

Ergodic Theory - A Summary

Ergodic Theory is the study of dynamical systems by means of invariant measures.

Definition: A measure is T -invariant if $\mu(T^{-1}(A)) = \mu(A)$ for every set A in the algebra of μ -measurable sets.

Equivalently: for every measurable function $f : X \rightarrow \mathbb{R}$,

$$\int_X f d\mu = \int_X f \circ T d\mu.$$

A measure is called **non-atomic** if $\mu(\{x\}) = 0$ for every $x \in X$.

A measure is called **absolutely continuous** (w.r.t. Lebesgue m) if $m(A) = 0$ implies $\mu(A) = 0$.

Existence of Invariant Measures

The existence of invariant measures is (usually) guaranteed by the:

Theorem of Krylov-Bogol'ubov: If $T : X \rightarrow X$ is a continuous map on a nonempty compact metric space X , then the set of invariant probability measures $\mathcal{M}(X, T) \neq \emptyset$.

Examples:

- ▶ If $T^p(x) = x$, then the equidistribution $\frac{1}{p} \sum_{j=0}^{p-1} \delta_{T^j(x)}$ is an invariant measure.
- ▶ If $P_T f(x) = \sum_{Ty=x} \frac{f(y)}{|DT(y)|}$ is the transfer operator w.r.t. Lebesgue measure, then

$$d\mu = f dx \text{ is invariant if } P_T f = f.$$

Ergodicity

Definition: A measure μ for a dynamical system (X, T) is **ergodic** if

$$\mu(A) = 0 \text{ or } \mu(A^c) = 0$$

for every measurable set $A \subset X$ such that $T^{-1}(A) = A \bmod \mu$.

This says that (mod μ), the space doesn't decompose into parts that don't communicate with each other.

Equivalent expression for ergodicity are:

- ▶ The only T -invariant functions $\psi \in L^1(\mu)$ i.e., $\psi \circ T = \psi$ μ -a.e., are constant μ -a.e.
- ▶ $\frac{1}{n} \sum_{j=0}^{n-1} \mu(A \cap T^j B) - \mu(A)\mu(B) \rightarrow 0$ for all measurable sets $A, B \subset X$.

Ergodicity

Examples:

- ▶ The **doubling map** $T : \mathbb{S}^1 \rightarrow \mathbb{S}^1$, $x \mapsto 2x \bmod 1$ preserves Lebesgue measure m , and it is ergodic. However, $\frac{1}{2}m + \frac{1}{2}\delta_0$ is invariant but not ergodic.
- ▶ The **Gauß map** $G : [0, 1) \rightarrow (0, 1]$, $x \mapsto \frac{1}{x} - \lfloor \frac{1}{x} \rfloor$ preserves the measure $d\mu = \frac{1}{\log 2} \frac{dx}{1+x}$, and it is ergodic (Folklore Theorem).
- ▶ **Circle rotations**

$$R_\alpha : \mathbb{S}^1 \rightarrow \mathbb{S}^1, \quad R_\alpha(x) = x + \alpha \bmod 1.$$

preserve Lebesgue measure.

- ▶ If $\alpha \in \mathbb{Q}$, then every orbit is periodic. Lebesgue measure is not ergodic.
- ▶ If $\alpha \notin \mathbb{Q}$, then every orbit is dense in \mathbb{S}^1 . Lebesgue measure is ergodic; in fact it is the only R_α -invariant probability measure.

Unique Ergodicity

Definition: A system (X, T) is called **uniquely ergodic** if there is exactly one T -invariant **probability** measure.

This measure is automatically ergodic.

Oxtoby's Theorem: Let X be a compact space and $T : X \rightarrow X$ continuous. A transformation (X, T) is uniquely ergodic if and only if, for every continuous function ψ and **every point** $x \in X$, the Birkhoff averages

$$\frac{1}{n} \sum_{i=0}^{n-1} \psi \circ T^i(x)$$

converge **uniformly** to a constant.

