

# An Asymptotical Variational Principle Associated with the Steepest Descent Method for a Convex Function

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**Dedicated to R. T. Rockafellar on his 60th Birthday**

The asymptotical limit of the trajectory defined by the continuous steepest descent method for a proper closed convex function  $f$  on a Hilbert space is characterized in the set of minimizers of  $f$  via an asymptotical variational principle of Brezis-Ekeland type. The implicit discrete analogue (prox method) is also considered.

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## 1. Introduction

Let  $X$  be a real Hilbert space endowed with inner product  $\langle \cdot, \cdot \rangle$  and associated norm  $\|\cdot\|$ , and let  $f$  be a proper closed convex function on  $X$ .

The paper considers the problem of minimizing  $f$ , that is, of finding  $\inf_X f$  and some element in the optimal set  $S := \text{Argmin } f$ , this set assumed being non empty.

Letting  $\partial f$  denote the subdifferential operator associated with  $f$ , we focus on the continuous steepest descent method associated with  $f$ , *i.e.*, the differential inclusion

$$-\frac{du}{dt} \in \partial f(u), \quad t > 0$$

with initial condition

$$u(0) = u_0.$$

This method is known to yield convergence under broad conditions summarized in the following theorem. Let us denote by  $\mathcal{A}$  the real vector space of continuous functions from  $[0, +\infty[$  into  $X$  that are absolutely continuous on  $[\delta, +\infty[$  for all  $\delta > 0$ .

**Theorem 1.1.**

- (i) ([2]) For all  $u_0$  in  $\text{cl dom } f$ , the closure of the effective domain of  $f$ , there exists a unique  $u$  in  $\mathcal{A}$  such that:
- $\forall t > 0, u(t)$  belongs to the domain of  $\partial f$ ;
  - the differential inclusion is verified for a.e.  $t > 0$  (recall that  $u$  is strongly derivable for a.e.  $t > 0$ );
  - $\forall \delta > 0, u$  is Lipschitz continuous on  $[\delta, +\infty[$  i.e.  $u'(\cdot) := \frac{du}{dt} \in \mathcal{L}^\infty([\delta, +\infty[; X)$ ;
  - $f \circ u$  is convex, non increasing, Lipschitz continuous on  $[\delta, +\infty[ \forall \delta > 0$ ;
  - $u$  has a derivative to the right for all  $t > 0$  and, if  $S := \text{Argmin } f$  is non empty, then  $\lim_{t \rightarrow +\infty} \|\frac{du}{dt}^+(t)\| = 0$ .
- (ii) ([5]) If  $S \neq \emptyset$  then the weak limit  $u_\infty := w - \lim_{t \rightarrow +\infty} u(t)$  exists and belongs to  $S$ , this asymptotic convergence being true for the norm topology if  $f$  is even ([5]) or if  $\text{Int } S \neq \emptyset$  ([2]).

Assuming henceforth that the optimal set  $S$  is non empty, Theorem 1.1 motivates us to consider the following problem:

For a given  $u_0 \in \text{cl dom } f$ , characterize the asymptotic limit  $u_\infty$  as the unique solution to some convex optimization problem with feasible set  $S$ .

This task amounts to show that, like regularization procedures *à la Tikhonov* ([15], [7], [1]), the steepest descent method enables us to select a particular optimal solution respectively to some auxiliary criterium. As we shall see, unlike regularization procedures, this auxiliary criterium is not at our disposal independently from the data  $f$  but is defined from this data. Indeed, the result is based upon a variational principle of Brezis-Ekeland type for an infinite horizon. An analogue holds true for the Euler's implicit discretization of the differential inclusion i.e. the proximal method of Martinet-Rockafellar.

The paper is organized as follows. Before reaching the main objective stated above, we show in section 2, that the trajectory  $u$  is always (even  $S$  is empty) minimizing and, if  $S$  is non empty, we give a localization result of  $u_\infty$  in  $S$  with respect to  $u_0$ . Section 3 is devoted to the variational principle for an infinite horizon. In section 4 we state the asymptotic variational principle characterizing  $u_\infty$  in  $S$ . The analogue for the discrete case is discussed in section 5.

In sections 2, 3, 4,  $u$  always denotes the solution mentioned in Theorem 1.1.

## 2. Localization of $u_\infty$

First let us recall an estimate.

