

# Quasi Bregman Monotonicity and Applications

Quang. V. NGUYEN  
(under the supervision of Patrick L. COMBETTES)

Laboratoire Jacques-Louis Lions  
Université Pierre et Marie Curie – Paris 6  
75005 Paris, France

Sestri Levante, September 2014

# Fejér Monotonicity in Hilbert spaces

- Notation: given  $C$  a nonempty subset and  $(x_n)_{n \in \mathbb{N}}$  a sequence in  $\mathcal{H}$ .
  - $\mathfrak{W}(x_n)_{n \in \mathbb{N}}$ : set of weak accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $\mathfrak{S}(x_n)_{n \in \mathbb{N}}$ : set of strong accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $d_C(x) = \inf \{ \|y - x\| \mid y \in C \}$ .

# Fejér Monotonicity in Hilbert spaces

- Notation: given  $C$  a nonempty subset and  $(x_n)_{n \in \mathbb{N}}$  a sequence in  $\mathcal{H}$ .
  - $\mathfrak{W}(x_n)_{n \in \mathbb{N}}$ : set of weak accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $\mathfrak{S}(x_n)_{n \in \mathbb{N}}$ : set of strong accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $d_C(x) = \inf \{ \|y - x\| \mid y \in C \}$ .
- Fejér monotonicity:

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|^2 \leq \|x - x_n\|^2.$$

# Fejér Monotonicity in Hilbert spaces

- Notation: given  $C$  a nonempty subset and  $(x_n)_{n \in \mathbb{N}}$  a sequence in  $\mathcal{H}$ .
  - $\mathfrak{W}(x_n)_{n \in \mathbb{N}}$ : set of weak accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $\mathfrak{S}(x_n)_{n \in \mathbb{N}}$ : set of strong accumulation points of  $(x_n)_{n \in \mathbb{N}}$ .
  - $d_C(x) = \inf \{ \|y - x\| \mid y \in C \}$ .
- Fejér monotonicity:

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|^2 \leq \|x - x_n\|^2.$$

- Properties:
  - $(d_C(x_n))_{n \in \mathbb{N}}$  converges.
  - $x_n \rightarrow \bar{x} \in C \Leftrightarrow \mathfrak{W}(x_n)_{n \in \mathbb{N}} \subset C$ .
  - $x_n \rightarrow \bar{x} \in C \Leftrightarrow \underline{\lim} d_C(x_n) = 0$ .

# Quasi Fejér Monotonicity in Hilbert spaces

- Quasi Fejér monotonicity:

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|^2 \leq (1 + \eta_n)\|x - x_n\|^2 + \varepsilon_n,$$

where  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$  and  $(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ .

# Quasi Fejér Monotonicity in Hilbert spaces

- Quasi Fejér monotonicity:

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|^2 \leq (1 + \eta_n)\|x - x_n\|^2 + \varepsilon_n,$$

where  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$  and  $(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ .

- Asymptotic properties: mostly preserved.

# Quasi Fejér Monotonicity in Hilbert spaces

- Quasi Fejér monotonicity:

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|^2 \leq (1 + \eta_n)\|x - x_n\|^2 + \varepsilon_n,$$

where  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$  and  $(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ .

- Asymptotic properties: mostly preserved.
- Stochastic version (initiated by Ermol'ev 1968; P. L. Combettes and J. C. Pesquet 2014): Given a separable Hilbert space  $\mathcal{H}$  together with its usual Borel  $\sigma$ -algebra, a probability space  $(\Omega, \mathcal{F}, \mu)$ ,  $(x_n)_{n \in \mathbb{N}}$  is a sequence of  $\mathcal{H}$ -valued random variables,  $\mathcal{X} = (\mathcal{X}_n)_{n \in \mathbb{N}}$  is canonic filtration of  $(x_n)_{n \in \mathbb{N}}$

$$(\forall z \in C)(\exists(\eta_n(z))_{n \in \mathbb{N}} \in \ell_+^1(\mathcal{X})(\exists(\varepsilon_n(z))_{n \in \mathbb{N}} \in \ell_+^1(\mathcal{X})(\forall n \in \mathbb{N}) \\ E(\|z - x_{n+1}\|^2 | \mathcal{X}_n) \leq (1 + \eta_n(z))\|z - x_n\|^2 + \varepsilon_n(z), \mu - \text{as},$$

where  $\ell_+^1(\mathcal{X})$  is the set of all summable  $[0, +\infty[$ -valued stochastic process which are adapted to  $\mathcal{X}$ .

# Variable metric quasi Fejér monotonicity: definition (P. L. Combettes and B. C. Vũ 2013)

- These notions have seen extensively used to obtain convergence proofs of wide array of algorithms in a unified and simplified fashion.



# Variable metric quasi Fejér monotonicity: definition (P. L. Combettes and B. C. Vũ 2013)

- These notions have seen extensively used to obtain convergence proofs of wide array of algorithms in a unified and simplified fashion.
- A self-adjoint operator  $U$  in  $\mathcal{B}(H)$  with property

$$(\forall x \in \mathcal{H}) \langle x | Ux \rangle \geq \alpha \|x\|^2, \text{ where } \alpha \in ]0, +\infty[$$

induces a scalar product and a norm

$$\langle x | y \rangle_U := \langle x | Uy \rangle \text{ and } \|\cdot\|_U := \sqrt{\langle \cdot | U \cdot \rangle}.$$

# Variable metric quasi Fejér monotonicity: analysis

- Variable metric Quasi-Fejér monotone sequence :

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|_{U_{n+1}}^2 \leq (1 + \eta_n) \|x - x_n\|_{U_n}^2 + \varepsilon_n,$$

where  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$  and  $(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ .

