

# Convex nonsmooth optimization

## Part I: Moreau subdifferential

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## Collaboration

This course is a direct adaptation of the course built by Jean-Christophe Pesquet (CentraleSupélec) and Nelly Pustelnik (LPENSL)



## Gradient descent in dimension $N$

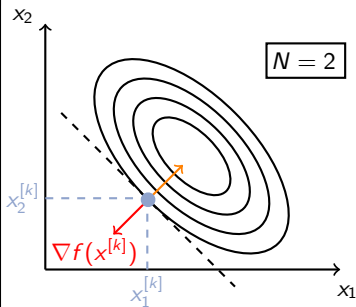
### Gradient descent

Let  $f : \mathbb{R}^N \rightarrow \mathbb{R}$  be convex, continuously differentiable on  $\mathbb{R}^N$  and with a  $\beta$ -Lipschitz gradient.

Let  $x_0 \in \mathbb{R}^N$  and  $\gamma_n \in ]0, 2/\beta[$

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n - \gamma_n \nabla f(x_n).$$

$(x_n)_{n \in \mathbb{N}}$  converges to a minimizer of  $f$ .



**$\beta$ -Lipschitz gradient** Let  $f : \mathbb{R}^N \rightarrow \mathbb{R}$  be convex, continuously differentiable on  $\mathbb{R}^N$ .  $f$  is gradient  $\beta$ -Lipschitz with  $\beta > 0$  if

$$(\forall (u, v) \in \mathbb{R}^N \times \mathbb{R}^N) \quad \|\nabla f(u) - \nabla f(v)\| \leq \beta \|u - v\|$$

## Gradient descent in dimension $N$

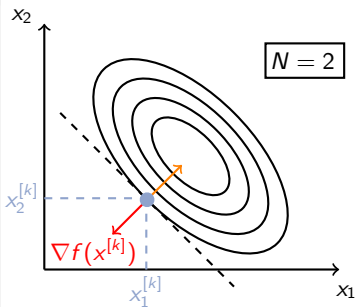
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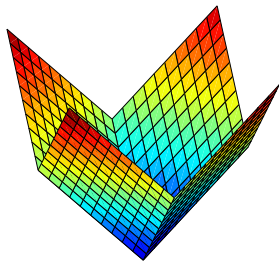


- ▶ Iterative method: build a sequence  $(x_n)_{n \in \mathbb{N}}$  s.t., at each iteration  $n$

$$f(x_{n+1}) < f(x_n)$$

- ▶ Choose  $\gamma_n$  for fast convergence: Newton method, ...
- ▶ Convergence proof: fixed point theorem.

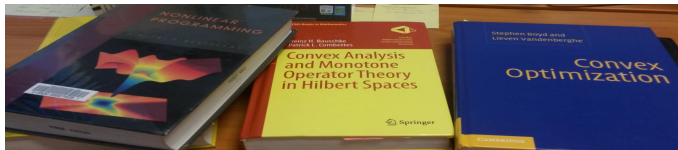
## Non-smooth convex optimization



$$\|\cdot\|_1 : \begin{cases} \mathbb{R}^2 & \rightarrow \mathbb{R} \\ (x, y) & \mapsto |x| + |y| \end{cases}$$

not differentiable on  
 $\{0\} \times \mathbb{R} \cup \mathbb{R} \times \{0\}$

## Reference books



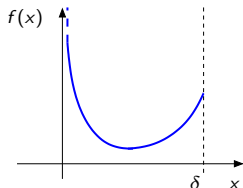
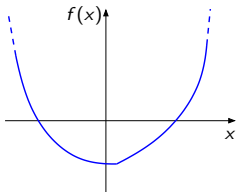
- ▶ **D. Bertsekas**, Nonlinear programming, Athena Scientific, Belmont, Massachussets, 1995.
- ▶ **Y. Nesterov**, Introductory Lectures on Convex Optimization: A Basic Course, Springer, 2004.
- ▶ **S. Boyd and L. Vandenberghe**, Convex optimization, Cambridge University Press, 2004.
- ▶ **H. H. Bauschke and P. L. Combettes**, Convex Analysis and Monotone Operator Theory in Hilbert Spaces, Springer, New York, 2011.

## Functional analysis: definitions

Let  $f : \mathcal{H} \rightarrow ]-\infty, +\infty]$  where  $\mathcal{H}$  is a Hilbert space.

