

From perspective maps to epi-projections

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Epigraphical projections via proximal map

Let $f \in \Gamma_0 := \{f : \mathbb{E} \rightarrow \mathbb{R} \cup \{+\infty\} \mid f \text{ closed, proper, convex}\}$, $\lambda > 0$:

$$P_\lambda f(x) := \operatorname{argmin}_{u \in \mathbb{E}} \left\{ f(u) + \frac{1}{2\lambda} \|x - u\|^2 \right\} \quad (\text{proximal map}).$$

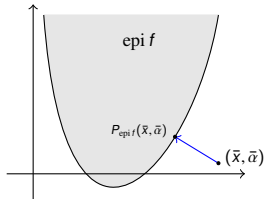
Let $f : \mathbb{E} \rightarrow \mathbb{R}$ convex, $(\bar{x}, \bar{\alpha}) \notin \operatorname{epi} f$.
Then

$$P_{\operatorname{epi} f}(\bar{x}, \bar{\alpha}) = (P_\lambda f(\bar{x}), \bar{\alpha} + \lambda^*),$$

where $\lambda^* > 0$ is a positive root of

$$0 < \lambda \mapsto f(P_\lambda f(\bar{x})) - \lambda - \bar{\alpha}.$$

(Bauschke/Combettes, Theorem 29.35)



- Proximal map vs. projection onto the epigraph in the literature: E.g. Bauschke/Combettes '10, Beck '17, Chierca et al. '15, Meng et al. '05/'08.
- Goal: Understanding the convex-analytic properties of the proximal map (and Moreau envelope) as a function of the prox parameter λ (and the base point x simultaneously)!

A general framework via perspective maps

For $f \in \Gamma_0$, $\omega = \frac{1}{2}\|\cdot\|^2$:

$$P_\lambda f(x) = \operatorname{argmin}_{u \in \mathbb{B}} \left\{ f(u) + \lambda \omega\left(\frac{x-u}{\lambda}\right) \right\} \quad \forall (x, \lambda) \in \mathbb{B} \times \mathbb{R}_{++}.$$

Definition 1 (Perspective map).

For $\omega \in \Gamma_0$, (the closure of) its *perspective map* is $\omega^\pi : \mathbb{B} \times \mathbb{R} \rightarrow \mathbb{R} \cup \{+\infty\}$,




$$\omega^\pi(z, \lambda) = \begin{cases} \lambda \omega\left(\frac{z}{\lambda}\right) & \text{if } \lambda > 0, \\ \omega^\infty(z) & \text{if } \lambda = 0, \\ +\infty & \text{else.} \end{cases}$$

- $\omega^\pi(z, \lambda) = \sigma_{\operatorname{epi} \omega^*}(z, -\lambda)$, hence ω^π is lsc, proper and sublinear with $\operatorname{dom} \omega^\pi = \mathbb{R}_+(\operatorname{dom} \omega \times \{1\})$.
- Literature: E.g. Rockafellar '70, Aravkin et al '18, Combettes/Müller '18/'19.

For $f, \omega \in \Gamma_0$ we want to study the solution map

$$P_{\omega, f} : \mathbb{B} \times \mathbb{R} \rightrightarrows \mathbb{B}, \quad P_{\omega, f}(x, \lambda) = \operatorname{argmin}_{u \in \mathbb{B}} \{f(u) + \omega^\pi(x - u, \lambda)\}.$$

Interlude: Tools from variational analysis

Name	Definition	Properties	Example
tangent cone	$T_A(\bar{x}) := \text{Lim sup}_{t \downarrow 0} \frac{A - \bar{x}}{t}$	closed	
regular normal cone	$\hat{N}_A(\bar{x}) := T_A(\bar{x})^\circ$	closed, convex	
limiting normal cone	$N_A(\bar{x}) := \text{Lim sup}_{x \rightarrow \bar{x}} \hat{N}_A(x)$	closed	

$S : \mathbb{E}_1 \rightrightarrows \mathbb{E}_2, (\bar{x}, \bar{y}) \in \text{gph } S.$

