



# Asymptotic behaviour of the composition of two prox operators

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# An Analysis Problem



Let  $r$  be a positive constant and  $c_0 \geq 0$ . Consider the iteration

$$c_{n+1} = c_n + r - \frac{c_n}{\sqrt{1 + c_n^2}}.$$

- (a) For which values of  $r$  does the sequence  $(c_n)$  converge?
- (b) In case of convergence to  $c$  with  $c \neq c_0$ , prove that  $\lim(c_{n+1} - c)/(c_n - c)$  exists and determine its value.
- (c) In case of divergence, find an asymptotic expression for  $c_n$ .



# An Analysis Problem



- This is D. Borwein and J. Borwein's Problem 10335 in *American Mathematical Monthly*, Vol. 100, 1993.
- Solution was used in a paper by HB and J. Borwein from 1994. In fact, the Acknowledgment of this paper reads:

*The authors wish to thank David Borwein for discussion of Example 5.3, Judith Borwein for preparing the manuscript, and two anonymous referees for helpful suggestions.*

And our affiliations were Dalhousie and Waterloo!



# Overview

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1. MOTIVATION
2. BREGMAN OBJECTS
3. BREGMAN RESULTS
4. REFERENCES



# Collaborators

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Based on joint works with:

- *Patrick L. Combettes* (Paris 6, France),
- *Dominikus Noll* (Toulouse, France).



# 1. MOTIVATION

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# Set up



Throughout,

$$X = \mathbb{R}^J$$

with

inner product  $\langle x, y \rangle$  and norm  $\|x\| = \sqrt{\langle x, x \rangle}$ ,

for  $x$  and  $y$  in  $X$ . Also, the proper lower semicontinuous convex functions on  $X$  are denoted by

$$\Gamma_0(X).$$



# Alternating projections



Suppose

$A$  and  $B$  are nonempty closed convex sets in  $X$ ,  
with corresponding **projectors** (nearest-point mappings)