- ▶ The **domain** of  $f$  is  $\text{dom } f = \{x \in \mathcal{H} \mid f(x) < +\infty\}$ .
- ▶ The function  $f$  is **proper** if  $\text{dom } f \neq \emptyset$ .

### Domains of the functions ?

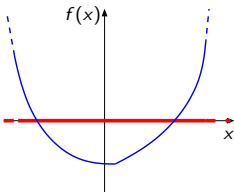


## Functional analysis: definitions

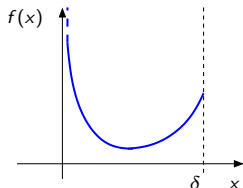
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$\text{dom } f = \mathbb{R}$   
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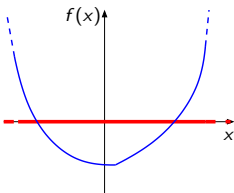


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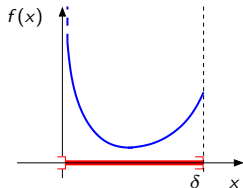
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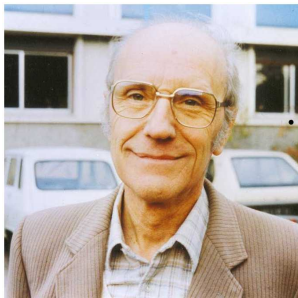


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$\text{dom } f = ]0, \delta]$   
(proper)

## A pioneer



Jean-Jacques Moreau  
(1923–2014)

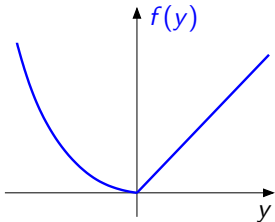
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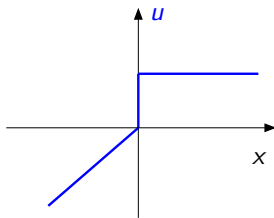
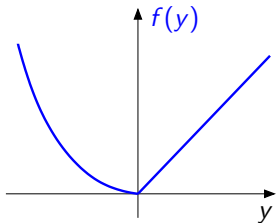
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$$\partial f : \mathcal{H} \rightarrow 2^{\mathcal{H}}$$

$$x \rightarrow \{u \in \mathcal{H} \mid (\forall y \in \mathcal{H}) \langle y - x | u \rangle + f(x) \leq f(y)\}$$



