

Inexact Version of the Entropic Proximal Point Algorithm

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Abstract: In this paper, we study an inexact version of the entropic proximal point algorithm defined by

$$x^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\}.$$

This study recovers the most of the algorithms of the proximal point.

Keywords: Convex optimization, Bregman's function, Entropic approximation.

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1. Introduction

Let's consider the problem of a convex optimization

$$(P) : \min \{f(x), x \in R^n\},$$

where $f \in \Gamma_0(R^n)$ set of lower semicontinuous, proper and convex functions on R^n . By using the approximation of Moreau-Yosida, Auslender [1], Martine [8], Rockafellar [9], Behraoui [2], and Lemaire [6, 7] have given the methods of resolution of (P) named proximal methods defined by:

$$x^k \in \varepsilon_k - \text{Argmin} \left\{ f(\cdot) + \frac{1}{2\lambda_k} \|\cdot - x^{k-1}\|^2 \right\}.$$

By using the entropic approximation [5], Eckstein [3], thus Chen and Teboulle [10] have given a classe of methods named entropics proximals defined by:

$$x^k := \text{argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\}$$

where $D_h(\cdot, \cdot)$ is the entropic distance defined by:

$$D_h(x, y) := h(x) - h(y) - \langle x - y, \nabla h(y) \rangle.$$

In this labor, we propose an inexact version defined by:

$$x^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\},$$

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where $h \in C(S)$ [5], we show that x^k converges to x^* solution of the problem (P). Our result of convergence obtained in section 4, recovers what has been given by Auslender (it's enough to take $h(\cdot) = \frac{1}{2}\|\cdot\|^2$), thus the result of convergence obtained by Eckstein (it's enough to take $\varepsilon_k = 0$).

Our notation is fairly standard; $\langle \cdot, \cdot \rangle$ is the scalar product on R^n ; and the associated norm $\|\cdot\|$. The closure of the set C (relative interior) is denoted by \overline{C} (riC, respectively), $\text{Adh } \{x^k\}$ is the set of adherence values of a sequence $\{x^k\}$. For any convex function f , we denote by:

- (1). $\text{dom} f = \{x \in R^n; f(x) < +\infty\}$ its effective domain,
- (2). $\partial_\varepsilon f(\cdot) = \{v, f(y) \geq f(\cdot) + \langle v, y - \cdot \rangle - \varepsilon, \forall y\}$ its ε -sub differential,
- (3). $\text{Arg min } f = \{x \in R^n; f(x) = \inf f\}$ its Argmin f ,
- (4). $\varepsilon - \text{Arg min } f = \{x \in R^n; f(x) \leq \inf f + \varepsilon\}$ its $\varepsilon - \text{Argmin} f$.

2. Preliminaries

Let S be an convex open subset of R^n and $h : \overline{S} \rightarrow R$. We define $D_h(\cdot, \cdot)$ by: $\forall x \in \overline{S}, \forall y \in S$:

$$D_h(x, y) := h(x) - h(y) - \langle x - y, \nabla h(y) \rangle.$$

Let us consider the following hypotheses:

H_1 : h is continuously differentiable on S .

H_2 : h is continuous and strictly convex on \overline{S} .

H_3 : $\forall r \geq 0, \forall x \in \overline{S}, \forall y \in S$, the sets $L_1(x, r)$ and $L_2(y, r)$ are bounded where:

$$L_1(x, r) = \{y \in S / D_h(x, y) \leq r\}$$

$$L_2(y, r) = \{x \in \overline{S} / D_h(x, y) \leq r\}.$$

H_4 : If $\{y^k\}_k \subset S$ is such as $y^k \rightarrow y^* \in \overline{S}$, so $D_h(y^*, y^k) \rightarrow 0$.

H_5 : If $\{x^k\}_k \subset \overline{S}$ is such as $\{y^k\}_k \subset S$ are such as:

$$y^k \rightarrow y^* \in \overline{S}, \{x^k\}_k \text{ is bounded, and } D_h(x^k, y^k) \rightarrow 0, \text{ then } x^k \rightarrow y^*.$$

H_6 : If $\{x^k\}_k$ and $\{y^k\}_k$ are two sequences of S such as:

$$D_h(x^k, y^k) \rightarrow 0 \text{ and } x^k \rightarrow x^* \in S, \text{ then } y^k \rightarrow x^*.$$

Definition 2.1.