# Variable metric quasi Fejér monotonicity: analysis

- Variable metric Quasi-Fejér monotone sequence :

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x - x_{n+1}\|_{U_{n+1}}^2 \leq (1 + \eta_n) \|x - x_n\|_{U_n}^2 + \varepsilon_n,$$

where  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$  and  $(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ .

- Properties: set  $d_C^U(x) = \inf \{ \|y - x\|_U \mid y \in C \}$ .

1  $(d_C^{U_n}(x_n))_{n \in \mathbb{N}}$  is quasi monotone.

2 Weak convergence:  $x_n \rightarrow \bar{x} \Leftrightarrow \mathfrak{W}(x_n)_{n \in \mathbb{N}} \subset C$ .

3 Strong convergence:  $x_n \rightarrow \bar{x} \Leftrightarrow \underline{\lim} d_C(x_n) = 0$ .

# From Hilbert spaces to Banach spaces

## ■ Notation:

- $\mathcal{X}, \mathcal{Y}$ : reflexive real Banach spaces.
- $\langle \cdot, \cdot \rangle$ : the canonic duality pairing,  $\Delta$  is the duality mapping.
- $\Gamma_0(\mathcal{X})$ : the set of all proper lower semicontinuous convex functions defined on  $\mathcal{X}$ ,  $f^*$  is the Fenchel conjugate of  $f$ .

# From Hilbert spaces to Banach spaces

- Notation:
  - $\mathcal{X}, \mathcal{Y}$ : reflexive real Banach spaces.
  - $\langle \cdot, \cdot \rangle$ : the canonic duality pairing,  $\Delta$  is the duality mapping.
  - $\Gamma_0(\mathcal{X})$ : the set of all proper lower semicontinuous convex functions defined on  $\mathcal{X}$ ,  $f^*$  is the Fenchel conjugate of  $f$ .
- It is too difficult to manipulate the terms of the form  $\|x - x_n\|^2$  in Banach spaces.

# From Hilbert spaces to Banach spaces

- Notation:
  - $\mathcal{X}, \mathcal{Y}$ : reflexive real Banach spaces.
  - $\langle \cdot, \cdot \rangle$ : the canonic duality pairing,  $\Delta$  is the duality mapping.
  - $\Gamma_0(\mathcal{X})$ : the set of all proper lower semicontinuous convex functions defined on  $\mathcal{X}$ ,  $f^*$  is the Fenchel conjugate of  $f$ .
- It is too difficult to manipulate the terms of the form  $\|x - x_n\|^2$  in Banach spaces.
- The operator  $P_C$  and its generalizations (proximal operators) becomes more complicated to manipulate, for example

$$p = P_C x \Leftrightarrow 0 \in N_C p + \Delta(p - x)$$

because the the gradient of the squared norm (the duality map) is not additive.

# Bregman monotonicity

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ .  
The Bregman distance (L. M. Bregman 1967) associated with  $f$  is

$$D_f: \mathcal{X} \times \mathcal{X} \rightarrow ]-\infty, +\infty]:$$

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

# Bregman monotonicity

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ .  
The Bregman distance (L. M. Bregman 1967) associated with  $f$  is

$$D_f: \mathcal{X} \times \mathcal{X} \rightarrow ]-\infty, +\infty]:$$

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

- $d_C^f(x) = \inf \{ D_f(y, x) \mid y \in C \}$ .



# Bregman monotonicity

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ .  
The Bregman distance (L. M. Bregman 1967) associated with  $f$  is

$$D_f: \mathcal{X} \times \mathcal{X} \rightarrow ]-\infty, +\infty]:$$

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

- $d_C^f(x) = \inf \{D_f(y, x) \mid y \in C\}$ .
- Bregman monotonicity (H. H. Bauschke, J. M. Borwein, and P. L. Combettes 2003):
  - $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{int dom } f$ .
  - $(\forall x \in C \cap \text{int dom } f)(\forall n \in \mathbb{N}) \quad D_f(x, x_{n+1}) \leq D_f(x, x_n)$ .

# Bregman Monotonicity: convergence

## ■ Conditions:

1  $D_f(x, \cdot)$  is coercive for  $x \in \text{int dom } f$ .

2  $\begin{cases} x' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}}, x'' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}} \\ (x_n)_{n \in \mathbb{N}} \text{ is Bregman monotone w.r.t } C \end{cases} \Rightarrow x' = x''$ .

3 For all bounded sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}}$  in  $\text{int dom } f$ ,

$$D_f(x_n, y_n) \rightarrow 0 \Rightarrow x_n - y_n \rightarrow 0.$$

# Bregman Monotonicity: convergence

## ■ Conditions:

1  $D_f(x, \cdot)$  is coercive for  $x \in \text{int dom } f$ .

2  $\begin{cases} x' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}}, x'' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}} \\ (x_n)_{n \in \mathbb{N}} \text{ is Bregman monotone w.r.t } C \end{cases} \Rightarrow x' = x''$ .