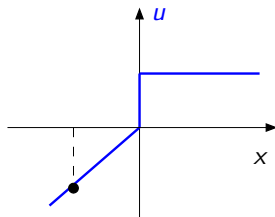
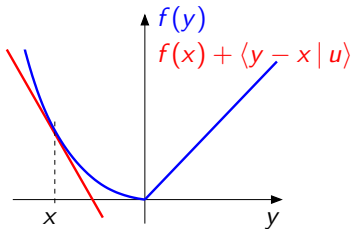
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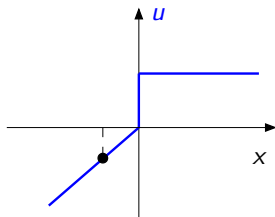
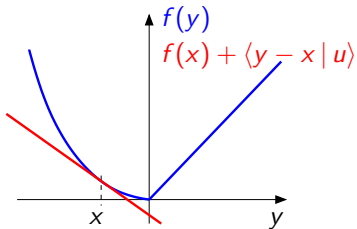
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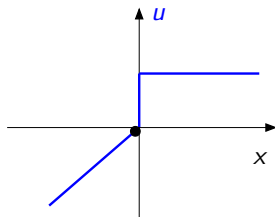
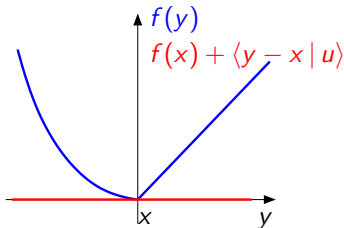
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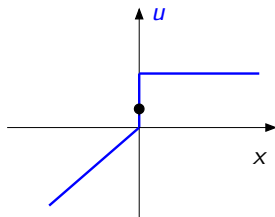
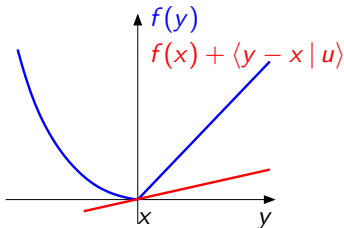
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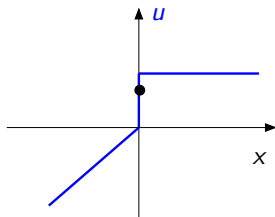
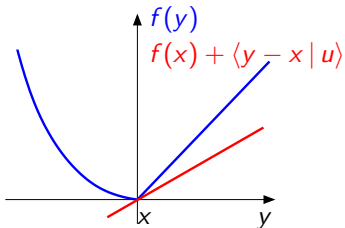
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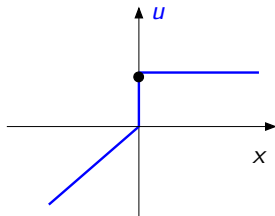
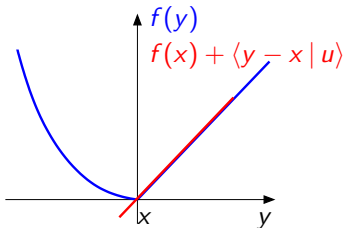
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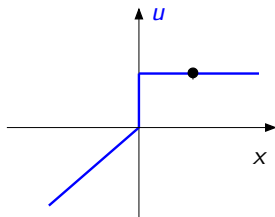
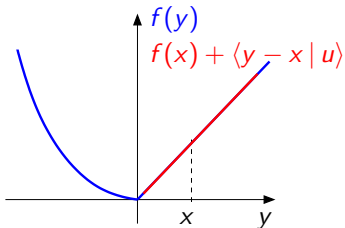
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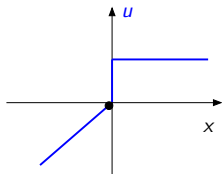
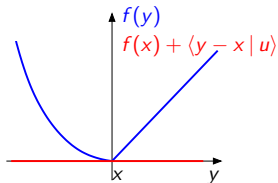


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**Fermat's rule**:  $0 \in \partial f(x) \Leftrightarrow x \in \text{Argmin } f$

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- ▶  $u \in \partial f(x)$  is a **subgradient** of  $f$  at  $x$ .

## Subdifferential of a convex function: properties

If  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  is convex and it is Gâteaux differentiable at  $x$ , then

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$$(\forall y \in \mathcal{H}) \quad \langle \nabla f(x) \mid y \rangle = \lim_{\substack{\alpha \rightarrow 0 \\ \alpha \neq 0}} \frac{f(x + \alpha y) - f(x)}{\alpha}.$$

Proof:

For every  $\alpha \in [0, 1]$  and  $y \in \mathcal{H}$ ,

$$\begin{aligned} f(x + \alpha(y - x)) &\leq (1 - \alpha)f(x) + \alpha f(y) \\ \Rightarrow \langle \nabla f(x) \mid y - x \rangle &= \lim_{\substack{\alpha \rightarrow 0 \\ \alpha \neq 0}} \frac{f(x + \alpha(y - x)) - f(x)}{\alpha} \leq f(y) - f(x) \end{aligned}$$

Then  $\nabla f(x) \in \partial f(x)$ .



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Proof:

Conversely, if  $u \in \partial f(x)$ , then, for every  $\alpha \in [0, +\infty[$  and  $y \in \mathcal{H}$ ,

$$\begin{aligned} f(x + \alpha y) &\geq f(x) + \langle u \mid x + \alpha y - x \rangle \\ \Rightarrow \langle \nabla f(x) \mid y \rangle &= \lim_{\substack{\alpha \rightarrow 0 \\ \alpha \neq 0}} \frac{f(x + \alpha y) - f(x)}{\alpha} \geq \langle u \mid y \rangle \end{aligned}$$

By selecting  $y = u - \nabla f(x)$ , it results that  $\|u - \nabla f(x)\|^2 \leq 0$  and then  $u = \nabla f(x)$ .

## Subdifferential of a convex function: properties

Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be Gâteaux differentiable on  $\text{dom } f$ , with  $\text{dom } f$  a convex subset of  $\mathcal{H}$ .

Then,  $f$  is convex if and only if

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Proof:

We have already seen that the gradient inequality holds when  $f$  is convex and differentiable at  $x \in \mathcal{H}$ .

