# A CONTINUOUS MAPPING THEOREM FOR THE ARGMIN-SET FUNCTIONAL WITH APPLICATIONS TO CONVEX STOCHASTIC PROCESSES

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For lower-semicontinuous and convex stochastic processes  $Z_n$  and nonnegative random variables  $\epsilon_n$  we investigate the pertaining random sets  $A(Z_n, \epsilon_n)$  of all  $\epsilon_n$ -approximating minimizers of  $Z_n$ . It is shown that, if the finite dimensional distributions of the  $Z_n$  converge to some Z and if the  $\epsilon_n$  converge in probability to some constant c, then the  $A(Z_n, \epsilon_n)$  converge in distribution to A(Z, c) in the hyperspace of Vietoris. As a simple corollary we obtain an extension of several argmin-theorems in the literature. In particular, in contrast to these argmin-theorems we do not require that the limit process has a unique minimizing point. In the non-unique case the limit-distribution is replaced by a Choquet-capacity.

Keywords: convex stochastic processes, sets of approximating minimizers, weak convergence, Vietoris hyperspace topologies, Choquet-capacity

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## 1. INTRODUCTION AND MAIN RESULTS

Let  $(\Omega, \mathcal{A}, \mathbb{P})$  be a probability space and let  $Z : \Omega \times \mathbb{R}^d \to \overline{\mathbb{R}}$  be a bivariate function with values in the extended real line  $\overline{\mathbb{R}}$  endowed with the Borel- $\sigma$  algebra  $\overline{\mathcal{B}}$ . Such a function is called *stochastic process* or *integrand*, if  $Z(\cdot, t) : (\Omega, \mathcal{A}) \to (\overline{\mathbb{R}}, \overline{\mathcal{B}})$  is measurable for every  $t \in \mathbb{R}^d$ . It is convenient to identify a stochastic process with a function-valued map  $Z : \Omega \to \overline{\mathbb{R}}^{\mathbb{R}^d}$ . So,  $Z(\omega) \equiv Z(\omega, \cdot)$  is a function from  $\mathbb{R}^d$  into  $\overline{\mathbb{R}}$ , which is called the *trajectory* or *path* of Z pertaining to the sample point  $\omega \in \Omega$ . It takes the value  $Z(\omega)(t) \equiv Z(\omega, t)$  at point  $t \in \mathbb{R}^d$ . Occasionally it is practical to write Z(t) instead of  $Z(\cdot, t)$  for this ambiguity in the notation explains in the context.

In this paper we focus on integrands Z which are lower-semicontinuous (lsc) and convex. For a lsc function  $f : \mathbb{R}^d \to \overline{\mathbb{R}}$  and a real number  $r \in \mathbb{R}_+ = [0, \infty)$  let

$$A(f,r) := \{t \in \mathbb{R}^d : f(t) \le \inf_{s \in \mathbb{R}^d} f(s) + r\}$$

and

$$\operatorname{Argmin}(f) := \{ t \in \mathbb{R}^d : f(t) = \inf_{s \in \mathbb{R}^d} f(s) \}$$

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Thus A(f,r) is the set of all *r*-approximating minimizers of f and  $\operatorname{Argmin}(f)$  consists of all minimizers of f. Obviously,  $\operatorname{Argmin}(f) = A(f, 0)$ . By lower-semicontinuity A(f, r)is a closed subset of  $\mathbb{R}^d$  (possibly empty), see Lemma 4.1 in the appendix. Consider the space S of all lower-semicontinuous functions from  $\mathbb{R}^d$  into the extended real line  $\overline{\mathbb{R}}$ , i.e.

$$S := \{ f : \mathbb{R}^d \to \overline{\mathbb{R}}; f \, \operatorname{lsc} \}.$$

Then the assignment  $(f, r) \mapsto A(f, r)$  defines a map

$$A: S \times \mathbb{R}_+ \to \mathcal{F}_d,$$

where

$$\mathcal{F}_d := \mathcal{F}(\mathbb{R}^d) := \{F \subseteq \mathbb{R}^d : F \text{ is closed }\}$$

is the family of all closed subsets of  $\mathbb{R}^d$ . For a fixed lsc integrand Z and a nonnegative random real variable  $\epsilon$  on  $(\Omega, \mathcal{A})$  we have that  $A(Z, \epsilon) := A \circ (Z, \epsilon)$  is a map from  $(\Omega, \mathcal{A})$ into  $\mathcal{F}_d$ , or in other words a  $\mathcal{F}_d$ -valued random element.