(1). $h : \overline{S} \rightarrow R$ is a Bregman function on S or "D-function" if h verify H_1, H_2, H_3, H_4 and H_5 .

(2). $D_h(\cdot, \cdot)$, is called entropic distance if h is a Bregman function.

We put:

$$A(S) = \{h : \bar{S} \rightarrow R \text{ verifying } H_1 \text{ and } H_2\}$$

$$B(S) = \{h : \bar{S} \rightarrow R \text{ verifying } H_1, H_2, H_3, H_4 \text{ and } H_5\}$$

$$C(S) = \{h : \bar{S} \rightarrow R \text{ verifying } H_1, H_2, H_3, H_4 \text{ and } H_6\}.$$

Lemma 2.2. $\forall h \in A(S), \forall a \in \bar{S}, \forall b, c \in S:$

$$D_h(a, b) + D_h(b, c) - D_h(a, c) = \langle a - b, \nabla h(c) - \nabla h(b) \rangle.$$

Proposition 2.3. Let $h \in A(S)$. Let $\{x^k\}$ and $\{y^k\}$ two sequences of S such that:

(1). $x^k \rightarrow x^* \in S$ and $\{y^k\}$ bounded such as $\text{Adh}\{y^k\} \subset S$

(2). $D_h(x^k, y^k) \rightarrow 0$.

Then $y^k \rightarrow x^*$.

Proof. $\{y^k\}$ is bounded, then we can extract a sub-sequence $\{y^{k_i}\}$ of $\{y^k\}$ such that $y^{k_i} \rightarrow y \in S$. We have, $D_h(x^{k_i}, y^{k_i}) := h(x^{k_i}) - h(y^{k_i}) - \langle x^{k_i} - y^{k_i}, \nabla h(y^{k_i}) \rangle$. So, when $k_i \rightarrow +\infty$, we obtain: $0 = h(x^*) - h(y) - \langle x^* - y, \nabla h(y) \rangle \Rightarrow D_h(x^*, y) = 0 \Rightarrow x^* = y \Rightarrow \text{Adh}\{y^k\} = \{x^*\} \Rightarrow y^k \rightarrow x^*$. \square

Proposition 2.4. We assume:

(1). $h \in A(S)$.

(2). $\{x^k\} \subset S$ is bounded.

(3). $\text{Adh}\{x^k\} \subset S$.

Then $\{D_h(u, x^k)\}_k$ is bounded for all $u \in \bar{S}$.

Proof. If the sequence $\{D_h(u, x^k)\}_k$ is not bounded for $u \in \bar{S}$, then exists a sub-sequence $\{D_h(u, x^{k_i})\}_{k_i}$ such that: $D_h(u, x^{k_i}) \rightarrow +\infty$. $\{x^{k_i}\}$ is bounded, then it exists a sub-sequence $\{x^{k_j}\}$ of $\{x^{k_i}\}$ such as $x^{k_j} \rightarrow x^*$. From (3), $x^* \in S$, so $\nabla h(x^{k_j}) \rightarrow \nabla h(x^*)$.

$$D_h(u, x^{k_j}) := h(u) - h(x^{k_j}) - \langle u - x^{k_j}, \nabla h(x^{k_j}) \rangle.$$

when, $k_j \rightarrow +\infty$, we obtain: $+\infty = D_h(u, x^*)$, which is contradictory with $h : \bar{S} \rightarrow R$. \square

Proposition 2.5. Let $h \in A(S)$ and $f \in \Gamma_0(R^n)$ such as :

(1). $\text{ri}(\text{dom } f) \cap S \neq \emptyset$,

(2). $\text{Im}\nabla h = R^n$.

