

## EXERCISE n°1 : Fréchet differentiability

**Definition :**  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is Fréchet differentiable at point  $x$  if there exists a linear form  $df_x : \mathbb{R}^n \rightarrow \mathbb{R}^m$  such that  $f(x+h) = f(x) + df_x(h) + \|h\|\varepsilon(h)$ ,  $\forall h \in \mathbb{R}^n$ , where  $\varepsilon(h) \rightarrow 0$  when  $h \rightarrow 0$  (instead of  $\|h\|\varepsilon(h)$ , one can write  $o(h)$ ).

The gradient of  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is the vector  $\nabla f = \left( \frac{\partial f}{\partial x_i} \right)$ . The Hessian of  $f$  is the matrix  $\nabla^2 f = \left( \frac{\partial^2 f}{\partial x_i \partial x_j} \right)$ . The Jacobian of  $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is the matrix  $JF = \left( \frac{\partial F_i}{\partial x_j} \right)$ .

### Exercise 1 : Computation of Fréchet differentials

- 1.1. Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  
 $x \rightarrow f(x) := \langle c, x \rangle + b$ , where  $c \in \mathbb{R}^n$ ,  $b \in \mathbb{R}$ .

Compute the differential  $df_x$  of  $f$  at point  $x$ , then the gradient  $\nabla f(x)$ . Compute the second order differential  $d^2 f_x$ , then the hessian  $\nabla^2 f(x)$ .

- 1.2. Let  $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  
 $x \rightarrow F(x) := Ax + b$ , where  $A \in \mathcal{M}_{m,n}(\mathbb{R})$ ,  $b \in \mathbb{R}^m$ .

Compute the differential  $dF_x$ , then the Jacobian  $JF(x)$ .

- 1.3. Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  
 $x \rightarrow f(x) := \frac{1}{2} \langle Ax, x \rangle + \langle b, x \rangle + c$ , where  $A \in \mathcal{M}_{n,n}(\mathbb{R})$ ,  $b \in \mathbb{R}^n$ ,  $c \in \mathbb{R}$ .

Compute the differential  $df_x$ , then the gradient  $\nabla f(x)$ . Compute the second order differential  $d^2 f_x$ , then the hessian  $\nabla^2 f(x)$ .

- 1.4. Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  
 $x \rightarrow g(x) := \sum_{i=1}^m (r_i(x))^2$ , where the  $r_i : \mathbb{R}^n \rightarrow \mathbb{R}$  are twice differentiable.

Compute the differential  $dg_x$ , then the gradient  $\nabla g(x)$ . Compute the second order differential  $d^2 g_x$ , then the hessian  $\nabla^2 g(x)$ .

### Exercise 2 : Differentiability

- 2.1. Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}$ , where the  $f_i : \mathbb{R}^n \rightarrow \mathbb{R}$  are twice differentiable and  
 $x \rightarrow g(x) := \sum_{i=1}^m (f_i^+(x))^2$ ,

$a^+ = \max(0, a)$ . Is the function  $g$  differentiable, twice differentiable?

- 2.2. Let  $\mathbb{R}^m \xrightarrow{A} \mathbb{R}^n$ , where  $A : u \rightarrow A_0 u + b$  with  $A_0 \in \mathcal{M}_{n,m}(\mathbb{R})$ ,  $b \in \mathbb{R}^n$  and  $f$  twice
- $$\begin{array}{ccc}
 \mathbb{R}^m & \xrightarrow{A} & \mathbb{R}^n \\
 & \searrow g := f \circ A & \downarrow f \\
 & & \mathbb{R}
 \end{array}$$

differentiable.

Determine  $\nabla g$  and  $\nabla^2 g$  in terms of  $\nabla f$  and  $\nabla^2 f$ .

*Application :* let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , twice differentiable,  $x_0 \in \mathbb{R}^n$  and  $d \in \mathbb{R}^n$ . One sets  $\varphi(t) := f(x_0 + td)$ .

Compute  $\varphi'(t)$  (resp.  $\varphi''(t)$ ) in terms of  $\nabla f(x_0 + td)$  (resp.  $\nabla^2 f(x_0 + td)$ ).