Now, let  $(Z_n)$  be a sequence of lsc and convex stochastic processes accompanied by a sequence  $(\epsilon_n)$  of nonnegative random variables. Assume that

$$(Z_n(t_1), \dots, Z_n(t_k)) \xrightarrow{\mathcal{D}} (Z(t_1), \dots, Z(t_k)) \quad \text{in } \overline{\mathbb{R}}^k \text{ as } n \to \infty,$$
 (1)

for all  $t_1, \ldots, t_k \in D$ , where D is any countable and dense subset of  $\mathbb{R}^d$  (convergence of the finite-dimensional distributions on D). This is denoted by  $Z_n \xrightarrow{fd}_D Z$ . Further, assume that the sequence  $(\epsilon_n)$  converges in probability:

$$\epsilon_n \xrightarrow{\mathbb{I}^{\nu}} c,$$
 (2)

where  $c \ge 0$  is a real constant.

We now state our main results. For that purpose let  $\mathcal{F}_d = \mathcal{F}(\mathbb{R}^d)$  be endowed with either the Vietoris topology  $\tau_V = \tau_V(\mathcal{F}_d)$  or the upper Vietoris topology  $\tau_{uV} = \tau_{uV}(\mathcal{F}_d)$ . Here, the Vietoris topology  $\tau_V(\mathcal{F}_d)$  is generated through the system  $\mathcal{S}_V := \{\mathcal{M}(F) : F \in \mathcal{F}_d\} \cup \{\mathcal{H}(G) : G \in \mathcal{G}_d\}$ , where  $\mathcal{G}_d$  denotes the class of all open subsets in  $\mathbb{R}^d$ ,  $\mathcal{M}(E) := \{F \in \mathcal{F}_d : F \cap E = \emptyset\}$  is the collection of all missing sets of a set  $E \subseteq \mathbb{R}^d$ and  $\mathcal{H}(E) := \{F \in \mathcal{F}_d : F \cap E \neq \emptyset\}$  is the collection of all hitting sets of E. The upper Vietoris topology  $\tau_{uV}$  is generated by the sub-system  $\mathcal{S}_{uV} := \{\mathcal{M}(F) : F \in \mathcal{F}_d\}$ , whence it is coarser than the Vietoris topology.

The issue is to give minimal conditions such that our basic assumptions (1) and (2) ensure distributional convergence of  $A(Z_n, \epsilon_n)$  to A(Z, c) in the topological space  $(\mathcal{F}_d, \tau_{uV})$  or  $(\mathcal{F}_d, \tau_V)$ , respectively. These conditions concern the path properties of Zand  $Z_n$ . For their description recall that a function  $f : \mathbb{R}^d \to \mathbb{R}$  is called proper if  $f(t) > -\infty$  for all  $t \in \mathbb{R}^d$  and  $f(t) < \infty$  for at least one  $t \in \mathbb{R}^d$ . The set dom  $f := \{t \in \mathbb{R}^d : f(t) < \infty\}$  is called the *effective domain* of f. Further, f is *level-bounded*, if for every  $\alpha \in \mathbb{R}$  the *level-set*  $\{t \in \mathbb{R}^d : f(t) \le \alpha\}$  is bounded (possibly empty). This is the same as having  $f(t) \to \infty$  as  $|t| \to \infty$ , where  $|\cdot|$  is the euclidian norm on  $\mathbb{R}^d$ . Henceforth, we can introduce the subspaces  $S_0 := \{f \in S : f \text{ convex}, \text{ proper and int}(\text{dom } f) \neq \emptyset\}$ , where int(E) denotes the interior of  $E \subseteq \mathbb{R}^d$  and  $S_1 := \{f \in S_0 : f \text{ level-bounded}\}$ . It is easy to see that  $S_0 = \{f \in S : f \text{ convex}$  and finite on some non-empty open subset}, see Lemma 4.4 in the appendix. **Theorem 1.1.** Assume that Z and every  $Z_n$  have trajectories in  $S_0$  and that  $Z \in S_1 \mathbb{P}$ almost surely (a.s.). Then  $Z_n \xrightarrow{fd}_D Z$  and  $\epsilon_n \xrightarrow{\mathbb{P}} c$  yield

$$A(Z_n, \epsilon_n) \twoheadrightarrow^{\sim} A(Z, c) \text{ in } (\mathcal{F}_d, \tau_{uV}), \tag{3}$$

where  $\twoheadrightarrow^{\sim}$  denotes *convergence in Borel law*. Moreover, A(Z, c) is a.s. non-empty and compact.