Birkhoff's Ergodic Theorem

Birkhoff's Ergodic Theorem formalizes a frequent observation in physics:

Space Average = Time Average (for typical points).

This is expressed in:

Birkhoff's Ergodic Theorem: Let μ be a probability measure and $\psi \in L^1(\mu)$. Then the **ergodic average**

$$\psi^*(x) := \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \psi \circ T^i(x)$$

exists μ -a.e., and ψ^* is T -invariant, i.e., $\psi^* \circ T = \psi^*$ μ -a.e.

If in addition μ is **ergodic** then

$$\psi^* = \int_X \psi \, d\mu \quad \mu\text{-a.e.}$$

Absolutely Continuous Measures

Definition: A measure μ is called **absolutely continuous** w.r.t. the measure ν (notation: $\mu \ll \nu$) if $\nu(A) = 0$ implies $\mu(A) = 0$. If both $\mu \ll \nu$ and $\nu \ll \mu$, then μ and ν are called **equivalent**.

Theorem of Radon-Nikodym: If μ is a probability measure and $\mu \ll \nu$ then there is a function $h \in L^1(\nu)$ (called **Radon-Nikodym derivative** or **density**) such that $\mu(A) = \int_A h(x) d\nu(x)$ for every measurable set A .

Notation: $h(x) = \frac{d\mu(x)}{d\nu(x)}$.

Suppose that $\mu \ll \nu$ are both T -invariant probability measures, with a common σ -algebra \mathcal{B} of measurable sets. If ν is ergodic, then $\mu = \nu$.

Poincaré Recurrence

The Poincaré Recurrence Theorem: If (X, T, μ) is a measure preserving system with $\mu(X) = 1$, then for every measurable set $Y \subset X$ of positive measure, μ -a.e. $y \in Y$ returns to Y , i.e., the **first return time** to Y : $\tau_Y(y) < \infty$ μ -a.e.

Definition: A system (X, T, \mathcal{B}, μ) is called **conservative** if for every set $A \in \mathcal{B}$ with $\mu(A) > 0$, there is $n \geq 1$ such that $\mu(T^n(A) \cap A) > 0$. The Poincaré Recurrence Theorem thus implies that probability measure preserving systems are conservative.

If not conservative, then the system is called **dissipative**. It is called **totally dissipative** if for every set $A \in \mathcal{B}$,

$$\mu(\{x \in A : T^n(x) \in A \text{ infinitely often}\}) = 0.$$

Kac' Lemma

Kac's Lemma quantifies the expected value of the first return time τ_Y to $Y \subset X$.

Kac' Lemma: Let (X, T) preserve an ergodic measure μ . Take $Y \subset X$ measurable such that $\mu(Y) > 0$, and let $\tau : Y \rightarrow \mathbb{N}$ be the first return time to Y . Take $Y \subset X$ measurable such that $\mu(Y) > 0$. Then

$$\mathbb{E}_\mu(\tau_Y) = \int_Y \tau_Y d\mu = \sum_{n \geq 1} n\mu(Y_n) = \mu(X)$$

for $Y_n := \{y \in Y : \tau(y) = n\}$.

Induced Systems

Proposition: Let (X, \mathcal{B}, T, μ) be an ergodic dynamical system and $Y \in \mathcal{B}$ a set with $\mu(Y) > 0$. Let $T_Y = T^{\tau_Y}$ be the first return map to Y .

If μ is T -invariant, then $\nu(A) := \frac{1}{\mu(Y)}\mu(A \cap Y)$ is T_Y -invariant.

Conversely, if ν is T_Y -invariant, and

$$\Lambda := \int_Y \tau(y) d\nu < \infty,$$

then

$$\mu(A) = \frac{1}{\Lambda} \sum_{j=1}^{\infty} \nu(T^{-j}(A) \cap \{y \in Y : \tau(y) \geq j\})$$

is a T -invariant probability measure. Moreover μ is ergodic for T if and only if ν is ergodic for T_Y .

