

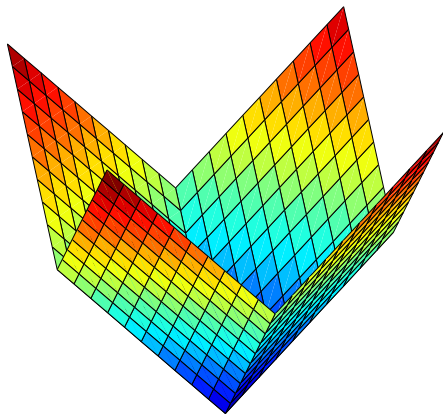
Numerical Optimization Methods in Imaging

Part I: Subdifferential and proximity operator

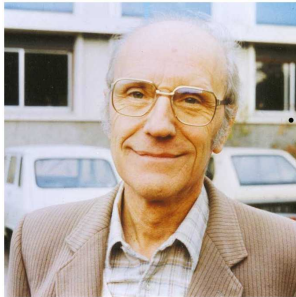
jean-christophe@pesquet.eu

PhD Summer School MMLIA – Bologna

Non-smooth convex optimization



A pioneer



Jean-Jacques Moreau
(1923–2014)

Subdifferential of function: definition

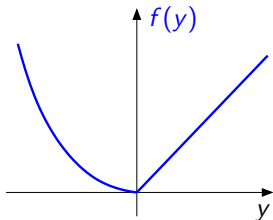
The (Moreau) subdifferential of f , denoted by ∂f ,

Subdifferential of function: definition

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

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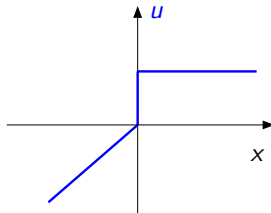
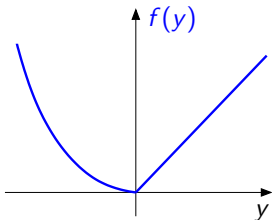
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$$\partial f : \mathcal{H} \rightarrow 2^{\mathcal{H}}$$

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



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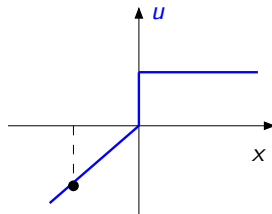
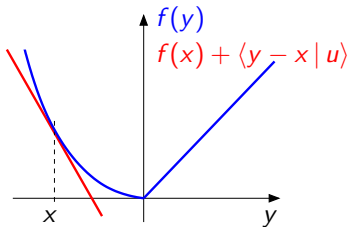
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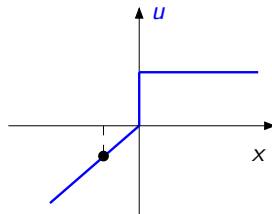
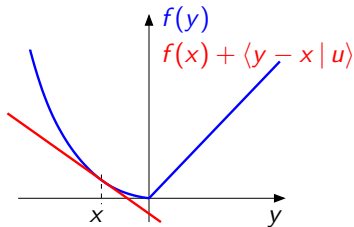
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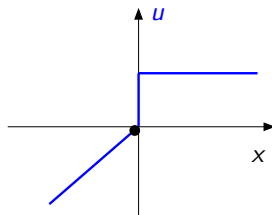
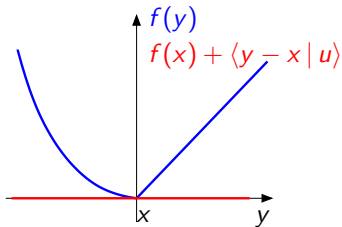
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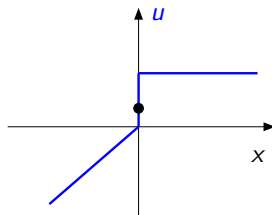
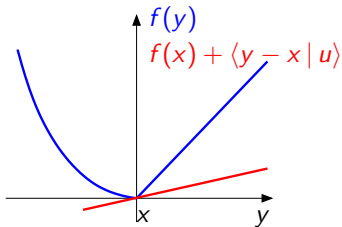
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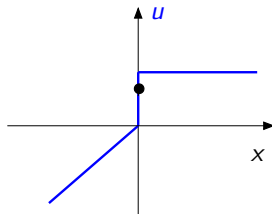
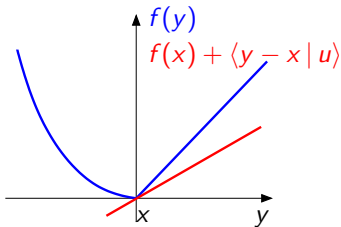
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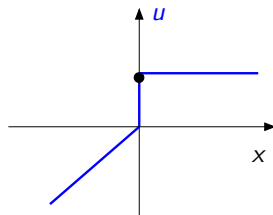
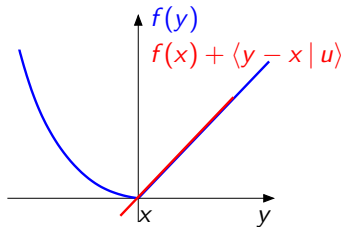
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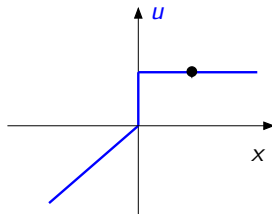
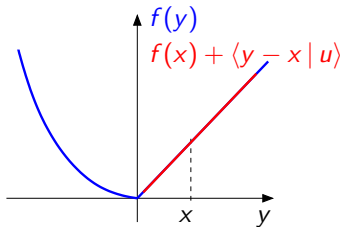
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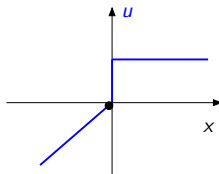
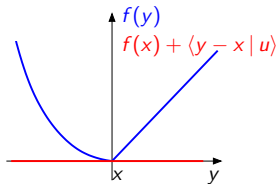
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Fermat's rule: $0 \in \partial f(x) \Leftrightarrow x \in \operatorname{Argmin} f$

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

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- ▶ $u \in \partial f(x)$ is a **subgradient** of f at x .
- ▶ If $x \notin \text{dom } f = \{x \in \mathcal{H} \mid f(x) < +\infty\}$, then $\partial f(x) = \emptyset$.

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- ▶ If $x \notin \text{dom } f = \{x \in \mathcal{H} \mid f(x) < +\infty\}$, then $\partial f(x) = \emptyset$.
- ▶ For every $x \in \text{dom } f$, $\partial f(x)$ is a closed and convex set.

Subdifferential of a function: properties

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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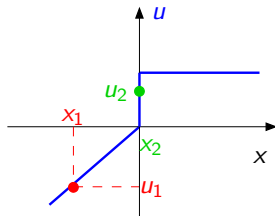
► Proof:

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 $\langle x_1 - x_2 \mid u_1 - u_2 \rangle \geq 0$.