- (Coderivative) $D^*S(\bar{x}|\bar{y}) : \mathbb{E}_2 \rightrightarrows \mathbb{E}_1, v \in D^*S(\bar{x}|\bar{y})(y) : \iff (v, -y) \in N_{\text{gph } S}(\bar{x}, \bar{y}).$
- (Graphical derivative) $DS(\bar{x}|\bar{y}) : \mathbb{E}_1 \rightrightarrows \mathbb{E}_2, v \in DS(\bar{x}|\bar{y})(u) : \iff (u, v) \in T_{\text{gph } S}(\bar{x}, \bar{y}).$

Variational Analysis of $P_{\omega,f}$: Part I

Recall: $P_{\omega,f} : \mathbb{E} \times \mathbb{R} \rightrightarrows \mathbb{E}$, $P_{\omega,f}(x, \lambda) = \operatorname{argmin}_{u \in \mathbb{E}} \{f(u) + \omega^\pi(x - u, \lambda)\}$.

Theorem 2 (Friedlander, Goodwin, H.).

Let $\omega : \mathbb{E} \rightarrow \mathbb{R}$ be strictly convex, level-bounded and C^2 . Then:

- $\operatorname{dom} P_{\omega,f} \subset \mathbb{E} \times \mathbb{R}_+$ and $P_{\omega,f}$ is single-valued for all $\lambda > 0$.
- For $\bar{\lambda} > 0$, if $\nabla^2 \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) > 0$, then $P_{\omega,f}$ is locally Lipschitz at $(\bar{x}, \bar{\lambda})$.
- Under the conditions in b), assume, in addition, that ∂f is proto-differentiable at $\left(P_{\omega,f}(\bar{x}, \bar{\lambda}), \nabla \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) \right)$. Then $P_{\omega,f}$ is directionally differentiable at $(\bar{x}, \bar{\lambda})$ with

$$P'_{\omega,f}((\bar{x}, \bar{\lambda}); (d, \Delta)) = \left[\bar{\lambda} D(\partial f) \left(P_{\omega,f}(\bar{x}, \bar{\lambda}) \middle| \nabla \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) \right) + \bar{V} \right]^{-1} \left(\bar{V} d - \frac{\Delta}{\bar{\lambda}} \bar{V} (\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})) \right)$$

with $\bar{V} = \nabla^2 \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right)$, for any $(d, \Delta) \in \mathbb{E} \times \mathbb{R}$.

- ω strongly convex \Rightarrow level-bounded (in fact, supercoercive) and strictly convex.
- ∂f is proto-differentiable e.g. if $f = g \circ F$, with g PLQ and $F \in C^2$ such that BCQ holds.

Directional normal cone and semismoothness* (Gfrerer et al.)

Directional normal cone¹ of A at \bar{x} in direction \bar{u} :

$$N_A(\bar{x}; \bar{u}) := \text{Lim sup}_{u \rightarrow \bar{u}, t \downarrow 0} \hat{N}_A(\bar{x} + tu).$$

- $N(\bar{x}; \bar{u}) = \emptyset$ if $\bar{u} \notin T_A(\bar{x})$;
- $N(\bar{x}; \bar{u}) \subset N_A(\bar{x})$ for all $u \in \mathbb{E}$.

Definition 3 (Semismoothness*).

- i) $A \subset \mathbb{E}$ *semismooth** at $\bar{x} \in A$: $\iff \langle x^*, u \rangle = 0 \quad \forall u \in \mathbb{E}, x^* \in N_A(\bar{x}; u)$.
- ii) $S : \mathbb{E}_1 \rightrightarrows \mathbb{E}_2$ *semismooth** at $(\bar{x}, \bar{y}) \in \text{gph } S$: $\iff \text{gph } S$ *semismooth** at (\bar{x}, \bar{y}) .

- A convex $\implies A$ *semismooth**;
- A_i *semismooth** at \bar{x} ($i=1, \dots, n$) $\implies \bigcup_{i=1}^n A_i$ *semismooth** at \bar{x} .

¹Mordukhovich and Ginchev

Semismoothness(*) of $P_{\omega,f}$

Proposition 4 (Gfrerer and Outrata '19).

Let $F : D \subset \mathbb{B}_1 \rightarrow \mathbb{B}_2$ be locally Lipschitz at $\bar{x} \in \text{int } D$. Then the following are equivalent:

- F semismooth (in the sense of Qi and Sun) at \bar{x} , i.e. $\lim_{\substack{V \in \partial_C F(\bar{x}+td) \\ t \downarrow 0, d \rightarrow \bar{d}}} Vd$ exists for all \bar{d} .
- F semismooth* and directionally differentiable at \bar{x} .