Then $\forall \varepsilon > 0, \forall \lambda > 0, \forall x \in S, \exists y \in S : y \in \varepsilon - \text{Argmin}\{f(\cdot) + \lambda^{-1}D_h(\cdot, x)\}$.

Proof. From the Corollary 5.8. [5], $\forall \varepsilon > 0, \forall \lambda > 0, \forall x \in S, y := h_\lambda^f(x) = \text{argmin}\{f(\cdot) + \lambda^{-1}D_h(\cdot, x)\} \in S$. So, $h_\lambda^f(x) \in \varepsilon - \text{Argmin}\{f(\cdot) + \lambda^{-1}D_h(\cdot, x)\}$. \square

3. Proximal Point Algorithm

Let $f \in \Gamma_0(R^n)$, Auslender [1] considered the proximal following method:

$$P(1) : \begin{cases} x^0 \in R^n, \varepsilon_k > 0 \text{ and } \lambda_k > 0. \\ x^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \frac{1}{2\lambda_k}\|\cdot - x^{k-1}\|^2\}. \end{cases}$$

The sequence $\{x^k\}$ generated by $P(1)$ verify the following properties:

Proposition 3.1. *It exists a sequence $\{\bar{x}^k\}$ such as : $\frac{x^{k-1} - \bar{x}^k}{\lambda_k} \in \partial_{\varepsilon_k} f(x^k)$ and $\|x^k - \bar{x}^k\| \leq \sqrt{2\lambda_k \varepsilon_k}$.*

Theorem 3.2. *We assume:*

- (1). $\lambda_k \geq \underline{\lambda} > 0$,
- (2). f is inf-compact, i.e., $\forall r \in R, \{x \in R^n, f(x) \leq r\}$ is compact, if it is not empty.
- (3). $\sum \varepsilon_k < +\infty$ or ($f(0)$ is finite and $\varepsilon_k \rightarrow 0$).

Then $\{x^k\}$ defined by $P(1)$ is bounded and $f(x^k) \rightarrow \inf f$.

Teboule and chen [10] have considered the algorithm defined by:

$$P(2)(h) : \begin{cases} x^0 \in S, \lambda_k > 0. \\ x^k = \text{argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\}. \end{cases}$$

Theorem 3.3. *If $f \in \Gamma_0(R^n)$ and $h \in B(S)$ such that*

- (1). $\text{Im} \nabla h = R^n$,
- (2). $\sum \lambda_k = +\infty$,
- (3). $\text{ri}(\text{dom } f) \subset S$.

Then:

- (a). $f(x^k) \rightarrow \inf f$.
- (b). If $\text{Argmin } f \neq \emptyset$, then $x^k \rightarrow x^* \in \text{Argmin } f$.

4. Inexact Entropic Proximal Point Algorithm

In this paragraph, we assume:

- (A) : $h \in C(S) : \text{Im} \nabla h = R^n$
- (B) : $f \in \Gamma_0(R^n) : \overline{\text{dom } f} \subset S$

From the Proposition 2.5, we can thus construct the sequence $\{x^k\}$ defined by:

$$P(3)(h) : \begin{cases} x^0 \in S \\ x^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\} \\ f(x^k) \text{ is decreasing} \\ \varepsilon_k \rightarrow 0 \text{ and } \lambda_k \geq \underline{\lambda} > 0. \end{cases}$$

To choose $\{x^k\}$ such that $\{f(x^k)\}$ is decreasing is possible. Indeed:

Let $\bar{x}^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1}D_h(\cdot, x^{k-1})\}$, for all $k \geq 1$. We take

$$x^k = \begin{cases} \bar{x}^k & \text{if } f(\bar{x}^k) \leq f(x^{k-1}) \\ x^{k-1} & \text{if not} \end{cases}$$

If $f(\bar{x}^k) > f(x^{k-1})$ then $x^{k-1} \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1}D_h(\cdot, x^{k-1})\}$, because:

$$\lambda_k^{-1}D_h(x^{k-1}, x^{k-1}) + f(x^{k-1}) < f(\bar{x}^k) \leq f(\bar{x}^k) + \lambda_k^{-1}D_h(\bar{x}^k, x^{k-1}) \leq f(x) + \lambda_k^{-1}D_h(x, x^{k-1}) + \varepsilon_k.$$