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 $y \in \mathcal{H}$,

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

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 $y \in \mathcal{H}$,

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

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: example

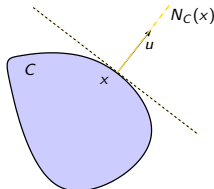
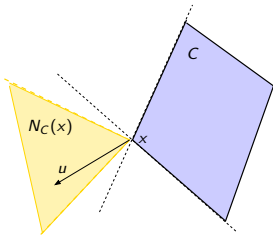
Let C be a nonempty subset of \mathcal{H} .

Its indicator function is

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

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

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



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- ▶ If $x \in \text{int } C$, then $N_C(x) = \{0\}$.
- ▶ If C is a vector space, then for every $x \in C$, $N_C(x) = C^\perp$.

Subdifferential calculus

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

- ▶ Let $f: \mathcal{H} \rightarrow]-\infty, +\infty]$ be proper, i.e., $\text{dom } f \neq \emptyset$, then for every $\lambda \in]0, +\infty[$ $\partial(\lambda f) = \lambda \partial f$.
- ▶ Let $f: \mathcal{H} \rightarrow]-\infty, +\infty]$, $g: \mathcal{G} \rightarrow]-\infty, +\infty]$, and $L \in \mathcal{B}(\mathcal{H}, \mathcal{G})$.
Let L^* denote the adjoint of L .
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 $x \in \mathcal{H}$

$$\partial f(x) + L^* \partial g(Lx) = \{u + L^*v \mid u \in \partial f(x), v \in \partial g(Lx)\}$$

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Proof: Let $x \in \mathcal{H}$ 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 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 $\text{int}(\text{dom } g - L(\text{dom } f)) \neq \emptyset$, then

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

Particular case:

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

Subdifferential calculus

Let I be a finite subset of \mathbb{N} .

Let $(\mathcal{H}_i)_{i \in I}$ be Hilbert spaces and let $\mathcal{H} = \prod_{i \in I} \mathcal{H}_i$.

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,

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

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

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

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

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

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Then,

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

Proof: Conversely,

$$t = (t_i)_{i \in I} \in \partial f(x)$$

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

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

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

Conjugate



Adrien-Marie Legendre
(1752–1833)



Werner Fenchel
(1905–1988)

Conjugate



Adrien-Marie Legendre
(1752–1833)



Werner Fenchel
(1905–1988)

Conjugate: reminders

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

The **conjugate** of f is the function $f^*: \mathcal{H} \rightarrow [-\infty, +\infty]$ such that

$$(\forall u \in \mathcal{H}) \quad f^*(u) = \sup_{x \in \mathcal{H}} (\langle x | u \rangle - f(x)) .$$

Moreau-Fenchel theorem

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

$$f \text{ is l.s.c. and convex} \Leftrightarrow f^{**} = f.$$

Conjugate: properties

Fenchel-Young inequality : If f is proper, then

1. $(\forall (x, u) \in \mathcal{H}^2) \quad f(x) + f^*(u) \geq \langle x \mid u \rangle$
2. $(\forall (x, u) \in \mathcal{H}^2) \quad u \in \partial f(x) \Leftrightarrow f(x) + f^*(u) = \langle x \mid u \rangle.$

Conjugate: properties

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Proof: Let $(x, u) \in \mathcal{H}^2$. We have

$$\begin{aligned}
 & f(x) + f^*(u) = \langle x \mid u \rangle \\
 \Leftrightarrow & f(x) + f^*(u) \leq \langle x \mid u \rangle \\
 \Leftrightarrow & \sup_{y \in \mathcal{H}} \langle y \mid u \rangle - f(y) \leq \langle x \mid u \rangle - f(x) \\
 \Leftrightarrow & (\forall y \in \mathcal{H}) \quad \langle y \mid u \rangle - f(y) \leq \langle x \mid u \rangle - f(x) \\
 \Leftrightarrow & (\forall y \in \mathcal{H}) \quad f(y) \geq f(x) + \langle y - x \mid u \rangle \\
 \Leftrightarrow & u \in \partial f(x).
 \end{aligned}$$

Conjugate: properties

Fenchel-Young inequality : If f is proper, then

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If $f \in \Gamma_0(\mathcal{H})$, then

$$(\forall (x, u) \in \mathcal{H}^2) \quad u \in \partial f(x) \Leftrightarrow x \in \partial f^*(u) .$$

Conjugate: properties

Fenchel-Young inequality : If f is proper, then

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If $f \in \Gamma_0(\mathcal{H})$, then

$$(\forall (x, u) \in \mathcal{H}^2) \quad u \in \partial f(x) \Leftrightarrow x \in \partial f^*(u).$$

Proof: Since $f \in \Gamma_0(\mathcal{H})$, we have

$$f^*(u) + f^{**}(x) = \langle x \mid u \rangle.$$

which is equivalent to $x \in \partial f^*(u)$.

Proximity operator: definition

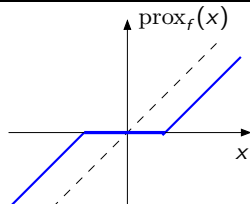
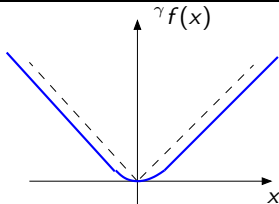
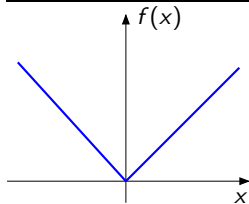
Let \mathcal{H} be a Hilbert space. Let $f \in \Gamma_0(\mathcal{H})$.

- ▶ The **Moreau envelope** of f of parameter $\gamma \in]0, +\infty[$ is

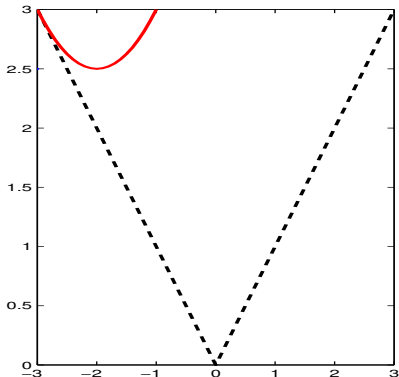
$$\gamma f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \inf_{y \in \mathcal{H}} f(y) + \frac{1}{2\gamma} \|y - x\|^2.$$

- ▶ The **proximity operator** of f is

$$\text{prox}_f: \mathcal{H} \rightarrow \mathcal{H}: x \mapsto \operatorname{argmin}_{y \in \mathcal{H}} f(y) + \frac{1}{2} \|y - x\|^2.$$