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Proof:

Conversely, if the gradient inequality is satisfied, we have, for every  $(x, y) \in (\text{dom } f)^2$  and  $\alpha \in [0, 1]$ ,  $\alpha x + (1 - \alpha)y \in \text{dom } f$ , and

$$f(x) \geq f(\alpha x + (1 - \alpha)y) + (1 - \alpha) \langle \nabla f(\alpha x + (1 - \alpha)y) \mid x - y \rangle$$

$$f(y) \geq f(\alpha x + (1 - \alpha)y) + \alpha \langle \nabla f(\alpha x + (1 - \alpha)y) \mid y - x \rangle.$$

By multiplying the first inequality by  $\alpha$  and the second one by  $1 - \alpha$  and summing them, we get

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

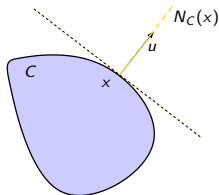
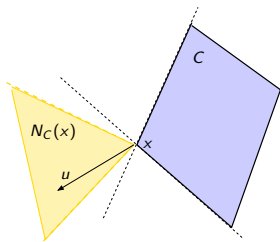
## Subdifferential of a convex function: example

Let  $C$  be a nonempty subset of  $\mathcal{H}$  with **indicator function** defined as

$$(\forall x \in \mathcal{H}) \quad \iota_C(x) = \begin{cases} 0 & \text{if } x \in C \\ +\infty & \text{otherwise.} \end{cases}$$

For every  $x \in \mathcal{H}$ ,  $\partial \iota_C(x)$  is the **normal cone** to  $C$  at  $x$  defined by

$$N_C(x) = \begin{cases} \{u \in \mathcal{H} \mid (\forall y \in C) \langle u \mid y - x \rangle \leq 0\} & \text{if } x \in C \\ \emptyset & \text{otherwise.} \end{cases}$$



## Subdifferential calculus

Let  $\mathcal{H}$  and  $\mathcal{G}$  be two real Hilbert spaces.

- ▶ Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, then  $\forall \lambda \in ]0, +\infty[$   $\partial(\lambda f) = \lambda \partial f$ .
- ▶ Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ ,  $g: \mathcal{G} \rightarrow ]-\infty, +\infty]$ , and  $L \in \mathcal{B}(\mathcal{H}, \mathcal{G})$ .

Define  $g \circ L(x) := g(Lx)$  and  $L^*$  the *adjoint* operator of  $L$ :

$$(\forall (x, y) \in \mathcal{H} \times \mathcal{G}) \quad \langle y \mid Lx \rangle = \langle L^*y \mid x \rangle.$$

If  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ , then

$$(\forall x \in \mathcal{H}) \quad \partial f(x) + L^* \partial g(Lx) \subset \partial(f + g \circ L)(x).$$

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Proof: Let  $x \in \mathcal{H}$ ,  $u \in \partial f(x)$  and  $v \in \partial g(Lx)$ . We have:

$u + L^*v \in \partial f(x) + L^* \partial g(Lx)$  and

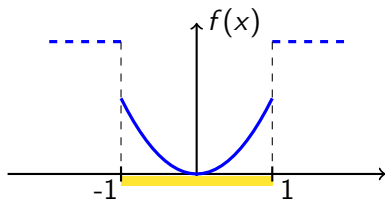
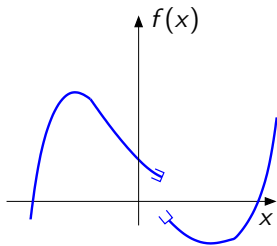
$$\begin{aligned} (\forall y \in \mathcal{H}) \quad f(y) &\geq f(x) + \langle y - x \mid u \rangle \\ g(Ly) &\geq g(Lx) + \langle L(y - x) \mid v \rangle. \end{aligned}$$

Therefore, by summing,

$$f(y) + g(Ly) \geq f(x) + g(Lx) + \langle y - x \mid u + L^*v \rangle.$$

We deduce that  $u + L^*v \in \partial(f + g \circ L)(x)$ .