Let's consider now the function $h_{u,\lambda}$ defined by: $h_{u,\lambda} : \bar{S} \rightarrow R, \forall \lambda > 0, \forall u \in S$.

$$h_{u,\lambda}(x) = \lambda^{-1}D_h(x, u), \forall x \in \bar{S}.$$

Proposition 4.1. $\forall \varepsilon > 0, \forall \lambda > 0, \forall x^* \in \bar{S}, \forall u \in S$:

$$\partial_\varepsilon h_{u,\lambda}(x^*) = \left\{ z/z = \frac{\nabla h(\bar{x}) - \nabla h(u)}{\lambda} \text{ with } \bar{x} \in S \text{ and } D_h(x^*, \bar{x}) \leq \lambda\varepsilon \right\}.$$

Proof.

$$\begin{aligned} z \in \partial_\varepsilon h_{u,\lambda}(x^*) &\Leftrightarrow \forall x \in \bar{S}, h_{u,\lambda}(x) - h_{u,\lambda}(x^*) \geq \langle z, x - x^* \rangle - \varepsilon \\ &\Leftrightarrow \forall x \in \bar{S}, \lambda^{-1}[h(x) - h(u) - \langle x - u, \nabla h(u) \rangle - h(x^*) + h(u) + \langle x^* - u, \nabla h(u) \rangle] \geq \langle z, x - x^* \rangle - \varepsilon \\ &\Leftrightarrow \forall x \in \bar{S}, h(x^*) - h(x) - \langle x^* - x, \nabla h(u) \rangle \leq \langle \lambda z, x^* - x \rangle + \lambda\varepsilon \\ &\Leftrightarrow \forall x \in \bar{S}, h(x^*) - h(x) - \langle x^* - x, \nabla h(u) + \lambda z \rangle \leq \lambda\varepsilon. \end{aligned} \tag{1}$$

According to (A), it exists $\bar{x} \in S$ such that $\nabla h(u) + \lambda z = \nabla h(\bar{x})$, which means:

$$\exists \bar{x} \in S, z = \frac{\nabla h(\bar{x}) - \nabla h(u)}{\lambda}.$$

Replacing in (1), \times by \bar{x} , we get

$$h(x^*) - h(\bar{x}) - \langle x^* - \bar{x}, \nabla h(\bar{x}) \rangle \leq \lambda\varepsilon \Leftrightarrow D_h(x^*, \bar{x}) \leq \lambda\varepsilon.$$

Finally

$$\partial_\varepsilon h_{u,\lambda}(x^*) \subset \left\{ z/z = \frac{\nabla h(\bar{x}) - \nabla h(u)}{\lambda} \text{ with } \bar{x} \in S \text{ and } D_h(x^*, \bar{x}) \leq \lambda\varepsilon \right\}.$$

Conversely, let z such that $z = \frac{\nabla h(\bar{x}) - \nabla h(u)}{\lambda}$ and $D_h(x^*, \bar{x}) \leq \lambda\varepsilon$.

$$\begin{aligned} D_h(x^*, \bar{x}) \leq \lambda\varepsilon &\Rightarrow h(x^*) - h(\bar{x}) - \langle x^* - \bar{x}, \nabla h(\bar{x}) \rangle \leq \lambda\varepsilon \leq \lambda\varepsilon + D_h(x, \bar{x}) \\ &\Rightarrow h(x^*) - h(\bar{x}) - \langle x^* - \bar{x}, \nabla h(\bar{x}) \rangle - h(x) + h(\bar{x}) + \langle x - \bar{x}, \nabla h(\bar{x}) \rangle \leq \lambda\varepsilon \\ &\Rightarrow h(x^*) - h(x) - \langle x^* - x, \nabla h(\bar{x}) \rangle \leq \lambda\varepsilon. \end{aligned}$$

Replacing $\nabla h(\bar{x})$ by $\nabla h(u) + \lambda z$, we get (1). According to what precedes (1) $\Leftrightarrow z \in \partial_\varepsilon h_{u,\lambda}(x^*)$, which establishes the desired equality. \square