Convergence in Borel law is introduced and investigated by Hoffmann-Jørgensen [10]. Now,  $S_{uV}$  is actually a base of  $\tau_{uV}$ , because  $\bigcap_{i=1}^{n} \mathcal{M}(F_i) = \mathcal{M}(\bigcup_{i=1}^{n} F_i)$  whenever  $F_1, \ldots, F_n$  are closed (or even arbitrary) subsets of  $\mathbb{R}^d$ . Therefore it follows from the Borel Law Portmanteau Theorem of [10] that (3) is equivalent to

$$\limsup_{n \to \infty} \mathbb{P}^* \Big( \bigcap_{F \in \mathcal{F}'} \{ A(Z_n, \epsilon_n) \cap F \neq \emptyset \} \Big) \le \mathbb{P}_* \Big( \bigcap_{F \in \mathcal{F}'} \{ A(Z, c) \cap F \neq \emptyset \} \Big)$$
(4)

for every sub-collection  $\mathcal{F}' \subseteq \mathcal{F}_d$  of closed sets in  $\mathbb{R}^d$ . Here,  $\mathbb{P}^*$  and  $\mathbb{P}_*$  denote the outer and inner probability of  $\mathbb{P}$ . An essential feature of convergence in Borel law is that the involved random elements are simply maps from  $\Omega$  into  $\mathcal{F}_d$  without any measurability requirement. In fact, there is no  $\sigma$ -algebra on  $\mathcal{F}_d$  so far. So, let us endow  $\mathcal{F}_d$  with the Borel- $\sigma$ -algebra  $\mathcal{B}_{uV} := \mathcal{B}_{uV}(\mathcal{F}_d) := \sigma(\tau_{uV}(\mathcal{F}_d))$  pertaining to the upper Vietoris topology. The following result sharpens the Borel law convergence (3) to classical weak convergence under the additional assumption that the  $Z_n$  in Theorem 1.1 are level bounded as well.

**Theorem 1.2.** Assume that Z and every  $Z_n$  have trajectories in  $S_1$ . Then  $Z_n \xrightarrow{fd}_D Z$  and  $\epsilon_n \xrightarrow{\mathbb{P}} c$  entail

$$A(Z_n, \epsilon_n) \xrightarrow{\mathcal{D}} A(Z, c) \text{ in } (\mathcal{F}_d, \tau_{uV}) \text{ as } n \to \infty.$$
 (5)

Furthermore, A(Z,c) and all  $A(Z_n, \epsilon_n)$  are non-empty and compact.

Note that  $(\mathcal{F}_d, \tau_{uV})$  is a topological space, which is not metrizable. Therefore, we need to say a few words about the meaning of (5). Firstly it means that the  $A(Z_n, \epsilon_n)$  and A(Z, c) are  $\mathcal{A} - \mathcal{B}_{uV}$  measurable maps from  $\Omega$  into  $\mathcal{F}_d$  and secondly that the induced distributions  $\mathbb{P} \circ A(Z_n, \epsilon_n)^{-1}$  converge in the weak topology to  $\mathbb{P} \circ A(Z, c)^{-1}$ . The classical Portmanteau Theorem, see Gänssler and Stute [6], Proposition 8.4.9, or Topsøe [21], Theorem 8.1, then gives that (5) is equivalent to

$$\limsup_{n \to \infty} \mathbb{P}\Big(\bigcap_{F \in \mathcal{F}'} \{A(Z_n, \epsilon_n, ) \cap F \neq \emptyset\}\Big) \le \mathbb{P}\Big(\bigcap_{F \in \mathcal{F}'} \{A(Z, c) \cap F \neq \emptyset\}\Big) \text{ for all } \mathcal{F}' \subseteq \mathcal{F}_d.$$
(6)

Our Theorems 1.1 and 1.2 can be viewed as *Continuous Mapping Theorems* for the functional A. They can easily be extended to asymptotic subsets  $C_n$  of  $A(Z_n, \epsilon_n)$ . By this we mean a sequence  $(C_n)$  of  $\mathcal{F}_d$ -valued random elements such that

$$\liminf_{n \to \infty} \mathbb{P}_*(C_n \subseteq A(Z_n, \epsilon_n)) = 1, \tag{7}$$

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which in fact is the same as

$$\lim_{n \to \infty} \mathbb{P}_*(C_n \subseteq A(Z_n, \epsilon_n)) = 1.$$
(8)

For example, if  $C_n \subseteq A(Z_n, \epsilon_n)$  a.s. for eventually all  $n \in \mathbb{N}$ , then the sequence  $(C_n)$  consists of asymptotic subsets.