3 For all bounded sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}}$  in  $\text{int dom } f$ ,

$$D_f(x_n, y_n) \rightarrow 0 \Rightarrow x_n - y_n \rightarrow 0.$$

## ■ Properties:

1  $(d_C^f(x_n))_{n \in \mathbb{N}}$  converges.

2 Suppose that Conditions (1) and (2) are satisfied. Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C \cap \text{dom } \bar{f}$  if and only if  $\mathfrak{W}(x_n)_{n \in \mathbb{N}} \subset C$ .

3 Suppose that  $x_n \rightarrow \bar{x} \in C \cap \text{int dom } f$  and Condition (3) is satisfied. Then  $x_n \rightarrow \bar{x}$  if and only if  $\mathfrak{G}(x_n)_{n \in \mathbb{N}} \neq \emptyset$ .

# Quasi Bregman monotonicity: Notation and objectives

- Notation: Given  $\alpha \in ]0, +\infty[$ , set

$$\mathcal{F}_\alpha(f) = \{g \in \Gamma_0(\mathcal{X}) \mid D_g \geq \alpha D_f\}.$$

Note that  $g \in \mathcal{F}_\alpha(f)$  if  $g - \alpha f$  is convex, i.e.,  $g$  is more convex than  $\alpha f$  in the terminology of J. J. MOREAU.

# Quasi Bregman monotonicity: Notation and objectives

- Notation: Given  $\alpha \in ]0, +\infty[$ , set

$$\mathcal{F}_\alpha(f) = \{g \in \Gamma_0(\mathcal{X}) \mid D_g \geq \alpha D_f\}.$$

Note that  $g \in \mathcal{F}_\alpha(f)$  if  $g - \alpha f$  is convex, i.e.,  $g$  is more convex than  $\alpha f$  in the terminology of J. J. MOREAU.

- Objectives:
  - Give an unified framework for variable metric quasi Fejér monotonicity and Bregman monotonicity.
  - Apply to propose a forward-backward splitting in reflexive Banach spaces.

# Quasi Bregman monotonicity: Definitions

- Quasi Bregman monotonicity :
  - $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{int dom } f$ .
  - $(\exists (\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})) (\forall x \in C \cap \text{int dom } f) (\exists (\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})) (\forall n \in \mathbb{N}) D_{f_{n+1}}(x, x_{n+1}) \leq (1 + \eta_n) D_{f_n}(x, x_n) + \varepsilon_n$ .

# Quasi Bregman monotonicity: Definitions

## ■ Quasi Bregman monotonicity :

- $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{int dom } f$ .
- $(\exists(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N}))(\forall x \in C \cap \text{int dom } f)(\exists(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N}))(\forall n \in \mathbb{N}) D_{f_{n+1}}(x, x_{n+1}) \leq (1 + \eta_n)D_{f_n}(x, x_n) + \varepsilon_n$ .

## ■ Stationary quasi Bregman monotonicity :

- $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{int dom } f$ .
- $(\exists(\varepsilon_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N}))(\exists(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N}))(\forall x \in C \cap \text{int dom } f)(\forall n \in \mathbb{N}) D_{f_{n+1}}(x, x_{n+1}) \leq (1 + \eta_n)D_{f_n}(x, x_n) + \varepsilon_n$ .

# Quasi Bregman monotonicity: weak convergence

- Assumptions:  $(f_n)_{n \in \mathbb{N}}$  is a sequence of functions in  $\mathcal{F}_\alpha(f)$ , and  $(x_n)_{n \in \mathbb{N}}$  is quasi Bregman monotone with respect to  $C$  relative to  $(f_n)_{n \in \mathbb{N}}$ .



# Quasi Bregman monotonicity: weak convergence

- Assumptions:  $(f_n)_{n \in \mathbb{N}}$  is a sequence of functions in  $\mathcal{F}_\alpha(f)$ , and  $(x_n)_{n \in \mathbb{N}}$  is quasi Bregman monotone with respect to  $C$  relative to  $(f_n)_{n \in \mathbb{N}}$ .
- $x_n \rightharpoonup \bar{x}$  if
  - 1  $D_f(x, \cdot)$  is coercive for every  $x \in \text{int dom } f$ .
  - 2  $(\exists g \in \Gamma_0(\mathcal{X}))(\exists \gamma \in ]0, +\infty[) \gamma g \in \bigcap_{n \in \mathbb{N}} \mathcal{F}_1(f_n)$ .
  - 3  $\left\{ \begin{array}{l} x' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}}, \quad x'' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}} \\ (\langle x' - x'', \nabla f_n(x_n) \rangle)_{n \in \mathbb{N}} \text{ converges} \end{array} \right. \Rightarrow x' = x''$ .
  - 4  $\mathfrak{W}(x_n)_{n \in \mathbb{N}} \subset C \cap \text{int dom } f$ .

# Quasi Bregman monotonicity: strong convergence

- Assumptions:  $(f_n)_{n \in \mathbb{N}}$  is a sequence of functions in  $\mathcal{F}_\alpha(f)$ , and  $(x_n)_{n \in \mathbb{N}}$  is stationarily quasi Bregman monotone with respect to  $C$  relative to  $(f_n)_{n \in \mathbb{N}}$ .