**Proposition 2.1.** [9]

$$\forall x \in X, \forall t > 0, \frac{\|u(t) - x\|^2}{2t} + f(u(t)) \leq f(x) + \frac{\|u_0 - x\|^2}{2t}$$

**Proof.** By definition of the subdifferential we have

$$\forall x \in X, \text{ for a.e. } s > 0, f(x) \geq f(u(s)) - \langle u'(s), x - u(s) \rangle$$

Then, integrating between  $\delta$  and  $t$  with  $0 < \delta < t$ , and thanks to the non increasingness of  $f(u(\cdot))$  and the absolute continuity of  $u$  on  $[\delta, +\infty[$ , we get

$$(t - \delta)f(x) \geq (t - \delta)f(u(t)) + (\|u(t) - x\|^2 - \|u(\delta) - x\|^2)/2$$

Finally, since  $u$  is continuous at 0 we get the result.  $\square$

**Corollary 2.2.**

- (i) In all cases, that is, whether  $\inf_X f$  be finite or not, achieved or not,  $\lim_{t \rightarrow +\infty} f(u(t)) = \inf_X f$ .
- (ii) If  $S \neq \emptyset$ , then  $\|u_\infty - \text{proj}_S u_0\| \leq d(u_0, S)$  and therefore,  $\|u_\infty - u_0\| \leq 2d(u_0, S)$ , where  $\text{proj}_S$  denotes the projection operator onto the closed convex set  $S$ .
- (iii) If  $S$  is an affine subspace then  $u_\infty = \text{proj}_S u_0$ .

**Proof.** (i) Trivial.

(ii) Taking  $x \in S$  in the estimate we get  $\|u(t) - x\| \leq \|u_0 - x\|$ . Passing to the  $\liminf$  as  $t \rightarrow +\infty$ , thanks to the weak lower semi-continuity of the norm, and taking as  $x$  the projection of  $u_0$  onto  $S$ , we get the result.

(iii) Let  $e := u_\infty - \text{proj}_S u_0$  and assume  $e \neq 0$ . As  $S$  is an affine subspace,  $x := \text{proj}_S u_0 - td(u_0, S)e/\|e\|$  is in  $S$  for all  $t \geq 0$ . We have  $\|u_\infty - x\| = td(u_0, S) + \|e\|$  and, thanks to Pythagore,  $\|u_0 - x\| = \sqrt{t^2 + 1}d(u_0, S)$ . Therefore,  $\|u_0 - x\| - \|u_\infty - x\| = (\sqrt{t^2 + 1} - t)d(u_0, S) - \|e\|$ . Referring to the proof of (ii) above we get  $\|e\| \leq (\sqrt{t^2 + 1} - t)d(u_0, S)$  for all  $t \geq 0$ , which is a contradiction.  $\square$

**Remark 2.3.** (iii) in corollary 2.2 gives a complete and simple characterization of  $u_\infty$  in  $S$ , actually the same than the Tikhonov regularization method. This characterization has yet been obtained in [4] for the discrete method (cf. section 5 below) when  $f$  is a quadratic form (a case where  $S$  is a subspace). Unfortunately this simple characterization fails to be true in general.

**3. Variational principle, infinite horizon**

First, let  $T > 0$  be fixed. We denote by  $\mathcal{L}_{sc}^1(0, T)$  the set of functions  $\psi$  from  $[0, T]$  into  $\mathbb{R}$  that belong to  $\mathcal{L}^1(\delta, T)$  for all  $\delta$ ,  $0 < \delta < T$ , and such that  $\int_0^T \psi(t)dt := \lim_{\delta \rightarrow 0} \int_\delta^T \psi(t)dt$  exists in  $\mathbb{R}$ .

We define the set of feasible trajectories which reach some  $v_\infty \in X$  from  $u_0$ , possibly using infinite time, arriving asymptotically with nul velocity:

$$K(u_0) := \{v \in \mathcal{A}; f \circ v + f^* \circ (-v') \in \mathcal{L}_{sc}^1(0, T) \ \forall T > 0,$$

$$v(0) = u_0, v_\infty := w - \lim_{t \rightarrow +\infty} v(t) \text{ exists, } s - \text{ess} - \lim_{t \rightarrow +\infty} v'(t) = 0\},$$

where  $w$  is short notation for *weak*,  $s$  is for *strong*, *ess* is for *essential*, and the  $*$  denotes the Fenchel conjugacy.