Proposition 5 (Friedlander, Goodwin, H.).

Let $f \in \Gamma_0$, let $(\bar{x}, \bar{\lambda}) \in \mathbb{B} \times \mathbb{R}_{++}$ such that ∂f is semismooth* at $\left(P_{\omega,f}(\bar{x}, \bar{\lambda}), \nabla \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) \right)$ and let ω be strongly convex and C^2 . Then:

- $P_{\omega,f}$ is semismooth* at $((\bar{x}, \bar{\lambda}), P_{\omega,f}(\bar{x}, \bar{\lambda}))$.
- If, also, ∂f is proto-differentiable at $\left(P_{\omega,f}(\bar{x}, \bar{\lambda}), \nabla \omega \left(\frac{\bar{x} - P_{\omega,f}(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) \right)$, then $P_{\omega,f}$ is semismooth at $(\bar{x}, \bar{\lambda})$.
- $f \in C^2$ or f PLQ $\implies \partial f$ proto-differentiable and semismooth*.
- $\omega = \langle H(\cdot), (\cdot) \rangle$ with $H > 0 \implies \omega$ strongly convex + C^2 .

${}^2\partial_C F(x) := \text{conv} \{ V \mid \exists \{x_k \in D_F\} \rightarrow x : F'(x_k) \rightarrow V \}$ (Clarke Jacobian).

The proximal map revisited I - continuity properties

Proposition 6 (Attouch, Rockafellar/Wets, Friedlander/Goodwin, H.).

Let $f \in \Gamma_0$ and $\bar{x} \in \mathbb{E}$. Then

$$\lim_{\substack{\lambda \downarrow 0, \\ x \rightarrow \bar{x}}} P_\lambda f(x) = P_{\text{cl}(\text{dom } f)}(\bar{x}).$$

Natural extension of the proximal operator of $f \in \Gamma_0$:

$$P_f : \mathbb{E} \times \mathbb{R} \rightrightarrows \mathbb{E}, \quad P_f(x, \lambda) := \begin{cases} P_\lambda f(x), & \lambda > 0, \\ P_{\text{cl}(\text{dom } f)}(x), & \lambda = 0, \\ \emptyset, & \lambda < 0. \end{cases}$$

Corollary 7 (Friedlander/Goodwin/H., Milzarek, Strömberg).

- a) P_f is continuous on $\text{dom } P_f = \mathbb{E} \times \mathbb{R}_+$, and locally Lipschitz on $\text{int}(\text{dom } P_f)$.
- b) For $\bar{x} \in \text{dom } \partial f$ (hence $P_f(\bar{x}, 0) = \bar{x} \in \text{dom } f$), P is calm at $(\bar{x}, 0)$, i.e.

$$\exists \kappa, \varepsilon > 0 : \|P_f(\bar{x}, 0) - P_f(x, \lambda)\| \leq \kappa \|(x - \bar{x}, \lambda)\| \quad ((x, \lambda) \in B_\varepsilon(\bar{x}, 0) \cap \text{dom } P_f).$$

- c) Under the assumptions of b) the map $P_f(\bar{x}, \cdot) : \mathbb{R}_+ \rightarrow \mathbb{E}$ is locally Lipschitz at 0, i.e.

$$\exists \kappa, \varepsilon > 0 : \|P_f(\bar{x}, \lambda) - P_f(\bar{x}, \mu)\| \leq \kappa |\mu - \lambda| \quad \forall \lambda, \mu \in [0, \varepsilon].$$

The proximal map revisited II - directional differentiability

Corollary 8 (Friedlander, Goodwin, H.).