# Quasi Bregman monotonicity: strong convergence

- Assumptions:  $(f_n)_{n \in \mathbb{N}}$  is a sequence of functions in  $\mathcal{F}_\alpha(f)$ , and  $(x_n)_{n \in \mathbb{N}}$  is stationarily quasi Bregman monotone with respect to  $C$  relative to  $(f_n)_{n \in \mathbb{N}}$ .
- $x_n \rightarrow \bar{x}$  if
  - 1  $C$  is a closed convex subset of  $\text{int dom } f$ .
  - 2  $f$  is uniformly convex on bounded subsets.
  - 3  $(\exists \beta \in ]0, +\infty[) \beta f \in \bigcap_{n \in \mathbb{N}} \mathcal{F}_1(f_n)$ .
  - 4  $\underline{\lim} d_C^f(x_n) = 0$ .

# Variable Bregman proximal point algorithm: Tools

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ ,  
 $\varphi \in \Gamma_0(\mathcal{X})$ , and  $\gamma \in ]0, +\infty[$ .

$$\text{prox}_{\gamma\varphi}^f : \mathcal{X} \rightarrow 2^{\mathcal{X}} : x \mapsto \underset{y \in \mathcal{X}}{\text{Argmin}} \varphi(y) + \frac{1}{\gamma} D_f(y, x).$$

# Variable Bregman proximal point algorithm: Tools

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ ,  $\varphi \in \Gamma_0(\mathcal{X})$ , and  $\gamma \in ]0, +\infty[$ .

$$\text{prox}_{\gamma\varphi}^f : \mathcal{X} \rightarrow 2^{\mathcal{X}} : x \mapsto \underset{y \in \mathcal{X}}{\text{Argmin}} \varphi(y) + \frac{1}{\gamma} D_f(y, x).$$

- Suppose that  $0 \in \text{sri}(\text{dom } f - \text{dom } \varphi)$  and  $\text{dom } \partial f \cap \text{dom } \partial \varphi \subset \text{int dom } f$ . Then
  - 1  $\text{dom prox}_{\gamma\varphi}^f \subset \text{int dom } f$ ;  $\text{ran prox}_{\gamma\varphi}^f \subset \text{int dom } f$ .
  - 2  $\text{prox}_{\gamma\varphi}^f = (\nabla f + \gamma \partial \varphi)^{-1} \circ \nabla f$ .
  - 3  $\text{Fix prox}_{\gamma\varphi}^f = \text{int dom } f \cap \text{Argmin } \varphi$ .
  - 4 Let  $x \in \text{Fix prox}_{\gamma\varphi}^f$  and  $(y, v) \in \text{gra prox}_{\gamma\varphi}^f$ . Then  $D_f(x, v) \leq D_f(x, y) - D_f(v, y)$ .
  - 5  $\text{prox}_{\gamma\varphi}^f$  is single-valued on its domain if  $f|_{\text{int dom } f}$  is strictly convex.

# Variable Bregman proximal point algorithm: Tools

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ ,  
 $\varphi \in \Gamma_0(\mathcal{X})$ , and  $\gamma \in ]0, +\infty[$ .

$$\text{Prox}_{\gamma\varphi}^f : \mathcal{X}^* \rightarrow 2^{\mathcal{X}} : x^* \mapsto \underset{x \in \mathcal{X}}{\text{Argmin}} \left( \varphi(x) + \frac{1}{\gamma} (f(x) - \langle x, x^* \rangle) \right).$$

# Variable Bregman proximal point algorithm: Tools

- $f \in \Gamma_0(\mathcal{X})$  Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ ,  $\varphi \in \Gamma_0(\mathcal{X})$ , and  $\gamma \in ]0, +\infty[$ .

$$\text{Prox}_{\gamma\varphi}^f : \mathcal{X}^* \rightarrow 2^{\mathcal{X}} : x^* \mapsto \underset{x \in \mathcal{X}}{\text{Argmin}} \varphi(x) + \frac{1}{\gamma} (f(x) - \langle x, x^* \rangle).$$

- Properties: suppose that  $0 \in \text{sri}(\text{dom } f - \text{dom } \varphi)$ . Then
  - 1  $\text{int}(\text{dom } f^* + \gamma \text{dom } \varphi^*) \subset \text{dom } \text{Prox}_{\gamma\varphi}^f$ .
  - 2  $\text{Prox}_{\gamma\varphi}^f$  is single-valued on its domain if  $f|_{\text{int dom } f}$  is strictly convex.
  - 3  $\text{ran } \text{Prox}_{\gamma\varphi}^f \subset \text{int dom } f$  and  $\text{Prox}_{\gamma\varphi}^f = (\nabla f + \gamma \partial \varphi)^{-1}$  if  $\text{dom } \partial f \cap \text{dom } \partial \varphi \subset \text{int dom } f$ .

# Variable Bregman proximal point algorithm: Algorithm

- $\varphi \in \Gamma_0(\mathcal{X})$  with  $\text{Argmin}\varphi \neq \emptyset$ .
- $f \in \Gamma_0(\mathcal{X})$  strictly convex and Gâteaux differentiable on  $\text{int dom } f \supset \text{dom } \partial\varphi$ .
- $\alpha \in ]0, +\infty[$ ,  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ ,  $(f_n)_{n \in \mathbb{N}}$  a sequence of functions in  $\mathcal{F}_\alpha(f)$  such that  $(\forall n \in \mathbb{N}) (1 + \eta_n)f_n \in \mathcal{F}_1(f_{n+1})$ .
- $(\gamma_n)_{n \in \mathbb{N}}$  a sequence in  $\mathbb{R}$  such that  $0 < \inf_{n \in \mathbb{N}} \gamma_n \leq \sup_{n \in \mathbb{N}} \gamma_n < +\infty$ ,
- Iteration:

$$\begin{aligned}
 (\forall n \in \mathbb{N}) \quad x_{n+1} &= \text{prox}_{\gamma_n \varphi}^{f_n} x_n \\
 &= \text{argmin}_{x \in \mathcal{X}} \varphi(x) + \frac{1}{\gamma_n} D_{f_n}(x, x_n).
 \end{aligned}$$



# Variable Bregman proximal point algorithm: analysis

The following hold for some  $\bar{x} \in \text{Argmin}\varphi$ .