Isomorphic Systems

Definition: Two measure preserving dynamical systems (X, \mathcal{B}, T, μ) and (Y, \mathcal{C}, S, ν) are called **isomorphic** if there are $X' \in \mathcal{B}$, $Y' \in \mathcal{C}$ and $\phi : Y' \rightarrow X'$ such that

- ▶ $\mu(X') = 1, \nu(Y') = 1$;
- ▶ $\phi : Y' \rightarrow X'$ is a bi-measurable bijection;
- ▶ ϕ is measure preserving: $\nu(\phi^{-1}(B)) = \mu(B)$ for all $B \in \mathcal{B}$.
- ▶ $\phi \circ S = T \circ \phi$.

That is, the below diagram commutes, and $\phi : Y \rightarrow X$ is one-to-one almost everywhere.

$$\begin{array}{ccc} (Y, \mathcal{C}, \nu) & \xrightarrow{S} & (Y, \mathcal{C}, \nu) \\ \phi \downarrow & & \downarrow \phi \\ (X, \mathcal{B}, \mu) & \xrightarrow{T} & (X, \mathcal{B}, \mu) \end{array}$$

The Bernoulli Property

Definition: Let (X, \mathcal{B}, μ, T) be a measure preserving dynamical system.

1. If T is invertible, then the system is called **Bernoulli** if it is isomorphic to a 2-sided Bernoulli shift.
2. If T is non-invertible, then the system is called **one-sided Bernoulli** if it is isomorphic to a 1-sided Bernoulli shift.
3. If T is non-invertible, then the system is called **Bernoulli** if its natural extension is isomorphic to a 2-sided Bernoulli shift.

Mixing

Definition: A probability measure preserving dynamical system (X, \mathcal{B}, μ, T) is **mixing** (or **strong mixing**) if

$$\mu(T^{-n}(A) \cap B) \rightarrow \mu(A)\mu(B) \text{ as } n \rightarrow \infty$$

for every $A, B \in \mathcal{B}$.

This says that the “events” A and B are asymptotically independent.

Equivalently, a probability preserving dynamical system (X, \mathcal{B}, T, μ) is mixing if and only if

$$\int_X f \circ T^n(x) \cdot \overline{g(x)} \, d\mu \rightarrow \int_X f(x) \, d\mu \cdot \int_X \overline{g(x)} \, d\mu \text{ as } n \rightarrow \infty$$

for all $f, g \in L^2(\mu)$

Measure-Theoretic Entropy

Given a finite partition \mathcal{P} of a probability space (X, μ) , let

$$H_\mu(\mathcal{P}) = - \sum_{P \in \mathcal{P}} \mu(P) \log(\mu(P)). \quad (1)$$

For a T -invariant probability measure μ on (X, \mathcal{B}, T) , and a partition \mathcal{P} , define the **entropy of μ w.r.t. \mathcal{P}** as

$$h_\mu(T, \mathcal{P}) = \lim_{n \rightarrow \infty} \frac{1}{n} H_\mu\left(\bigvee_{k=0}^{n-1} T^{-k} \mathcal{P}\right). \quad (2)$$

Finally, the **measure theoretic entropy** of μ is

$$h_\mu(T) = \sup\{h_\mu(T, \mathcal{P}) : \mathcal{P} \text{ is a finite partition of } X\}. \quad (3)$$

Measure-Theoretic Entropy

Remarks concerning measure-theoretic entropy

- ▶ The existence of the limit in (2) depends on:

Fekete's Lemma: If $(a_n)_{n \geq 1}$ is subadditive, then

$$\lim_{n \rightarrow \infty} \frac{a_n}{n} = \inf_{q \geq 1} \frac{a_q}{q}.$$

- ▶ By Sinaĭ's Theorem, instead of taking the supremum over all partitions, it suffices to take a **generating partition**.
- ▶ Entropy is preserved under isomorphism (and is non-increasing under taking measure-theoretical factors).
- ▶ The entropy of the (p_1, \dots, p_n) -Bernoulli shift (both one-sided and two-sided) is $h(\mu_p) = -\sum_i p_i \log p_i$.