Let $f \in \Gamma_0$ and let $(\bar{x}, \bar{\lambda}) \in \mathbb{E} \times \mathbb{R}_{++}$. If ∂f is proto-differentiable at $(P_f(\bar{x}, \bar{\lambda}), \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}})$, then P_f is directionally differentiable at $(\bar{x}, \bar{\lambda})$ with

$$P_f'((\bar{x}, \bar{\lambda}); (d, \Delta)) = \left[\bar{\lambda} D(\partial f) \left(P_f(\bar{x}, \bar{\lambda}) \left| \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right. \right) + I \right]^{-1} \left(d - \Delta \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right) \quad \forall (d, \Delta) \in \mathbb{E} \times \mathbb{R}.$$

In particular, for any $\bar{\lambda} > 0$, we have

$$(P_{\bar{\lambda}} f)'(\bar{x}; d) = \left[\bar{\lambda} D(\partial f) \left(P_f(\bar{x}, \bar{\lambda}) \left| \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}} \right. \right) + I \right]^{-1} (d) \quad \forall d \in \mathbb{E}.$$

Observe: Projection onto a closed, convex set is not necessarily directionally differentiable. For

$$S = \text{conv} \left[\left\{ \left(\cos \frac{\pi}{2^k}, \sin \frac{\pi}{2^k} \right) \mid k \in \mathbb{N} \right\} \cup \{(0,0), (1,0)\} \right],$$

the projection P_S is not directionally differentiable at $(2, 0)$ in direction $(0, 1)$.

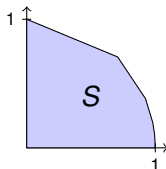


Figure: Shapiro's example (Shapiro '94)

The proximal map revisited III - semismoothness(*) properties

Proposition 9 (Friedlander, Goodwin, H.).

Let $f \in \Gamma_0$ and let $(\bar{x}, \bar{\lambda}) \in \mathbb{B} \times \mathbb{R}_{++}$. Then:

- If ∂f semismooth* at $\left(P_f(\bar{x}, \bar{\lambda}), \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}}\right)$ then P_f is semismooth* at $(\bar{x}, \bar{\lambda})$.
- The map $P_{\bar{\lambda}}f (= P_f(\cdot, \bar{\lambda}))$ is semismooth* at \bar{x} if and only if ∂f is semismooth* at $\left(P_{\bar{\lambda}}f(\bar{x}), \frac{\bar{x} - P_{\bar{\lambda}}f(\bar{x})}{\bar{\lambda}}\right)$.
- If ∂f is proto-differentiable and semismooth* at $\left(P_f(\bar{x}, \bar{\lambda}), \frac{\bar{x} - P_f(\bar{x}, \bar{\lambda})}{\bar{\lambda}}\right)$, then P_f is semismooth at $(\bar{x}, \bar{\lambda})$.

- The assumption in a) are satisfied e.g. if $f = g \circ H$ with g PLQ and $H \in C^2$ such that BCQ holds.
- The assumptions in c) are satisfied if f is PLQ or twice continuously differentiable at $P_f(\bar{x}, \bar{\lambda})$ (in which case P_f is continuously differentiable at $(\bar{x}, \bar{\lambda})$).

The proximal value map $\lambda \mapsto f(P_\lambda f(\bar{x}))$

For $f \in \Gamma_0$ and $\bar{x} \in \mathbb{E}$, consider $0 < \lambda \mapsto f(P_\lambda f(\bar{x}))$.

Example: Let $f := |\cdot| + \delta_{[-1,1]} \in \Gamma_0(\mathbb{R})$. Then

$$P_\lambda f(x) = \min\{\max\{|x| - \lambda, 0\}, 1\} \cdot \text{sgn}(x) \quad (x \in \mathbb{R}, \lambda > 0).$$

For $\bar{x} = 2$:

$$f(P_\lambda f(\bar{x})) = \begin{cases} 1, & \lambda \in (0, 1], \\ 2 - \lambda, & \lambda \in (1, 2], \\ 0, & \lambda > 2. \end{cases}$$

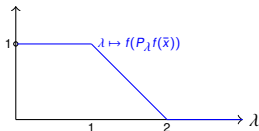


Figure: Nonconvex example

Proposition 10 (Friedlander/Goodwin/H., Attouch '84).

Let $f \in \Gamma_0$ and $\bar{x} \in \mathbb{E}$. Then:

- The function $0 < \lambda \mapsto f(P_\lambda f(\bar{x}))$ is decreasing (i.e. increasing as $\lambda \downarrow 0$).
- $\lim_{\lambda \rightarrow 0} f(P_\lambda f(\bar{x})) = f(P_{\text{cl}(\text{dom } f)}(\bar{x}))$.

