associated norm we obtain the Sobolev space $H^1(U, \mathbb{R})$. Similarly, taking the completion of $C^1_c(U, \mathbb{R})$ with respect to the same norm, we obtain the Sobolev space $H^1_0(U, \mathbb{R})$. Here $C^r_c(U, Y)$ denotes the set of functions in $C^r(U, Y)$ with compact support.

This construction of $H^1(U, \mathbb{R})$ implies that a sequence $u_k$ in $C^1(U, \mathbb{R})$ converges to $u \in H^1(U, \mathbb{R})$ if and only if $u_k$ and all its first order derivatives $\partial_j u_k$ converge in $L^2(U, \mathbb{R})$. Hence we can assign each $u \in H^1(U, \mathbb{R})$ its first order derivatives $\partial_j u$ by taking the limits from above. In order to show that this is a useful generalization of the ordinary derivative, we need to show that the derivative depends only on the limiting function $u \in L^2(U, \mathbb{R})$. To see this we need the following lemma.

**Lemma 18.1 (Integration by parts).** Suppose $u \in H^1_0(U, \mathbb{R})$ and $v \in H^1(U, \mathbb{R})$, then

$$\int_U u(\partial_j v)dx = -\int_U (\partial_j u)v dx.$$  \hspace{1cm} (18.9)

**Proof.** By continuity it is no restriction to assume $u \in C^1_c(U, \mathbb{R})$ and $v \in C^1(U, \mathbb{R})$. Moreover, we can find a function $\phi \in C^1_c(U, \mathbb{R})$ which is 1 on the support of $u$. Hence by considering $\phi v$ we can even assume $v \in C^1_c(U, \mathbb{R})$.

Moreover, we can replace $U$ by a rectangle $K$ containing $U$ and extend $u, v$ to $K$ by setting it 0 outside $U$. Now use integration by parts with respect to the $j$-th coordinate. \hfill \Box

In particular, this lemma says that if $u \in H^1(U, \mathbb{R})$, then

$$\int_U (\partial_j u)\phi dx = -\int_U u(\partial_j \phi) dx, \quad \phi \in C^\infty_c(U, \mathbb{R}).$$  \hspace{1cm} (18.10)

And since $C^\infty_c(U, \mathbb{R})$ is dense in $L^2(U, \mathbb{R})$, the derivatives are uniquely determined by $u \in L^2(U, \mathbb{R})$ alone. Moreover, if $u \in C^1(U, \mathbb{R})$, then the derivative in the Sobolev space corresponds to the usual derivative. In summary, $H^1(U, \mathbb{R})$ is the space of all functions $u \in L^2(U, \mathbb{R})$ which have first order derivatives (in the sense of distributions, i.e., (18.10)) in $L^2(U, \mathbb{R})$.

Next, we want to consider some additional properties which will be used later on. First of all, the **Poincaré–Friedrichs inequality**.

**Lemma 18.2 (Poincaré–Friedrichs inequality).** Suppose $u \in H^1_0(U, \mathbb{R})$, then

$$\int_U u^2 dx \leq d_j^2 \int_U (\partial_j u)^2 dx,$$  \hspace{1cm} (18.11)

where $d_j = \sup\{|x_j - y_j| |(x_1, \ldots, x_n), (y_1, \ldots, y_n) \in U\}$.

**Proof.** Again we can assume $u \in C^1_c(U, \mathbb{R})$ and we assume $j = 1$ for notational convenience. Replace $U$ by a set $K = [a, b] \times \tilde{K}$ containing $U$ and
extend $u$ to $K$ by setting it 0 outside $U$. Then we have
\[ u(x_1, x_2, \ldots, x_n)^2 = \left( \int_a^{x_1} 1 \cdot (\partial_1 u)(\xi, x_2, \ldots, x_n) d\xi \right)^2 \leq (b - a) \int_a^b (\partial_1 u)^2(\xi, x_2, \ldots, x_n) d\xi, \]
where we have used the Cauchy–Schwarz inequality. Integrating this result over $[a,b]$ gives
\[ \int_a^b u^2(\xi, x_2, \ldots, x_n) d\xi \leq (b - a)^2 \int_a^b (\partial_1 u)^2(\xi, x_2, \ldots, x_n) d\xi \]
and integrating over $\tilde{K}$ finishes the proof. □

Hence, from the viewpoint of Banach spaces, we could also equip $H^1_0(U, \mathbb{R})$ with the scalar product
\[ \langle u, v \rangle = \int_U (\partial_j u)(x)(\partial_j v)(x) dx. \] (18.12)
This scalar product will be more convenient for our purpose and hence we will use it from now on. (However, all results stated will hold in either case.) The norm corresponding to this scalar product will be denoted by $|.|$.

Next, we want to consider the embedding $H^1_0(U, \mathbb{R}) \hookrightarrow L^2(U, \mathbb{R})$ a little closer. This embedding is clearly continuous since by the Poincaré–Friedrichs inequality we have
\[ \|u\|_2 \leq \frac{d(U)}{\sqrt{n}} \|u\|, \quad d(U) = \sup\{|x - y| : x, y \in U\}. \] (18.13)
Moreover, by a famous result of Rellich, it is even compact. To see this we first prove the following inequality.

**Lemma 18.3** (Poincaré inequality). Let $Q \subset \mathbb{R}^n$ be a cube with edge length $\rho$. Then
\[ \int_Q u^2 dx \leq \frac{1}{\rho^n} \left( \int_Q |u| dx \right)^2 + \frac{n\rho^2}{2} \int_Q (\partial_k u)(\partial_k u) dx \] (18.14)
for all $u \in H^1(Q, \mathbb{R})$.

**Proof.** After a scaling we can assume $Q = (0,1)^n$. Moreover, it suffices to consider $u \in C^1(Q, \mathbb{R})$.

Now observe
\[ u(x) - u(\bar{x}) = \sum_{i=1}^n \int_{x_{i-1}}^{x_i} (\partial_i u) dx_i, \]
where \( x^i = (\tilde{x}_1, \ldots, \tilde{x}_i, x_{i+1}, \ldots, x_n) \). Squaring this equation and using Cauchy–Schwarz on the right-hand side we obtain

\[
    u(x)^2 - 2u(x)u(\tilde{x}) + u(\tilde{x})^2 \leq \left( \sum_{i=1}^{n} \int_{0}^{1} |\partial_i u| dx_i \right)^2 \leq n \sum_{i=1}^{n} \left( \int_{0}^{1} |\partial_i u| dx_i \right)^2 \leq n \sum_{i=1}^{n} \int_{0}^{1} (\partial_i u)^2 dx_i.
\]

Now we integrate over \( x \) and \( \tilde{x} \), which gives

\[
    2 \int_{Q} u^2 dx - 2 \left( \int_{Q} u dx \right)^2 \leq n \int_{Q} (\partial_i u)(\partial_i u) dx
\]

and finishes the proof. \( \square \)

Now we are ready to show Rellich’s compactness theorem.

**Theorem 18.4 (Rellich’s compactness theorem).** Let \( U \) be a bounded open subset of \( \mathbb{R}^n \). Then the embedding

\[
    H^1_0(U, \mathbb{R}) \hookrightarrow L^2(U, \mathbb{R}) \quad (18.15)
\]

is compact.