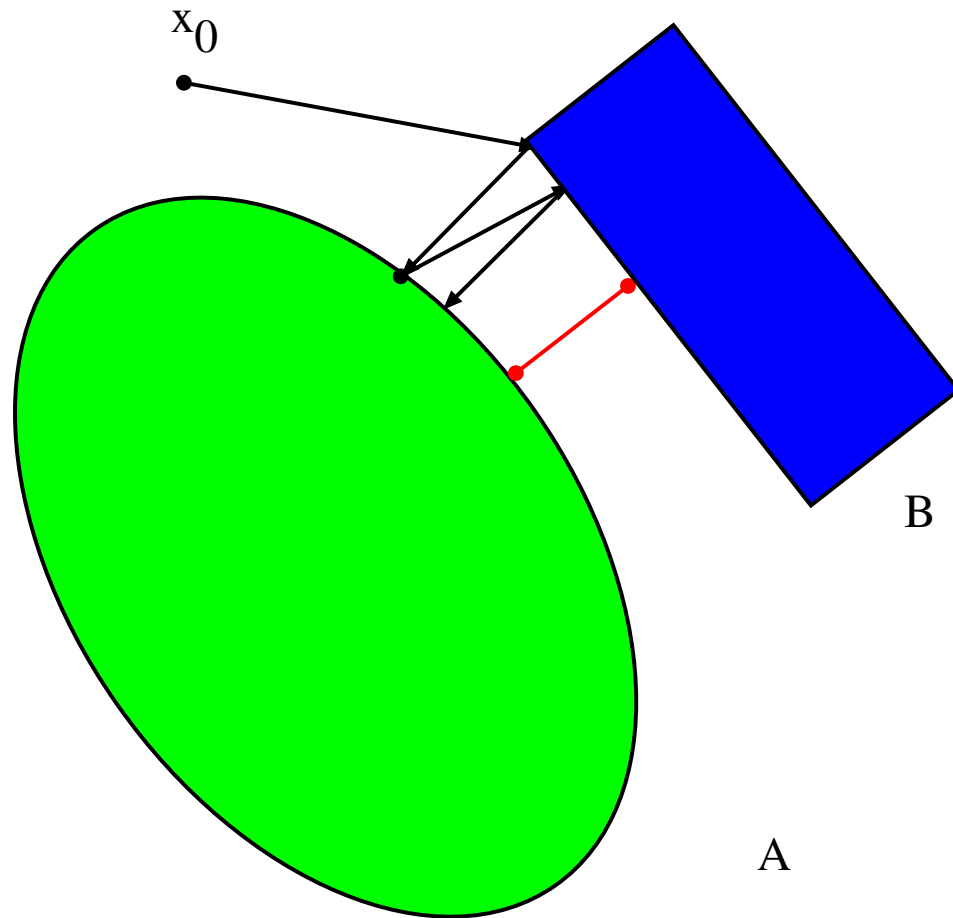
$$P_A \text{ and } P_B.$$

Given a starting point  $x_0 \in X$ , the method of **alternating projections** generates sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}}$  by

$$(\forall n \in \mathbb{N}) \quad y_n = P_B(x_n) \quad x_{n+1} = P_A(y_n).$$







# Basic convergence result



**Theorem.** (Cheney-Goldstein 1959, ...) Suppose the **gap**

$$\gamma := \inf \|A - B\|$$

between  $A$  and  $B$  is attained. Then:

$(x_n)_{n \in \mathbb{N}}$  converges to  $\bar{x} \in A$ ,  $(y_n)_{n \in \mathbb{N}}$  converges to  $\bar{y} \in B$ , and  $\|\bar{x} - \bar{y}\| = \gamma$ .

*Remark.* True in Hilbert space with weak convergence — but not norm convergence, thanks to Hundal.



# Observations on the limits



- Fixed point characterization:  $\bar{x}$  and  $\bar{y}$  satisfy  $\bar{x} = P_A P_B \bar{x}$  and  $\bar{y} = P_B \bar{x} = P_B P_A \bar{y}$
- The dual solution is

$$v := P_{\overline{B-A}}(0) \equiv \bar{y} - \bar{x},$$

i.e., the nearest point to 0 in the closure of the Minkowski difference  $B - A$ . Note that  $\|v\|$  is exactly the gap  $\gamma$ !



# Observations on the limits



- $(\bar{x}, \bar{y})$  solves the **optimization problem**

$$\text{minimize } (x, y) \mapsto \iota_A(x) + \iota_B(y) + \frac{1}{2}\|x - y\|^2.$$

Here  $\iota_A$  and  $\iota_B$  are **indicator functions**, defined by

$$\iota_C(x) := \begin{cases} 0, & \text{if } x \in C; \\ +\infty, & \text{otherwise.} \end{cases}$$



# Moreau envelope



Let  $\theta \in \Gamma_0(X)$ . The **Moreau envelope** of  $\theta$  at  $z$  is

$$\text{env}_\theta(z) := (\theta \square \frac{1}{2} \|\cdot\|^2)(z) := \inf_{w \in X} \theta(w) + \frac{1}{2} \|z - w\|^2.$$

This operation regularizes  $\theta$ . For instance, if  $\theta = \iota_C$ , then

$$\text{env}_{\iota_C}(z) = \inf_{w \in X} \iota_C(w) + \frac{1}{2} \|z - w\|^2 = \inf_{c \in C} \frac{1}{2} \|z - c\|^2$$

is  $\frac{1}{2} \cdot$  the square of the *distance* of  $z$  to  $C$ .



# Proximity operator



The infimum in the definition of  $\text{env}_\theta(z)$ , i.e.,

$$\text{env}_\theta(z) = \inf_{w \in X} \theta(w) + \frac{1}{2} \|z - w\|^2,$$

is *always uniquely* attained! The induced map

$$\text{prox}_\theta: X \rightarrow X: z \mapsto w_z := \underset{w \in X}{\text{argmin}} \theta(w) + \frac{1}{2} \|z - w\|^2$$

is called the **proximity operator** or **proximal map** of  $\theta$ .





Note that  $0 \in \partial\theta(w_z) - (z - w_z) \Leftrightarrow z \in (\text{Id} + \partial\theta)(w_z)$ ;  
equivalently,

$$w_z = \text{prox}_\theta(z) = (\text{Id} + \partial\theta)^{-1}(z).$$

Since  $\partial\theta$  is *maximal monotone*, the operator

$\text{prox}_\theta$  is *firmly nonexpansive*

and everywhere defined.

