# Re-exam Continuous Optimization 

22 February 2021, 14.00-17.00

The exam consists of 5 questions. In total you can obtain 90 points. The final grade is $1+$ \#points/10 rounded to the nearest integer.

This is an open-book exam. It is NOT allowed to discuss with anyone else. If you have any questions regarding the exam, or technical questions regarding uploading of your answer, please contact David de Laat at d.delaat@tudelft.nl.

Please review the instructions posted on the announcement page for the course. The most important points are repeated below:

- Write your answers by hand and start each exercise on a new sheet.
- On your first answer sheet, you should write the following statement: "This exam will be solely undertaken by myself, without any assistance from others, and without use of sources other than those allowed."
- When scanning your work place your student ID on the first page. If you do not have a student ID please use some other form of identification but in that case make sure only your name and photo are visible.
- Scan your work and submit it as one single pdf-file at 17.00 .
- You should keep an eye on your email from 17.00-17.30 because you can be asked to join the zoom call for a random check.

Good luck!

1. Let $\alpha>0$ and consider the function $f$ defined by

$$
f\left(x_{1}, x_{2}\right)=x_{1}^{2}+2 x_{2}^{2}+\alpha \sin \left(x_{2}\right) .
$$

(a) (6 points) Find the gradient descent search direction $\Delta x$ at $x=(1,0)$.
(b) ( 6 points) For which values of $\alpha$ is the function $f$ convex?
(c) (6 points) Suppose $\alpha=1$. What is the next iterate after one step of Newton's method starting from $x=(1,0)$ ?
2. Let $n$ be a positive integer and consider the optimization problem

$$
\begin{aligned}
\operatorname{minimize} & \sum_{i=1}^{n} x_{i}^{2} \\
\text { subject to } & \sum_{i=1}^{n} x_{i} \geq 1
\end{aligned}
$$

(a) (12 points) Derive the Lagrangian, Lagrange dual function, and Lagrange dual problem. (b) (6 points) Give optimal primal and dual solutions and show they are optimal.
3. (a) (12 points) Suppose

$$
F(x)=f(g(x)+h(x))
$$

with $f(x)=x^{2}, g(x)=1 / x$, and $h(x)=x^{3}$. Show how $F^{\prime}(1)$ can be computed using reversemode automatic differentiation (a.k.a. backpropagation). Show two separate diagrams for the forward and backward phases.
(b) (6 points) Explain why reverse-mode automatic differentiation can be much faster than computing and evaluating the symbolic derivative even for problems with only 1 variable.
4. Consider the primal-dual interior point method as discussed in class (and in the book) with the parameter $\mu=2$. We apply this to the problem

$$
\begin{aligned}
\text { minimize } & x_{1}^{4}+x_{2} \\
\text { subject to } & 1-x_{1}-x_{2} \leq 0 .
\end{aligned}
$$

Suppose the the current primal-dual iterate is $(x, \lambda)$ with $x=(1,1)$ and $\lambda=1$.
(a) (6 points) Compute the surrogate duality gap. Explain why this implies $x$ and $\lambda$ as given above cannot both be optimal.
(b) (9 points) Compute the primal-dual search direction $\Delta y_{\mathrm{pd}}=\left(\Delta x_{\mathrm{pd}}, \Delta \lambda_{\mathrm{pd}}\right)$.
(c) (3 points) Show that $\Delta x_{\mathrm{pd}}$ is a primal descent direction and explain the value of $\Delta \lambda_{\mathrm{pd}}$.
5. Let $Q \in \mathbb{R}^{n \times n}$ be a symmetric matrix and consider the nonconvex optimization problem

$$
\begin{gathered}
\operatorname{minimize} x^{\top} Q x \\
\text { subject to }\|x\|_{2}^{2}=1 .
\end{gathered}
$$

(a) (6 points) Express the optimal value of this problem in terms of the eigenvalue(s) of $Q$.
(b) (6 points) We apply the penalty method to obtain the unconstrained problem

$$
\operatorname{minimize} x^{\top} Q x+\alpha\left(\|x\|_{2}^{2}-1\right)^{2}
$$

where $\alpha>0$ is the penalty parameter. Show the optimal solution is an eigenvector of $Q$.
(c) (6 points) Show that as $\alpha \rightarrow \infty$, the vector $x_{\alpha}^{*}$ converges to a feasible solution of the original constrained problem by showing that the following inequalities hold:

$$
\frac{\lambda_{\min }(Q)}{2 \alpha} \leq 1-\left\|x_{\alpha}^{*}\right\|_{2} \leq \frac{\lambda_{\max }(Q)}{2 \alpha} .
$$

Here $\lambda_{\max }(Q)$ and $\lambda_{\min }(Q)$ are the largest and smallest eigenvalues of $Q$.