**Proof.** Pick a cube \( Q \) (with edge length \( \rho \)) containing \( U \) and a bounded sequence \( u^k \in H^1_0(U, \mathbb{R}) \). Since bounded sets are weakly compact, it is no restriction to assume that \( u^k \) is weakly convergent in \( L^2(U, \mathbb{R}) \). By setting \( u^k(x) = 0 \) for \( x \notin U \) we can also assume \( u^k \in H^1(Q, \mathbb{R}) \) (show this). Next, subdivide \( Q \) into \( N \) subcubes \( Q_i \) with edge lengths \( \rho/N \). On each subcube \( Q_i \) holds and hence

\[
    \int_{U} u^2 dx = \int_{Q} u^2 dx = \sum_{i=1}^{N} \frac{N}{\rho} \left( \int_{Q_i} u dx \right)^2 + \frac{n\rho^2}{2N^2} \int_{U} (\partial_k u)(\partial_k u) dx
\]

for all \( u \in H^1_0(U, \mathbb{R}) \). Hence we infer

\[
    \|u^k - u^\ell\|_2^2 \leq \sum_{i=1}^{N} \frac{N}{\rho} \left( \int_{Q_i} (u^k - u^\ell) dx \right)^2 + \frac{n\rho^2}{2N^2} \|u^k - u^\ell\|_2^2.
\]

The last term can be made arbitrarily small by picking \( N \) large. The first term converges to 0 since \( u^k \) converges weakly and each summand contains the \( L^2 \) scalar product of \( u^k - u^\ell \) and \( \chi_{Q_i} \) (the characteristic function of \( Q_i \)). \( \square \)

In addition to this result we will also need the following interpolation inequality.
Lemma 18.5 (Ladyzhenskaya inequality). Let \( U \subset \mathbb{R}^3 \). For all \( u \in H^1_0(U, \mathbb{R}) \) we have
\[
\| u \|_4 \leq \sqrt[4]{8} \| u \|_2^{1/4} \| u \|^{3/4}.
\] (18.16)

**Proof.** We first prove the case where \( u \in C_c^1(U, \mathbb{R}) \). The key idea is to start with \( U \subset \mathbb{R}^1 \) and then work ones way up to \( U \subset \mathbb{R}^2 \) and \( U \subset \mathbb{R}^3 \).

If \( U \subset \mathbb{R}^1 \) we have
\[
\int_u^2 = \int^x \partial_1 u^2(x_1)dx_1 \leq 2 \int |u\partial_1 u|dx_1
\]
and hence
\[
\max_{x \in U} u(x)^2 \leq 2 \int |u\partial_1 u|dx_1.
\]
Here, if an integration limit is missing, it means that the integral is taken over the whole support of the function.

If \( U \subset \mathbb{R}^2 \) we have
\[
\int u^4dx_1dx_2 \leq \int u^2(x, x_2)dx_1 \int \max_x u(x, x_2)dx_1
\]
\[
\leq 4 \int |u\partial_1 u|dx_1dx_2 \int |u\partial_2 u|dx_1dx_2
\]
\[
\leq 4 \left( \int u^2dx_1dx_2 \right)^{2/2} \left( \int (\partial_1 u)^2dx_1dx_2 \right)^{1/2} \left( \int (\partial_2 u)^2dx_1dx_2 \right)^{1/2}
\]
\[
\leq 4 \int u^2dx_1dx_2 \int ((\partial_1 u)^2 + (\partial_2 u)^2)dx_1dx_2
\]
Now let \( U \subset \mathbb{R}^3 \), then
\[
\int u^4dx_1dx_2dx_3 \leq 4 \int dx_3 \int u^2dx_1dx_2 \int ((\partial_1 u)^2 + (\partial_2 u)^2)dx_1dx_2
\]
\[
\leq 4 \int \max_z u(x_1, x_2, z)dx_1dx_2 \int ((\partial_1 u)^2 + (\partial_2 u)^2)dx_1dx_2dx_3
\]
\[
\leq 8 \int u^2dx_1dx_2dx_3 \int ((\partial_1 u)^2 + (\partial_2 u)^2)dx_1dx_2dx_3
\]
and applying Cauchy–Schwarz finishes the proof for \( u \in C^1_c(U, \mathbb{R}) \).

If \( u \in H^1_0(U, \mathbb{R}) \) pick a sequence \( u_k \) in \( C^1_c(U, \mathbb{R}) \) which converges to \( u \) in \( H^1_0(U, \mathbb{R}) \) and hence in \( L^2(U, \mathbb{R}) \). By our inequality, this sequence is Cauchy in \( L^4(U, \mathbb{R}) \) and converges to a limit \( v \in L^4(U, \mathbb{R}) \). Since \( \| u \|_2 \leq \sqrt[4]{8} \| u \|_4 \) \( (\int 1 \cdot u^2dx \leq \sqrt[4]{8} \int 1 dx \int u^4dx) \), \( u_k \) converges to \( v \) in \( L^2(U, \mathbb{R}) \) as well and hence \( u = v \). Now take the limit in the inequality for \( u_k \). \( \square \)

As a consequence we obtain
\[
\| u \|_4 \leq \left( \frac{8d(U)}{\sqrt[4]{3}} \right)^{1/4} \| u \|, \quad U \subset \mathbb{R}^3,
\] (18.17)
Corollary 18.6. The embedding
\[ H^1_0(U, \mathbb{R}) \hookrightarrow L^4(U, \mathbb{R}), \quad U \subset \mathbb{R}^3, \] (18.18)
is compact.

Proof. Let \( u_k \) be a bounded sequence in \( H^1_0(U, \mathbb{R}) \). By Rellich’s theorem there is a subsequence converging in \( L^2(U, \mathbb{R}) \). By the Ladyzhenskaya inequality this subsequence converges in \( L^4(U, \mathbb{R}) \). \( \square \)

Our analysis clearly extends to functions with values in \( \mathbb{R}^n \) since we have \( H^1_0(U, \mathbb{R}^n) = \bigoplus_{j=1}^n H^1_0(U, \mathbb{R}) \).

18.3. Existence and uniqueness of solutions

Now we come to the reformulation of our original problem (18.5). We pick as underlying Hilbert space \( H^1_0(U, \mathbb{R}^3) \) with scalar product
\[ \langle u, v \rangle = \int_U (\partial_j u_i)(\partial_j v_i) dx. \] (18.19)

Let \( X \) be the closure of \( X \) in \( H^1_0(U, \mathbb{R}^3) \), that is,
\[ X := \{ v \in C^2(\overline{U}, \mathbb{R}^3) | \partial_j v_j = 0 \text{ and } v|_{\partial U} = 0 \} = \{ v \in H^1_0(U, \mathbb{R}^3) | \partial_j v_j = 0 \}. \] (18.20)

Now we multiply (18.5) by \( w \in X \) and integrate over \( U \)
\[ \int_U \left( \eta \partial_k \partial_{k} v_j - (v_k \partial_{k} v_j) + K_j \right) w_j dx = \int_U (\partial_j p) w_j dx = 0. \] (18.21)

Using integration by parts this can be rewritten as
\[ \int_U \left( \eta (\partial_k v_j)(\partial_{k} w_j) - v_k v_j (\partial_{k} w_j) - K_j w_j \right) dx = 0. \] (18.22)

Hence if \( v \) is a solution of the Navier–Stokes equation, then it is also a solution of
\[ \eta \langle v, w \rangle - a(v, v, w) - \int_U K w dx = 0, \quad \text{for all } w \in X, \] (18.23)

where
\[ a(u, v, w) = \int_U u_k v_j (\partial_{k} w_j) dx. \] (18.24)

In other words, (18.23) represents a necessary solubility condition for the Navier–Stokes equations. A solution of (18.23) will also be called a weak solution of the Navier–Stokes equations. If we can show that a weak solution is in \( C^2 \), then we can read our argument backwards and it will be also a classical solution. However, in general this might not be true and it will
only solve the Navier–Stokes equations in the sense of distributions. But let us try to show existence of solutions for (18.23) first.

For later use we note
\[
\int_{U} v_{k} \partial_{k} v_{j} \, dx = 1/2 \int_{U} v_{j} \partial_{k} v_{k} \, dx = 0, \quad v \in \mathcal{X}.
\] (18.25)

We proceed by studying (18.23). Let \( K \in L^{2}(U, \mathbb{R}^{3}) \), then \( \int_{U} Kw \, dx \) is a linear functional on \( \mathcal{X} \) and hence there is a \( \tilde{K} \in \mathcal{X} \) such that
\[
\int_{U} Kw \, dx = \langle \tilde{K}, w \rangle, \quad w \in \mathcal{X}.
\] (18.26)
Moreover, the same is true for the map \( a(u,v,w), u,v \in \mathcal{X} \), and hence there is an element \( B(u,v) \in \mathcal{X} \) such that
\[
a(u,v,w) = \langle B(u,v), w \rangle, \quad w \in \mathcal{X}.
\] (18.27)
In addition, the map \( B : \mathcal{X}^{2} \to \mathcal{X} \) is bilinear. In summary we obtain
\[
\langle \eta v - B(v,v) - \tilde{K}, w \rangle = 0, \quad w \in \mathcal{X},
\] (18.28)
and hence
\[
\eta v - B(v,v) = \tilde{K}.
\] (18.29)
So in order to apply the theory from our previous chapter, we need a Banach space \( Y \) such that \( \mathcal{X} \hookrightarrow Y \) is compact.