## Subdifferential: the case of discontinuous functions





## Epigraph

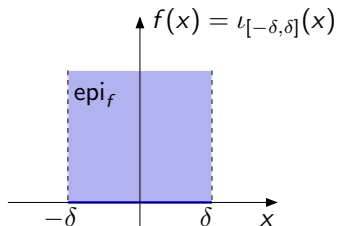
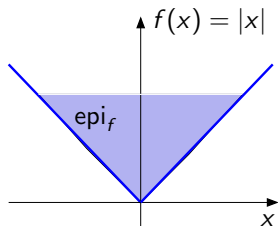
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## Lower semi-continuity

Let  $f : \mathcal{H} \rightarrow ]-\infty, +\infty]$ .

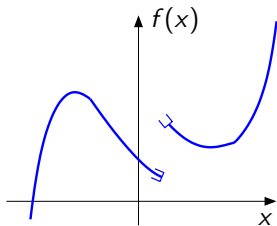
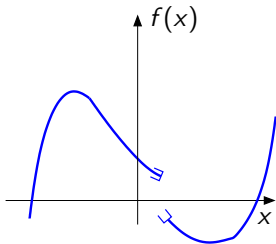
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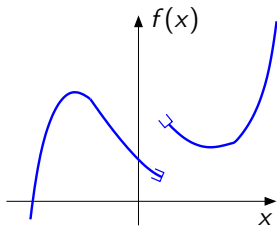
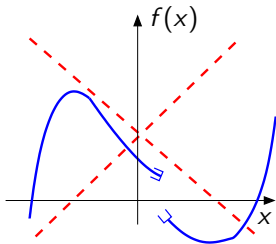


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## Lower semi-continuity

- ▶ Every continuous function on  $\mathcal{H}$  is l.s.c.
- ▶ Every finite sum of l.s.c. functions is l.s.c.
- ▶ Let  $(f_i)_{i \in I}$  be a family of l.s.c functions.  
Then,  $\sup_{i \in I} f_i$  is l.s.c.

## A class of convex functions

- ▶  $\Gamma_0(\mathcal{H})$ : class of convex, l.s.c., and proper functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$ .
- ▶  $\iota_C \in \Gamma_0(\mathcal{H}) \Leftrightarrow C$  is a nonempty closed convex set.  
Proof:  $\text{epi}_{\iota_C} = C \times [0, +\infty[$ .

## Subdifferential calculus

Let  $\mathcal{H}$  and  $\mathcal{G}$  be two real Hilbert spaces.

Let  $f \in \Gamma_0(\mathcal{H})$ ,  $g \in \Gamma_0(\mathcal{G})$ , and  $L \in \mathcal{B}(\mathcal{H}, \mathcal{G})$ .

If  $\text{int}(\text{dom } g) \cap L(\text{dom } f) \neq \emptyset$  or  $\text{dom } g \cap \text{int}(L(\text{dom } f)) \neq \emptyset$ , then

$$\partial f + L^* \partial g L = \partial(f + g \circ L).$$

Particular case:

- ▶ If  $f \in \Gamma_0(\mathcal{H})$ ,  $g \in \Gamma_0(\mathcal{G})$ , and  $f$  is finite valued, then  $\partial f + \partial g = \partial(f + g)$ .
- ▶ If  $g \in \Gamma_0(\mathcal{G})$ ,  $L \in \mathcal{B}(\mathcal{G}, \mathcal{H})$ , and  $\text{int}(\text{dom } g) \cap \text{ran } L \neq \emptyset$ , then  $L^* \partial g L = \partial(g \circ L)$ .



## Subdifferential calculus

Let  $(\mathcal{H})_{i \in I}$  where  $I \subset \mathbb{N}$  be Hilbert spaces and let  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ .  
 For every  $i \in I$ , let  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$  be a proper function. Let

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty] : x = (x_i)_{i \in I} \mapsto \sum_{i \in I} f_i(x_i)$$

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Proof: Let  $x = (x_i)_{i \in I} \in \mathcal{H}$ . We have

$$t = (t_i)_{i \in I} \in \times_{i \in I} \partial f_i(x_i)$$

$$\Leftrightarrow (\forall i \in I)(\forall y_i \in \mathcal{H}_i) \quad f_i(y_i) \geq f_i(x_i) + \langle t_i | y_i - x_i \rangle$$

$$\Rightarrow (\forall y = (y_i)_{i \in I} \in \mathcal{H}) \quad \sum_{i \in I} f_i(y_i) \geq \sum_{i \in I} f_i(x_i) + \sum_{i \in I} \langle t_i | y_i - x_i \rangle$$

$$\Leftrightarrow (\forall y \in \mathcal{H}) \quad f(y) \geq f(x) + \langle t | y - x \rangle.$$

