## Numerical Optimization

Homework 2

Due 14.05.2014

Give your answers with logical and/or mathematical explanations. Handin your homework in the beginning of a lecture on due date. Late submissions will not be accepted.

**1.** For  $f(x) = \frac{1}{2} ||y - Ax||_2^2$  where  $x \in \mathbb{R}^n$ ,  $y \in \mathbb{R}^k$ , and  $A \in \mathbb{R}^{k \times n}$ , derive the expressions of  $\nabla f(x)$  and  $\nabla^2 f(x)$  using the chain rule.

**2.** Compute the gradient  $\nabla f(x)$  and the Hessian  $\nabla^2 f(x)$  of the Rosenbrock function,

$$f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2.$$

Show that  $x^* = (1, 1)^T$  is the only local minimizer of this function, and that the Hessian matrix at that point is positive definite.

**3.** Consider  $f(x) = x^T H x$  where  $H \in \mathbb{R}^{n \times n}$  is a symmetric positive semidefinite matrix. Show using the definition of convex functions that f(x) is convex on the domain  $\mathbb{R}^n$ . Hint: it may be convenient to prove the following equivalent inequality for all  $x, y \in \mathbb{R}^n$  and all  $\alpha \in [0, 1]$ ,

$$f(y + \alpha(x - y)) - \alpha f(x) - (1 - \alpha)f(y) \le 0.$$

**4.** Suppose that  $f : \mathbb{R}^n \to \mathbb{R}$  is a convex function. Show that the set of global minimizers of f is a convex set.

**5.** For a square symmetric matrix  $A \in \mathbb{R}^{n \times n}$  and a vector  $x \in \mathbb{R}^n$ ,  $x \neq 0$ , the quantity

$$\frac{x^T A x}{x^T x}, \ x \in \mathbb{R}^n$$

is called the *Rayleigh quotient*. Find a vector  $x^*$  that minimizes the Rayleigh quotient for a given A, and the corresponding minimum value. Also, find  $x^*$  that maximizes the quotient, and the corresponding maximum value.