Lemma 4.2 ([4]). Let f_1, f_2 two functions of $\Gamma_0(R^n)$ if it exists $\bar{x} \in \text{dom}f_1$ in which f_2 is finite and continuous, then for $\varepsilon > 0$, for all $y \in \text{dom}f_1 \cap \text{dom}f_2$,

$$\partial_\varepsilon(f_1 + f_2)(y) = \bigcup_{\varepsilon_1 + \varepsilon_2 = \varepsilon, \varepsilon_1 \geq 0, \varepsilon_2 \geq 0} \partial_{\varepsilon_1}f_1(y) + \partial_{\varepsilon_2}f_2(y).$$

Definition 4.3. The sequence $\{(\lambda_k; a_k; b_k; c_k; d_k)\}_k \in R^{+*}XS^4$ verifies the K -property if only if the following properties are verified:

$$K_1 : \exists \underline{\lambda} > 0, \forall k, \lambda_k \geq \underline{\lambda}.$$

$$K_2 : \{a_k\} \text{ is bounded and } \text{Adh}\{a_k\} \subset S.$$

$$K_3 : D_h(a_k, b_k) \rightarrow 0.$$

$$K_4 : D_h(a_k, c_k) \rightarrow 0.$$

$$K_5 : d_k = \frac{\nabla h(b_k) - \nabla h(c_k)}{\lambda_k}.$$

Lemma 4.4. If the sequence $\{(\lambda_k; a_k; b_k; c_k; d_k)\}_k$ verifies the K -property then $d_k \rightarrow 0$.

Proof. If the sequence $\{d_k\}$ does not tend to zero, then it exists $M > 0$ and the sub-sequence $\{d_{k_i}\}$ of $\{d_k\}$ such that:

$$\forall k_i, \|d_{k_i}\| > M. \quad (2)$$

The sequence $\{a_{k_i}\}$ is bounded and $\text{Adh}\{a_{k_i}\} \subset S$, it exists then the sub-sequence $\{a_{k_j}\}$ of $\{a_{k_i}\}$ and $u^* \in S$ such as: $a_{k_j} \rightarrow u^*$. $D_h(a_{k_j}, b_{k_j}) \rightarrow 0$ and $D_h(a_{k_j}, c_{k_j}) \rightarrow 0$ allow to write, from H_6 , $b_{k_j} \rightarrow u^*$ and $c_{k_j} \rightarrow u^*$. On the other hand,

$$0 \leq \|d_{k_j}\| = \left\| \frac{\nabla h(b_{k_j}) - \nabla h(c_{k_j})}{\lambda_{k_j}} \right\| \leq \underline{\lambda}^{-1} \|\nabla h(b_{k_j}) - \nabla h(c_{k_j})\|.$$

∇h is continuous on S , then $\nabla h(b_{k_j}) - \nabla h(c_{k_j}) \rightarrow 0$. It follows that $\|d_{k_j}\| \rightarrow 0$. $\{d_{k_j}\}$ being a sub-sequence of $\{d_{k_i}\}$, from (2) we have: $0 \geq M > 0$, so $d_k \rightarrow 0$. \square

Theorem 4.5. If the sequence $\{x^k\}$ generated by $P(3)(h)$ is bounded, then:

(1). $f(x^k) \rightarrow \text{inf}f$.

(2). $\text{Adh}\{x^k\} \subset \text{Argmin}f$.