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Proof: Conversely,

$$\begin{aligned} t &= (t_i)_{i \in I} \in \partial f(x) \\ \Leftrightarrow (\forall y = (y_i)_{i \in I} \in \mathcal{H}) \quad \sum_{i \in I} f_i(y_i) &\geq \sum_{i \in I} f_i(x_i) + \sum_{i \in I} \langle t_i \mid y_i - x_i \rangle. \end{aligned}$$

Let  $j \in I$ . By setting  $(\forall i \in I \setminus \{j\}) y_i = x_i \in \text{dom } f_i$ , we get

$$(\forall y_j \in \mathcal{H}_j) \quad f_j(y_j) \geq f_j(x_j) + \langle t_j \mid y_j - x_j \rangle.$$

## Exercise 1: Huber function

Let  $\rho > 0$  and set

$$f: \mathbb{R} \rightarrow \mathbb{R}: \mapsto \begin{cases} \frac{x^2}{2}, & \text{if } |x| \leq \rho \\ \rho|x| - \frac{\rho^2}{2}, & \text{otherwise.} \end{cases}$$

1. What is the domain of  $f$  ?
2. Plot the subdifferential of  $f$ .
3. Is  $f$  differentiable ? Prove that  $f$  is convex.

## Exercise 2

Let  $\mathcal{H}$  be a Hilbert space. Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $C \subset \mathcal{H}$  such that  $\text{dom } f \cap C \neq \emptyset$ . Give a sufficient condition for  $x \in \mathcal{H}$  to be a global minimizer of  $f + \iota_C$ .

## Exercice 3: Monotony of the subdifferential of a function

Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper function.  
Its subdifferential is a monotone operator, i.e.

$$(\forall (x_1, x_2) \in \mathcal{H}^2) (\forall u_1 \in \partial f(x_1)) (\forall u_2 \in \partial f(x_2)) \quad \langle u_1 - u_2 \mid x_1 - x_2 \rangle \geq 0.$$

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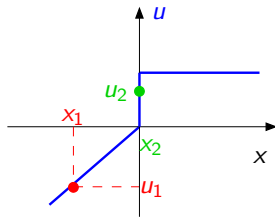
► Proof:

By definition:

$$\langle x_2 - x_1 \mid u_1 \rangle + f(x_1) \leq f(x_2)$$

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► It results that  $\langle x_1 - x_2 \mid u_1 - u_2 \rangle \geq 0$ .



## Exercice 4: Convexity and monotony

Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be Gâteaux differentiable on  $\text{dom } f$ , which is convex.

Then,  $f$  is convex if and only if  $\nabla f$  is **monotone** on  $\text{dom } f$ , i.e.

$$(\forall (x, y) \in (\text{dom } f)^2) \quad \langle \nabla f(y) - \nabla f(x) \mid y - x \rangle \geq 0.$$



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Proof:

When  $f$  is convex, we have seen that its subdifferential is monotone and, for every  $x \in \text{dom } f$ ,  $\partial f(x) = \{\nabla f(x)\}$ .

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Proof:

Conversely, assume that  $\nabla f$  is monotone on  $\text{dom } f$ . For every  $(x, y) \in (\text{dom } f)^2$ , let  $\varphi: [0, 1] \rightarrow \mathbb{R}: \alpha \mapsto f(x + \alpha(y - x))$ .  $\varphi$  is differentiable on  $[0, 1]$  and

$$(\forall \alpha \in [0, 1]) \quad \varphi'(\alpha) = \langle \nabla f(x + \alpha(y - x)) \mid y - x \rangle.$$

On the other hand, for every  $\alpha \in ]0, 1[$

$$\begin{aligned} & \langle \nabla f(x + \alpha(y - x)) - \nabla f(x) \mid y - x \rangle \geq 0 \\ \Leftrightarrow & \varphi'(\alpha) \geq \langle \nabla f(x) \mid y - x \rangle \\ \Rightarrow & \varphi(1) - \varphi(0) = \int_0^1 \varphi'(\alpha) d\alpha \geq \langle \nabla f(x) \mid y - x \rangle \\ \Leftrightarrow & f(y) - f(x) \geq \langle \nabla f(x) \mid y - x \rangle. \end{aligned}$$