If  $\theta = \iota_C$ , then  $\text{prox}_{\iota_C} = P_C$  (projector onto the set  $C$ ).



# Alternating prox operators



Let  $\varphi, \psi$  be in  $\Gamma_0(X)$  and  $x_0 \in X$ . Consider the method of *alternating prox operators*:

$$(\forall n \in \mathbb{N}) \quad y_n := \text{prox}_\psi(x_n) \quad x_{n+1} := \text{prox}_\varphi(y_n).$$

**Theorem.** (Acker-Prestel 1980)

$x_n \xrightarrow{\text{weak}} \bar{x} \in X$  and  $y_n \xrightarrow{\text{weak}} \bar{y} \in X$ , where  $(\bar{x}, \bar{y})$  is a solution of the optimization problem

$$\text{minimize } (x, y) \mapsto \varphi(x) + \psi(y) + \frac{1}{2} \|x - y\|^2.$$

*Remark.*  $\varphi = \iota_A, \psi = \iota_B$  yields alternating projections.





# Purpose of this talk



- Consider the objective function

$$X \rightarrow ]-\infty, +\infty] : (x, y) \mapsto \varphi(x) + \psi(y) + \frac{1}{2}\|x - y\|^2.$$

What happens if we replace

$$\frac{1}{2}\|x - y\|^2$$

by some other “distance-like” term?



# 2. BREGMAN OBJECTS

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Suppose that

$$X = \mathbb{R}^J$$

and

$f \in \Gamma_0(X)$  is differentiable on  $U := \text{int dom } f \neq \emptyset$ .

Then the **Bregman distance**  $D = D_f: X \times X \rightarrow [0, +\infty]$  corresponding to  $f$  is defined by

$$(x, y) \mapsto \begin{cases} f(x) - f(y) - \langle f'(y), x - y \rangle, & \text{if } y \in U; \\ +\infty, & \text{otherwise.} \end{cases}$$

 *Remark.*  $D$  is not a distance in the sense of topology.

# Further assumptions on $f$



We assume that  $f$  satisfies the following:

**A1**  $f$  is of *Legendre* type;

**A2**  $f''$  exists and is continuous on  $U$ ;

**A3**  $D$  is *jointly convex*, i.e., it is convex on  $X \times X$ ;

**A4**  $(\forall x \in U) D(x, \cdot)$  is strictly convex on  $U$ ;

**A5**  $(\forall x \in U) D(x, \cdot)$  is *coercive*, i.e., it has bounded lower level sets.



# Examples



Write  $x = (\xi_j)$  and  $y = (\eta_j)$ . Then the following functions satisfy all assumptions on  $f$ :

(i) If  $f$  is the **energy**  $x \mapsto \frac{1}{2}\|x\|^2$ , then  $U = X$  and

$$D(x, y) = \frac{1}{2}\|x - y\|^2.$$

(ii) If  $f$  is the **negative entropy**  $x \mapsto \sum_j \xi_j \ln(\xi_j) - \xi_j$ , then  $U = \{x \in X : x > 0\}$  and

$$D(x, y) = \begin{cases} \sum_j \xi_j \ln(\xi_j/\eta_j) - \xi_j + \eta_j, & \text{if } x \geq 0 \text{ and } y > 0; \\ +\infty, & \text{otherwise.} \end{cases}$$



# Remarks



- Note that the setting considered earlier is covered, since  $D(x, y) = \frac{1}{2} \|x - y\|^2$  when  $f$  is the energy.
- When  $f$  is the negative entropy, the term  $D(x, y)$  is known as the **Kullback-Leibler information divergence** in statistics and information theory.
- Other examples are:
  - (iii) the *Fermi-Dirac entropy* and
  - (iv) the *log-quad function*.
- The function  $f = -\ln$  has many good properties, but it does *not* satisfy all our assumptions.



# Exploiting joint convexity



Since  $D$  is jointly convex (**A3**), its Bregman distance  $D_D$  is nonnegative.