**Corollary 1.3.** Let the assumptions of Theorem 1.1 be fulfilled. If  $C_n, n \in \mathbb{N}$ , are asymptotic subsets of  $A(Z_n, \epsilon_n)$ , then

$$C_n \twoheadrightarrow^{\sim} A(Z,c) \text{ in } (\mathcal{F}_d, \tau_{uV}).$$
 (9)

If additionally the  $C_n$  are  $\mathcal{A} - \mathcal{B}_{uV}$  measurable and  $Z \in S_1$ , then actually

$$C_n \xrightarrow{\mathcal{D}} A(Z,c) \quad \text{in } (\mathcal{F}_d, \tau_{uV}).$$
 (10)

Measurability of  $C_n$  is guaranteed for instance, if  $C_n$  is a convex and bounded random closed set, see Proposition 2.11 below. (The notion of random closed set will be explained later on). Also, we show that  $C_n$  is  $\mathcal{A} - \mathcal{B}_{uV}$  measurable, if it consists of finitely many random variables in  $\mathbb{R}^d$ , see Lemma 4.6 in the appendix.

Next, we ask for convergence in Borel law if  $\mathcal{F}_d$  is equipped with the Vietoris-topology  $\tau_V$ . Since  $\tau_V$  is finer (stronger) than the upper Vietoris-topology  $\tau_{uV}$  it does not surprise that here additional assumptions are necessary. In short we need that c = 0 and that Z has at most one minimizing point with probability one.

**Theorem 1.4.** Let the assumptions of Theorem 1.1 be fulfilled with c = 0. Further, assume that

$$\operatorname{Argmin}(Z) \subseteq \{\xi\} \text{ a.s. for some random variable } \xi.$$
(11)

Then actually

$$\operatorname{Argmin}(Z) = \{\xi\} \quad \text{a.s.} \tag{12}$$

and

$$A(Z_n, \epsilon_n) \twoheadrightarrow^{\sim} \operatorname{Argmin}(Z) \text{ in } (\mathcal{F}_d, \tau_V).$$
 (13)

If the  $Z_n$  in Theorem 1.4 are level-bounded one might expect that (13) could be sharpened to classical weak convergence. However, for this we needed that the underlying random sets are  $\mathcal{A} - \mathcal{B}_V$  measurable, which is not self-evident and in fact questionable. Therefore we consider the *Fell-topology*  $\tau_F = \tau_F(\mathcal{F}_d)$  on  $\mathcal{F}_d$ , which is generated by the system  $\mathcal{S}_F := \{\mathcal{M}(K), K \in \mathcal{K}_d\} \cup \{\mathcal{H}(G), G \in \mathcal{G}_d\}$ , where  $\mathcal{K}_d$  is the family of all compact sets in  $\mathbb{R}^d$ . Since  $\mathcal{S}_F \subseteq \mathcal{S}_V$  the Fell-topology is coarser than the Vietoris-topology. The hyperspace  $(\mathcal{F}_d, \tau_F)$  is known to be compact, second-countable and Hausdorff and hence it is metrizable. In fact one can specify a metrization  $\delta$ , e. g., the *Painlevé–Kuratowskimetric*, see Pflug [17]. **Theorem 1.5.** Let the assumptions of Theorem 1.2 be fulfilled with c = 0 and assume that (11) holds. Then (12) is true and

$$A(Z_n, \epsilon_n) \xrightarrow{\mathcal{D}} \operatorname{Argmin}(Z) \text{ in } (\mathcal{F}_d, \tau_F).$$
 (14)

In applications, e.g., in statistics or in stochastic optimization, one considers measurable selections  $\xi_n$  of  $A(Z_n, \epsilon_n)$ , that means  $\xi_n : (\Omega, \mathcal{A}) \to (\mathbb{R}^d, \mathcal{B}_d)$  is measurable with  $\xi_n \in A(Z_n, \epsilon_n)$  a.s. Here,  $\mathcal{B}_d = \mathcal{B}(\mathbb{R}^d)$  is the Borel- $\sigma$  algebra on  $\mathbb{R}^d$ .