The map $\bar{\phi}_{\bar{x}}^f$

Proposition 11 (Friedlander, Goodwin, H.).

Let $f \in \Gamma_0$, $\bar{x} \in \mathbb{B}$ and define $\bar{\phi}_{\bar{x}}^f : \mathbb{R} \rightarrow \bar{\mathbb{R}}$ by

$$\bar{\phi}_{\bar{x}}^f(\lambda) := \begin{cases} -\lambda e_{\lambda} f(\bar{x}), & \lambda > 0,^3 \\ -\frac{1}{2} d_{\text{cl}(\text{dom } f)}^2(\bar{x}), & \lambda = 0, \\ -f(P_{\text{cl}(\text{dom } f)}(\bar{x}))\lambda - \frac{1}{2} d_{\text{cl}(\text{dom } f)}^2(\bar{x}), & \lambda < 0. \end{cases}$$

Then the following hold:

- a) $\bar{\phi}_{\bar{x}}^f$ is proper, convex and continuous (possibly in an extended real-valued sense), and continuously differentiable on \mathbb{R}_{++} with $\frac{d}{d\lambda} \bar{\phi}_{\bar{x}}^f(\lambda) = -f(P_{\lambda} f(\bar{x}))$ locally Lipschitz for all $\lambda > 0$.
- b) If $\bar{x} \in \text{dom } f$, then $\bar{\phi}_{\bar{x}}^f$ is continuously differentiable (on \mathbb{R}) with derivative given by

$$\frac{d}{d\lambda} \bar{\phi}_{\bar{x}}^f(\lambda) = \begin{cases} -f(P_{\lambda} f(\bar{x})), & \lambda > 0, \\ -f(P_{\text{cl}(\text{dom } f)}(\bar{x})), & \lambda \leq 0. \end{cases}$$

If, more strictly, $\bar{x} \in \text{dom } \partial f$, then $\frac{d}{d\lambda} \bar{\phi}_{\bar{x}}^f(\lambda)$ is locally Lipschitz on all of \mathbb{R} .

- c) If $P_{\text{cl}(\text{dom } f)}(\bar{x}) \notin \text{dom } f$, then $\text{dom } \bar{\phi}_{\bar{x}}^f = \mathbb{R}_+$ and $\partial \bar{\phi}_{\bar{x}}^f(\lambda) = \begin{cases} -f(P_{\lambda} f(\bar{x})), & \lambda > 0, \\ \emptyset, & \lambda \leq 0. \end{cases}$

³For $\varphi \in \Gamma_0$, $\lambda > 0$, the function $e_{\lambda} \varphi(\bar{x}) = \varphi(P_{\lambda} \varphi(\bar{x})) + \frac{1}{2\lambda} \|\bar{x} - P_{\lambda} \varphi(\bar{x})\|^2$ is the Moreau envelope of φ at \bar{x} .

Proximal operators for post-compositions

Recall that

$$\phi_{\bar{x}}^{\psi}(\lambda) = \begin{cases} -\lambda e_{\lambda} \psi(\bar{x}), & \lambda > 0, \\ -\frac{1}{2} d_{\text{cl}(\text{dom } \psi)}^2(\bar{x}), & \lambda = 0. \end{cases}$$

Proposition 12 (Friedlander, Goodwin, H.).

Let $g \in \Gamma_0(\mathbb{R})$ be increasing and let $\psi \in \Gamma_0(\mathbb{B})$ such that

$$\text{ri}(\text{dom } g) \cap \psi(\text{ri}(\text{dom } \psi)) \neq \emptyset. \quad (1)$$

Then⁴ $g \circ \psi \in \Gamma_0$ and:

- a) $e_1(g \circ \psi)(\bar{x}) = -\min_{\lambda \geq 0} g^*(\lambda) + \bar{\phi}_{\bar{x}}^{\psi}(\lambda)$;
- b) $P_1(g \circ \psi)(\bar{x}) = P_1(\bar{\lambda} \cdot \psi)(\bar{x})$ for every $\bar{\lambda} \in \text{argmin}_{\lambda \geq 0} \{g^*(\lambda) + \bar{\phi}_{\bar{x}}^{\psi}(\lambda)\} \neq \emptyset$;
- c) If $\psi(P_{\text{cl}(\text{dom } \psi)}(\bar{x})) \notin \partial g^*(0)$, then $\text{argmin}_{\lambda \geq 0} \{g^*(\lambda) + \bar{\phi}_{\bar{x}}^{\psi}(\lambda)\} \subset \mathbb{R}_{++}$. This is, in particular, the case if $P_{\text{cl}(\text{dom } \psi)}(\bar{x}) \notin \text{dom } \psi$.