Proof. $x^k \in \varepsilon_k - \text{Argmin}\{f(\cdot) + \lambda_k^{-1}D_h(\cdot, x^{k-1})\} \Leftrightarrow 0 \in \partial_{\varepsilon_k}[f(\cdot) + \lambda_k^{-1}D_h(\cdot, x^{k-1})](x^k)$. From the Lemma 4.2, it exists $\varepsilon_{k_1}, \varepsilon_{k_2} \geq 0$ such as: $\varepsilon_{k_1} + \varepsilon_{k_2} = \varepsilon_k$ and $0 \in \partial_{\varepsilon_{k_1}}f(x^k) + \partial_{\varepsilon_{k_2}}(\lambda_k^{-1}D_h(\cdot, x^{k-1}))(x^k)$. Since $\partial_\varepsilon f$ increases with ε , we have: $0 \in \partial_{\varepsilon_k}f(x^k) + \partial_{\varepsilon_k}(\lambda_k^{-1}D_h(\cdot, x^{k-1}))(x^k)$. Therefore, it exists $z_k \in \partial_{\varepsilon_k}f(x^k)$ such as $-z_k \in \partial_{\varepsilon_k}(\lambda_k^{-1}D_h(\cdot, x^{k-1}))(x^k)$. From the Proposition 4.1, it exists $\bar{x}^k \in S$ such as

$$-z_k = \frac{\nabla h(\bar{x}^k) - \nabla h(x^{k-1})}{\lambda_k} \text{ and } D_h(x^k, \bar{x}^k) \leq \lambda_k \varepsilon_k.$$

Finally, it exists $\{\bar{x}^k\}$ such that:

$$\begin{cases} z_k = \frac{\nabla h(x^{k-1}) - \nabla h(\bar{x}^k)}{\lambda_k} \in \partial_{\varepsilon_k}f(x^k) \\ D_h(x^k, \bar{x}^k) \leq \lambda_k \varepsilon_k. \end{cases}$$

$f(x^k)$ is decreasing, so, $f(x^k) \rightarrow l \in [-\infty, +\infty[$. If $l = -\infty$, then $\lim f(x^k) = \text{inf}f = -\infty$. In what follows, we assume $f(x^k) \rightarrow l > -\infty$. The sequence is such that

$$\forall u \in \bar{S}, \forall k \geq 1, f(x^k) + \lambda_k^{-1}D_h(x^k, x^{k-1}) \leq f(u) + \lambda_k^{-1}D_h(u, x^{k-1}) + \varepsilon_k.$$

Replacing u with x^{k-1} we get:

$$f(x^k) + \lambda_k^{-1} D_h(x^k, x^{k-1}) \leq f(x^{k-1}) + \varepsilon_k.$$

When $k \rightarrow +\infty$,

$$\lambda_k^{-1} D_h(x^k, x^{k-1}) \rightarrow 0. \quad (3)$$

At this level, we distinguish two cases:

Case 1: $0 < \underline{\lambda} \leq \lambda_k \leq \bar{\lambda} < +\infty$.

$$(3) \Rightarrow D_h(x^k, x^{k-1}) \rightarrow 0,$$

on the other hand, $D_h(x^k, \bar{x}^k) \leq \lambda_k \varepsilon_k \leq \bar{\lambda} \varepsilon_k$. So the sequence $\{x^k\}$ verifies then:

$$\begin{cases} z_k = \frac{\nabla h(x^{k-1}) - \nabla h(\bar{x}^k)}{\lambda_k} \in \partial_{\varepsilon_k} f(x^k) \\ D_h(x^k, \bar{x}^k) \rightarrow 0 \\ D_h(x^k, x^{k-1}) \rightarrow 0. \end{cases}$$

Since $\text{Adh}\{x^k\} \subset \overline{\text{dom}f} \subset S$, the sequence $\{(\lambda_k; x^k; \bar{x}^k; x^{k-1}; z_k)\}_k$ verifies then the K-property. From the Lemma 4.4, $z_k \rightarrow 0$. On the other hand, $f(x) \geq f(x^k) + \langle z_k, x - x^k \rangle - \varepsilon_k$, which leads to $f(x) \geq \lim f(x^k)$, i.e., $\lim f(x^k) = \inf f$.

Case 2: $\{\lambda_k\}$ is not bounded. If $\{\lambda_k\}$ is not bounded, then it exists a sub-sequence $\{\lambda_{k_i}\}$ of $\{\lambda_k\}$ such that $\lambda_{k_i} \rightarrow +\infty$.