**Fact.** (B-Noll 2002).

Take  $\{x, y, u, v\} \subset U$ . Then:

$$0 \leq D_D((x, y), (u, v)) = D(x, y) + D(x, u) - D(x, v) \\ + \langle f''(v)(u - v), y - v \rangle.$$





Moreover:

(i) If  $f$  is the *energy*, then

$$D_D((x, y), (u, v)) = D(x, y + (u - v)).$$

(ii) If  $f$  is the *negative entropy*, then

$$D_D((x, y), (u, v)) = D(x, yu/v),$$

where the product and quotient is taken coordinate-wise.







In general,  $D$  is *not* symmetric; consequently, we expect **two** envelopes for a given function  $\theta \in \Gamma_0(X)$ .

The **backward Bregman envelope** of  $\theta$  is

$$\overleftarrow{\text{env}}_{\theta}: X \rightarrow [-\infty, +\infty] : z \mapsto \inf_{w \in X} \theta(w) + D(w, z),$$

and the **forward Bregman envelope** of  $\theta$  is

$$\overrightarrow{\text{env}}_{\theta}: X \rightarrow [-\infty, +\infty] : z \mapsto \inf_{w \in X} \theta(w) + D(z, w).$$



# Examples



- If  $f$  is the *energy*, then backward & forward Bregman envelope coincide with the Moreau envelope.
- If  $\theta = \iota_C$  for some closed convex set  $C$ , then we obtain the **backward Bregman distance**

$$\overleftarrow{D}_C := \overleftarrow{\text{env}}_{\iota_C} : z \mapsto \inf_{c \in C} D(c, z)$$

and the **forward Bregman distance**

$$\overrightarrow{D}_C := \overrightarrow{\text{env}}_{\iota_C} : z \mapsto \inf_{c \in C} D(z, c).$$



# Definitions



Let  $\theta \in \Gamma_0(X)$  such that  $\text{dom } \theta \cap U \neq \emptyset$ . Under reasonable assumptions, we have:

(i) The **backward proximity operator** is well-defined by

$$\overleftarrow{\text{prox}}_{\theta} : U \rightarrow U : y \mapsto \underset{x \in X}{\text{argmin}} \theta(x) + D(x, y).$$

(ii) The **forward proximity operator** is well-defined by

$$\overrightarrow{\text{prox}}_{\theta} : U \rightarrow U : x \mapsto \underset{y \in X}{\text{argmin}} \theta(y) + D(x, y).$$





**Proposition.** (“the backward prox is very nice”)

Suppose  $\theta$  is nice and  $(x, y) \in U \times U$ . Then TFAE:

- $x = \overleftarrow{\text{prox}}_{\theta}(y)$ ;
- $0 \in \partial\theta(x) + f'(x) - f'(y)$ ;
- $(\forall z \in X) \quad \langle f'(y) - f'(x), z - x \rangle + \theta(x) \leq \theta(z)$ .

Moreover,

$$\overleftarrow{\text{prox}}_{\theta} = (f' + \partial\theta)^{-1} \circ f'$$

is continuous on  $U$ , and

$$\nabla \overleftarrow{\text{env}}_{\theta}(y) = f''(y)(y - \overleftarrow{\text{prox}}_{\theta}(y)).$$





**Proposition.** (“the forward prox is just nice”)

Suppose  $\theta$  is nice and  $(x, y) \in U \times U$ . Then TFAE:

- $y = \overrightarrow{\text{prox}}_{\theta}(x)$ ;
- $0 \in \partial\theta(y) + f''(y)(y - x)$ ;
- $(\forall z \in X) \quad \langle f''(y)(x - y), z - y \rangle + \theta(y) \leq \theta(z)$ .

Moreover,  $\overrightarrow{\text{prox}}_{\theta}$  is continuous on  $U$ , and

$$\nabla \overrightarrow{\text{env}}_{\theta}(x) = f'(x) - f'(\overrightarrow{\text{prox}}_{\theta}(x)).$$

*Remark.* Both propositions extend Moreau’s results.



# 3. BREGMAN RESULTS

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# Optimization problem



Throughout, let  $\varphi, \psi$  in  $\Gamma_0(X)$  be sufficiently “nice”, and consider the *optimization problem*

minimize  $\Lambda : (x, y) \mapsto \varphi(x) + \psi(y) + D(x, y)$  over  $U \times U$ .