**Theorem 1.6.** Assume that  $Z_n \in S_0$  for every  $n \in \mathbb{N}$  and  $Z \in S_1$ . For every  $n \in \mathbb{N}$  let  $\xi_n$  be a measurable selection of  $A(Z_n, \epsilon_n)$ . Then  $Z_n \xrightarrow{fd}_D Z$  and  $\epsilon_n \xrightarrow{\mathbb{P}} c$  ensure that

$$\{\xi_n\} \xrightarrow{\mathcal{D}} A(Z,c) \text{ in } (\mathcal{F}_d, \tau_{uV}).$$
 (15)

Moreover, it follows that

$$\limsup_{n \to \infty} \mathbb{P}(\xi_n \in F) \le T(F) \text{ for all closed subsets } F \subseteq \mathbb{R}^d,$$
(16)

where T is the Choquet-capacity functional of the random closed set A(Z, c), that is

$$T(F) = \mathbb{P}(A(Z,c) \cap F \neq \emptyset), \ F \in \mathcal{F}_d.$$
(17)

Here, a random closed set (in  $\mathbb{R}^d$ ) is a measurable map  $C : (\Omega, \mathcal{A}) \to (\mathcal{F}_d, \mathcal{B}_{uF})$ , where  $\mathcal{B}_{uF} = \mathcal{B}_{uF}(\mathcal{F}_d)$  is the Borel- $\sigma$  algebra induced by the upper Fell-topology  $\tau_{uF} =$  $\tau_{uF}(\mathcal{F}_d)$ . This is the topology on  $\mathcal{F}_d$  generated by  $\{\mathcal{M}(K): K \in \mathcal{K}_d\} \subseteq \mathcal{S}_{uV}$ . Thus  $\tau_{uF}$ is coarser than  $\tau_{uV}$ , whence  $\mathcal{B}_{uF} \subseteq \mathcal{B}_{uV}$ . Therefore A(Z, c) is a random closed set in  $\mathbb{R}^d$ , since it is even  $\mathcal{A} - \mathcal{B}_{uV}$  measurable by Theorem 1.2. As any capacity functional our T can be extended to the Borel- $\sigma$  algebra  $\mathcal{B}_d$  such that (17) holds for all Borel-sets  $F \in \mathcal{B}_d$ , see, e.g., Molchanov [16]. So, formally (16) looks exactly like the characterization of weak convergence given in the Portmanteau-Theorem. However,  $T: \mathcal{B}_d \to [0,1]$  in general is not a probability measure, since it lacks additivity. Consequently, we can not deduce weak convergence for the random points  $\xi_n$  at least as long as T is not a probability measure. On the other hand, if c = 0 and  $A(Z, 0) = \operatorname{Argmin}(Z)$  consists of a single random variable  $\xi$ , which means that Z has a **unique minimizer**, then T is equal to the distribution of  $\xi$  and (16) is the same as  $\xi_n \xrightarrow{\mathcal{D}} \xi$ . To sum up, in the unique case we obtain classical weak convergence, whereas in the **non-unique case** the  $\xi_n$  converge weakly to a Choquet-capacity under which we exactly mean (16), see Ferger [5] for a detailed characterization of this generalized concept of weak convergence. A distinction between the two cases is no longer necessary when considering the sets  $\{\xi_n\}$  instead of the single points  $\xi_n$ . In either case we have weak convergence of the singletons  $\{\xi_n\}$ in the hyperspace  $\mathcal{F}_d$  endowed with the upper Vietoris topology. Thus this topology matches perfectly in our framework.

As our short discussion of Theorem 1.6 reveals the special case c = 0 plays a peculiar role. The uniqueness condition occurring there can be slightly weakened:

**Theorem 1.7.** Let Z and  $Z_n, n \in \mathbb{N}$ , be with trajectories in  $S_0$ . Further assume that  $Z \in S_1$  a.s. and that

$$\operatorname{Argmin}(Z) \subseteq \{\xi\} \text{ a.s. for some random variable } \xi.$$
(18)

If  $Z_n \xrightarrow{fd}_D Z$  and  $\epsilon_n \xrightarrow{\mathbb{P}} 0$ , then for every sequence  $(\xi_n)$  of random variables with  $\xi_n \in A(Z_n, \epsilon_n)$  a.s. we can infer that

$$\xi_n \xrightarrow{\mathcal{D}} \xi \quad \text{in } \mathbb{R}^d. \tag{19}$$

**Notice:** If  $Z_n \in S_1$  then  $A(Z_n, \epsilon_n)$  is non-empty and  $\mathcal{A} - \mathcal{B}_{uV}$  measurable by Theorem 1.2. In particular, it is a non-empty random closed set. The Fundamental selection theorem, see Molchanov [16], then guarantees the existence of a measurable selection  $\xi_n$ .