The Shannon-Breiman-McMillan Theorem

The Shannon-Breiman-McMillan Theorem uses entropy to measure how large sets in the n -th joint \mathcal{P}_n are. Typically, they decrease exponentially and the exponential rate is exactly the measure-theoretical entropy.

Shannon-McMillan-Breiman Theorem: Let (X, \mathcal{B}, μ, T) be a measure-preserving transformation and \mathcal{P} a (countable or finite) partition with $H(\mathcal{P}) < \infty$. Let $\mathcal{P}_n = \bigvee_{k=0}^{n-1} T^{-k}(\mathcal{P})$ and $\mathcal{P}_n(x)$ the element of \mathcal{P}_n containing x . Then

$$-\lim_{n \rightarrow \infty} \frac{1}{n} \log \mu(\mathcal{P}_n(x)) = h(\mathcal{P}, T) \quad \mu\text{-a.e.}$$

Topological Entropy

Definition: Topological entropy can be defined as the exponential growth rate of the

- ▶ (Adler-Konheim-McAndrew) minimal cardinality of subcovers of joints:

$$h_{top}(T) = \lim_{\varepsilon \rightarrow 0} \sup_{\mathcal{U}} \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathcal{N}(\mathcal{U}^n), \quad (4)$$

where the supremum is taken over all open ε -covers \mathcal{U} .

- ▶ (Bowen) maximal cardinality of n, ε -separated sets:

$$h_{top}(T) = \lim_{\varepsilon \rightarrow 0} \limsup_{n \rightarrow \infty} \frac{1}{n} \log s_n(\varepsilon). \quad (5)$$

- ▶ (Bowen) minimal cardinality of n, ε -spanning sets:

$$h_{top}(T) = \lim_{\varepsilon \rightarrow 0} \limsup_{n \rightarrow \infty} \frac{1}{n} \log r_n(\varepsilon). \quad (6)$$

Topological Entropy

For maps T on the interval we have:

Theorem of Szlenk & Misiurewicz Let $T : [0, 1] \rightarrow [0, 1]$ has finitely many laps. Then

$$\begin{aligned}h_{top}(T) &= \lim_{n \rightarrow \infty} \frac{1}{n} \log \ell(T^n) \\&= \limsup_{n \rightarrow \infty} \frac{1}{n} \log \#\{\text{clusters of } n\text{-periodic points}\} \\&= \max\{0, \lim_{n \rightarrow \infty} \frac{1}{n} \log \text{Var}(T^n)\}.\end{aligned}$$

where two n -periodic points are in the same cluster if they belong to the same lap of T^n .

The Variational Principle

Topological and measure-theoretical entropy are related by the **Variational Principle** which says that (for continuous map on compact metric space)

$$h_{top}(T) = \sup\{h_{\mu}(T) : \mu \text{ is } T\text{-invariant probability measure}\}$$

If μ is such that $h_{\mu}(T) = h_{top}(T)$, then μ is called a **measure of maximal entropy**.

If there is a **unique** measure of maximal entropy μ_{max} , then (X, T) is called **intrinsically ergodic**. In this case, μ_{max} is ergodic.

The Variational Principle

- ▶ The full shift on N symbols (one-sided or two-sided) has entropy $\log N$, and the measure of maximal entropy is then $(\frac{1}{N}, \dots, \frac{1}{N})$ -Bernoulli measure.
- ▶ A subshift of finite type with transition matrix A has the logarithm of the leading eigenvalue as entropy. The Parry measure is the measure of maximal entropy.
- ▶ Interval maps with constant slope $\pm s$ for $s > 1$ have an absolutely continuous measure, which is the measure of maximal entropy.