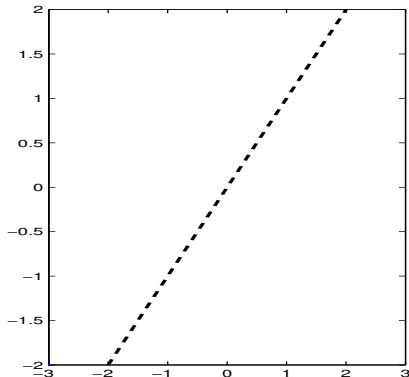


Proximity operator: definition



Moreau envelope

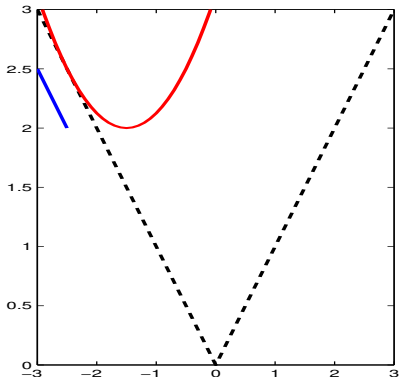
$$\gamma f(x) = \inf_{y \in \mathcal{H}} f(y) + \frac{1}{2\gamma} \|y - x\|^2$$



Proximity operator

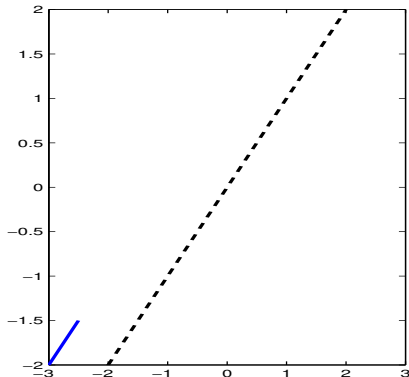
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Proximity operator: definition



Moreau envelope

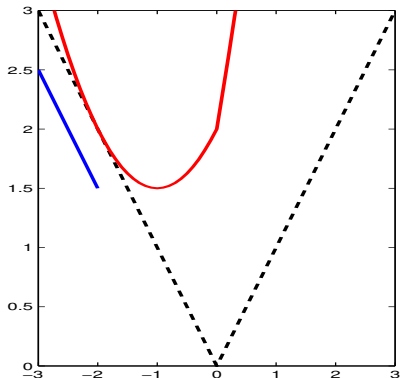
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Proximity operator

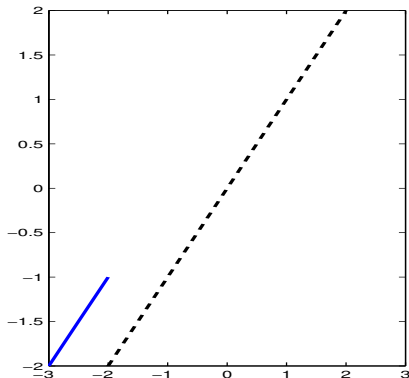
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Proximity operator: definition



Moreau envelope

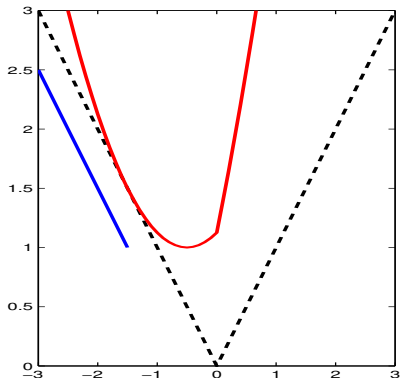
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Proximity operator

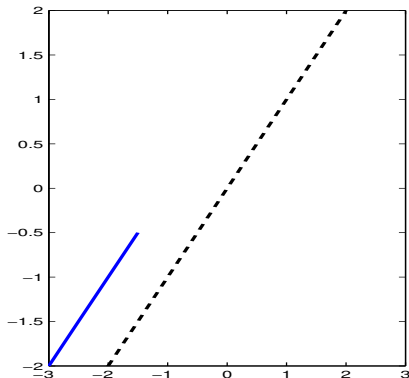
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Proximity operator: definition



Moreau envelope

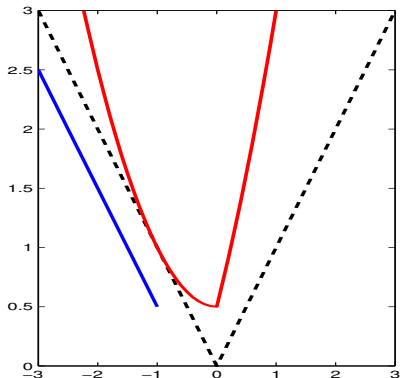
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Proximity operator

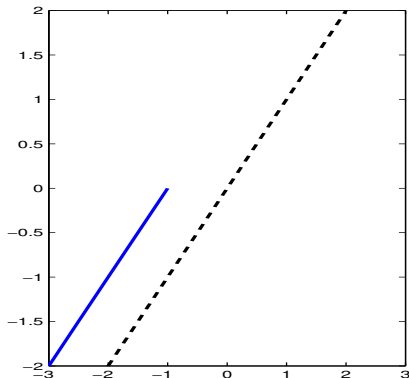
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Proximity operator: definition



Moreau envelope

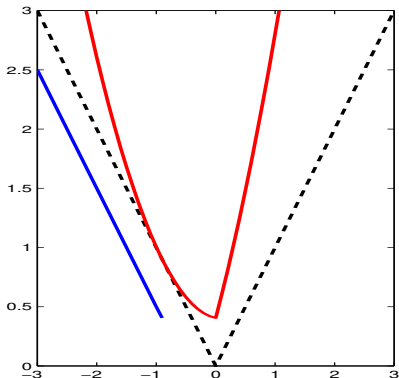
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Proximity operator

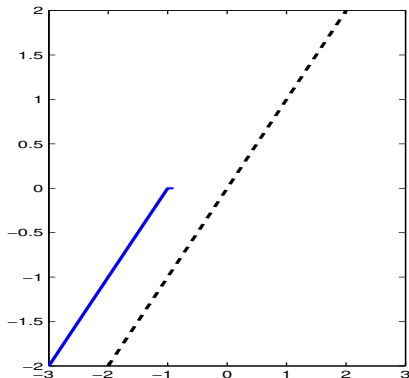
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Proximity operator: definition



Moreau envelope

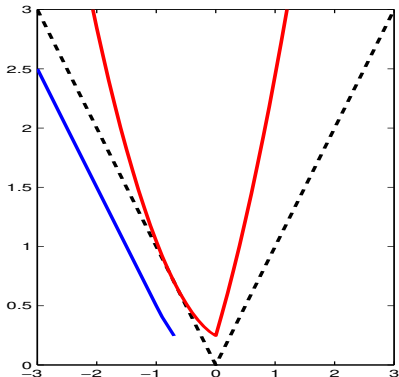
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Proximity operator

