

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

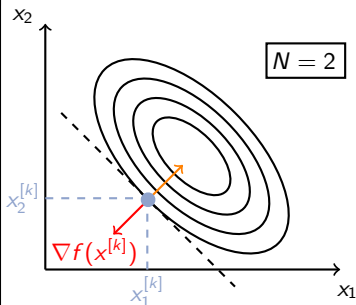
Gradient descent

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



β -Lipschitz gradient Let $f : \mathbb{R}^N \rightarrow \mathbb{R}$ be convex, continuously differentiable on \mathbb{R}^N . f is gradient β -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

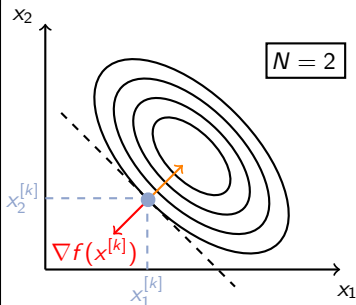
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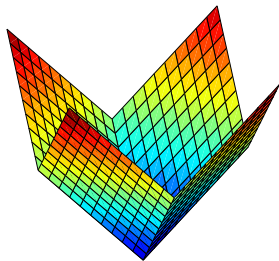


- Iterative method: build a sequence $(x_n)_{n \in \mathbb{N}}$ s.t., at each iteration n

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

- Choose γ_n for fast convergence: Newton method, ...
- Convergence proof: fixed point theorem.

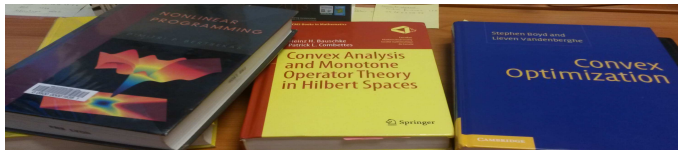
Non-smooth convex optimization



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

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

Reference books



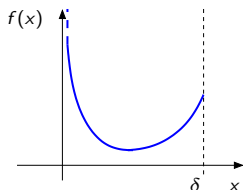
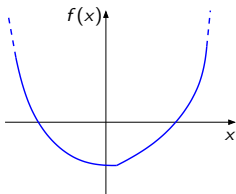
- ▶ **D. Bertsekas**, Nonlinear programming, Athena Scientific, Belmont, Massachussets, 1995.
- ▶ **Y. Nesterov**, Introductory Lectures on Convex Optimization: A Basic Course, Springer, 2004.
- ▶ **S. Boyd and L. Vandenberghe**, Convex optimization, Cambridge University Press, 2004.
- ▶ **H. H. Bauschke 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, e.g., $\mathcal{H} = \mathbb{R}^N$.

- ▶ 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 ?

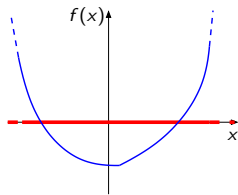


Functional analysis: definitions

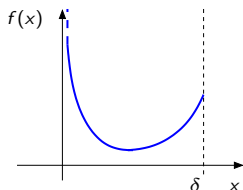
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Domains of the functions ?



$\text{dom } f = \mathbb{R}$
(proper)

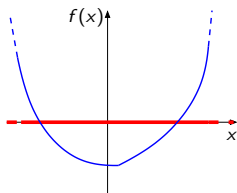


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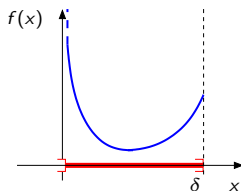
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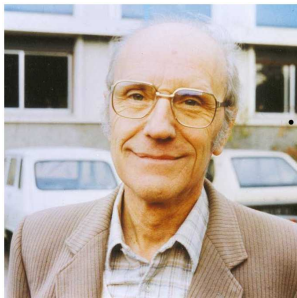


$\text{dom } f = \mathbb{R}$
(proper)



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

A pioneer



Jean-Jacques Moreau
(1923–2014)

Subdifferential of function: definition

The (Moreau) subdifferential of f , denoted by ∂f ,

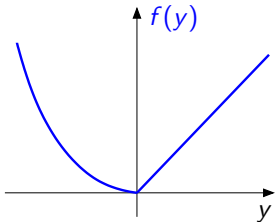
Example: $\mathcal{H} = \mathbb{R}$

Subdifferential of function: definition

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

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Subdifferential of function: definition