We note that, if  $v$  is in  $K(u_0)$  then  $v(t)$  is in  $\text{dom } f$  for all  $t > 0$  and  $-v'(t)$  is in  $\text{dom } f^*$  for a.e.  $t > 0$ .

**Proposition 3.1.**  $K(u_0)$  is a convex subset of  $\mathcal{A}$  containing  $u$ .

**Proof.** Convexity is easy. For the second assertion, integrating the equality

$$f(u(t)) + f^*(-u'(t)) = -\langle u'(t), u(t) \rangle \text{ for a.e. } t > 0$$

between  $\delta$  and  $T$  for  $\delta > 0$  and passing to the limit as  $\delta$  tends to 0, we get

$$\int_0^T [f(u(t)) + f^*(-u'(t))] dt = \frac{1}{2}(\|u_0\|^2 - \|u(T)\|^2)$$

The last condition is fulfilled since (cf. Theorem 1.1)  $\lim_{t \rightarrow +\infty} \|\frac{du}{dt}^+(t)\| = 0$ .  $\square$

Motivated by Proposition 3.1 we define the cost function for the finite horizon  $T$ ,  $J_T : K(u_0) \rightarrow \mathbb{R}$  by

$$\begin{aligned} J_T(v) &:= \int_0^T [f(v(t)) + f^*(-v'(t)) + \langle v'(t), v(t) \rangle] dt \\ &= \int_0^T [f(v(t)) + f^*(-v'(t))] dt + \frac{1}{2}(\|v(T)\|^2 - \|u_0\|^2) \end{aligned}$$

As  $u_0$  is fixed, we note that  $J_T$  is convex. Moreover, via Fenchel's inequality we have  $J_T \geq 0$ .

**Remark 3.2.** According to Moreau's theory for elastoplastic systems [11,12,13],  $J_T(v)$  can be interpreted as the total energy of such a system in the time interval  $[0, T]$  if it moves with the velocity  $v$ . Indeed,  $f$  is then the support function of some closed convex subset  $C$  containing the origin (therefore  $\text{Argmin } f$  is the normal cone to  $C$  at 0),  $f(v(t))$  is the dissipated power and  $\langle v'(t), v(t) \rangle$  is the kinetic power.

Now let us define the (convex) cost function for an infinite horizon  $J : K(u_0) \rightarrow \mathbb{R}_+ \cup \{+\infty\}$  by

$$J(v) := \lim_{T \rightarrow +\infty} J_T(v) = \sup_{T > 0} J_T(v)$$

**Theorem 3.3.** (Variational principle, variant of [3]).  $u$  is the unique minimizer of  $J$  on  $K(u_0)$  and the minimum value is nul.

**Proof.** Clearly  $J(u) = 0$  and  $J(v) = 0$  if and only if  $J_T(v) = 0$  for all  $T > 0$  i.e.  $v$  satisfies the differential inclusion on  $[0, T]$  for all  $T > 0$  i.e.  $v = u$ .  $\square$

The following result will be of importance in section 5.

**Proposition 3.4.**

$$\forall v \in K(u_0), \quad J(v) \geq \frac{1}{2} \|v_\infty - u_\infty\|^2.$$

**Proof.** From the definition of a subgradient and as  $-u' \in \partial f(u) \Leftrightarrow u \in \partial f^*(-u')$ , we have

$$f(v) - f(u) + f^*(-v') - f^*(-u') \geq -\langle u', v - u \rangle + \langle u, u' - v' \rangle$$

Therefore, adding  $\langle v', v \rangle - \langle u', u \rangle$  to both sides, we obtain

$$\forall T > 0, \quad J_T(v) = J_T(v) - J_T(u) \geq \lim_{\delta \rightarrow 0} \int_{\delta}^T \langle v' - u', v - u \rangle dt = \frac{1}{2} \|v(T) - u(T)\|^2$$

Then use the weak lower semi-continuity of the norm as  $T \rightarrow +\infty$ . □

We end this section by studying some properties of the cost function  $J$  that make the optimization problem involved in theorem 3.3 non trivial in the sense that  $J$  is finite not only at  $u$ .