Special cases of Theorem 1.7 include former results of the literature. We start with:

**Corollary 1.8.** (Geyer [7]) Let Z and  $Z_n, n \in \mathbb{N}$ , be with trajectories in  $S_0$ , where Z a.s. possesses the random variable  $\xi$  as its unique minimizing point. Consider non-negative constants  $c_n$  converging to zero and random variables  $\xi_n$  which are the  $c_n$ -approximating minimizers of  $Z_n$ . Then  $Z_n \xrightarrow{fd}_D Z$  implies

$$\xi_n \xrightarrow{\mathcal{D}} \xi \quad \text{in } \mathbb{R}^d.$$
 (20)

This result goes back to Geyer [7]. It is well-known in the statistical literature and has been cited in more than 100 contributions even though the paper of Geyer [7] is an unpublished manuscript. For the special choice  $c_n = 0$  the  $\xi_n$  in (20) are the minimizers of the  $Z_n$ . The great utility of Corollary 1.8 has been demonstrated, e.g., by Chernozhukov [3], Geyer [7], Knight [12], [13], [14] or Wagener and Dette [22] to mention only a few. For example Knight [14] rediscovers Smirnov's [20] four types of all possible limiting distributions for quantile-estimators. Here, it is inevitable that the limit process Z may assume the value infinity. Indeed, stochastic processes taking the value infinity arise canonically in stochastic optimization problems with constraints, see Pflug [17], [18] and Knight [13]. In contrast, Davis, Knight and Liu [4] exclude this profitable case, since they only investigate *real-valued* stochastic processes  $Z_n, Z : \Omega \times \mathbb{R}^d \to \mathbb{R}$  with convex trajectories.

**Corollary 1.9.** (Davis et al. [4]) Let Z and  $Z_n, n \in \mathbb{N}$ , be real-valued and convex stochastic processes and let  $\xi_n$  minimize  $Z_n$  and  $\xi$  minimize Z, where  $\xi$  is unique with probability 1. If

$$(Z_n(t_1), \dots, Z_n(t_k)) \xrightarrow{\mathcal{D}} (Z(t_1), \dots, Z(t_k)) \quad \text{in } \mathbb{R}^k \text{ as } n \to \infty,$$
 (21)

for all  $t_1, \ldots, t_k \in \mathbb{R}^d$ , then

$$\xi_n \xrightarrow{\mathcal{D}} \xi \quad \text{in } \mathbb{R}^d.$$
 (22)

Haberman [8] investigates a very broad class of M-estimators based on convex criterion functions. His proof of asymptotic normality (Theorem 6.1) is rather long, but using the above corollary can make it much less difficult.

A significant simplification of assumption (21) is possible, if the  $Z_n$  allow a second-order expansion.

**Corollary 1.10.** (Hjort and Pollard [9]) Let  $Z_n, n \in \mathbb{N}$ , be real-valued and convex stochastic processes and let  $\xi_n$  minimize  $Z_n$ . Assume there exists a sequence  $(U_n)$  of random vectors with  $U_n \xrightarrow{\mathcal{D}} U$  in  $\mathbb{R}^d$ , and a sequence  $(V_n)$  of matrices with  $V_n \xrightarrow{\mathbb{P}} V$ , where V is positive definite. If  $Z_n$  has the representation

$$Z_n(t) = U'_n t + \frac{1}{2}t' V_n t + r_n(t),$$

where  $r_n(t) \xrightarrow{\mathbb{P}} 0$  for every  $t \in \mathbb{R}^d$ , then

$$\xi_n \xrightarrow{\mathcal{D}} -V^{-1}U \quad \text{in } \mathbb{R}^d.$$
 (23)

The paper of Hjort and Pollard [9] is also an unpublished manuscript, whence we refer to Theorem 7.133 of Liese and Mieschke [15], who present a proof by following the ideas of Hjort and Pollard [9]. Also notice that in contrast to Hjort and Pollard [9] we do not require that the matrices  $V_n$  are positive definite.