Let us pick \( Y = L^{4}(U, \mathbb{R}^{3}) \). Then, applying the Cauchy–Schwarz inequality twice to each summand in \( a(u,v,w) \) we see
\[
|a(u,v,w)| \leq \sum_{j,k} \left( \int_{U} (u_{k}v_{j})^{2} \, dx \right)^{1/4} \left( \int_{U} (\partial_{k}w_{j})^{2} \, dx \right)^{1/4}
\]
\[
\leq \|w\| \sum_{j,k} \left( \int_{U} (u_{k})^{4} \, dx \right)^{1/4} \left( \int_{U} (v_{j})^{4} \, dx \right)^{1/4} = \|u\|_{4} \|v\|_{4} \|w\|.
\] (18.30)

Moreover, by Corollary 18.6 the embedding \( \mathcal{X} \hookrightarrow Y \) is compact as required.

Motivated by this analysis we formulate the following theorem.

**Theorem 18.7.** Let \( \mathcal{X} \) be a Hilbert space, \( Y \) a Banach space, and suppose there is a compact embedding \( \mathcal{X} \hookrightarrow Y \). In particular, \( \|u\|_{Y} \leq \beta\|u\| \). Let \( a : \mathcal{X}^{3} \to \mathbb{R} \) be a multilinear form such that
\[
|a(u,v,w)| \leq \alpha\|u\|_{Y} \|v\| \|w\|
\] (18.31)
and \( a(v,v,v) = 0 \). Then for any \( \tilde{K} \in \mathcal{X} \), \( \eta > 0 \) we have a solution \( v \in \mathcal{X} \) to the problem
\[
\eta \langle v, w \rangle - a(v,v,w) = \langle \tilde{K}, w \rangle, \quad w \in \mathcal{X}.
\] (18.32)
Moreover, if $2\alpha\beta|\tilde{K}| < \eta^2$ this solution is unique.

**Proof.** It is no loss to set $\eta = 1$. Arguing as before we see that our equation is equivalent to

$$v - B(v, v) + \tilde{K} = 0,$$

where our assumption (18.31) implies

$$\|B(u, v)\| \leq \alpha \|u\|_Y \|v\|_Y \leq \alpha \beta^2 \|u\| \|v\|.$$

Here the second equality follows since the embedding $X \hookrightarrow Y$ is continuous.

Abbreviate $F(v) = B(v, v)$. Observe that $F$ is locally Lipschitz continuous since if $\|u\|, \|v\| \leq \rho$ we have

$$\|F(u) - F(v)\| = \|B(u - v, u) - B(v, u - v)\| \leq 2\alpha \rho \|u - v\|_Y \leq 2\alpha \beta^2 \rho \|u - v\|.$$

Moreover, let $v_n$ be a bounded sequence in $X$. After passing to a subsequence we can assume that $v_n$ is Cauchy in $Y$ and hence $F(v_n)$ is Cauchy in $X$ by $\|F(u) - F(v)\| \leq 2\alpha \rho \|u - v\|_Y$. Thus $F : X \to X$ is compact.

Hence all we need to apply the Leray–Schauder principle is an a priori estimate. Suppose $v$ solves $v = tF(v) + t\tilde{K}$, $t \in [0, 1]$, then

$$\langle v, v \rangle = ta(v, v, v) + t\langle \tilde{K}, v \rangle = t\langle \tilde{K}, v \rangle.$$

Hence $\|v\| \leq \|\tilde{K}\|$ is the desired estimate and the Leray–Schauder principle yields existence of a solution.

Now suppose there are two solutions $v_i$, $i = 1, 2$. By our estimate they satisfy $\|v_i\| \leq \|\tilde{K}\|$ and hence $\|v_1 - v_2\| = \|F(v_1) - F(v_2)\| \leq 2\alpha \beta^2 \|\tilde{K}\| \|v_1 - v_2\|$ which is a contradiction if $2\alpha \beta^2 \|\tilde{K}\| < 1$. \qed

Hence we have found a solution $v$ to the generalized problem (18.23). This solution is unique if $2(\frac{2d(U)}{\sqrt{3}})3/2\|\tilde{K}\|_2 < \eta^2$. Under suitable additional conditions on the outer forces and the domain, it can be shown that weak solutions are $C^2$ and thus also classical solutions. However, this is beyond the scope of this introductory text.
19.1. Monotone maps

The Leray–Schauder theory can only be applied to compact perturbations of the identity. If $F$ is not compact, we need different tools. In this section we briefly present another class of maps, namely monotone ones, which allow some progress.

If $F : \mathbb{R} \to \mathbb{R}$ is continuous and we want $F(x) = y$ to have a unique solution for every $y \in \mathbb{R}$, then $f$ should clearly be strictly monotone increasing (or decreasing) and satisfy $\lim_{x \to \pm \infty} F(x) = \pm \infty$. Rewriting these conditions slightly such that they make sense for vector valued functions the analogous result holds.

**Lemma 19.1.** Suppose $F : \mathbb{R}^n \to \mathbb{R}^n$ is continuous and satisfies

$$
\lim_{|x| \to \infty} \frac{F(x)x}{|x|} = \infty.
$$

(19.1)

Then the equation

$$
F(x) = y
$$

(19.2)

has a solution for every $y \in \mathbb{R}^n$. If $F$ is strictly monotone

$$
(F(x) - F(y))(x - y) > 0, \quad x \neq y,
$$

(19.3)

then this solution is unique.

**Proof.** Our first assumption implies that $G(x) = F(x) - y$ satisfies $G(x)x = F(x)x - yx > 0$ for $|x|$ sufficiently large. Hence the first claim follows from Theorem 16.13. The second claim is trivial. \qed
Now we want to generalize this result to infinite dimensional spaces. Throughout this chapter, $\mathcal{H}$ will be a Hilbert space with scalar product $\langle \cdot, \cdot \rangle$. A map $F : \mathcal{H} \to \mathcal{H}$ is called **monotone** if
\[
\langle F(x) - F(y), x - y \rangle \geq 0, \quad x, y \in \mathcal{H},
\] (19.4)
strictly monotone if
\[
\langle F(x) - F(y), x - y \rangle > 0, \quad x \neq y \in \mathcal{H},
\] (19.5)
and finally **strongly monotone** if there is a constant $C > 0$ such that
\[
\langle F(x) - F(y), x - y \rangle \geq C\|x - y\|^2, \quad x, y \in \mathcal{H}.
\] (19.6)

Note that the same definitions can be made for a Banach space $X$ and mappings $F : X \to X^*$.

Observe that if $F$ is strongly monotone, then it automatically satisfies
\[
\lim_{|x| \to \infty} \frac{\langle F(x), x \rangle}{\|x\|} = \infty.
\] (19.7)
(Just take $y = 0$ in the definition of strong monotonicity.) Hence the following result is not surprising.

**Theorem 19.2** (Zarantonello). Suppose $F \in C(\mathcal{H}, \mathcal{H})$ is (globally) Lipschitz continuous and strongly monotone. Then, for each $y \in \mathcal{H}$ the equation
\[
F(x) = y
\] (19.8)
has a unique solution $x(y) \in \mathcal{H}$ which depends continuously on $y$.

**Proof.** Set
\[
G(x) := x - t(F(x) - y), \quad t > 0,
\]
then $F(x) = y$ is equivalent to the fixed point equation
\[
G(x) = x.
\]
It remains to show that $G$ is a contraction. We compute
\[
\|G(x) - G(\tilde{x})\|^2 = \|x - \tilde{x}\|^2 - 2t\langle F(x) - F(\tilde{x}), x - \tilde{x} \rangle + t^2\|F(x) - F(\tilde{x})\|^2
\leq (1 - 2\frac{C}{L}(Lt + (Lt)^2))\|x - \tilde{x}\|^2,
\]
where $L$ is a Lipschitz constant for $F$ (i.e., $\|F(x) - F(\tilde{x})\| \leq L\|x - \tilde{x}\|$). Thus, if $t \in (0, \frac{C}{2L})$, $G$ is a uniform contraction and the rest follows from the uniform contraction principle. \(\square\)

Again observe that our proof is constructive. In fact, the best choice for $t$ is clearly $t = \frac{C}{2L}$ such that the contraction constant $\theta = 1 - (\frac{C}{L})^2$ is minimal. Then the sequence
\[
x_{n+1} = x_n - \frac{C}{L^2}(F(x_n) - y), \quad x_0 = y,
\] (19.9)
converges to the solution.