- $(x_n)_{n \in \mathbb{N}}$  is stationarily quasi Bregman monotone w.r.t  $\text{Argmin}\varphi$  relative to  $(f_n)_{n \in \mathbb{N}}$  and is a minimizing sequence of  $\varphi$ .

# Variable Bregman proximal point algorithm: analysis

The following hold for some  $\bar{x} \in \text{Argmin}\varphi$ .

- $(x_n)_{n \in \mathbb{N}}$  is stationarily quasi Bregman monotone w.r.t  $\text{Argmin}\varphi$  relative to  $(f_n)_{n \in \mathbb{N}}$  and is a minimizing sequence of  $\varphi$ .
- $x_n \rightarrow \bar{x}$  if:
  - 1  $D_f(x, \cdot)$  is coercive for every  $x \in \text{int dom } f$ .
  - 2  $(\exists g \in \Gamma_0(\mathcal{X}))(\exists \beta \in ]0, +\infty[) \beta g \in \bigcap_{n \in \mathbb{N}} \mathcal{F}_1(f_n)$ .
  - 3  $\begin{cases} x' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}}, x'' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}} \\ (\langle x' - x'', \nabla f_n(x_n) \rangle)_{n \in \mathbb{N}} \text{ converges} \end{cases} \Rightarrow x' = x''$ .

# Variable Bregman proximal point algorithm: analysis

The following hold for some  $\bar{x} \in \text{Argmin}\varphi$ .

- $(x_n)_{n \in \mathbb{N}}$  is stationarily quasi Bregman monotone w.r.t  $\text{Argmin}\varphi$  relative to  $(f_n)_{n \in \mathbb{N}}$  and is a minimizing sequence of  $\varphi$ .
- $x_n \rightarrow \bar{x}$  if:
  - 1  $D_f(x, \cdot)$  is coercive for every  $x \in \text{int dom } f$ .
  - 2  $(\exists g \in \Gamma_0(\mathcal{X}))(\exists \beta \in ]0, +\infty[) \beta g \in \bigcap_{n \in \mathbb{N}} \mathcal{F}_1(f_n)$ .
  - 3  $\begin{cases} x' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}}, x'' \in \mathfrak{W}(x_n)_{n \in \mathbb{N}} \\ (\langle x' - x'', \nabla f_n(x_n) \rangle)_{n \in \mathbb{N}} \text{ converges} \end{cases} \Rightarrow x' = x''$ .
- $x_n \rightarrow \bar{x}$  if:
  - 1  $f$  is uniformly convex on bounded subsets.
  - 2  $(\exists \beta \in ]0, +\infty[) \beta f \in \bigcap_{n \in \mathbb{N}} \mathcal{F}_1(f_n)$ .
  - 3  $\varliminf_{\text{Argmin}\varphi} d_{\text{Argmin}\varphi}^f(x_n) = 0$ .

# Application: forward-backward splitting

- Problem:  $\varphi \in \Gamma_0(\mathcal{X}), \psi \in \Gamma_0(\mathcal{Y})$  is differentiable,  $L \in \mathcal{B}(\mathcal{X}, \mathcal{Y})$ . The problem is to

$$\underset{x \in \mathcal{X}}{\text{minimize}} \quad \varphi(x) + \psi(Lx).$$

# Application: forward-backward splitting

- Problem:  $\varphi \in \Gamma_0(\mathcal{X}), \psi \in \Gamma_0(\mathcal{Y})$  is differentiable,  $L \in \mathcal{B}(\mathcal{X}, \mathcal{Y})$ . The problem is to

$$\underset{x \in \mathcal{X}}{\text{minimize}} \quad \varphi(x) + \psi(Lx).$$

- Forward-Backward splitting in Hilbert spaces:
  - Algorithm :

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad x_{n+1} &= \text{prox}_{\gamma_n \varphi}(x_n - \gamma_n L^* \nabla \psi(Lx_n)) \\ &= \underset{x \in \mathcal{H}}{\text{argmin}} \quad \varphi(x) + \langle x - x_n \mid L^* \nabla \psi(Lx_n) \rangle + \psi(Lx_n) \\ &\quad + \frac{1}{\gamma_n} \|x - x_n\|^2 \\ &= \text{prox}_{\gamma_n \Phi_{x_n}} x_n, \end{aligned}$$

where  $\Phi_Y(x) = \langle x - y \mid L^* \nabla \psi(Ly) \rangle + \psi(Ly)$ .

- Algorithm requires that  $\psi$  is differentiable on the whole space with a Lipschitz gradient and the closed form of the proximal operators.