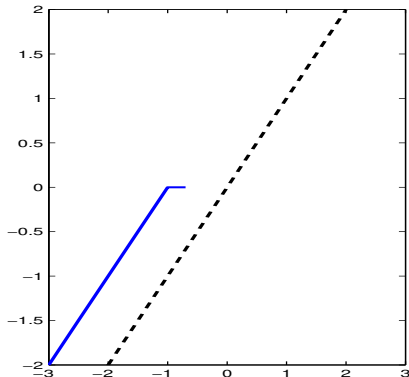
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Proximity operator: definition



Moreau envelope

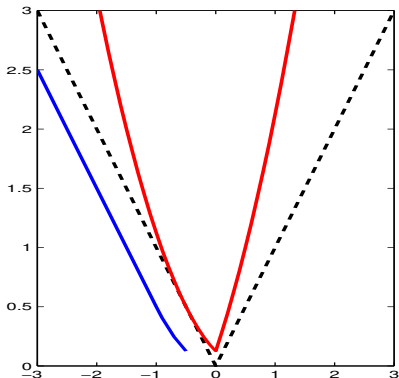
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Proximity operator

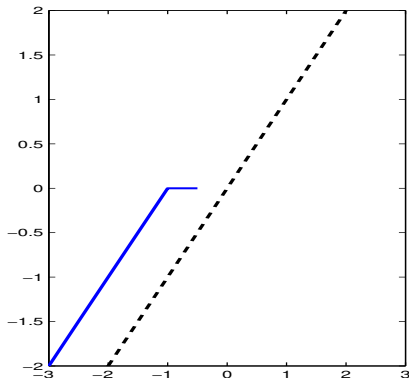
$$\text{prox}_f(x) = \underset{y \in \mathcal{H}}{\operatorname{argmin}} f(y) + \frac{1}{2} \|y - x\|^2$$

Proximity operator: definition



Moreau envelope

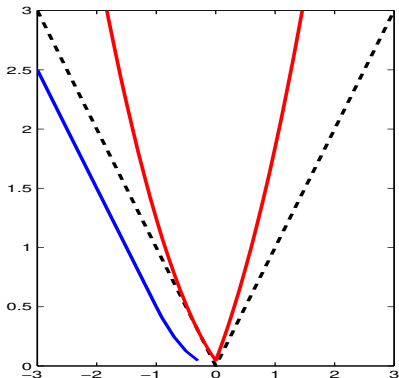
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Proximity operator

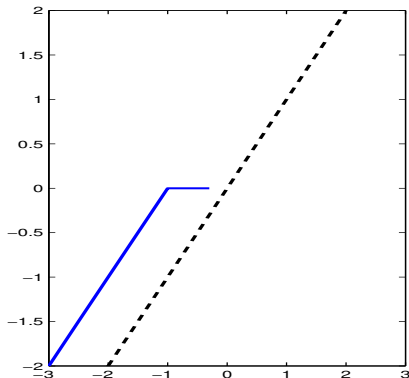
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Moreau envelope

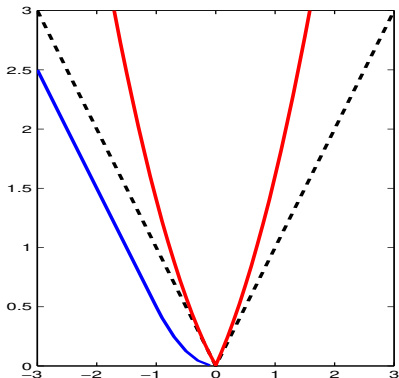
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Proximity operator

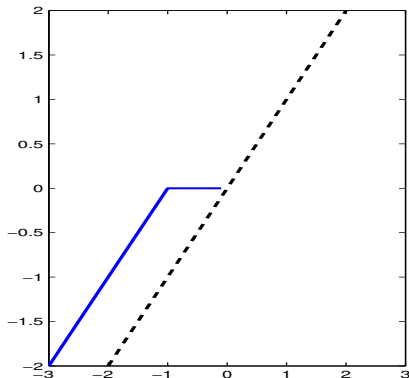
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Proximity operator: definition



Moreau envelope

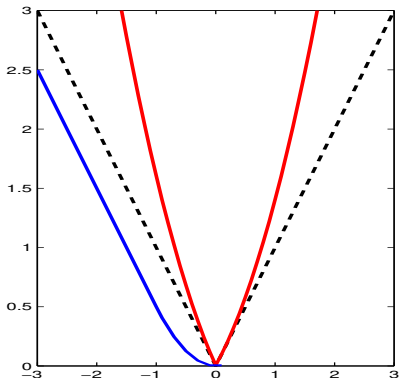
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Proximity operator

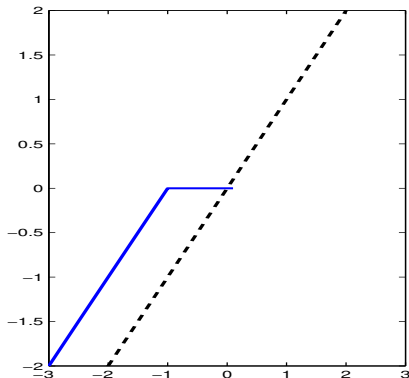
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Proximity operator: definition



Moreau envelope

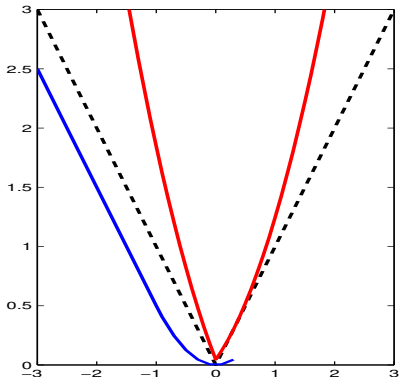
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Proximity operator

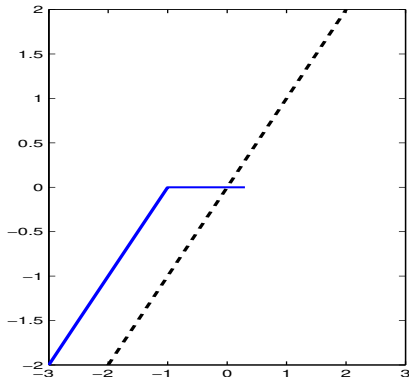
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Proximity operator: definition



Moreau envelope

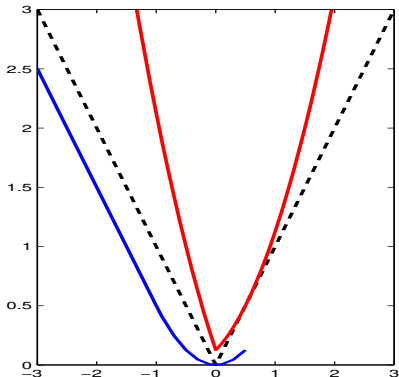
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Proximity operator