19.2. The nonlinear Lax–Milgram theorem

As a consequence of the last theorem we obtain a nonlinear version of the Lax–Milgram theorem. We want to investigate the following problem:

\[ a(x, y) = b(y), \quad \text{for all } y \in \mathcal{H}, \quad (19.10) \]

where \( a : \mathcal{H}^2 \to \mathbb{R} \) and \( b : \mathcal{H} \to \mathbb{R} \). For this equation the following result holds.

**Theorem 19.3** (Nonlinear Lax–Milgram theorem). Suppose \( b \in \mathcal{L}(\mathcal{H}, \mathbb{R}) \) and \( a(x, \cdot) \in \mathcal{L}(\mathcal{H}, \mathbb{R}) \), \( x \in \mathcal{H} \), are linear functionals such that there are positive constants \( L \) and \( C \) such that for all \( x, y, z \in \mathcal{H} \) we have

\[ a(x, x - y) - a(y, x - y) \geq C|x - y|^2 \quad (19.11) \]

and

\[ |a(x, z) - a(y, z)| \leq L|x - y|. \quad (19.12) \]

Then there is a unique \( x \in \mathcal{H} \) such that (19.10) holds.

**Proof.** By the Riez lemma (Theorem 2.10) there are elements \( F(x) \in \mathcal{H} \) and \( z \in \mathcal{H} \) such that \( a(x, y) = b(y) \) is equivalent to \( \langle F(x) - z, y \rangle = 0, \quad y \in \mathcal{H}, \) and hence to

\[ F(x) = z. \]

By (19.11) the map \( F \) is strongly monotone. Moreover, by (19.12) we infer

\[ \|F(x) - F(y)\| = \sup_{\tilde{x} \in \mathcal{H}, \|\tilde{x}\|=1} |\langle F(x) - F(y), \tilde{x} \rangle| \leq L\|x - y\| \]

that \( F \) is Lipschitz continuous. Now apply Theorem 19.2. \( \square \)

The special case where \( a \in \mathcal{L}^2(\mathcal{H}, \mathbb{R}) \) is a bounded bilinear form which is strongly coercive, that is,

\[ a(x, x) \geq C\|x\|^2, \quad x \in \mathcal{H}, \quad (19.13) \]

is usually known as (linear) Lax–Milgram theorem (Theorem 2.15).

The typical application of this theorem is the existence of a unique weak solution of the Dirichlet problem for **elliptic equations**

\[ -\partial_i A_{ij}(x) \partial_j u(x) + b_j(x) \partial_j u(x) + c(x) u(x) = f(x), \quad x \in U, \]
\[ u(x) = 0, \quad x \in \partial U, \quad (19.14) \]

where \( U \) is a bounded open subset of \( \mathbb{R}^n \). By elliptic we mean that all coefficients \( A, b, c \) plus the right-hand side \( f \) are bounded and \( a_0 > 0 \), where

\[ a_0 = \inf_{e \in S^n, x \in U} e_i A_{ij}(x) e_j, \quad b_0 = \sup_{x \in U} |b(x)|, \quad c_0 = \inf_{x \in U} c(x). \quad (19.15) \]
As in Section 18.3 we pick $H^1_0(U, \mathbb{R})$ with scalar product

$$\langle u, v \rangle = \int_U (\partial_j u)(\partial_j v)dx$$  \hspace{1cm} (19.16)

as underlying Hilbert space. Next we multiply (19.14) by $v \in H^1_0$ and integrate over $U$

$$\int_U \left(- \partial_i A_{ij}(x) \partial_j u(x) + b_j(x)\partial_j u(x) + c(x)u(x)\right)v(x)dx = \int_U f(x)v(x)dx.$$  \hspace{1cm} (19.17)

After integration by parts we can write this equation as

$$a(v, u) = f(v), \quad v \in H^1_0,$$  \hspace{1cm} (19.18)

where

$$a(v, u) = \int_U \left(\partial_i v(x)A_{ij}(x)\partial_j u(x) + b_j(x)v(x)\partial_j u(x) + c(x)v(x)u(x)\right)dx$$

$$f(v) = \int_U f(x)v(x)dx,$$  \hspace{1cm} (19.19)

We call a solution of (19.18) a **weak solution** of the elliptic Dirichlet problem (19.14).

By a simple use of the Cauchy–Schwarz and Poincaré–Friedrichs inequalities we see that the bilinear form $a(u, v)$ is bounded. To be able to apply the (linear) Lax–Milgram theorem we need to show that it satisfies

$$a(u, u) \geq C \int |\partial_j u|^2 dx.$$  \hspace{1cm} (19.20)

Using (19.15) we have

$$a(u, u) \geq \int_U \left(a_0|\partial_j u|^2 - b_0|u||\partial_j u| + c_0|u|^2\right),$$  \hspace{1cm} (19.21)

and we need to control the middle term. If $b_0 = 0$ there is nothing to do and it suffices to require $c_0 \geq 0$.

If $b_0 > 0$ we distribute the middle term by means of the elementary inequality

$$|u||\partial_j u| \leq \frac{\varepsilon}{2}|u|^2 + \frac{1}{2\varepsilon} |\partial_j u|^2$$

which gives

$$a(u, u) \geq \int_U \left((a_0 - \frac{b_0}{2\varepsilon})|\partial_j u|^2 + (c_0 - \frac{\varepsilon b_0}{2})|u|^2\right).$$  \hspace{1cm} (19.22)

Since we need $a_0 - \frac{b_0}{2\varepsilon} > 0$ and $c_0 - \frac{\varepsilon b_0}{2} \geq 0$, or equivalently $\frac{2c_0}{b_0} \geq \varepsilon > \frac{b_0}{2a_0}$, we see that we can apply the Lax–Milgram theorem if $4a_0c_0 > b_0^2$. In summary, we have proven

**Theorem 19.4.** The elliptic Dirichlet problem (19.14) has a unique weak solution $u \in H^1_0(U, \mathbb{R})$ if $a_0 > 0$, $b_0 = 0$, $c_0 \geq 0$ or $a_0 > 0$, $4a_0c_0 > b_0^2$.  

19.3. The main theorem of monotone maps

Now we return to the investigation of $F(x) = y$ and weaken the conditions of Theorem 19.2. We will assume that $\mathcal{H}$ is a separable Hilbert space and that $F: \mathcal{H} \to \mathcal{H}$ is a continuous monotone map satisfying
\[
\lim_{|x| \to \infty} \frac{\langle F(x), x \rangle}{\|x\|} = \infty.
\] (19.23)

In fact, if suffices to assume that $F$ is weakly continuous
\[
\lim_{n \to \infty} \langle F(x_n), y \rangle = \langle F(x), y \rangle, \quad \text{for all } y \in \mathcal{H}
\] (19.24)
whenever $x_n \to x$.

The idea is as follows: Start with a finite dimensional subspace $\mathcal{H}_n \subset \mathcal{H}$ and project the equation $F(x) = y$ to $\mathcal{H}_n$ resulting in an equation
\[
F_n(x_n) = y_n, \quad x_n, y_n \in \mathcal{H}_n.
\] (19.25)

More precisely, let $P_n$ be the (linear) projection onto $\mathcal{H}_n$ and set $F_n(x_n) = P_n F(x_n), y_n = P_n y$ (verify that $F_n$ is continuous and monotone!).

Now Lemma 19.1 ensures that there exists a solution $u_n$. Now chose the subspaces $\mathcal{H}_n$ such that $\mathcal{H}_n \to \mathcal{H}$ (i.e., $\mathcal{H}_n \subset \mathcal{H}_{n+1}$ and $\bigcup_{n=1}^{\infty} \mathcal{H}_n$ is dense). Then our hope is that $u_n$ converges to a solution $u$.

This approach is quite common when solving equations in infinite dimensional spaces and is known as Galerkin approximation. It can often be used for numerical computations and the right choice of the spaces $\mathcal{H}_n$ will have a significant impact on the quality of the approximation.