# Forward-backward splitting in Banach spaces

## ■ Assumptions:

- 1  $f \in \Gamma_0(\mathcal{X})$  is strictly convex and Gâteaux differentiable.
- 2  $L(\text{int dom } f) \subset \text{int dom } \psi$  and  $\text{int dom } f \supset \text{dom } (\varphi + \psi \circ L)$ .
- 3 Either  $f$  is cofinite or  $-L^*(\text{ran } \nabla \psi) \subset \text{dom } \varphi^*$ .
- 4  $(\exists \beta \in ]0, +\infty[) \beta f \in \mathcal{F}_1(\psi \circ L)$ .

# Forward-backward splitting in Banach spaces

## ■ Assumptions:

- 1  $f \in \Gamma_0(\mathcal{X})$  is strictly convex and Gâteaux differentiable.
- 2  $L(\text{int dom } f) \subset \text{int dom } \psi$  and  $\text{int dom } f \supset \text{dom } (\varphi + \psi \circ L)$ .
- 3 Either  $f$  is cofinite or  $-L^*(\text{ran } \nabla \psi) \subset \text{dom } \varphi^*$ .
- 4  $(\exists \beta \in ]0, +\infty[) \beta f \in \mathcal{F}_1(\psi \circ L)$ .

## ■ Algorithm:

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = \text{Prox}_{\gamma_n \varphi}^f (\nabla f(x_n) - \gamma_n L^* \nabla \psi(Lx_n)).$$

where,  $\varepsilon \in ]0, \beta/(\beta + 1)[$ ,  $(\gamma_n)_{n \in \mathbb{N}} \in \mathbb{R}$ ,  $(\eta_n)_{n \in \mathbb{N}} \in \ell_+^1(\mathbb{N})$ ,

$$(\forall n \in \mathbb{N}) \quad \varepsilon \leq \gamma_n \leq \beta(1 - \varepsilon) \text{ and } (1 + \eta_n)\gamma_n - \gamma_{n+1} \leq \beta\eta_n.$$

## Forward-backward splitting in Banach spaces

$$\begin{aligned}
x_{n+1} &= \operatorname{argmin}_{x \in \mathcal{X}} \varphi(x) + \langle x - x_n, L^* \nabla \psi(Lx_n) \rangle + \psi(Lx_n) + \frac{1}{\gamma_n} D_f(x, x_n) \\
&= \operatorname{argmin}_{x \in \mathcal{X}} (\varphi(x) + \psi(Lx)) - (\psi(Lx) - \langle x - x_n, L^* \nabla \psi(Lx_n) \rangle) \\
&\quad - \psi(Lx_n) + \frac{1}{\gamma_n} D_f(x, x_n) \\
&= \operatorname{argmin}_{x \in \mathcal{X}} (\varphi(x) + \psi(Lx)) + \frac{1}{\gamma_n} (D_f(x, x_n) - \gamma_n D_\psi(Lx, Lx_n)) \\
&= \operatorname{argmin}_{x \in \mathcal{X}} (\varphi(x) + \psi(Lx)) + \frac{1}{\gamma_n} D_{f - \gamma_n \psi \circ L}(x, x_n) \\
&= \operatorname{prox}_{\gamma_n(\varphi + \psi \circ L)}^{f_n} x_n,
\end{aligned}$$

where  $(\forall n \in \mathbb{N}) f_n = f - \gamma_n \psi \circ L$ .



# Forward-backward: Convergence

The following hold for some  $\bar{x} \in \text{Argmin}(\varphi + \psi \circ L)$ :

- Weak convergence: suppose that
  - 1  $D_f(x, \cdot)$  is coercive for every  $x \in \text{int dom } f$ .
  - 2  $\nabla f$  and  $\nabla \psi$  are weakly sequentially continuous.

Then  $x_n \rightharpoonup \bar{x}$ .

# Forward-backward: Convergence

The following hold for some  $\bar{x} \in \text{Argmin}(\varphi + \psi \circ L)$ :

- Weak convergence: suppose that

- 1  $D_f(x, \cdot)$  is coercive for every  $x \in \text{int dom } f$ .
- 2  $\nabla f$  and  $\nabla \psi$  are weakly sequentially continuous.

Then  $x_n \rightharpoonup \bar{x}$ .

- Strong convergence: suppose that

- 1  $f$  is uniformly convex on bounded subsets.
- 2 One of the following conditions holds:
  - 1  $\varphi$  is uniformly convex at  $\bar{x}$ .
  - 2  $\psi$  is uniformly convex at  $L\bar{x}$  and  $L$  is bounded below.
  - 3  $\liminf_{x_n \in \text{Argmin}(\varphi + \psi \circ L)} d_f^x(x_n) = 0$ .

Then  $x_n \rightarrow \bar{x}$ .

# Examples: notation

- Notation:  $m$  and  $K$  are strictly positive integers,  $\mathbb{R}^m$  and  $\mathbb{R}^K$  are usual Euclidean spaces,  $(a_{ik})_{1 \leq i \leq m, 1 \leq k \leq K} \in \mathbb{R}_+^{m \times K} \setminus \{0\}$ ,  
 $\alpha = \min \{a_{ik} \neq 0 \mid 1 \leq i \leq m, 1 \leq j \leq K\}$ ,  
 $\beta = \max \{a_{ik} \mid 1 \leq i \leq m, 1 \leq j \leq K\}$ .
- Some functions of interest:

- 1 Burg entropy:  $b: \xi \mapsto \begin{cases} -\ln \xi, & \text{if } \xi > 0; \\ +\infty, & \text{otherwise.} \end{cases}$

- 2 Boltzmann-Shannon entropy:

$$s: \xi \mapsto \begin{cases} \xi \ln \xi - \xi, & \text{if } \xi \geq 0; \\ +\infty, & \text{otherwise.} \end{cases}$$

- The Bregman distance associated with Burg entropy is the Itakura-Saito divergence and the Bregman distance associated with Boltzmann-Shannon entropy is the Kullback-Leibler divergence.