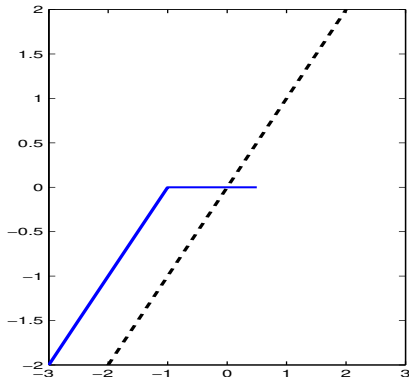
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Proximity operator: definition



Moreau envelope

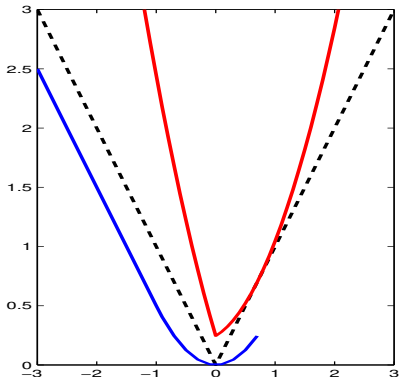
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Proximity operator

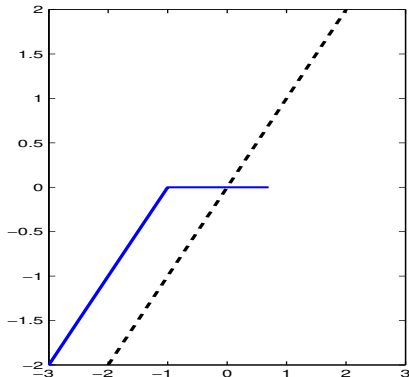
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Proximity operator: definition



Moreau envelope

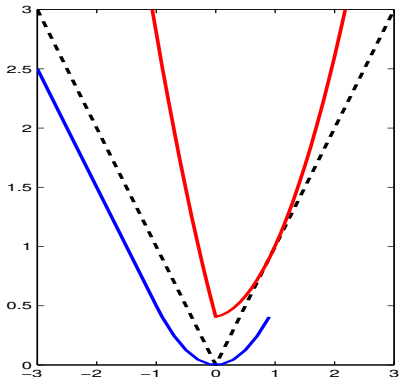
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Proximity operator

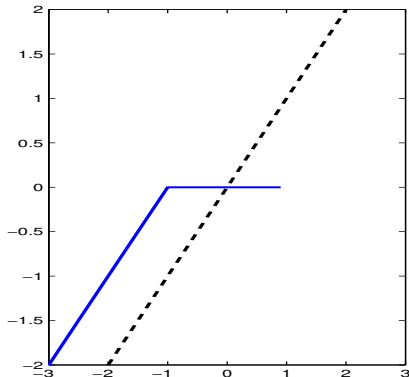
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Proximity operator: definition



Moreau envelope

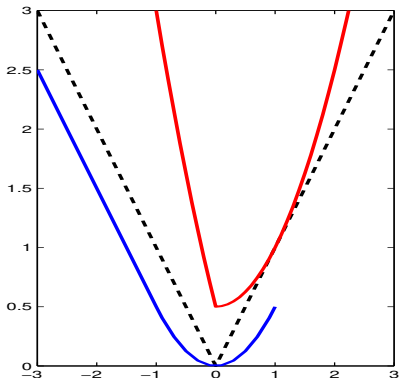
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Proximity operator

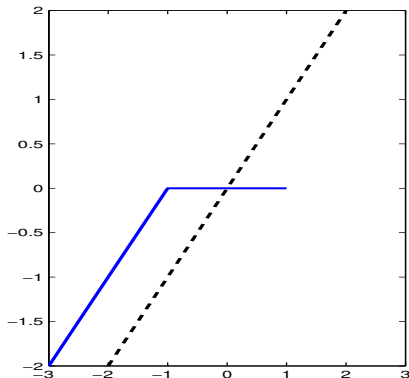
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Proximity operator: definition



Moreau envelope

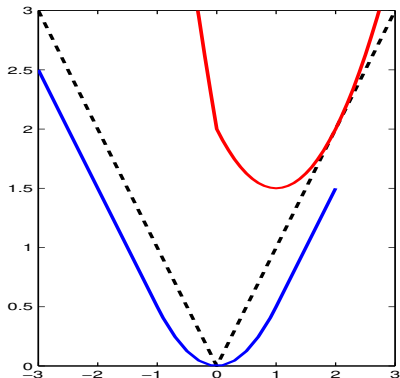
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Proximity operator

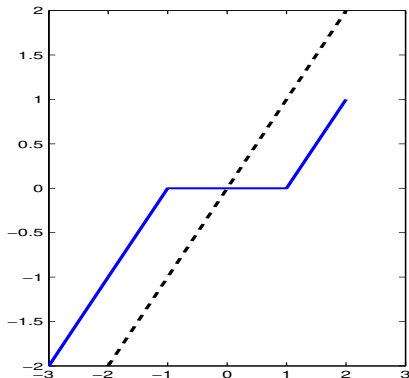
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Proximity operator: definition



Moreau envelope

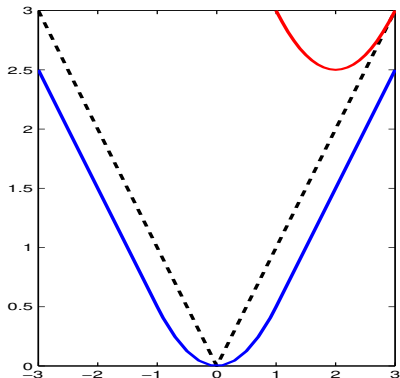
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Proximity operator

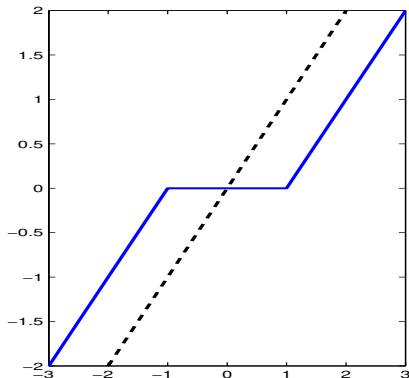
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Proximity operator: definition



Moreau envelope

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Proximity operator

$$\text{prox}_f(x) = \underset{y \in \mathcal{H}}{\operatorname{argmin}} f(y) + \frac{1}{2} \|y - x\|^2$$

Proximity operator: existence and uniqueness

Let $f \in \Gamma_0(\mathcal{H})$ and $\gamma \in]0, +\infty[$.