**Proposition 3.5.**

- (i) For all  $v \in K(u_0)$  such that  $v_{\infty} \notin \text{Argmin } f$  then  $J(v) = +\infty$ .
- (ii) If  $f$  is coercive then for all  $x \in \text{Argmin } f$  there exists  $v$  in  $K(u_0)$  such that  $v_{\infty} = x$  and  $J(v) < +\infty$ .

**Proof.** (i) By (weak and strong) lower semi-continuity of  $f$  and  $f^*$ , we get

$$\liminf_{t \rightarrow +\infty} [f(v(t)) + f^*(-v'(t)) + \langle v'(t), v(t) \rangle] \geq f(v_{\infty}) + f^*(0) > 0$$

where the last inequality follows from  $0 \notin \partial f(v_{\infty})$ . Evidently, this inequality implies  $J(v) = +\infty$ .

(ii) Let  $0 < \delta < \tau$ . For  $x \in X$ , let us define the function  $v$  from  $[0, +\infty[$  into  $X$  by

$$v(t) := \begin{cases} u(t) & 0 \leq t \leq \delta \\ u(\delta) + (t - \delta) \frac{x - u(\delta)}{\tau - \delta} & \delta \leq t \leq \tau \\ x & \tau \leq t \end{cases}$$

Clearly  $v$  is in  $\mathcal{A}$ ,  $v_{\infty} = v(\tau) = x$  and  $v'(t) = 0$  for all  $t > \tau$ .

Since  $x$  is in  $\text{dom } f$  then  $v(t)$  is in  $\text{dom } f$  for all  $t > 0$ . Since  $f$  is weakly inf-compact *i.e.*  $f^*$  is strongly continuous at 0, if  $\tau$  is large enough then  $-v'(t)$  is in  $\text{dom } f^*$  for *a.e.*  $t > 0$ . Finally, because  $f(x) + f^*(0) = 0$  and  $f \circ v$  is integrable on  $[\delta, \tau]$ , we have

$$\forall T \geq \tau, \quad J(v) = J_T(v) = J_{\tau}(v) = \int_{\delta}^{\tau} f(v(t)) dt + (\tau - \delta) f^*\left(\frac{u(\delta) - x}{\tau - \delta}\right) + \frac{1}{2} (\|x\|^2 - \|u(\delta)\|^2) < +\infty,$$

taking into account that  $J_{\delta}(v) = J_{\delta}(u) = 0$ . □

**4. Asymptotical variational principle**

Let us define the asymptotic cost function  $\varphi_{u_0} : X \rightarrow \mathbb{R}_+ \cup \{+\infty\}$  by

$$\varphi_{u_0}(x) := \inf \{ J(v); \quad v \in K(u_0), \quad v_{\infty} = x \}$$

**Proposition 4.1.**  $\varphi_{u_0}$  is convex;  
 $\varphi_{u_0}(u_\infty) = J(u) = 0$ ;  
 $\forall x \in X, \varphi_{u_0}(x) \geq \frac{1}{2}\|x - u_\infty\|^2$ .

**Proof.** Let  $x^1, x^2 \in X$ ,  $\theta \in [0, 1]$  and  $v^i \in K(u_0)$  such that  $v_\infty^i = x^i$ ,  $i = 1, 2$ . As  $K(u_0)$  is convex,  $v^\theta := \theta v^1 + (1 - \theta)v^2 \in K(u_0)$ . Moreover  $v_\infty^\theta = x^\theta := \theta x^1 + (1 - \theta)x^2$ . Then

$$\varphi_{u_0}(x^\theta) \leq J(v^\theta) \leq \theta J(v^1) + (1 - \theta)J(v^2)$$

As this holds for all  $v^i \in K(u_0)$  such that  $v_\infty^i = x^i$ , we are done for the first statement. The second statement is immediate, the third one comes directly from proposition 3.4.  $\square$

Now, as a direct consequence of proposition 4.1, we can state the announced asymptotic variational principle.

**Theorem 4.2.**  $u_\infty$  is the unique minimizer of  $\varphi_{u_0}$  on  $\text{Argmin } f$  (or on  $X$ ) and the minimum value is zero.  $\square$

Finally, in order that the optimization problem involved in theorem 4.2 be non trivial, referring to proposition 3.5, we get

**Proposition 4.3.** If  $f$  is coercive then  $\text{dom } \varphi_{u_0} = \text{Argmin } f$ .  $\square$

## 5. Discrete case

For the same convex function  $f$  there is an intimate relationship between the continuous steepest descent method and its Euler's *implicit* discrete version, *i.e.*, the proximal method of Martinet-Rockafellar [14], [8]. Thanks to this connexion, the proximal method inherits many of the nice asymptotical properties of the continuous steepest descent method [9]. For the problem under consideration here, concerning characterization of the asymptotical limit, a similar discussion for the proximal method is entirely parallel to the one above. The statements are the same, albeit now "time" is discrete.