So how should we show that $x_n$ converges? First of all observe that our construction of $x_n$ shows that $x_n$ lies in some ball with radius $R_n$, which is chosen such that
\[
\langle F_n(x), x \rangle > \|y_n\| \|x\|, \quad \|x\| \geq R_n, \ x \in \mathcal{H}_n.
\] (19.26)

Since $\langle F_n(x), x \rangle = \langle P_n F(x), x \rangle = \langle F(x), P_n x \rangle = \langle F(x), x \rangle$ for $x \in \mathcal{H}_n$ we can drop all $n$’s to obtain a constant $R$ which works for all $n$. So the sequence $x_n$ is uniformly bounded
\[
\|x_n\| \leq R.
\] (19.27)

Now by a well-known result there exists a weakly convergent subsequence. That is, after dropping some terms, we can assume that there is some $x$ such that $x_n \rightharpoonup x$, that is,
\[
\langle x_n, z \rangle \to \langle x, z \rangle, \quad \text{for every } z \in \mathcal{H}.
\] (19.28)

And it remains to show that $x$ is indeed a solution. This follows from
Lemma 19.5. Suppose $F : \mathcal{H} \to \mathcal{H}$ is weakly continuous and monotone, then
\[ \langle y - F(z), x - z \rangle \geq 0 \quad \text{for every } z \in \mathcal{H} \] (19.29) implies $F(x) = y$.

**Proof.** Choose $z = x \pm tw$, then $\mp \langle y - F(x \pm tw), w \rangle \geq 0$ and by continuity $\mp \langle y - F(x), w \rangle \geq 0$. Thus $\langle y - F(x), w \rangle = 0$ for every $w$ implying $y - F(x) = 0$. \[ \square \]

Now we can show

**Theorem 19.6** (Browder, Minty). Suppose $F : \mathcal{H} \to \mathcal{H}$ is weakly continuous, monotone, and satisfies
\[ \lim_{|x| \to \infty} \frac{\langle F(x), x \rangle}{\|x\|} = \infty. \] (19.30)

Then the equation
\[ F(x) = y \] (19.31)
has a solution for every $y \in \mathcal{H}$. If $F$ is strictly monotone then this solution is unique.

**Proof.** Abbreviate $y_n = F(x_n)$, then we have $\langle y - F(z), x_n - z \rangle = \langle y_n - F_n(z), x_n - z \rangle \geq 0$ for $z \in \mathcal{H}_n$. Taking the limit implies $\langle y - F(z), x - z \rangle \geq 0$ for every $z \in \mathcal{H}_\infty = \bigcup_{n=1}^{\infty} \mathcal{H}_n$. Since $\mathcal{H}_\infty$ is dense, $\langle y - F(z), x - z \rangle \geq 0$ for every $z \in \mathcal{H}$ by continuity and hence $F(x) = y$ by our lemma. \[ \square \]

Note that in the infinite dimensional case we need monotonicity even to show existence. Moreover, this result can be further generalized in two more ways. First of all, the Hilbert space $\mathcal{H}$ can be replaced by a reflexive Banach space $X$ if $F : X \to X^*$. The proof is almost identical. Secondly, it suffices if
\[ t \mapsto \langle F(x + ty), z \rangle \] (19.32)
is continuous for $t \in [0, 1]$ and all $x, y, z \in \mathcal{H}$, since this condition together with monotonicity can be shown to imply weak continuity.
Appendix A

Some set theory

At the beginning of the 20th century Russel showed with his famous paradox that naive set theory can lead into contradictions. Hence it was replaced by axiomatic set theory, more specific we will take the Zermelo–Fraenkel set theory (ZF), which assumes existence of some sets (like the empty set and the integers) and defines what operations are allowed. Somewhat informally (i.e. without writing them using the symbolism of first order logic) they can be stated as follows:

- **Axiom of empty set.** There is a set $\emptyset$ which contains no elements.

- **Axiom of extensionality.** Two sets $A$ and $B$ are equal $A = B$ if they contain the same elements. If a set $A$ contains all elements from a set $B$, it is called a subset $A \subseteq B$. In particular $A \subseteq B$ and $B \subseteq A$ if and only if $A = B$.

The last axiom implies that the empty set is unique and that any set which is not equal to the empty set has at least one element.

- **Axiom of pairing.** If $A$ and $B$ are sets, then there exists a set $\{A, B\}$ which contains $A$ and $B$ as elements. One writes $\{A, A\} = \{A\}$. By the axiom of extensionality we have $\{A, B\} = \{B, A\}$.

- **Axiom of union.** Given a set $F$ whose elements are again sets, there is a set $A = \bigcup F$ containing every element that is a member of some member of $F$. In particular, given two sets $A, B$ there exists a set $A \cup B = \bigcup \{A, B\}$ consisting of the elements of both sets. Note that this definition ensures that the union is commutative $A \cup B = B \cup A$ and associative $(A \cup B) \cup C = A \cup (B \cup C)$. Note also $\bigcup \{A\} = A$. 

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• **Axiom schema of specification.** Given a set \(A\) and a logical statement \(\phi(x)\) depending on \(x \in A\) we can form the set \(B = \{x \in A | \phi(x)\}\) of all elements from \(A\) obeying \(\phi\). For example, given two sets \(A\) and \(B\) we can define their intersection as \(A \cap B = \{x \in A \cup B | (x \in A) \land (x \in B)\}\) and their complement as \(A \setminus B = \{x \in A | x \notin B\}\). Or the intersection of a family of sets \(F\) as \(\bigcap F = \{x \in \bigcup F | \forall F \in F : x \in F\}\).

• **Axiom of power set.** For any set \(A\), there is a **power set** \(\mathcal{P}(A)\) that contains every subset of \(A\).

From these axioms one can define ordered pairs as \((x, y) = \{\{x\}, \{x, y\}\}\) and the Cartesian product as \(A \times B = \{z \in \mathcal{P}(A \cup \mathcal{P}(A \cup B)) | \exists x \in A, y \in B : z = (x, y)\}\). Functions \(f : A \to B\) are defined as single valued relations, that is \(f \subseteq A \times B\) such that \((x, y) \in f\) and \((x, \tilde{y}) \in f\) implies \(y = \tilde{y}\).

• **Axiom schema of replacement.** For every function \(f\) the image of a set \(A\) is again a set \(B = \{f(x) | x \in A\}\).

So far the previous axioms were concerned with ensuring that the usual set operations required in mathematics yield again sets. In particular, we can start constructing sets with any given finite number of elements starting from the empty set: \(\emptyset\) (no elements), \(\{\emptyset\}\) (one element), \(\{\emptyset, \{\emptyset\}\}\) (two elements), etc. However, while existence of infinite sets (like e.g. the integers) might seem obvious at this point, it cannot be deduced from the axioms we have so far. Hence it has to be added as well.

• **Axiom of infinity.** There exists a set \(A\) which contains the empty set and for every element \(x \in A\) we also have \(x \cup \{x\} \in A\). The smallest such set \(\{\emptyset, \{\emptyset\}, \{\emptyset, \{\emptyset\}\}, \ldots\}\) can be identified with the integers via \(0 = \emptyset, 1 = \{\emptyset\}, 2 = \{\emptyset, \{\emptyset\}\}, \ldots\) now we finally have the integers and thus everything we need to start constructing the rational, real, and complex numbers in the usual way. Hence we only add one more axiom to exclude some pathological objects which will lead to contradictions.

• **Axiom of Regularity.** Every nonempty set \(A\) contains an element \(x\) with \(x \cap A = \emptyset\). This excludes for example the possibility that a set contains itself as an element (apply the axiom to \(\{A\}\)). Similarly, we can only have \(A \in B\) or \(B \in A\) but not both (apply it to the set \(\{A, B\}\)).

Hence a set is something which can be constructed from the above axioms. Of course this raises the question if these axioms are consistent but as has been shown by Gödel this question cannot be answered: If ZF contains a statement of its own consistency then ZF is inconsistent. In fact, the
same holds for any other sufficiently rich (such that one can do basic math) system of axioms. In particular, it also holds for ZFC defined below. So we have to live with the fact that someday someone might come and prove that ZFC is inconsistent.