For every $x \in \mathcal{H}$, there exists a unique vector $p \in \mathcal{H}$ such that

$$f(p) + \frac{1}{2\gamma} \|p - x\|^2 = \inf_{y \in \mathcal{H}} f(y) + \frac{1}{2\gamma} \|y - x\|^2.$$

Proof: $f \in \Gamma_0(\mathcal{H}) \Rightarrow f^* \in \Gamma_0(\mathcal{H})$. Thus, there exists $u \in \mathcal{H}$ such that $f^*(u) \in \mathbb{R}$. According to Fenchel-Young inequality, we have

$$(\forall y \in \mathcal{H}) \quad f(y) \geq \langle u \mid y \rangle - f^*(u).$$

Then, $f(y) + (2\gamma)^{-1} \|y - x\|^2 \rightarrow +\infty$ when $\|y\| \rightarrow +\infty$.

Furthermore $(2\gamma)^{-1} \|\cdot - x\|^2$ being strictly convex, $f + (2\gamma)^{-1} \|\cdot - x\|^2$ is a strictly convex coercive function.

Proximity operator: characterization

Let \mathcal{H} be a Hilbert space and $f \in \Gamma_0(\mathcal{H})$.

$$(\forall x \in \mathcal{H}) \quad p = \text{prox}_f(x) \Leftrightarrow x - p \in \partial f(p) .$$

Proximity operator: characterization

Let \mathcal{H} be a Hilbert space and $f \in \Gamma_0(\mathcal{H})$.

$$(\forall x \in \mathcal{H}) \quad p = \text{prox}_f(x) \Leftrightarrow x - p \in \partial f(p).$$

Proof: By using Fermat's rule, for every $x \in \mathcal{H}$, $p = \text{prox}_f(x)$ if and only if

$$\begin{aligned} p &= \arg \min_{y \in \mathcal{H}} f(y) + \frac{1}{2} \|y - x\|^2 \\ \Leftrightarrow 0 &\in \partial \left(f + \frac{1}{2} \|\cdot - x\|^2 \right)(p) \\ \Leftrightarrow 0 &\in \partial f(p) + p - x \\ \Leftrightarrow x &\in (\text{Id} + \partial f)(p). \end{aligned}$$

Remark:

prox_f is the resolvent of ∂f :

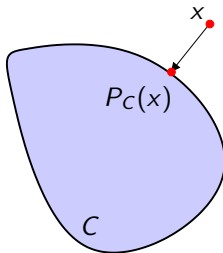
$$\text{prox}_f = (I + \partial f)^{-1} = J_{\partial f}$$

Proximity operator: examples

Projection :

Let \mathcal{H} be a Hilbert space. Let C be a nonempty closed convex subset of \mathcal{H} .

$$(\forall x \in \mathcal{H}) \quad \text{prox}_{\iota_C}(x) = \operatorname{argmin}_{y \in C} \frac{1}{2} \|y - x\|^2 = P_C(x).$$



Proximity operator: examples

Projection :

Let \mathcal{H} be a Hilbert space. Let C be a nonempty closed convex subset of \mathcal{H} .

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Remark :

- ▶ $p = P_C(x) \Leftrightarrow x - p \in \partial \iota_C(p) = N_C(p)$
 $\Leftrightarrow p \in C \text{ and } (\forall y \in C) \langle y - p \mid x - p \rangle \leq 0.$

Particular case: if C is a vector space: $p = P_C(x) \Leftrightarrow \begin{cases} p \in C \\ x - p \in C^\perp \end{cases}.$

- ▶ $\gamma_{\iota_C} = (2\gamma)^{-1} d_C^2$ where d_C distance to the convex set C is defined by $d_C: x \mapsto \inf_{y \in C} \|y - x\| = \|x - P_C x\|.$

Proximity operator: examples

Power q function with $q \geq 1$:

Let $\chi > 0$, $q \in [1, +\infty[$ and $\varphi: \mathbb{R} \rightarrow]-\infty, +\infty] : \xi \mapsto \chi|\xi|^q$.

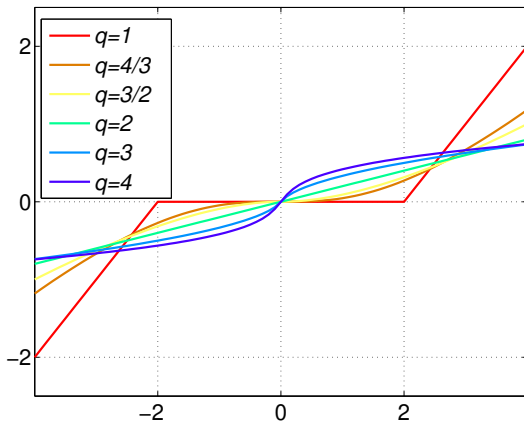
Then, for every $\xi \in \mathbb{R}$,

$$\text{prox}_{\varphi}\xi = \begin{cases} \text{sign}(\xi) \max\{|\xi| - \chi, 0\} & \text{if } q = 1 \\ \xi + \frac{4\chi}{3 \cdot 2^{1/3}} \left((\epsilon - \xi)^{1/3} - (\epsilon + \xi)^{1/3} \right) & \text{if } q = \frac{4}{3} \\ \quad \text{where } \epsilon = \sqrt{\xi^2 + 256\chi^3/729} \\ \xi + \frac{9\chi^2 \text{sign}(\xi)}{8} \left(1 - \sqrt{1 + \frac{16|\xi|}{9\chi^2}} \right) & \text{if } q = \frac{3}{2} \\ \frac{\xi}{1+2\chi} & \text{if } q = 2 \\ \text{sign}(\xi) \frac{\sqrt{1+12\chi|\xi|}-1}{6\chi} & \text{if } q = 3 \\ \left(\frac{\epsilon+\xi}{8\chi} \right)^{1/3} - \left(\frac{\epsilon-\xi}{8\chi} \right)^{1/3} & \text{where } \epsilon = \sqrt{\xi^2 + 1/(27\chi)} \text{ if } q = 4 \end{cases}$$

Proximity operator: examples

Plot of the graphs of these proximity operator on the same figure.

Power q function with $q \geq 1$ and $\chi = 2$.



Proximity operator: examples

Quadratic function :

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

Let $L \in \mathcal{B}(\mathcal{H}, \mathcal{G})$, $\gamma \in]0, +\infty[$ and $z \in \mathcal{G}$.

$$f = \gamma \|L \cdot - z\|^2 / 2 \quad \Rightarrow \quad \text{prox}_f = (\text{Id} + \gamma L^* L)^{-1}(\cdot + \gamma L^* z).$$

► Exercise : Prove this property.