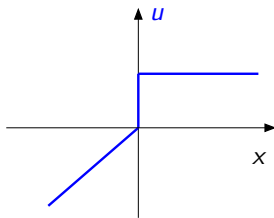
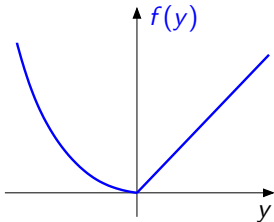
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The (Moreau) **subdifferential of f** , denoted by ∂f , is such that

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

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

Example: $\mathcal{H} = \mathbb{R}$



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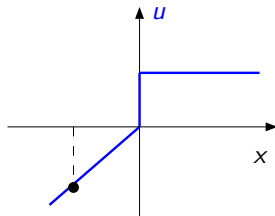
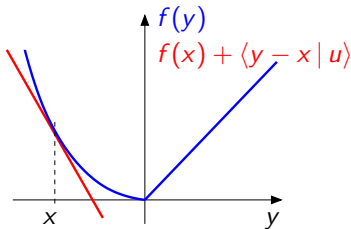
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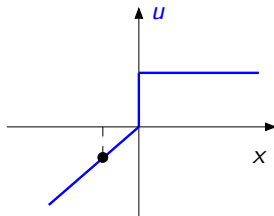
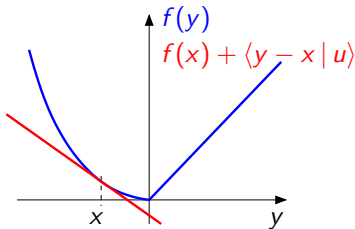
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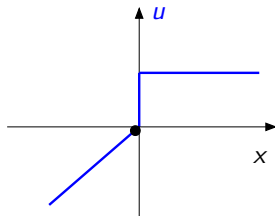
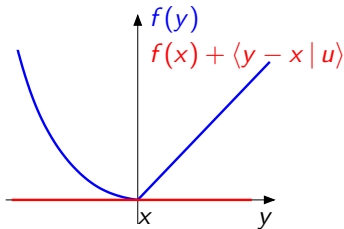
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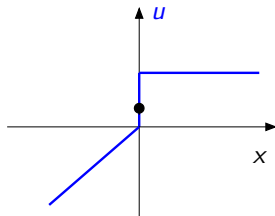
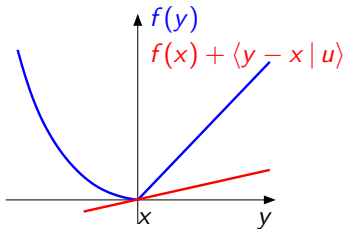
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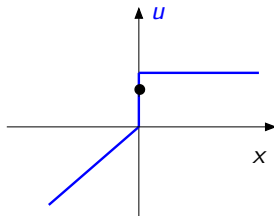
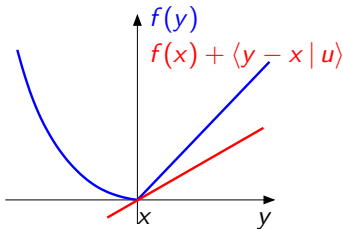
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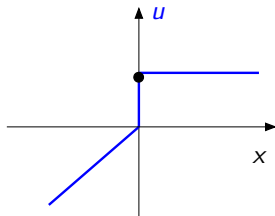
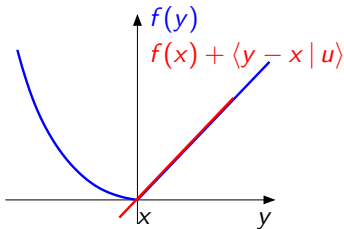
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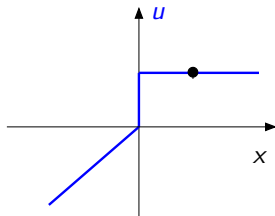
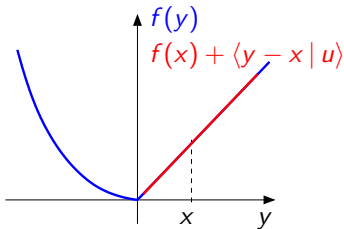
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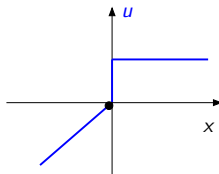
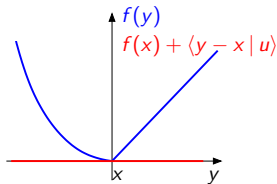
Subdifferential of a function: properties

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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 .