# Example 1

## ■ Problem:

$$\begin{aligned} \text{minimize}_{(\xi_1, \dots, \xi_m) \in ]0, +\infty[^m} & \sum_{i=1}^m \varphi_i(\xi_i) + \sum_{k=1}^K \left( -\ln \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} \right. \\ & \left. + \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} - 1 \right), \end{aligned}$$

which becomes

$$\text{minimize}_{x \in \mathbb{R}^m} \varphi(x) + D_\psi(Lx, r),$$

where  $\varphi: (\xi_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m \varphi_i(\xi_i)$ ,  $L: (\xi_i)_{1 \leq i \leq m} \mapsto (\sum_{i=1}^m a_{ik} \xi_i)_{1 \leq k \leq K}$ ,  
 $(\varrho_k)_{1 \leq k \leq K}$ ,  $\psi: (\rho_k)_{1 \leq k \leq K} \mapsto \begin{cases} -\sum_{k=1}^K \ln \rho_k & \text{if } \min_{1 \leq k \leq K} \rho_k > 0, \\ +\infty & \text{otherwise.} \end{cases}$

# Example 1

## ■ Problem:

$$\begin{aligned} \text{minimize}_{(\xi_1, \dots, \xi_m) \in ]0, +\infty[^m} & \sum_{i=1}^m \varphi_i(\xi_i) + \sum_{k=1}^K \left( -\ln \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} \right. \\ & \left. + \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} - 1 \right), \end{aligned}$$

which becomes

$$\text{minimize}_{x \in \mathbb{R}^m} \varphi(x) + D_\psi(Lx, r),$$

where  $\varphi: (\xi_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m \varphi_i(\xi_i)$ ,  $L: (\xi_i)_{1 \leq i \leq m} \mapsto (\sum_{i=1}^m a_{ik} \xi_i)_{1 \leq k \leq K}$ ,  
 $(\varrho_k)_{1 \leq k \leq K}$ ,  $\psi: (\rho_k)_{1 \leq k \leq K} \mapsto \begin{cases} -\sum_{k=1}^K \ln \rho_k & \text{if } \min_{1 \leq k \leq K} \rho_k > 0, \\ +\infty & \text{otherwise.} \end{cases}$

- Because  $D_\psi(L \cdot, r)$  is not differentiable on  $\mathbb{R}^m$ , the classical forward-backward can not be applied.

# Example 1

## ■ Algorithm:

For  $n = 0, 1, \dots$

    For  $i = 1 \dots, m$

$$\xi_{i,n+1} = \text{Prox}_{\gamma_n \varphi_i}^b \left( \frac{-1}{\xi_{i,n}} - \gamma_n \sum_{k=1}^K a_{ik} \left( \frac{-1}{\sum_{j=1}^m a_{jk} \xi_{j,n}} + \frac{1}{\rho_k} \right) \right).$$

Then  $(\xi_{i,n})_{1 \leq i \leq m}$  converges to a solution.

# Example 1

## ■ Algorithm:

For  $n = 0, 1, \dots$

For  $i = 1 \dots, m$

$$\xi_{i,n+1} = \text{Prox}_{\gamma_n \varphi_i}^b \left( \frac{-1}{\xi_{i,n}} - \gamma_n \sum_{k=1}^K a_{ik} \left( \frac{-1}{\sum_{j=1}^m a_{jk} \xi_{j,n}} + \frac{1}{\rho_k} \right) \right).$$

Then  $(\xi_{i,n})_{1 \leq i \leq m}$  converges to a solution.

- In some cases,  $\text{Prox}_{\gamma_n \varphi_i}^b$  is easier to calculate than the classic proximal operator.

## Example 2

### ■ Problem:

$$\begin{aligned} \text{minimize}_{(\xi_1, \dots, \xi_m) \in [0, +\infty]^m} & \sum_{i=1}^m \varphi_i(\xi_i) + \sum_{k=1}^K \left( \sum_{i=1}^m a_{ik} \xi_i \right) \ln \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} \\ & - \sum_{k=1}^K \sum_{i=1}^m a_{ik} \xi_i + \sum_{k=1}^K \varrho_k, \end{aligned}$$

which becomes

$$\text{minimize}_{x \in \mathbb{R}^m} \varphi(x) + D_\psi(Lx, r),$$

where  $\varphi: (\xi_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m \varphi_i(\xi_i)$ ,  $L: (\xi_i)_{1 \leq i \leq m} \mapsto (\sum_{i=1}^m a_{ik} \xi_i)_{1 \leq k \leq K}$   
 $(\varrho_k)_{1 \leq k \leq K}, \psi: (\rho_k)_{1 \leq k \leq K} \mapsto \begin{cases} \sum_{k=1}^K \rho_k \ln \rho_k - \rho_k & \text{if } \min_{1 \leq k \leq K} \rho_k \\ +\infty & \text{otherwise.} \end{cases}$