Proximity operator: examples

Quadratic function :

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

Let $L \in \mathcal{B}(\mathcal{H}, \mathcal{G})$, $\gamma \in]0, +\infty[$ and $z \in \mathcal{G}$.

$$f = \gamma \|L \cdot - z\|^2 / 2 \quad \Rightarrow \quad \text{prox}_f = (\text{Id} + \gamma L^* L)^{-1}(\cdot + \gamma L^* z).$$

► Proof: We have, for every $x \in \mathcal{H}$,

$$p = \text{prox}_f x \quad \Leftrightarrow \quad x - p \in \partial f(p).$$

In addition, f is Gâteaux differentiable and its gradient at p is

$$\nabla f(p) = \gamma L^*(Lp - z).$$

Therefore,

$$x - p = \gamma L^*(Lp - z) \quad \Leftrightarrow \quad p = (\text{Id} + \gamma L^* L)^{-1}(x + \gamma L^* z).$$

Proximity operator: properties

Let \mathcal{H} be a Hilbert space, $x \in \mathcal{H}$ and $f \in \Gamma_0(\mathcal{H})$.

Properties	$g(x)$	$\text{prox}_g x$
Translation	$f(x - z), z \in \mathcal{H}$	$z + \text{prox}_f(x - z)$
Quadratic perturbation	$f(x) + \alpha \ x\ ^2 / 2 + \langle z x \rangle + \gamma$ $z \in \mathcal{H}, \alpha > 0, \gamma \in \mathbb{R}$	$\text{prox}_{\frac{f}{\alpha+1}}\left(\frac{x-z}{\alpha+1}\right)$
Scaling	$f(\rho x), \rho \in \mathbb{R}^*$	$\frac{1}{\rho} \text{prox}_{\rho^2 f}(\rho x)$
Reflection	$f(-x)$	$-\text{prox}_f(-x)$
Moreau envelope	$\gamma f(x) = \inf_{y \in \mathcal{H}} f(y) + \frac{1}{2\gamma} \ x - y\ ^2$ $\gamma > 0$	$\frac{1}{1+\gamma} \left(\gamma x + \text{prox}_{(1+\gamma)f}(x) \right)$

Proximity operator: properties

For every $i \in \{1, \dots, n\}$, let \mathcal{H}_i be a Hilbert space and let $f_i \in \Gamma_0(\mathcal{H}_i)$.
If

$$(\forall x = (x_1, \dots, x_n) \in \mathcal{H}_1 \times \dots \times \mathcal{H}_n) \quad f(x) = \sum_{i=1}^n f_i(x_i),$$

then

$$(\forall x = (x_1, \dots, x_n) \in \mathcal{H}_1 \times \dots \times \mathcal{H}_n) \quad \text{prox}_f(x) = (\text{prox}_{f_i}(x_i))_{1 \leq i \leq n}.$$

Proximity operator: properties

Let \mathcal{H} be a separable Hilbert space.

Let $(b_i)_{i \in I}$ be an orthonormal basis of \mathcal{H} .

For every $i \in I$, let $\varphi_i \in \Gamma_0(\mathbb{R})$ such that $\varphi_i \geq 0$. For every $x \in \mathcal{H}$, if

$$f(x) = \sum_{i \in I} \varphi_i(\langle x \mid b_i \rangle)$$

then

$$\text{prox}_f(x) = \sum_{i \in I} \text{prox}_{\varphi_i}(\langle x \mid b_i \rangle) b_i.$$

Remark: The assumption $(\forall i \in I) \varphi_i \geq 0$ can be relaxed if \mathcal{H} is finite dimensional.

Proximity operator: properties

Let \mathcal{H} be a separable Hilbert space.

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For every $i \in I$, let $\varphi_i \in \Gamma_0(\mathbb{R})$ such that $\varphi_i \geq 0$. For every $x \in \mathcal{H}$, if

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then

$$\text{prox}_f(x) = \sum_{i \in I} \text{prox}_{\varphi_i}(\langle x | b_i \rangle) b_i.$$

Example: $\mathcal{H} = \mathbb{R}^N$, $(b_i)_{1 \leq i \leq N}$ canonical basis of \mathbb{R}^N , $f = \lambda \|\cdot\|_1$ with $\lambda \in [0, +\infty[$.

$$(\forall x = (x^{(i)})_{1 \leq i \leq N} \in \mathbb{R}^N) \quad \text{prox}_{\lambda \|\cdot\|_1}(x) = (\text{prox}_{\lambda|\cdot|}(x^{(i)}))_{1 \leq i \leq N}$$

Proximity operator: properties

Moreau decomposition formula

Let \mathcal{H} be a Hilbert space, $f \in \Gamma_0(\mathcal{H})$ and $\gamma \in]0, +\infty[$.

$$(\forall x \in \mathcal{H}) \quad \text{prox}_{\gamma f^*} x = x - \gamma \text{prox}_{\gamma^{-1} f}(\gamma^{-1} x).$$

Proof:

$$\begin{aligned} p = \text{prox}_{\gamma f^*} x &\Leftrightarrow x - p \in \gamma \partial f^*(p) \\ &\Leftrightarrow p \in \partial f\left(\frac{x - p}{\gamma}\right) \\ &\Leftrightarrow \frac{x}{\gamma} - \frac{x - p}{\gamma} \in \frac{1}{\gamma} \partial f\left(\frac{x - p}{\gamma}\right) \\ &\Leftrightarrow \frac{x - p}{\gamma} = \text{prox}_{\gamma^{-1} f}(\gamma^{-1} x) \\ &\Leftrightarrow p = x - \gamma \text{prox}_{\gamma^{-1} f}(\gamma^{-1} x). \end{aligned}$$

Proximity operator: properties

Moreau decomposition formula

Let \mathcal{H} be a Hilbert space, $f \in \Gamma_0(\mathcal{H})$ and $\gamma \in]0, +\infty[$.

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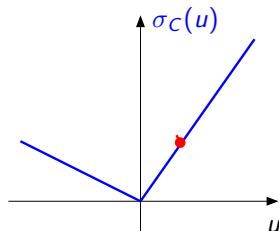
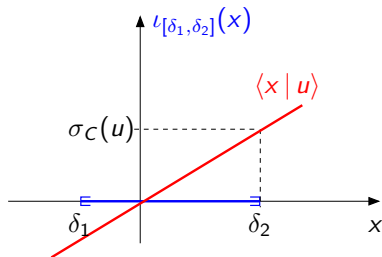
Example: If $\mathcal{H} = \mathbb{R}^N$, $f = \frac{1}{q} \|\cdot\|_q^q$ with $q \in]1, +\infty[$, then $f^* = \frac{1}{q^*} \|\cdot\|_{q^*}^{q^*}$ with $1/q + 1/q^* = 1$, and

$$(\forall x \in \mathbb{R}^N) \quad \text{prox}_{\frac{\gamma}{q^*} \|\cdot\|_{q^*}^{q^*}} x = x - \gamma \text{prox}_{\frac{1}{\gamma q} \|\cdot\|_q^q}(\gamma^{-1} x).$$