Starting from ZF one can develop basic analysis (including the construction of the real numbers). However, it turns out that several fundamental results require yet another construction for their proof:

Given an index set \( A \) and for every \( \alpha \in A \) some set \( M_\alpha \) the product \( \prod_{\alpha \in A} M_\alpha \) is defined to be the set of all functions \( \varphi : A \rightarrow \bigcup_{\alpha \in A} M_\alpha \) which assign each element \( \alpha \in A \) some element \( m_\alpha \in M_\alpha \). If all sets \( M_\alpha \) are nonempty it seems quite reasonable that there should be such a choice function which chooses an element from \( M_\alpha \) for every \( \alpha \in A \). However, no matter how obvious this might seem, it cannot be deduced from the ZF axioms alone and hence has to be added:

- **Axiom of Choice**: Given an index set \( A \) and nonempty sets \( \{M_\alpha\}_{\alpha \in A} \) their product \( \prod_{\alpha \in A} M_\alpha \) is nonempty.

ZF augmented by the axiom of choice is known as ZFC and we accept it as the fundament upon which our functional analytic house is built.

Note that the axiom of choice is not only used to ensure that infinite products are nonempty but also in many proofs! For example, suppose you start with a set \( M_1 \) and recursively construct some sets \( M_n \) such that in every step you have a nonempty set. Then the axiom of choice guarantees the existence of a sequence \( x = (x_n)_{n \in \mathbb{N}} \) with \( x_n \in M_n \).

The axiom of choice has many important consequences (many of which are in fact equivalent to the axiom of choice and it is hence a matter of taste which one to choose as axiom).

A **partial order** is a binary relation \( \preceq \) over a set \( \mathcal{P} \) such that for all \( A, B, C \in \mathcal{P} \):

- \( A \preceq A \) (reflexivity),
- if \( A \preceq B \) and \( B \preceq A \) then \( A = B \) (antisymmetry),
- if \( A \preceq B \) and \( B \preceq C \) then \( A \preceq C \) (transitivity).

It is custom to write \( A \prec B \) if \( A \preceq B \) and \( A \neq B \).

**Example.** Let \( \mathcal{P}(X) \) be the collections of all subsets of a set \( X \). Then \( \mathcal{P} \) is partially ordered by inclusion \( \subseteq \).

It is important to emphasize that two elements of \( \mathcal{P} \) need not be comparable, that is, in general neither \( A \preceq B \) nor \( B \preceq A \) might hold. However, if any two elements are comparable, \( \mathcal{P} \) will be called **totally ordered**. A set with a total order is called **well-ordered** if every nonempty subset has
a least element, that is some $A \in \mathcal{P}$ with $A \preceq B$ for every $B \in \mathcal{P}$. Note that the least element is unique by antisymmetry.

Example. $\mathbb{R}$ with $\leq$ is totally ordered and $\mathbb{N}$ with $\leq$ is well-ordered. \hfill \diamond

On every well-ordered set we have the

**Theorem A.1 (Induction principle).** Let $K$ be well ordered and let $S(k)$ be a statement for arbitrary $k \in K$. Then, if $A(l)$ true for all $l < k$ implies $A(k)$ true, then $A(k)$ is true for all $k \in K$.

**Proof.** Otherwise the set of all $k$ for which $A(k)$ is false had a least element $k_0$. But by our choice of $k_0$, $A(l)$ holds for all $l < k_0$ and thus for $k_0$ contradicting our assumption. \hfill \Box

The induction principle also shows that in a well-ordered set functions $f$ can be defined recursively, that is, by a function $\varphi$ which computes the value of $f(k)$ from the values $f(l)$ for all $l < k$. Indeed, the induction principle implies that on the set $M_k = \{l \in K | l < k\}$ there is at most one such function $f_k$. Since $k$ is arbitrary, $f$ is unique. In case of the integers existence of $f_k$ is also clear provided $f(1)$ is given. In general, one can prove existence provided $f_k$ is given for some $k$ but we will not need this.

If $\mathcal{P}$ is partially ordered, then every totally ordered subset is also called a chain. If $\mathcal{Q} \subseteq \mathcal{P}$, then an element $M \in \mathcal{P}$ satisfying $A \preceq M$ for all $A \in \mathcal{Q}$ is called an upper bound.

**Example.** Let $\mathcal{P}(X)$ as before. Then a collection of subsets $\{A_n\}_{n \in \mathbb{N}} \subseteq \mathcal{P}(X)$ satisfying $A_n \subseteq A_{n+1}$ is a chain. The set $\bigcup_n A_n$ is an upper bound. \hfill \diamond

An element $M \in \mathcal{P}$ for which $M \preceq A$ for some $A \in \mathcal{P}$ is only possible if $M = A$ is called a maximal element.

**Theorem A.2 (Zorn’s lemma).** Every partially ordered set in which every chain has an upper bound contains at least one maximal element.

**Proof.** Suppose it were false. Then to every chain $C$ we can assign an element $m(C)$ such that $m(C) \succ x$ for all $x \in C$ (here we use the axiom of choice). We call a chain $C$ distinguished if it is well-ordered and if for every segment $C_x = \{y \in C | y \prec x\}$ we have $m(C_x) = x$. We will also regard $C$ as a segment of itself.

Then (since for the least element of $C$ we have $C_x = \emptyset$) every distinguished chain must start like $m(\emptyset) < m(m(\emptyset)) < \cdots$ and given two segments $C, D$ we expect that always one must be a segment of the other.

So let us first prove this claim. Suppose $D$ is not a segment of $C$. Then we need to show $C = D_z$ for some $z$. We start by showing that $x \in C$ implies $x \in D$ and $C_x = D_z$. To see this suppose it were wrong and let
$x$ be the least $x \in C$ for which it fails. Then $y \in K_x$ implies $y \in L$ and hence $K_x \subset L$. Then, since $C_x \neq D$ by assumption, we can find a least $z \in D \setminus C_x$. In fact we must even have $z \succ C_x$ since otherwise we could find a $y \in C_x$ such that $x \succ y \succ z$. But then, using that it holds for $y$, $y \in D$ and $C_y = D_y$ so we get the contradiction $z \in D_y = C_y \subset C_x$. So $z \succ C_x$ and thus also $C_x = D_z$ which in turn shows $x = m(C_x) = m(D_z) = z$ and proves that $x \in C$ implies $x \in D$ and $C_x = D_x$. In particular $C \subset D$ and as before $C = D_z$ for the least $z \in D \setminus C$. This proves the claim.

Now using this claim we see that we can take the union over all distinguished chains to get a maximal distinguished chain $C_{\max}$. But then we could add $m(C_{\max}) \not\in C_{\max}$ to $C_{\max}$ to get a larger distinguished chain $C_{\max} \cup \{m(C_{\max})\}$ contradicting maximality.

We will also frequently use the **cardinality** of sets: Two sets $A$ and $B$ have the same cardinality, written as $|A| = |B|$, if there is a bijection $\varphi : A \to B$. We write $|A| \leq |B|$ if there is an injective map $\varphi : A \to B$. Note that $|A| \leq |B|$ and $|B| \leq |C|$ implies $|A| \leq |C|$. A set $A$ is called infinite if $|A| \geq |\mathbb{N}|$, countable if $|A| \leq |\mathbb{N}|$, and countably infinite if $|A| = |\mathbb{N}|$.

**Theorem A.3** (Schröder–Bernstein). $|A| \leq |B|$ and $|B| \leq |A|$ implies $|A| = |B|$.

**Proof.** Suppose $\varphi : A \to B$ and $\psi : B \to A$ are two injective maps. Now consider sequences $x_n$ defined recursively via $x_{2n+1} = \varphi(x_{2n})$ and $x_{2n+1} = \psi(x_{2n})$. Given a start value $x_0 \in A$ the sequence is uniquely defined but might terminate at a negative integer since our maps are not surjective. In any case, if an element appears in two sequences, the elements to the left and to the right must also be equal (use induction) and hence the two sequences differ only by an index shift. So the ranges of such sequences form a partition for $A \cup B$ and it suffices to find a bijection between elements in one partition. If the sequence stops at an element in $A$ we can take $\varphi$. If the sequence stops at an element in $B$ we can take $\psi^{-1}$. If the sequence is doubly infinite either of the previous choices will do. \qed

**Theorem A.4** (Zerlem). Either $|A| \leq |B|$ or $|B| \leq |A|$.

**Proof.** Consider the set of all bijective functions $\varphi_\alpha : A_\alpha \to B$ with $A_\alpha \subseteq A$. Then we can define a partial ordering via $\varphi_\alpha \preceq \varphi_\beta$ if $A_\alpha \subseteq A_\beta$ and $\varphi_\beta|_{A_\alpha} = \varphi_\alpha$. Then every chain has an upper bound (the unique function defined on the union of all domains) and by Zorn’s lemma there is a maximal element $\varphi_m$. For $\varphi_m$ we have either $A_m = A$ or $\varphi_m(A_m) = B$ since otherwise there is some $x \in A \setminus A_m$ and some $y \in B \setminus f(A_m)$ which could be used to extend $\varphi_m$ to $A_m \cup \{x\}$ by setting $\varphi(x) = y$. But if $A_m = A$ we have $|A| \leq |B|$ and if $\varphi_m(A_m) = B$ we have $|B| \leq |A|$. \qed
The cardinality of the power set $\mathcal{P}(A)$ is strictly larger than the cardinality of $A$.

