## Scientific Computing 2

Summer term 2017
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## Sheet 1

Submission on Thursday, 27.4.2017.

Exercise 1. (multivariate quadratic polynomial)
Consider the function $f: \mathbb{R}^{n} \rightarrow \mathbb{R}, \quad f(x)=\frac{1}{2} x^{T} H x+b^{T} x+c$.
a) Calculate the gradient $\nabla f$ and Hessian $\nabla^{2} f$ for an arbitrary $H \in \mathbb{R}^{n \times n}, b \in \mathbb{R}^{n}$, $c \in \mathbb{R}$ and for symmetric $H$.
b) Show that $H$ can be replaced by a symmetric matrix without changing the objective function $f$.
c) Show that if $H$ is positive definite, then $f$ is strictly convex and $\lim _{\|x\| \rightarrow \infty} f(x)=\infty$.
d) Rewrite the function

$$
g\left(x_{1}, x_{2}\right)=5 x_{1}^{2}+5 x_{2}^{2}+8 x_{1} x_{2}-4 x_{1}-2 x_{2}+3
$$

as $g(x)=\frac{1}{2} x^{T} H x+b^{T} x+c$ with symmetric $H \in \mathbb{R}^{n \times n}$. Is $H$ positive definite? If yes, what does this imply for existence and uniqueness of minima? Calculate the global minimum of $g$.
(6 points)
Exercise 2. (saddle point property)
Consider the minimization problem from the lecture:

$$
\begin{array}{r}
\min f\left(x_{1}, x_{2}\right)=x_{1}^{2}+x_{2}^{2} \\
x_{2} \leq x_{1}-1
\end{array}
$$

Derive the corresponding Lagrangian $L\left(x_{1}, x_{2}, \lambda\right)$. Show that the solution $\left(\overline{x_{1}}, \overline{x_{2}}, \bar{\lambda}\right)$ satisfies the saddle point property

$$
\begin{equation*}
L\left(\overline{x_{1}}, \overline{x_{2}}, \lambda\right) \leq L\left(\overline{x_{1}}, \overline{x_{2}}, \bar{\lambda}\right) \leq L\left(x_{1}, x_{2}, \bar{\lambda}\right) \quad \forall \lambda \geq 0, \forall x_{1}, x_{2} \tag{4points}
\end{equation*}
$$

Exercise 3. (3-norm)
Calculate all minima of $f\left(x_{1}, x_{2}\right)=x_{1}^{3}+x_{2}^{3}$ on the unit circle, using the Lagrange formalism.
(4 points)
Exercise 4. (parabolic graph)
Find the point lying on the parabolic graph $x_{1}^{2}-4 x_{2}=0$ which minimizes the euclidean distance to $(0,1)$. Try to solve this problem using Gaussian elimination.

- Eliminate $x_{1}^{2}$. What happens, and why?
- Eliminate $x_{2}$ instead.

Afterwards, use the Lagrange formalism to solve this minimization problem.