Mixing

Review of ergodic properties:

- ▶ **Bernoulli** if it is isomorphic to a two-sided Bernoulli shift.
- ▶ **strong mixing** if for all $A, B \in \mathcal{B}$

$$\mu(T^{-n}(A) \cap B) - \mu(A)\mu(B) \rightarrow 0.$$

- ▶ **weak mixing** if for all $A, B \in \mathcal{B}$ the average

$$\frac{1}{n} \sum_{i=0}^{n-1} |\mu(T^{-i}(A) \cap B) - \mu(A)\mu(B)| \rightarrow 0.$$

- ▶ **ergodic** if $T^{-1}(A) = A \bmod \mu$ implies $\mu(A) = 0$ or $\mu(A^c) = 0$.
- ▶ **conservative** if for all $A \in \mathcal{B}$ with $\mu(A) > 0$ there is $n \geq 1$ such that $\mu(T^n(A) \cap A) > 0$.

Bernoulli - Mixing - Ergodic - Conservative

Theorem We have the implications:

Bernoulli \Rightarrow mixing \Rightarrow weak mixing \Rightarrow ergodic \Rightarrow conservative.

None of the reverse implications holds.

Classifying Systems up to Isomorphism

Definition: Two dynamical systems (X, T) and (Y, S) are **topologically conjugate** if there is a homeomorphism $\phi : Y \rightarrow X$ such that $\phi \circ S = T \circ \phi$.

We can classify dynamical systems up to conjugacy, and measure preserving systems up to isomorphism.

The one does not imply the other or vice versa:

- ▶ The doubling map (with Lebesgue measure) and the one-sided $(\frac{1}{2}, \frac{1}{2})$ -Bernoulli shift are isomorphic. They are not conjugate (note: they are defined on **non-homeomorphic** spaces).
- ▶ The doubling map with Lebesgue measure is conjugate to the doubling map with δ_0 , but they are not isomorphic. (Conjugacies are topological objects; they don't care about measures).

Classifying Systems up to Isomorphism

The following properties and quantities are preserved under isomorphisms.

- ▶ ergodicity, weak and strong mixing, the Bernoulli property
- ▶ measure-theoretic entropy. Moreover,

Ornstein's Theorem: Two **two-sided** Bernoulli shifts (X, μ_p, σ) and $(X', \mu_{p'}, \sigma)$ are isomorphic if and only if $h(\mu_p) = h(\mu_{p'})$.

This fails for one-sided Bernoulli shifts.

- ▶ eigenvalues of the Koopman operator $U_T f = f \circ T$. Moreover,

Theorem (Halmos & Von Neumann) Two measure-preserving dynamical systems with pure point spectra are isomorphic if and only if their eigenvalues are the same.

This fails without the assumption of pure point spectrum.

Toral Automorphisms

Definition: A **toral automorphism** $T : \mathbb{T}^d \rightarrow \mathbb{T}^d$ is an invertible linear map on the (d -dimensional) torus \mathbb{T}^d . Each such T is of the form $T_A(x) = Ax \pmod{1}$, where the matrix A satisfies:

- ▶ A is an integer matrix with $\det(A) = \pm 1$;
- ▶ To avoid degenerate examples including $A = Id$, we assume that A is **primitive**, i.e., A^n is strictly positive for some $n \geq 1$.
- ▶ If the eigenvalues of A are not on the unit circle, then the toral automorphism is called **hyperbolic**.

Toral Automorphisms

Properties of the toral automorphisms T_A :

- ▶ A preserves the integer lattice \mathbb{Z}^d , so T_A is well-defined and continuous.
- ▶ $\det(A) = \pm 1$, so **Lebesgue measure m is preserved** (both by A and T_A). Also A and T_A are invertible, and A^{-1} is still an integer matrix (so T_A^{-1} is well-defined and continuous too).
- ▶ One can show that Lebesgue measure is ergodic if and only if A has no eigenvalues that are roots of unity.
- ▶ Hyperbolic toral automorphisms have a Markov partition w.r.t. which the symbolic dynamics is a **subshift of finite type**, and Lebesgue measure is the measure of maximal entropy.
- ▶ Hyperbolic toral automorphisms are mixing w.r.t. Lebesgue.