The paper is organized as follows: In section 2 we endow the function space S with the *epi-metric* e, which corresponds to *epi-convergence*. This type of convergence is known to be most suitable for minimization problems. According to Attouch [1] the metric space (S, e) is second countable (and compact). If  $\mathcal{B}_e(S)$  denotes the Borel- $\sigma$ algebra induced be e, it turns out that measurability of a map  $Z: (\Omega, \mathcal{A}) \to (S, \mathcal{B}_e(S))$ is exactly the same as being a *normal integrand* in the sense of Rockafellar and Wets [19]. This link to the theory of normal integrands enables us to deduce that every lsc and convex stochastic process (which has an effective domain with non-empty interior) is Borel-measurable. A first fundamental result in section 2 gives conditions under which the map A is  $\tau_{uV}$ -continuous or  $\tau_V$ -continuous, respectively. As a consequence we obtain that for a stochastic process Z with trajectories in  $S_1$  and a non-negative random variable  $\epsilon$  the random set  $A(Z, \epsilon)$  is  $\mathcal{A} - \mathcal{B}_{uV}$ -measurable and in particular this holds for Argmin(Z). Next, for a countable and dense subset  $D = \{t_i : i \in \mathbb{N}\}$  of  $\mathbb{R}^d$  we consider the projection  $\pi_D(f) := (f(t_i) : i \in \mathbb{N}), f \in S_0$ , and show that it is a homeomorphism from  $(S_0, e)$  onto its range equipped with the metric  $\rho$  of coordinatewise convergence. This leads to our second fundamental result in section 2, namely that  $Z_n \xrightarrow{fd} Z$  with  $Z_n$ and Z in  $S_0$  guarantees epi-convergence in distribution, i.e.,  $Z_n \xrightarrow{\mathcal{D}} Z$  in  $(S_0, e)$ . Finally, section 3 contains the proofs of our main theorems, where we just combine the results of section 2 with the Continuous Mapping Theorem. Several technical lemmas, mainly about convex functions, are deferred in the appendix (section 4).

# 2. CONTINUITY OF THE FUNCTIONAL A AND EPI-CONVERGENCE IN DISTRIBUTION

A sequence  $(f_n) \subseteq S$  of lsc functions *epi-converges* to some  $f \in S$   $(f_n \to_{epi} f)$  if at each  $x \in \mathbb{R}^d$  one has

$$\liminf_{n \to \infty} f_n(x_n) \ge f(x) \quad \text{for every sequence } x_n \to x, \tag{24}$$

and

 $\limsup f_n(x_n) \le f(x) \quad \text{for at least one sequence } x_n \to x.$ (25)

Epi-convergence can equivalently be described by convergence of the pertaining *epigraphs* in the hyperspace  $(\mathcal{F}_{d+1}, \tau_F)$ . To see this recall that for a function  $f : \mathbb{R}^d \to \overline{\mathbb{R}}$ , the *epigraph* of f is the set

$$epi(f) := \{ (x, \alpha) \in \mathbb{R}^d \times \mathbb{R} : f(x) \le \alpha \}.$$

The crucial point is that every function f is uniquely determined by its epigraph. Indeed, we have that:

**Lemma 2.1.** If f and g are functions from  $\mathbb{R}^d$  into  $\overline{\mathbb{R}}$  with  $\operatorname{epi}(f) = \operatorname{epi}(g)$ , then f = g. In other words the map with  $\phi(f) := \operatorname{epi}(f)$  is an injection from  $\overline{\mathbb{R}}^{\mathbb{R}^d}$  into the power set of  $\mathbb{R}^d \times \mathbb{R}$ .

A proof is given at the end of the appendix. Another well-known fact says that f is lsc if and only if epi(f) is a closed subset of  $\mathbb{R}^d \times \mathbb{R} \equiv \mathbb{R}^{d+1}$ . Let  $\mathcal{F}_{d+1} = \mathcal{F}(\mathbb{R}^{d+1})$  be equipped with the Fell-topology  $\tau_F = \tau_F(\mathcal{F}_{d+1})$ . The next result follows from Theorem 2.78 and Proposition 1.14 of Attouch [1].

**Theorem 2.2.** (Attouch [1]) For every sequence  $(f_n)_{n \in \mathbb{N}}$  in S the following equivalence holds:

$$f_n \to_{epi} f \quad \Leftrightarrow \quad \operatorname{epi}(f_n) \to \operatorname{epi}(f) \quad \operatorname{in} \ (\mathcal{F}_{d+1}, \tau_F).$$
 (26)

Let  $\mathcal{E} := \{ \operatorname{epi}(f) : f \in S \}$  be the system of all epigraphs of lsc functions from  $\mathbb{R}^d$  into  $\overline{\mathbb{R}}$ . As mentioned above  $\mathcal{E} \subseteq \mathcal{F}_{d+1}$  and from Lemma 2.1 it follows that the map  $\phi : S \to \mathcal{E}$  given by  $\phi(f) := \operatorname{epi}(f), f \in S$ , is a bijection. Attouch [1], p.254-255, proves that  $\mathcal{E}$  is compact for the Fell-topology  $\tau_F(\mathcal{F}_{d+1})$  or in other words that  $(\mathcal{E}, \delta)$  is a compact metric space. Recall that  $\delta$  is a metrization for  $\tau_F(\mathcal{F}_{d+1})$ . Since  $(\mathcal{F}_{d+1}, \delta)$  is second countable and therefore separable, this property applies to the subspace  $(\mathcal{E}, \delta)$  as well.