⁴Set $g(+\infty) := +\infty$.

Level-set and epigraphical projections

Setting $g := \delta_{\mathbb{R}_-}$ and $\psi := f - \bar{\alpha}$ in in Theorem 12 gives:

Corollary 13 (Level set case).

Let $f \in \Gamma_0$, $\bar{\alpha} \in \mathbb{R}$, $\bar{x} \in \mathbb{B}$, and assume that there exists $\hat{x} \in \mathbb{B}$ such that $f(\hat{x}) < \bar{\alpha}$. Then

$$P_{\text{lev}_{\bar{\alpha}} f}(\bar{x}) = \begin{cases} P_{\text{cl}(\text{dom } f)}(\bar{x}), & f(P_{\text{cl}(\text{dom } f)}(\bar{x})) \leq \bar{\alpha}, \\ P_{\bar{\lambda}} f(\bar{x}), & \text{else} \end{cases}$$

for any $0 < \bar{\lambda} \in \text{argmin}_{\lambda \geq 0} \{\bar{\phi}_{\bar{x}}^f(\lambda) + \bar{\alpha}\lambda\} = \{\lambda \geq 0 \mid f(P_{\lambda} f(\bar{x})) = \bar{\alpha}\} \neq \emptyset$.

Letting $g := \delta_{\mathbb{R}_-}$ and $\psi : (x, \alpha) \in \mathbb{B} \times \mathbb{R} \mapsto f(x) - \alpha$ in Theorem 12 yields:

Corollary 14 (Projection onto epigraph).

Let $f \in \Gamma_0$, $\bar{\alpha} \in \mathbb{R}$ and $\bar{x} \in \mathbb{B}$. Then

$$P_{\text{epi } f}(\bar{x}, \bar{\alpha}) = \begin{cases} (P_{\text{cl}(\text{dom } f)}(\bar{x}), \bar{\alpha}), & f(P_{\text{cl}(\text{dom } f)}(\bar{x})) \leq \bar{\alpha}, \\ (P_{\bar{\lambda}} f(\bar{x}), \bar{\alpha} + \bar{\lambda}), & \text{else,} \end{cases}$$

where $\bar{\lambda} > 0$ is the unique minimizer of the (strongly convex) optimization problem

$$\min_{\lambda \geq 0} \frac{1}{2} \lambda^2 + \bar{\alpha} \lambda + \bar{\phi}_{\bar{x}}^f(\lambda).$$

SC¹ optimization framework

For computing level-set and epigraphical projections, respectively, we have to solve the scalar problems

$$\min_{\lambda \geq 0} \theta_{\xi}(\lambda) = \begin{cases} \bar{\alpha} \cdot \lambda + \bar{\phi}_x^f(\lambda), & \xi = \text{lev}, \\ \frac{1}{2}\lambda^2 + \bar{\alpha} \cdot \lambda + \bar{\phi}_x^f(\lambda), & \xi = \text{epi}. \end{cases}$$

The variational properties of ϕ_x^f suggest the SC¹ optimization framework à la Pang and Qi (1995).

Algorithm 1 SC¹ Newton method for minimizing θ_{ξ}

- (S.0) Choose $\lambda_0, \delta > 0, \{\varepsilon_k\} \downarrow 0$, and let $\beta, \sigma \in (0, 1)$. Set $k := 0$.
- (S.1) If $|\theta'(\lambda_k)| \leq \delta$: STOP.
- (S.2) Choose $g_k \in \partial_C(\theta'_{\xi})(\lambda_k)$ and set

$$\Delta_k := P_{[-\lambda_k, \infty)} \left(-\frac{\theta'_{\xi}(\lambda_k)}{g_k + \varepsilon_k} \right).$$

(S.3) Set

$$t_k := \max_{l \in \mathbb{N}_0} \left\{ \beta^l \mid \theta_{\xi}(\lambda_k + \beta^l \Delta_k) \leq \theta_{\xi}(\lambda_k) + \beta^l \sigma \theta'_{\xi}(\lambda) \Delta_k \right\}.$$

(S.4) Set $\lambda_{k+1} := \lambda_k + t_k \Delta_k, k \leftarrow k + 1$, and go to (S.1).

