



- 1 (See *N&W, Exercise 2.1*). The *Rosenbrock function* is defined as

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

- a) Compute the gradient and the Hessian of the Rosenbrock function.  
b) Show that the point  $(1, 1)$  is the unique (global and local) minimizer of  $f$ .

- 2 Let  $I$  be any index set and let  $f_i: \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  be lower semi-continuous. Show that the function  $f: \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  given as

$$f(x) = \sup_{i \in I} f_i(x)$$

is lower semi-continuous.

- 3 For the following functions, decide whether they are lower semi-continuous or coercive, and whether they attain a global minimizer:

- a) The function  $f: \mathbb{R} \rightarrow \mathbb{R}$  given by

$$f(x) = x^4 - 20x^3 + \sup_{k \in \mathbb{N}} \sin(kx).$$

- b) The function  $g: \mathbb{R} \rightarrow \mathbb{R}$  given by

$$g(x) = e^x - \frac{1}{x^2 + 1}.$$

- c) The function  $h: \mathbb{R}^2 \rightarrow \mathbb{R}$  given by

$$h(x) = x_1^2(1 + x_2^3) + x_1^2.$$

- 4 (See *N&W, Exercise 2.8*) Assume that  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is a convex function. Show that the set of minimizers of  $f$  is convex.

- 5 Show that a strictly convex function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  has at most one global minimizer. In addition, find a strictly convex function that has no global minimizer at all.

- 6 Assume that  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex, continuous, and bounded below. Show that for every  $\lambda > 0$  the function

$$f_\lambda(x) := f(x) + \lambda\|x\|^2$$

has a unique global minimizer.

*(Additional challenge: Is it possible to drop the condition that  $f$  is bounded below?)*

- 7 Show that the requirement that  $0 < c_1 < c_2 < 1$  in the definition of the Wolfe conditions is essential.

More precisely: Show that it is possible that no positive step length satisfies the Wolfe conditions if either  $c_2 \geq 1$  or  $c_1 > c_2$ .