Denote the *optimal value* and the set of *solutions* by

$p := \inf \Lambda(U \times U)$  and  $S := \{(x, y) \in U \times U : \Lambda(x, y) = p\}$ ,

respectively. We assume that

$$p \in \mathbb{R}.$$



# Alternating prox operators



In view of the characterization

$$(x, y) \in S \Leftrightarrow (x = \overleftarrow{\text{prox}}_{\varphi}(y) \text{ and } y = \overrightarrow{\text{prox}}_{\psi}(x)),$$

for any  $(x, y) \in U \times U$ , we propose to find a solution in  $S$  via the **method of alternating prox operators** with starting point  $x_0 \in X$ :

$$(\forall n \in \mathbb{N}) \quad y_n := \overrightarrow{\text{prox}}_{\psi}(x_n), \quad x_{n+1} := \overleftarrow{\text{prox}}_{\varphi}(y_n). \quad (\text{APO})$$





# Some inequalities



Suppose  $((x_n, y_n))_{n \in \mathbb{N}}$  is generated by (APO),  $n \in \mathbb{N}$ , and  $\{x, y\} \subset U$ . Then

$$\Lambda(x_{n+1}, y_{n+1}) \leq \Lambda(x_{n+1}, y_n) \leq \Lambda(x_n, y_n) \rightarrow \lambda, \quad (1)$$

and

$$\begin{aligned} D(x, x_{n+1}) \leq D(x, x_n) - D_D((x, y), (x_n, y_n)) \\ - (\Lambda(x_{n+1}, y_n) - \Lambda(x, y)). \end{aligned} \quad (2)$$

(*Proof.* Combine prox characterizations with Fact on  $D_D$ .)



# Convergence in value



**Corollary.**

$$\lambda = \lim \Lambda(x_n, y_n) = \lim \Lambda(x_{n+1}, y_n) = p.$$

*Proof.* Clearly,

$$\lambda = \inf_{n \in \mathbb{N}} \Lambda(x_n, y_n) = \inf_{n \in \mathbb{N}} \Lambda(x_{n+1}, y_n) \geq \inf \Lambda(U \times U) = p.$$

Assume  $\lambda > p$ . Then obtain  $(x, y) \in U \times U$  such that  $\lambda = \Lambda(x, y) + \epsilon$ , where  $\epsilon > 0$ . Now (2) implies

$$(\forall n \in \mathbb{N}) \quad \epsilon = \lambda - \Lambda(x, y) \leq D(x, x_n) - D(x, x_{n+1}).$$

 Telescoping this yields a contradiction. Hence  $\lambda = p$ . 

# Bregman convergence result



**Theorem.** (B-Combettes-Noll 2004).

Suppose  $(x, y) \in S \neq \emptyset$ . Then:

$$\sum_{n \in \mathbb{N}} (\Lambda(x_{n+1}, y_n) - p) < +\infty,$$

$$\sum_{n \in \mathbb{N}} D_D((x, y), (x_n, y_n)) < +\infty,$$

and  $((x_n, y_n))_{n \in \mathbb{N}}$  converges to some point in  $S$ .



# Geometry of solutions



Let  $(x, y)$ ,  $(\tilde{x}, \tilde{y})$  be in  $S$ , and  $x_0 = \tilde{x}$ . Then  $(x_n)_{n \in \mathbb{N}} = (\tilde{x})_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}} = (\tilde{y})_{n \in \mathbb{N}}$ . Thus the Theorem yields the invariance

$$D_D((x, y), (\tilde{x}, \tilde{y})) = 0$$

(This does *not* imply  $(x, y) = (\tilde{x}, \tilde{y})$ .) In particular:

(i) If  $f$  is the energy, then  $D(x, y + (\tilde{x} - \tilde{y})) = 0$ , i.e.,