**Theorem A.5** (Cantor). $|A| < |\mathcal{P}(A)|$.

**Proof.** Suppose there were a bijection $\varphi : A \to \mathcal{P}(A)$. Then, for $B = \{ x \in A | x \notin \varphi(x) \}$ there must be some $y$ such that $B = \varphi(y)$. But $y \in B$ if and only if $y \notin \varphi(y) = B$, a contradiction. □

This innocent looking result also caused some grief when announced by Cantor as it clearly gives a contradiction when applied to the set of all sets (which is fortunately not a legal object in ZFC).

The following result and its corollary will be used to determine the cardinality of unions and products.

**Lemma A.6.** Any infinite set can be written as a disjoint union of countably infinite sets.

**Proof.** Consider collections of disjoint countably infinite subsets. Such collections can be partially ordered by inclusion and hence there is a maximal collection by Zorn’s lemma. If the union of such a maximal collection falls short of the whole set the complement must be finite. Since this finite remainder can be added to a set of the collection we are done. □

**Corollary A.7.** Any infinite set can be written as a disjoint union of two disjoint subsets having the same cardinality as the original set.

**Proof.** By the lemma we can write $A = \bigcup A_\alpha$, where all $A_\alpha$ are countably infinite. Now split $A_\alpha = B_\alpha \cup C_\alpha$ into two disjoint countably infinite sets (map $A$ bijective to $\mathbb{N}$ and the split into even and odd elements). Then the desired splitting is $A = B \cup C$ with $B = \bigcup B_\alpha$ and $C = \bigcup C_\alpha$. □

**Theorem A.8.** Suppose $A$ or $B$ is infinite. Then $|A \cup B| = \max\{|A|, |B|\}$.

**Proof.** Suppose $A$ is infinite and $|B| \leq |A|$. Then $|A| \leq |A \cup B| \leq |A \cup A| = |A|$ by the previous corollary. Here $\cup$ denotes the disjoint union. □

A standard theorem proven in every introductory course is that $\mathbb{N} \times \mathbb{N}$ is countable. The generalization of this result is also true.

**Theorem A.9.** Suppose $A$ is infinite and $B \neq \emptyset$. Then $|A \times B| = \max\{|A|, |B|\}$.

**Proof.** Without loss of generality we can assume $|B| \leq |A|$ (otherwise exchange both sets). Then $|A| \leq |A \times B| \leq |A \times A|$ and it suffices to show $|A \times A| = |A|$. 

We proceed as before and consider the set of all bijective functions \( \varphi_\alpha : A_\alpha \to A_\alpha \times A_\alpha \) with \( A_\alpha \subseteq A \) with the same partial ordering as before. By Zorn’s lemma there is a maximal element \( \varphi_m \). Let \( A_m \) be its domain and let \( A'_m = A \setminus A_m \). We claim that \( |A'_m| < |A_m| \). If not, \( A'_m \) had a subset \( A''_m \) with the same cardinality of \( A_m \) and hence we had a bijection from \( A''_m \to A''_m \times A''_m \) which could be used to extend \( \varphi \). So \( |A'_m| < |A_m| \) and thus \( |A| = |A_m \cup A'_m| = |A_m| \). Since we have shown \( |A_m \times A_m| = |A_m| \) the claim follows. 

\[ \Box \]

**Example.** Note that for \( A = \mathbb{N} \) we have \( |\mathcal{P}(\mathbb{N})| = |\mathbb{R}| \). Indeed, since \( |\mathbb{R}| = |\mathbb{Z} \times [0,1)| = |[0,1)| \) it suffices to show \( |\mathcal{P}(\mathbb{N})| = |[0,1)| \). To this end note that \( \mathcal{P}(\mathbb{N}) \) can be identified with the set of all sequences with values in \( \{0,1\} \) (the value of the sequence at a point tells us whether it is in the corresponding subset). Now every point in \( [0,1) \) can be mapped to such a sequence via its binary expansion. This map is injective but not surjective since a point can have different binary expansions: \( |[0,1)| \leq |\mathcal{P}(\mathbb{N})| \). Conversely, given a sequence \( a_n \in \{0,1\} \) we can map it to the number \( \sum_{n=1}^{\infty} a_n 4^{-n} \). Since this map is again injective (note that we avoid expansions which are eventually 1) we get \( |\mathcal{P}(\mathbb{N})| \leq |[0,1)| \). 

Hence we have \( |\mathbb{N}| < |\mathcal{P}(\mathbb{N})| = |\mathbb{R}| \) (A.1) and the **continuum hypothesis** states that there are no sets whose cardinality lies in between. It was shown by Gödel and Cohen that it, as well as its negation, is consistent with ZFC and hence cannot be decided within this framework.

**Problem A.1.** Show that Zorn’s lemma implies the axiom of choice. (Hint: Consider the set of all partial choice functions defined on a subset.)

**Problem A.2.** Show \( |\mathbb{R}^\mathbb{N}| = |\mathbb{R}| \). (Hint: Without loss we can replace \( \mathbb{R} \) by \( (0,1) \) and identify each \( x \in (0,1) \) with its decimal expansion. Now the digits in a given sequence are indexed by two countable parameters.)
Bibliography


