



## Tutorial 1 : Introduction to optimization and refresher course

**Exercise 1** (Differentiation of some functions). Compute the gradient and Hessian of the following functions:

1.  $f : \mathbb{R}^d \rightarrow \mathbb{R} : x \mapsto \|\mathbf{A}x - b\|_2^2$  ( $\mathbf{A}$  and  $b$  are constant matrix and vector).
2.  $f : \mathbb{R}^d \rightarrow \mathbb{R} : x \mapsto x^\top \mathbf{A}x - b^\top x + c$  ( $\mathbf{A}, b, c$  are constant matrix, vector and scalar).
3.  $f : \mathbb{R}^d \rightarrow \mathbb{R} : x \mapsto \|x\|_2^a$  (where  $a > 2$ ).
4.  $g : \mathbb{R} \rightarrow \mathbb{R} : t \mapsto f(x + t(y - x))$  ( $x, y$  are two fixed vectors,  $f$  is a fixed  $C^2$  function). Express the gradient and the Hessian matrix of  $g$  by those of  $f$ .

**Exercise 2** (Differentiable but not  $C^1$ ). Consider the function:

$$f(x) = \begin{cases} x^2 \sin(\frac{1}{x}), & x \neq 0 \\ 0, & \text{otherwise} \end{cases}$$

1. Is  $f$  differentiable?
2. Is  $f$  continuously differentiable?
3. Based on this function, can you construct a function  $f$  such that  $f$  is twice differentiable but not  $C^2$ ?

**Exercise 3** (Necessary conditions of optimal solution revisited). If  $f$  is only differentiable and not  $C^1$ , is it still necessary that  $\nabla f(x^*) = 0$  for any local solution  $x^*$ ?

**Exercise 4** (Properties of derivatives and gradient). Given two differentiable functions  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  and  $g : \mathbb{R}^n \rightarrow \mathbb{R}$ , we have:

$$\begin{aligned} \nabla(f + g)(x) &= \nabla f(x) + \nabla g(x) \\ \nabla(\alpha f)(x) &= \alpha \nabla f(x), \forall \alpha > 0 \\ \nabla(f \cdot g)(x) &= g(x) \nabla f(x) + f(x) \nabla g(x) \\ \nabla\left(\frac{f}{g}\right) &= \frac{g(x) \nabla f(x) - f(x) \nabla g(x)}{g(x)^2}, \quad \text{assuming that } g(x) > 0. \end{aligned} \tag{1}$$

**Exercise 5** (Chain rule). Given two differentiable functions  $f : \mathbb{R}^k \rightarrow \mathbb{R}^\ell$  and  $g : \mathbb{R}^d \rightarrow \mathbb{R}^k$ , prove that the composition  $f \circ g : \mathbb{R}^d \rightarrow \mathbb{R}^\ell$  is also differentiable and its Jacobian matrix is given by:

$$J_{f \circ g}(x) = J_f(g(x)) J_g(x).$$

**Exercise 6** (Two Taylor formulations). Given a  $C^1$  (resp.  $C^2$ ) function  $f : \mathbb{R}^d \rightarrow \mathbb{R}$ , we have:

$$\begin{aligned} f(y) &= f(x) + \int_0^1 \nabla f(x + t(y - x))^\top (y - x) dt && , \forall x, y \in \mathbb{R}^d \\ (\text{resp.}) f(y) &= f(x) + (y - x)^\top \nabla f(x) + \frac{1}{2}(y - x)^\top \nabla^2 f(x)(y - x) + R_2(x - y) && , \forall x, y \in \mathbb{R}^d, \end{aligned} \quad (2)$$

where  $R_2(x - y)$  is a reminder satisfying  $\lim_{y \rightarrow x} \frac{R_2(x - y)}{\|y - x\|^2} = 0$ .

Hint: you might need to use the fundamental theorem of calculus, i.e., if  $f : \mathbb{R} \rightarrow \mathbb{R}$  is differentiable, we have:

$$f(b) = f(a) + \int_a^b f'(t) dt.$$