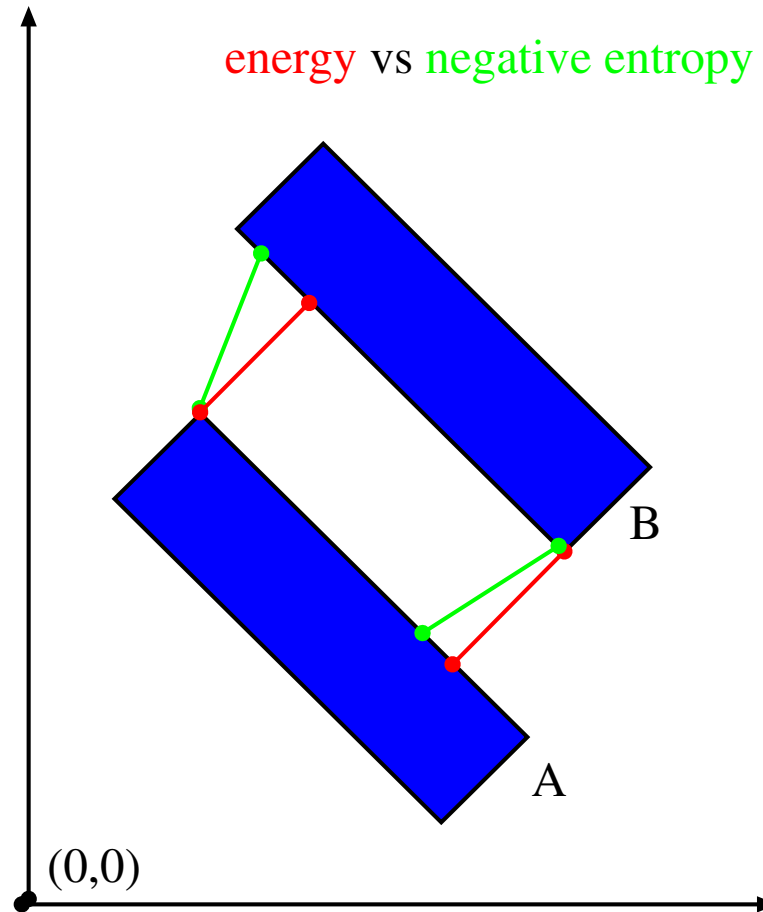
$$x - y = \tilde{x} - \tilde{y}.$$

(ii) If  $f$  is the negative entropy, then  $D(x, y\tilde{x}/\tilde{y}) = 0$ , i.e.,

$$x/y = \tilde{x}/\tilde{y}.$$



# The invariance visualized



# Applications



**Corollary.** (Acker-Prestel 1980)  
Alternating (regular) prox operators . . . .  
(*Proof.* Let  $f$  be the energy. ■)

**Corollary.** (Csiszár-Tusnády 1984)  
Alternating “entropic projections”:  $x_0 > 0$  and

$$(\forall n \in \mathbb{N}) \quad y_n := \overrightarrow{P}_B(x_n), \quad x_{n+1} := \overleftarrow{P}_A(y_n).$$

Then  $(x_n, y_n) \rightarrow (\tilde{x}, \tilde{y})$ , a Kullback-Leibler gap pair.  
(*Proof.* Let  $f$  be the negative entropy,  $\varphi = \iota_A$ ,  $\psi = \iota_B$ . ■)

*Remark.* Related to *Expectation-Maximization method*.



# Applications



Suppose  $\theta \in \Gamma_0(X)$  “nice” and assume

$$\emptyset \neq M := \text{minimizers of } \theta \text{ over } U.$$

**Corollary.** (Censor-Zenios 1992)

The sequence  $(z_n)_{n \in \mathbb{N}}$  generated by the **backward proximal point iteration**

$$z_0 \in U, \quad (\forall n \in \mathbb{N}) \quad z_{n+1} = \overleftarrow{\text{prox}}_{\theta}(z_n)$$

converges to a point in  $M$ .

(*Proof.* Set  $\varphi = \theta$  and  $\psi = 0$ , then  $\overrightarrow{\text{prox}}_{\psi} = \text{Id}$ . ■)



# Applications



**Corollary.** (B-Combettes-Noll 2004)

The sequence  $(z_n)_{n \in \mathbb{N}}$  generated by the **forward proximal point iteration**

$$z_0 \in U, \quad (\forall n \in \mathbb{N}) \quad z_{n+1} = \overrightarrow{\text{prox}}_{\theta}(z_n)$$

converges to a point in  $M$ .

(*Proof.* Set  $\varphi = 0$  and  $\psi = \theta$ , then  $\overleftarrow{\text{prox}}_{\varphi} = \text{Id}$ . ■)

*Remark.* New parallel applications arise via a product space technique!





# 4. REFERENCES



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