We have

$$f(x^{k_i}) + \lambda_{k_i}^{-1} D_h(x^{k_i}, x^{k_i-1}) \leq f(u) + \lambda_{k_i}^{-1} D_h(u, x^{k_i-1}) + \varepsilon_{k_i}.$$

$\{x^k\}$ is bounded, so from the Proposition 2.4, $\{D_h(u, x^{k_i-1})\}$ is bounded, which leads to:

$$l \leq f(u), \forall u \in \bar{S}.$$

Then:

$$\inf f = l = \lim f(x^k).$$

(b) Let $x^* \in \text{Adh}\{x^k\}$, it exists then the sub-sequence $\{x^{k_i}\}$ of $\{x^k\}$ such that $x^{k_i} \rightarrow x^*$. In any case, we have shown that: $\lim f(x^{k_i}) \leq \inf f$.

$$x^{k_i} \rightarrow x^* \Rightarrow f(x^*) \leq \lim f(x^{k_i}) \Rightarrow \inf f \leq f(x^*) \leq \lim f(x^{k_i}) \leq \inf f.$$

Finally we have:

$$f(x^*) = \lim f(x^k) = \inf f.$$

By additionating the hypothesis on h , we can establish the convergence of $\{x^k\}$ towards $x^* \in \text{Argmin}f$. We put

$$y^k := \text{prox}_{\lambda_k^h f}^h(x^{k-1}) = \text{argmin}\{f(\cdot) + \lambda_k^{-1} D_h(\cdot, x^{k-1})\}.$$

□

Proposition 4.6. $D_h(x^k, y^k) \leq \lambda_k \varepsilon_k$.

Proof.

$$y^k = \text{prox}_{\lambda_k f}^h(x^{k-1}) \Rightarrow \frac{\nabla h(x^{k-1}) - \nabla h(y^k)}{\lambda_k} \in \partial f(y^k) \Rightarrow \forall u, \lambda_k(f(u) - f(y^k)) \geq \langle u - y^k, \nabla h(x^{k-1}) - \nabla h(y^k) \rangle.$$

Thanks to the Lemma 2.2,

$$\langle u - y^k, \nabla h(x^{k-1}) - \nabla h(y^k) \rangle = D_h(u, y^k) + D_h(y^k, x^{k-1}) - D_h(u, x^{k-1}).$$

So:

$$\lambda_k(f(y^k) - f(u)) \leq D_h(u, x^{k-1}) - D_h(u, y^k) - D_h(y^k, x^{k-1}). \quad (4)$$

On the other hand,

$$\forall u \in \bar{S}, f(x^k) + \lambda_k^{-1} D_h(x^k, x^{k-1}) \leq f(u) + \lambda_k^{-1} D_h(u, x^{k-1}) + \varepsilon_k. \quad (5)$$

Replacing u with y^k in (5) we obtain:

$$\lambda_k(f(x^k) - f(y^k)) \leq D_h(y^k, x^{k-1}) - D_h(x^k, x^{k-1}) + \lambda_k \varepsilon_k. \quad (6)$$

Replacing u with x^k in (4) we have:

$$\lambda_k(f(y^k) - f(x^k)) \leq D_h(x^k, x^{k-1}) - D_h(x^k, y^k) - D_h(y^k, x^{k-1}). \quad (7)$$

$$(6) \text{ and } (7) \Rightarrow 0 \leq -D_h(x^k, y^k) + \lambda_k \varepsilon_k \Rightarrow D_h(x^k, y^k) \leq \lambda_k \varepsilon_k. \quad \square$$

Corollary 4.7. *If ∇h is a strongly convex operator, then it exists $\alpha > 0$ such that*

$$\|x^k - y^k\| \leq \sqrt{\frac{2}{\alpha} \lambda_k \varepsilon_k}.$$

Proof.

(b) From the Proposition 4.6, $D_h(x^k, y^k) \leq \lambda_k \varepsilon_k$. ∇h is a strongly convex operator, from the Proposition 3.9. [5], it exists $\alpha > 0$ such that:

$$\frac{\alpha}{2} \|x^k - y^k\|^2 \leq D_h(x^k, y^k).$$

$$\text{Then } \|x^k - y^k\| \leq \sqrt{\frac{2}{\alpha} \lambda_k \varepsilon_k}. \quad \square$$

Proposition 4.8. *We assume:*

(a). $\{x^k\}$ generated by $P_3(h)$ is bounded.