More precisely, let  $X$  and  $f$  be as in section 1 and  $\{\lambda_k\}$  be a sequence of positive reals. Let  $u_0 \in X$  be given. The proximal method [14] generates a sequence  $u = \{u_k\} \in X$ ,  $u_k$  being the unique solution of the iterative scheme

$$\frac{u_{k-1} - u_k}{\lambda_k} \in \partial f(u_k) \quad \forall k \geq 1.$$

Thus this method coincides with the Euler's *implicit* discretization of the differential inclusion arising in the continuous steepest descent (section 1). So, for all  $n \in \mathbb{N}$ ,  $u_n$  approximates the continuous steepest descent trajectory at the point  $t_n := \sum_1^n \lambda_k$ .

Let us recall some basic known results about the asymptotic behaviour.

**Proposition 5.1.** [14] If  $\lambda_k$  is bounded away from 0 and  $\text{Argmin } f \neq \emptyset$  then  $u_k$  weakly converges to  $u_\infty$  some minimizer of  $f$ .

Note that here the asymptotic limit  $u_\infty$  may be different from that one of the continuous case (for the same  $u_0$ ).

**Remark 5.2.** In proposition 5.1, the convergence holds true for the norm topology if  $f$  is even [4] or if  $\text{Int Argmin } f \neq \emptyset$  [9] or if  $f$  is well-posed [10].

Proposition 2.1 holds true replacing  $u(t)$  by  $u_n$  and  $t$  by  $t_n := \sum_1^n \lambda_k$  ([6]), implying  $\lim_{n \rightarrow +\infty} f(u_n) = \inf_X f$  and the same localization results as in corollary 2.2.

Now, all statements in sections 3 and 4 still hold replacing  $\mathcal{A}$  by  $X^{\mathbb{N}}$  and with the following new definitions.

**Set of discrete feasible trajectories from  $u_0$ :**

$$\begin{aligned} K(u_0) := \{ & v = (v_0, v_1, \dots); \quad \forall k \geq 1 \\ & v_k \in \text{dom } f, \\ & -d_k := \frac{v_{k-1} - v_k}{\lambda_k} \in \text{dom } f^*, \\ & v_0 := u_0, \quad v_k \xrightarrow{w} v_\infty, \quad \|d_k\| \rightarrow 0 \} \end{aligned}$$

**Cost function for a finite horizon:**

$$\begin{aligned} J_n(v) &:= \sum_{k=1}^n \lambda_k [f(v_k) + f^*(-d_k) + \langle v_k, d_k \rangle] \\ &= \sum_{k=1}^n \lambda_k [f(v_k) + f^*(-d_k) + \frac{1}{2} \lambda_k \|d_k\|^2] \\ &\quad + \frac{1}{2} (\|v_n\|^2 - \|u_0\|^2) \end{aligned}$$

**Cost function for an infinite horizon:**

$$J(v) := \lim_{n \rightarrow +\infty} J_n(v) = \sup_n J_n(v)$$

**Asymptotic cost function:** exactly the same as in section 4.

The adaptation of proofs is left out as a simple exercise. The crucial trick is

$$2\langle u_{k-1} - u_k, x - u_k \rangle = \|u_{k-1} - u_k\|^2 + \|x - u_k\|^2 - \|x - u_{k-1}\|^2$$

For the proof of the analogue of proposition 3.5 (ii), define  $v \in K(u_0)$  by:

$$v_k := \begin{cases} u_k & 0 \leq k \leq 1 \\ u_1 + (t_k - t_1) \frac{x - u_1}{t_N - t_1} & 1 < k \leq N \\ x & k > N \end{cases}$$

where  $N > 1$  large enough in order that, for  $1 < k \leq N$ ,  $-d_k = \frac{u_1 - x}{t_N - t_1}$  be in  $\text{dom } f^*$  since  $\lim_{n \rightarrow +\infty} t_n = +\infty$ .

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