- For epigraphical projections ($\xi = \text{epi}$) use $\varepsilon_k := 0$ to improve accuracy.
- Computational bottleneck: Backtracking in (S.3).

A full Newton step algorithm

Algorithm 2 Full step SC¹ Newton method

- (S.0) Choose $\lambda_0 > 0, \delta > 0$, and $\{\varepsilon_k\} \downarrow 0$. Set $k := 0$.
- (S.1) If $|\theta'_\xi(\lambda_k)| \leq \delta$: STOP.
- (S.2) Choose $g_k \in \partial_C(\theta'_\xi)(\lambda_k)$ and set

$$\Delta_k := \max \left\{ \frac{-\lambda_k}{2}, -\frac{\theta'_\xi(\lambda_k)}{g_k + \varepsilon_k} \right\}.$$

- (S.3) Set $\lambda_{k+1} := \lambda_k + \Delta_k, k \leftarrow k + 1$, and go to (S.1).
-

Slogan: *Take full Newton step in every iteration while respecting positivity of the iterates!*

Set $[\lambda_l, \lambda_u] := \{ \lambda > 0 \mid \theta'_\xi(\lambda) = 0 \}$.

- Assume θ'_ξ concave on $(0, \lambda_l)$.

Then Algorithm 2 converges to a minimizer of θ_ξ .

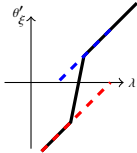


Figure: Cycling without concavity assumption

Projection onto the l_1 -unit ball

For $\text{lev}_1 \|\cdot\|_1 = \{x \in \mathbb{R}^n \mid \|x\|_1 \leq 1\}$ we have

$$\theta'_{\text{lev}}(\lambda) = \begin{cases} 1 - \sum_{i=1}^n \max\{|x_i| - \lambda, 0\}, & \lambda \geq 0, \\ 1 - \|x\|_1, & \lambda < 0, \end{cases}$$

which has θ'_{lev} concave on $(0, \lambda_l)$.

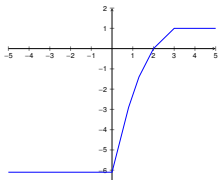


Figure: The function θ'_{ξ} for projecting onto the l_1 -unit ball.

Table: Time (seconds) for projecting onto the l_1 -unit ball in dimension N with coordinates chosen using a (component-wise) Gaussian distribution with $\sigma = 0.1$.

N	Warm Newton	Condat	IBIS
20	1.44×10^{-6}	1.53×10^{-6}	1.83×10^{-6}
10^3	1.83×10^{-5}	2.11×10^{-5}	3.65×10^{-5}
10^6	1.38×10^{-2}	1.44×10^{-2}	2.89×10^{-2}
10^7	1.51×10^{-1}	1.43×10^{-1}	2.85×10^{-1}

Takeaway: Our algorithm performs better than alternative methods when the starting point is good!

Future directions

- Understand semismoothness* of ∂f better:
 - Establish more powerful checkable conditions;
 - extend to non-convex functions f ;
 - extend to (maximally) monotone operators.
- Generalize the result for post-composition setting to the convex convex-composite setting (e.g. Burke, H., Nguyen, MOR 2020): $f : \mathbb{E}_1 \rightarrow \mathbb{R} \cup \{+\infty\}$ with $f = g \circ H$
 - $K \subset \mathbb{E}_2$ (closed) convex cone;
 - $H : \mathbb{E}_1 \rightarrow \mathbb{E}_2^*$ K -convex, i.e. $\{(X, Y) \mid Y - H(x) \in K\}$ (K -epigraph) is convex;
 - $g \in \Gamma_0(\mathbb{E}_2)$ K -increasing.

Goal: Establish theory and algorithms for computing proximal operators of $g \circ F$ and K -epigraphical projections.

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