Support function: reminders

Let \mathcal{H} be a Hilbert space and $C \subset \mathcal{H}$.
 σ_C is the **support function** of C if

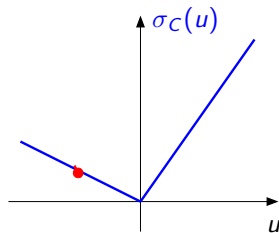
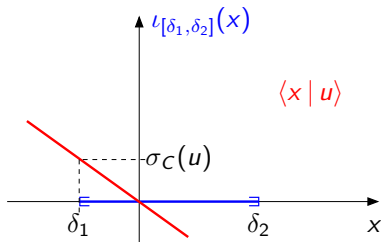
$$(\forall u \in \mathcal{H}) \quad \sigma_C(u) = \sup_{x \in C} \langle x | u \rangle \\ = \iota_C^*(u).$$



Support function: reminders

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Proximity operator: examples

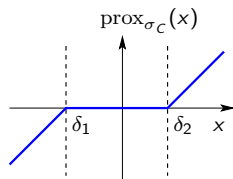
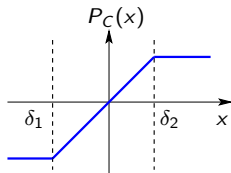
Support function :

Let \mathcal{H} be a Hilbert space and $C \subset \mathcal{H}$ be nonempty closed convex.

$$(\forall x \in \mathcal{H}) \quad \text{prox}_{\sigma_C} = \text{Id} - P_C.$$

Soft-thresholding : $\mathcal{H} = \mathbb{R}$, $\delta_1 = \inf C$ and $\delta_2 = \sup C$. For every $x \in \mathbb{R}$,

$$\sigma_C(x) = \begin{cases} \delta_1 x & \text{if } x < 0 \\ 0 & \text{if } x = 0 \\ \delta_2 x & \text{if } x > 0 \end{cases} \Rightarrow \text{prox}_{\sigma_C}(x) = \text{soft}_C(x) = \begin{cases} x - \delta_1 & \text{if } x < \delta_1 \\ 0 & \text{if } x \in C \\ x - \delta_2 & \text{if } x > \delta_2. \end{cases}$$



Proximity operator: properties

Let \mathcal{H} be a Hilbert spaces. Let $f \in \Gamma_0(\mathcal{H})$ and $L \in \mathcal{B}(\mathcal{H}, \mathcal{H})$ be unitary. Then

$$\text{prox}_{f \circ L} = L^* \circ \text{prox}_f \circ L.$$

Proximity operator: properties

Let \mathcal{H} be a Hilbert spaces. Let $f \in \Gamma_0(\mathcal{H})$ and $L \in \mathcal{B}(\mathcal{H}, \mathcal{H})$ be unitary. Then

$$\text{prox}_{f \circ L} = L^* \circ \text{prox}_f \circ L.$$

Proof: $LL^* = \text{Id} \Rightarrow \text{ran } L = \mathcal{H}.$

Thus $(\forall x \in \mathcal{H}) \ p = \text{prox}_{f \circ L} x \Leftrightarrow x - p \in \partial(f \circ L)(p) = L^* \partial f(Lp).$

This yields

$$Lx - Lp \in \partial f(Lp)$$

$$\Leftrightarrow Lp = \text{prox}_f(Lx)$$

$$\Rightarrow p = L^* Lp = L^* \text{prox}_f(Lx).$$

Proximity operator: properties

Let \mathcal{H} be a Hilbert spaces. Let $f \in \Gamma_0(\mathcal{H})$ and $L \in \mathcal{B}(\mathcal{H}, \mathcal{H})$ be unitary. Then

$$\text{prox}_{f \circ L} = L^* \circ \text{prox}_f \circ L.$$

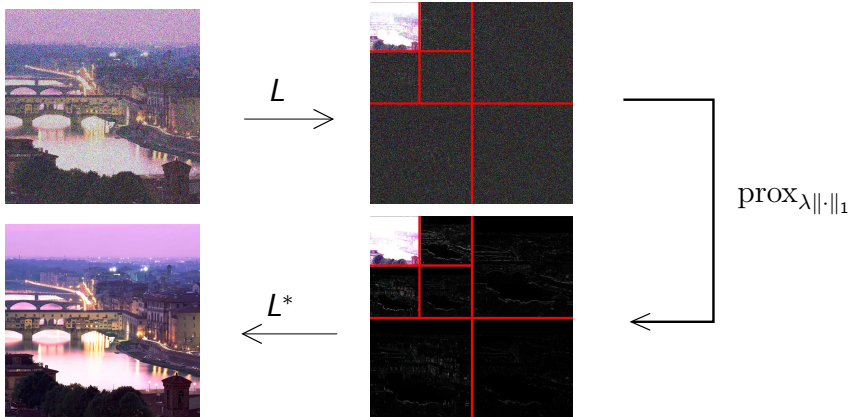
Let \mathcal{H} and \mathcal{G} be two Hilbert spaces. Let $f \in \Gamma_0(\mathcal{H})$ and $L \in \mathcal{B}(\mathcal{G}, \mathcal{H})$ such that $LL^* = \mu \text{Id}$ where $\mu \in]0, +\infty[$. Then

$$\text{prox}_{f \circ L} = \text{Id} - \mu^{-1} L^* \circ (\text{Id} - \text{prox}_{\mu f}) \circ L.$$

Proximity operator: properties

Illustration : $L \in \mathcal{B}(\mathcal{H}, \mathcal{H})$ unitary, $\text{prox}_{f \circ L} = L^* \text{prox}_f L$.

- ▶ Application: denoising using an ℓ_1 penalty on the coefficients resulting from an orthogonal wavelet transform L .



Proximity operator: Bayesian interpretation

- ▶ If $\mathcal{H} = \mathbb{R}^N$ and

$$x = \bar{y} + w$$

where \bar{y} is a realization of a random vector with probability density function $\exp(-f)$ and w is a realization of a $\mathcal{N}(0, I)$ noise, then $\text{prox}_f(x)$ is a Maximum A Posteriori estimate of \bar{y} .

- ▶ Explicit form for objective functions associated with usual log-concave probability densities

- | | |
|--------------------------------|--------------------|
| ▶ Laplace | ▶ Gaussian |
| ▶ Generalized Gaussian | ▶ Huber |
| ▶ maximum entropy | ▶ Smoothed Laplace |
| ▶ gamma | ▶ chi |
| ▶ uniform | ▶ triangular |
| ▶ Weibull | ▶ Pearson type I |
| ▶ Generalized inverse Gaussian | ... |

- ▶ And many other functions ! <http://proximity-operator.net>