(b). h is twice Continuously Differentiable on S and $D_h(\cdot, \cdot)$ is convex jointly.

(c). ∇h is a strongly convex operator

Then

$$D_h(x^*, x^k) \rightarrow l \geq 0, \forall x^* \in \text{Argmin} f.$$

Proof. From Theorem 4.5, let $Argmin f$. Let $x^* \in Argmin f$, from (4) we have:

$$D_h(x^*, y^k) + D_h(y^k, x^{k-1}) - D_h(x^*, x^{k-1}) \leq 0.$$

whence

$$D_h(y^k, x^{k-1}) \leq D_h(x^*, x^{k-1}) - D_h(x^*, x^k) + D_h(x^*, x^k) - D_h(x^*, y^k).$$

(2) $\Rightarrow \partial(D_h(x^*, \cdot))(y) = H(y)(y - x^*)$, where $H(y) = \nabla^2 h(y)$. So

$$D_h(x^*, y^k) - D_h(x^*, x^k) \geq \langle y^k - x^k, H(x^k)(x^k - x^*) \rangle.$$

It follows that

$$D_h(y^k, x^{k-1}) \leq D_h(x^*, x^{k-1}) - D_h(x^*, x^k) + \|y^k - x^k\| \cdot \|x^* - x^k\| \cdot \|H(x^k)\|$$

h is twice Continuously Differentiable on S and $\{x^k\}$ is bounded, so $\{H(x^k)\}$ is bounded, therefore

$$\exists K > 0, \forall k, \sqrt{\frac{2}{\alpha}} \|x^* - x^k\| \cdot \|H(x^k)\| \leq K.$$

Whence

$$D_h(y^k, x^{k-1}) \leq D_h(x^*, x^{k-1}) - D_h(x^*, x^k) + K\sqrt{\lambda_k \varepsilon_k}. \quad (8)$$

Since $D_h(y^k, x^{k-1}) \geq 0$, we have:

$$D_h(x^*, x^k) \leq D_h(x^*, x^{k-1}) + K\sqrt{\lambda_k \varepsilon_k},$$

$$\sum_k \sqrt{\lambda_k \varepsilon_k} < +\infty \Rightarrow D_h(x^*, x^k) \rightarrow l \geq 0. \quad \square$$

Theorem 4.9. *We assume:*

- (1). $0 < \underline{\lambda} \leq \lambda_k$,
- (2). $\sum \sqrt{\lambda_k \varepsilon_k} < +\infty$,
- (3). $\{x^k\}$ generated by $P_3(h)$ is bounded,
- (4). h is twice Continuously Differentiable on S and $D_h(\cdot, \cdot)$ is convex jointly,
- (5). ∇h is a strongly convex operator.

Then

$$(a). f(x^k) \rightarrow \inf f.$$

$$(b). x^k \rightarrow x^* \in Argmin f.$$

Proof.

(a). is verified from the Theorem 4.5.

(b). Let $x^* \in Adh\{x^k\}$, it exists then the sub-sequence $\{x^{k_i}\}$ of $\{x^k\}$ such that $x^{k_i} \rightarrow x^*$. From H_4 , $D_h(x^*, x^{k_i}) \rightarrow 0$. From the Proposition 4.8. and (B), we have: $D_h(x^*, x^{k_i}) \rightarrow l$. So

$$D_h(x^*, x^k) \rightarrow 0$$

According to the Proposition 2.3, $x^k \rightarrow x^*$. □

5. Conclusion

The entropic proximal algorithm proposed in this labor constitute a unified framework for the existing algorithms and provide others, thus:

- If $h(\cdot) = \frac{1}{2}\|\cdot\|^2$, then $P_1 \Leftrightarrow P_3(h)$.
- If $\varepsilon_k = 0, \forall k$ then $P_2(h) \Leftrightarrow P_3(h)$.

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