

Nonlinear Optimization
Exercise session 6

38. (Frobenius norm) Let be $A = (a_{ij}) \in \mathbb{R}^{n \times n}$.

(a) Prove that

$$\|A\|_F := \left(\sum_{i,j=1}^n a_{ij}^2 \right)^{1/2}$$

defines a matrix norm and that this matrix norm is not induced by a vector norm.

(b) We denote by

$$\text{Spur}(A) := \sum_{i=1}^n a_{ii}$$

the trace of the matrix A . Prove that

$$\text{Spur}(AB) = \text{Spur}(BA) \quad \forall B \in \mathbb{R}^{n \times n},$$

$$\text{Spur}(S^{-1}AS) = \text{Spur}(A) \quad \forall S \in \mathbb{R}^{n \times n} \text{ with } S \text{ regular,}$$

and

$$\|A\|_F^2 = \text{Spur}(A^T A).$$

(c) If $\{v^1, \dots, v^n\}$ is an orthonormal basis of \mathbb{R}^n , then it holds

$$\|A\|_F^2 = \sum_{i=1}^n \|Av^i\|^2.$$

(4 points)

39. Let be $u, v \in \mathbb{R}^n$. Show that the following statements are true:

(a) $\|uv^T\| = \|uv^T\|_F = \|u\| \|v\|.$

(b) $\det(I + uv^T) = 1 + u^T v.$

(4 points)

40. (Lemma 9.10 in the lecture course)

Let be $s, y \in \mathbb{R}^n$ with $s \neq 0$. Prove that $s^T y > 0$ if and only if there exists a symmetric and positive definite matrix $Q \in \mathbb{R}^{n \times n}$ such that $Qs = y$.

Hint. " \Rightarrow " For $v := \sqrt{\frac{y^T s}{s^T s}} s$, prove that the matrix

$$Q := \left(I + \frac{1}{v^T v} (y - v)v^T \right) \left(I + \frac{1}{v^T v} (y - v)v^T \right)^T$$

has the desired properties. (4 points)

41. Let $H \in \mathbb{R}^{n \times n}$ be a regular matrix and $u, v \in \mathbb{R}^n$. Prove that the matrix $H + uv^T$ is regular if and only if $1 + v^T H^{-1} u \neq 0$. Show that under these assumptions the so-called *Sherman-Morrison formula* holds:

$$(H + uv^T)^{-1} = \left(I - \frac{1}{1 + v^T H^{-1} u} H^{-1} uv^T \right) H^{-1}.$$

Hint. Use Exercise 39(b). (3 points)

42. Let be $s, y \in \mathbb{R}^n$ with $s^T y > 0$ and $H \in \mathbb{R}^{n \times n}$ a symmetric and positive definite matrix. The *direct BFGS update formula* reads

$$H_+^{BFGS} := H + \frac{1}{s^T y} yy^T - \frac{1}{s^T H s} H s s^T H.$$

Show that $B = H^{-1}$ implies $B_+^{BFGS} = (H_+^{BFGS})^{-1}$, whereby B_+^{BFGS} is given by the inverse BFGS update formula (Theorem 9.12 in the lecture course). (3 points)

43. Let $M \in \mathbb{R}^{n \times n}$ be a regular matrix and $\{M_k\}_{k \geq 0} \in \mathbb{R}^{n \times n}$ a sequence of matrices which converges to M as $k \rightarrow +\infty$. Show that there exists $k_0 \geq 0$ such that M_k is regular for all $k \geq k_0$, and that the sequence $\{M_k^{-1}\}_{k \geq k_0}$ converges to M^{-1} . (3 points)

44. Consider the quadratic optimization problem

$$(P) \quad \begin{aligned} \min \quad & f(x) := \gamma + c^T x + \frac{1}{2} x^T Q x, \\ \text{s.t.} \quad & h(x) := b^T x = 0 \end{aligned}$$

with $Q \in \mathbb{R}^{n \times n}$ a symmetric and positive definite matrix, $b, c \in \mathbb{R}^n$, $b \neq 0$, and $\gamma \in \mathbb{R}$. For given $\alpha > 0$, find the minimum $x^*(\alpha)$ of the penalty function

$$P(x; \alpha) := f(x) + \frac{\alpha}{2} (h(x))^2,$$

determine $x^* := \lim_{\alpha \rightarrow +\infty} x^*(\alpha)$, and prove that x^* is the unique optimal solution of the optimization problem (P).

Hint. Use the *Sherman-Morrison formula*. (3 points)

45. Implement the local inverse BFGS algorithm in MATLAB. To do this, create the file `quasi_newton_method.m` with

```
function [X, Iter] = quasi_newton_method(fhandle, x0, B0, kmax, eps)
```

as first line. Here, `func` denotes a function handle, `x0` the starting vector, `B0` the starting matrix, `kmax` the maximal number of allowed iterations, and `epsilon` the parameter for the stopping criterion. A matrix $X = [x_0, x_1, x_2, \dots]$ containing the iteration history and the number of performed iterations `Iter` should be returned.

The implemented algorithm should be tested for `epsilon`= 10^{-6} , `kmax`=500 and `B0` the identity matrix on the following functions and starting values:

- (a) $f(x_1, x_2) = (1 - x_1)^2 + 100(x_2 - x_1^2)^2$ (Rosenbrock function) for $\mathbf{x}_0 = (-1.2, 1)^T$.
- (b) $f(x_1, x_2, x_3, x_4) = \sum_{i=1}^4 \left(4 - \sum_{j=1}^4 \cos x_j + i(1 - \cos x_i) - \sin x_i \right)^2$ (trigonometric function) for $\mathbf{x}_0 = \left(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4} \right)^T$.
- (c) $f(x_1, x_2) = (x_1 - 10^6)^2 + (x_2 - 2 \cdot 10^6)^2 + (x_1 x_2 - 2)^2$ (Brown function) for $\mathbf{x}_0 = (1, 1)^T$.
- (d) $f(x_1, x_2, x_3, x_4) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2 + 90(x_4 - x_3^2)^2 + (1 - x_3)^2 + 10(x_2 + x_4 - 2)^2 + \frac{1}{10}(x_2 - x_4)^2$ (Wood function) for $\mathbf{x}_0 = (-3, -1, -3, -1)^T$.

The generated files should be sent by January 17, 2020 to `radu.bot@univie.ac.at`.