## Example 2


■ Problem:

$$\begin{aligned} \text{minimize}_{(\xi_1, \dots, \xi_m) \in [0, +\infty]^m} & \sum_{i=1}^m \varphi_i(\xi_i) + \sum_{k=1}^K \left( \sum_{i=1}^m a_{ik} \xi_i \right) \ln \frac{\sum_{i=1}^m a_{ik} \xi_i}{\varrho_k} \\ & - \sum_{k=1}^K \sum_{i=1}^m a_{ik} \xi_i + \sum_{k=1}^K \varrho_k, \end{aligned}$$

which becomes

$$\text{minimize}_{x \in \mathbb{R}^m} \varphi(x) + D_\psi(Lx, r),$$

where  $\varphi: (\xi_i)_{1 \leq i \leq m} \mapsto \sum_{i=1}^m \varphi_i(\xi_i)$ ,  $L: (\xi_i)_{1 \leq i \leq m} \mapsto (\sum_{i=1}^m a_{ik} \xi_i)_{1 \leq k \leq K}$   
 $(\varrho_k)_{1 \leq k \leq K}, \psi: (\rho_k)_{1 \leq k \leq K} \mapsto \begin{cases} \sum_{k=1}^K \rho_k \ln \rho_k - \rho_k & \text{if } \min_{1 \leq k \leq K} \rho_k \\ +\infty & \text{otherwise.} \end{cases}$

- Because  $D_\psi(L \cdot, r)$  is not differentiable on  $\mathbb{R}^m$ , the classical forward-backward can not be applied. 

## Example 2

■ Algorithm:

For  $n = 0, 1, \dots$

    For  $i = 1, \dots, m$

$$\left[ \left[ \xi_{i,n+1} = \text{prox}_{\gamma_n \varphi_i}^s \left( \xi_{i,n} \prod_{k=1}^K \left( \frac{a_k}{\sum_{j=1}^m a_{jk} \xi_{j,n}} \right)^{\gamma_n a_{ik}} \right) \right. \right.$$

Then  $(\xi_{i,n})_{1 \leq i \leq m}$  converges to a solution.

## Example 2

- Algorithm:

For  $n = 0, 1, \dots$

    For  $i = 1 \dots, m$

$$\left[ \left[ \xi_{i,n+1} = \text{prox}_{\gamma_n \varphi_i}^s \left( \xi_{i,n} \prod_{k=1}^K \left( \frac{a_k}{\sum_{j=1}^m a_{jk} \xi_{j,n}} \right)^{\gamma_n a_{ik}} \right) \right. \right.$$

Then  $(\xi_{i,n})_{1 \leq i \leq m}$  converges to a solution.

- In some cases,  $\text{Prox}_{\gamma_n \varphi_i}^s$  is easier to calculate than the classic proximal operator.

## Some remarks

- Our algorithm can be applied for some problems set in non-hilbertian reflexive real Banach spaces, such as  $\ell^p(\mathbb{N})$  for some  $p \in ]1, +\infty[$ .




## Some remarks

- Our algorithm can be applied for some problems set in non-hilbertian reflexive real Banach spaces, such as  $\ell^p(\mathbb{N})$  for some  $p \in ]1, +\infty[$ .
- Stochastic version: Given a probability space  $(\Omega, \mathcal{F}, \mu)$ . Suppose that  $D_f(x, \cdot)$  is  $\mathcal{F}$ -measurable. We can generalize the above Bregman monotonicity to a stochastic process.



## Some remarks

- Our algorithm can be applied for some problems set in non-hilbertian reflexive real Banach spaces, such as  $\ell^p(\mathbb{N})$  for some  $p \in ]1, +\infty[$ .
- Stochastic version: Given a probability space  $(\Omega, \mathcal{F}, \mu)$ . Suppose that  $D_f(x, \cdot)$  is  $\mathcal{F}$ -measurable. We can generalize the above Bregman monotonicity to a stochastic process.
- Open problems:
  - Forward-Backward splitting for sum of two monotone operators.
  - Recently, M. A. Alghamdi, A. Alotaibi, P. L. Combettes, and N. Shahzad 2014 proposed a primal-dual method of partial inverses for composite inclusions, which is, in fact, can be rewritten as a proximal algorithm. We don't know if there exists a suitable generalization of partial inverse, initiated by Spingarn, in Banach spaces.

# References

-  H. H. Bauschke, J. M. Borwein, and P. L. Combettes, Bregman monotone optimization algorithms, *SIAM J. Control Optim.*, vol. 42, pp. 596–636, 2003.
-  M. A. Alghamdi, A. Alotaibi, P. L. Combettes, and N. Shahzad, A primal-dual method of partial inverses for composite inclusions, *Optimization Letters*, 2014.
-  L. M. Bregman, The relaxation method of finding the common point of convex sets and its application to the solution of problems in convex programmings, *USSR Comp. Math. Math. Phys.*, vol 7, pp. 200–217, 1967.

# References

-  P. L. Combettes and J. C. Pesquet, Stochastic quasi-Fejér block-coordinate fixed point iterations with random sweeping, *arXiv:1404.7536*, 2014.
-  P. L. Combettes and B. C. Vũ, Variable metric quasi-Fejér monotonicity, *Nonlinear Anal.*, vol. 78, pp. 17–31, 2013.