Glossary of notation

\[ AC[a, b] \] \ldots absolutely continuous functions, 369
\[
\arg(z) \quad \ldots \text{argument of } z \in \mathbb{C}; \quad \arg(z) \in (-\pi, \pi], \quad \arg(0) = 0
\]
\[ B_r(x) \] \ldots open ball of radius \( r \) around \( x \), 4
\[ B(X) \] \ldots Banach space of bounded measurable functions
\[ BV[a,b] \] \ldots functions of bounded variation, 366
\[ \mathcal{B} = \mathcal{B}^1 \]
\[ \mathcal{B}^n \] \ldots Borel \( \sigma \)-algebra of \( \mathbb{R}^n \), 247
\[ \mathbb{C} \] \ldots the set of complex numbers
\[ C(U) \] \ldots set of continuous functions from \( U \) to \( \mathbb{C} \)
\[ C_0(U) \] \ldots set of continuous functions vanishing on the boundary \( \partial U \), 29
\[ C_c(U) \] \ldots set of compactly supported continuous functions
\[ C_{\text{per}}[a,b] \] \ldots set of periodic continuous functions (i.e. \( f(a) = f(b) \))
\[ C^k(U) \] \ldots set of \( k \) times continuously differentiable functions
\[ C_{\text{ps}}^k(U) \] \ldots set of smooth functions with at most polynomial growth, 414
\[ C_c^\infty(U) \] \ldots set of compactly supported smooth functions
\[ C(U,Y) \] \ldots set of continuous functions from \( U \) to \( Y \), 435
\[ C^r(U,Y) \] \ldots set of \( r \) times continuously differentiable functions, 440
\[ C^r_c(U,Y) \] \ldots functions in \( C^r \) with derivatives bounded, 446
\[ C^r_c(U,Y) \] \ldots functions in \( C^r \) with compact support, 497
\[ c_0(\mathbb{N}) \] \ldots set of sequences converging to zero, 44
\[ \mathcal{C}(X,Y) \] \ldots set of compact linear operators from \( X \) to \( Y \), 93
\[ C(U,Y) \] \ldots set of compact maps from \( U \) to \( Y \), 488
\[ \text{CP}(f) \] \ldots critical points of \( f \), 465
\[ \text{CS}(K) \] \ldots nonempty convex subsets of \( K \), 479
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<td>critical values of f, 465</td>
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<tr>
<td>$\chi_\Omega(.)$</td>
<td>characteristic function of the set $\Omega$</td>
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<td>$\mathcal{D}(.)$</td>
<td>domain of an operator</td>
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<tr>
<td>$\delta_{n,m}$</td>
<td>Kronecker delta, 46</td>
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<tr>
<td>deg(D, f, y)</td>
<td>mapping degree, 465, 474</td>
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<tr>
<td>det</td>
<td>determinant</td>
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<tr>
<td>dim</td>
<td>dimension of a linear space</td>
</tr>
<tr>
<td>div</td>
<td>divergence</td>
</tr>
<tr>
<td>diam(U)</td>
<td>diameter of a set</td>
</tr>
<tr>
<td>dist(U, V)</td>
<td>distance of two sets</td>
</tr>
<tr>
<td>$D^y(U,Y)$</td>
<td>functions in $C^r(U,Y)$ which do not attain $y$ on the boundary, 465</td>
</tr>
<tr>
<td>$\mathcal{D}_y(U,Y)$</td>
<td>functions in $C(U,Y)$ which do not attain $y$ on the boundary, 489</td>
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<tr>
<td>e</td>
<td>Napier's constant, $e^2 = \exp(z)$</td>
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<td>$dF$</td>
<td>derivative of $F$, 435</td>
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<td>$\mathcal{F}(X,Y)$</td>
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<tr>
<td>$\mathcal{S}(X,Y)$</td>
<td>set of all linear Fredholm operators from $X$ to $Y$, 209</td>
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<tr>
<td>GL(n)</td>
<td>general linear group in $n$ dimensions</td>
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<tr>
<td>$\Gamma(z)$</td>
<td>gamma function, 292</td>
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<tr>
<td>$\mathcal{H}$</td>
<td>a Hilbert space</td>
</tr>
<tr>
<td>hull(.)</td>
<td>convex hull</td>
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<td>$\mathcal{H}(U)$</td>
<td>set of holomorphic functions on a domain $U \subseteq \mathbb{C}$, 463</td>
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<td>$H^0_0(U)$</td>
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<tr>
<td>i</td>
<td>complex unity, $i^2 = -1$</td>
</tr>
<tr>
<td>Im(.)</td>
<td>imaginary part of a complex number</td>
</tr>
<tr>
<td>inf</td>
<td>infimum</td>
</tr>
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<td>$J_f(x)$</td>
<td>= $\text{det } df(x)$ Jacobi determinant of $f$ at $x$, 465</td>
</tr>
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<td>Ker(A)</td>
<td>kernel of an operator $A$, 58</td>
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<td>$\lambda^n$</td>
<td>Lebesgue measure in $\mathbb{R}^n$, 286</td>
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<td>= $\mathcal{L}(X,X)$</td>
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<td>$\mathcal{L}^2_{\text{cont}}(I)$</td>
<td>space of continuous square integrable functions, 54</td>
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<td>$\ell^p(\mathbb{N})$</td>
<td>Banach space of $p$ summable sequences, 43</td>
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<td>max</td>
<td>maximum</td>
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<td>finite complex measures on $X$, 345</td>
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<tr>
<td>$\mathcal{M}_{\text{reg}}(X)$</td>
<td>finite regular complex measures on $X$, 347</td>
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<tr>
<td>$\mathcal{M}(f)$</td>
<td>Hardy–Littlewood maximal function, 429</td>
</tr>
<tr>
<td>$\mathbb{N}$</td>
<td>the set of positive integers</td>
</tr>
<tr>
<td>$\mathbb{N}_0$</td>
<td>= $\mathbb{N} \cup {0}$</td>
</tr>
<tr>
<td>$n(\gamma, z_0)$</td>
<td>winding number</td>
</tr>
</tbody>
</table>
O(·) \ldots \text{Landau symbol, } f = O(g) \text{ iff } \limsup_{x \to x_0} |f(x)/g(x)| < \infty

o(·) \ldots \text{Landau symbol, } f = o(g) \text{ iff } \lim_{x \to x_0} |f(x)/g(x)| = 0

\mathbb{Q} \ldots \text{the set of rational numbers}

\mathbb{R} \ldots \text{the set of real numbers}

\text{RV}(f) \ldots \text{regular values of } f, 465

\text{Ran}(A) \ldots \text{range of an operator } A, 58

\text{Re}(·) \ldots \text{real part of a complex number}

\mathcal{R}(I,X) \ldots \text{set of regulated functions}, 218

\sigma_{n-1} \ldots \text{surface measure on } S^{n-1}, 291

S^{n-1} = \{x \in \mathbb{R}^n | |x| = 1\} \text{ unit sphere in } \mathbb{R}^n

S_n = n\pi^{n/2}/\Gamma\left(\frac{n}{2} + 1\right), \text{ surface area of the unit sphere in } \mathbb{R}^n, 290

\text{sign}(z) = z/|z| \text{ for } z \neq 0 \text{ and } 1 \text{ for } z = 0; \text{ complex sign function}

S(I,X) \ldots \text{simple functions } f : I \to X, 443

\mathcal{S}^n \ldots \text{semialgebra of rectangles in } \mathbb{R}^n, 239

\mathcal{S}^n \ldots \text{algebra of finite unions of rectangles in } \mathbb{R}^n, 239

\mathcal{S}(\mathbb{R}^n) \ldots \text{Schwartz space, 389}

\text{sup} \ldots \text{supremum}

\text{supp}(f) \ldots \text{support of a function } f, 13

\text{supp}(\mu) \ldots \text{support of a measure } \mu, 257

\text{span}(M) \ldots \text{set of finite linear combinations from } M, 45

W^{k,p}(U) \ldots \text{Sobolev space, 329}

W_0^{k,p}(U) \ldots \text{Sobolev space, 330}

V_n = \pi^{n/2}/\Gamma\left(\frac{n}{2} + 1\right), \text{ volume of the unit ball in } \mathbb{R}^n, 291

\mathbb{Z} \ldots \text{the set of integers}

I \ldots \text{identity operator}

\sqrt{z} \ldots \text{square root of } z \text{ with branch cut along } (-\infty,0)

z^* \ldots \text{complex conjugation}

A^* \ldots \text{adjoint of } A, 81

\overline{A} \ldots \text{closure of } A, 131

\hat{f} = \mathcal{F}f, \text{ Fourier transform/coefficients of } f, 88, 387

\check{f} = \mathcal{F}^{-1}f, \text{ inverse Fourier transform of } f, 390

|x| = \sqrt{\sum_{j=1}^{n} |x_j|^2} \text{ Euclidean norm in } \mathbb{R}^n \text{ or } \mathbb{C}^n

|\Omega| \ldots \text{Lebesgue measure of a Borel set } \Omega

\|\cdot\| \ldots \text{norm, 51}

\|\cdot\|_p \ldots \text{norm in the Banach space } \ell^p \text{ and } L^p, 43, 303

\langle \cdot, \cdot \rangle \ldots \text{scalar product in } \mathcal{H}, 51

\oplus \ldots \text{orthogonal sum of vector spaces or operators, 85}

\otimes \ldots \text{tensor product, 86}

\cup \ldots \text{union of disjoint sets, 240}

\lfloor x \rfloor = \max\{n \in \mathbb{Z} | n \leq x\} \text{ floor function}

\lceil x \rceil = \min\{n \in \mathbb{Z} | n \geq x\}, \text{ ceiling function}

\partial = (\partial_1 f, \ldots, \partial_n f) \text{ gradient in } \mathbb{R}^n

\partial_\alpha \ldots \text{partial derivative in multi-index notation, 68}

\partial_x F(x,y) \ldots \text{partial derivative with respect to } x, 439
\( \partial U = \overline{U} \setminus U^o \) boundary of the set \( U \), 4
\( \overline{U} \) closure of the set \( U \), 7
\( U^o \) interior of the set \( U \), 7
\( M^\perp \) orthogonal complement, 78
\( (\lambda_1, \lambda_2) = \{ \lambda \in \mathbb{R} \mid \lambda_1 < \lambda < \lambda_2 \} \), open interval
\( [\lambda_1, \lambda_2] = \{ \lambda \in \mathbb{R} \mid \lambda_1 \leq \lambda \leq \lambda_2 \} \), closed interval
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\( x_n \rightharpoonup x \) weak convergence, 149
\( x_n \rightharpoonup^* x \) weak-\(^*\) convergence, 156
\( A_n \to A \) norm convergence
\( A_n \rightharpoonup A \) strong convergence, 154
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