Gâteaux differentiability: definition

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

$$\forall y \in \mathcal{H}, \quad \lim_{\alpha \rightarrow 0^+} \frac{f(x + \alpha y) - f(x)}{\alpha}$$

exists.

In particular, if f is differentiable at x , this limit exists and

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

Subdifferential of a convex function: properties

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

$$\partial f(x) = \{\nabla f(x)\}$$

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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 $y \in \mathcal{H}$,

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

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 $y \in \mathcal{H}$,

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

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

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

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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 α and the second one by $1 - \alpha$ and summing them, we get

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

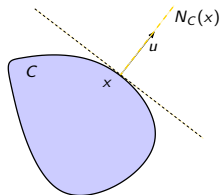
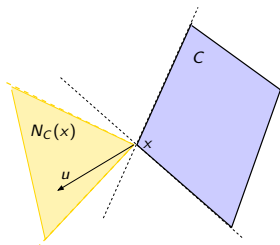
Subdifferential of a convex function: example

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 $\forall \lambda \in]0, +\infty[\quad \partial(\lambda f) = \lambda \partial f$.
- ▶ 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 :

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

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

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

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

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

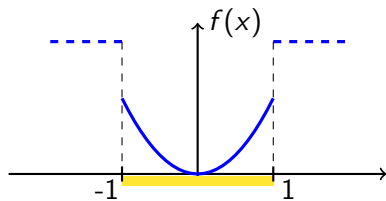
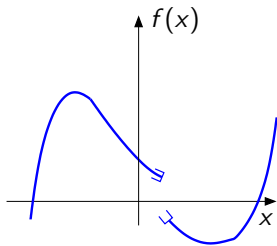
$$\begin{aligned} (\forall y \in \mathcal{H}) \quad f(y) &\geq f(x) + \langle y - x \mid u \rangle \\ g(Ly) &\geq g(Lx) + \langle L(y - x) \mid v \rangle. \end{aligned}$$

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

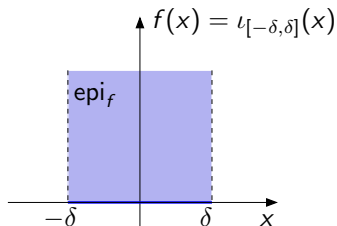
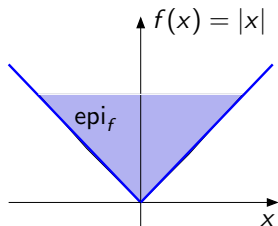
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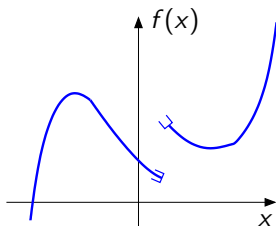
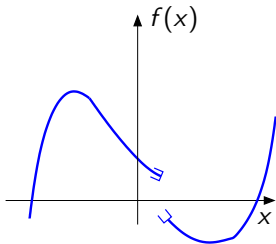
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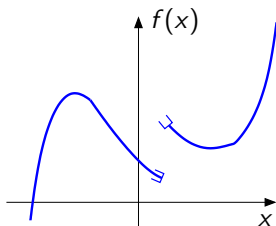
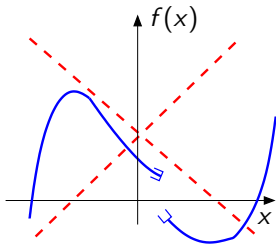


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Lower semi-continuity

- ▶ 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

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

Then,

$$(\forall x = (x_i)_{i \in I} \in \mathcal{H}) \quad \partial f(x) = \times_{i \in I} \partial f_i(x_i).$$

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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 sufficient conditions for $x \in \mathcal{H}$ to be a global minimizer of $f + \iota_C$.

Exercise 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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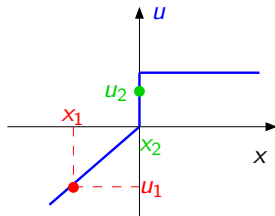
► Proof:

By definition:

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

$$\langle x_1 - x_2 \mid u_2 \rangle + f(x_2) \leq f(x_1)$$

► 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 ∇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 ∇f is monotone on $\text{dom } f$. For every $(x, y) \in (\text{dom } f)^2$, let $\varphi: [0, 1] \rightarrow \mathbb{R}: \alpha \mapsto f(x + \alpha(y - x))$. φ 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 $\alpha \in]0, 1]$

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