Define the *epi-metric*  $e: S \times S \to \mathbb{R}$  by  $e(f,g) := \delta(\phi(f), \phi(g))$ . Summing up we obtain from Lemma 2.1 and Theorem 2.2:

**Proposition 2.3.** The epi-metric e is a metric on S such that convergence in (S, e) coincides with epi-convergence, i. e., e is a metrization of epi-convergence. Moreover,  $\phi : (S, e) \to (\mathcal{E}, \delta)$  is a homeomorphism, and in particular (S, e) and  $(\mathcal{E}, \delta)$  are compact and separable metric spaces.

Rockafellar and Wets [19] define a normal integrand (on  $(\Omega, \mathcal{A})$ ) as follows: it is a function-valued map  $Z : (\Omega, \mathcal{A}) \to \mathbb{R}^{\mathbb{R}^d}$  such that  $\phi \circ Z$  is a random closed set in  $\mathbb{R}^{d+1}$ . Especially it follows that  $\operatorname{epi}(Z(\omega)) = \phi(Z(\omega)) = \phi \circ Z(\omega)$  is closed in  $\mathbb{R}^{d+1}$  and thus  $Z(\omega)$  is lsc for every  $\omega \in \Omega$ . In fact, by Proposition 14.28 of Rockafellar and Wets [19] Z is not only lsc but actually a stochastic process (integrand). The other direction needs not to be true: Not every lsc integrand is a normal integrand. However, the following lemma gives a sufficient condition for normality.

**Lemma 2.4.** Assume that Z is a lsc convex stochastic process with  $\operatorname{int}(\operatorname{dom} Z(\omega)) \neq \emptyset$  for all  $\omega \in \Omega$  with dom  $Z(\omega) \neq \emptyset$  (as for instance when  $Z \in S_0$ ). Then Z is a normal integrand.

Proof. This is the second part of Theorem 14.39 of Rockafellar and Wets [19].  $\Box$ 

We shall see that a normal integrand is nothing else but a Borel-measurable map from  $(\Omega, \mathcal{A})$  into the metric space (S, e).

**Lemma 2.5.** Let  $\mathcal{B}_e(S)$  be the Borel- $\sigma$  algebra on (S, e). Then Z is a normal integrand if and only if  $Z : (\Omega, \mathcal{A}) \to (S, \mathcal{B}_e(S))$  is measurable.

Proof. Let  $\tau_{\delta}$  denote the topology on  $\mathcal{F}_{d+1}$  induced by the Painlevé–Kuratowski metric  $\delta$ . We already mentioned above that  $\tau_{\delta}$  coincides with the Fell-topology  $\tau_F$  on  $\mathcal{F}_{d+1}$ , whence the corresponding Borel- $\sigma$  algebras  $\mathcal{B}_{\delta}(\mathcal{F}_{d+1}) := \sigma(\tau_{\delta})$  and  $\mathcal{B}_F(\mathcal{F}_{d+1}) := \sigma(\tau_F)$  coincide as well. Further, recall that  $\mathcal{E}$  is compact in  $(\mathcal{F}_{d+1}, \tau_F)$  and in particular  $\mathcal{E} \in \mathcal{B}_F(\mathcal{F}_{d+1})$ . For the Borel- $\sigma$  algebra  $\mathcal{B}_{\delta}(\mathcal{E})$  on  $(\mathcal{E}, \delta)$  we therefore obtain

$$\mathcal{B}_{\delta}(\mathcal{E}) = \sigma(\mathcal{E} \cap \tau_{\delta}) = \mathcal{E} \cap \sigma(\tau_{\delta}) = \mathcal{E} \cap \mathcal{B}_F(\mathcal{F}_{d+1}) \subseteq \mathcal{B}_F(\mathcal{F}_{d+1}).$$
(27)

Here, the first equality holds by definition and the second one is valid according to Lemma 1.6 in Kallenberg [11]. It is well-known, see, e.g., Molchanov [16], that  $\mathcal{B}_F(\mathcal{F}_{d+1}) = \sigma(\{\mathcal{M}(K) : K \in \mathcal{K}_{d+1}\})$ , whence  $\mathcal{B}_{uF}(\mathcal{F}_{d+1}) = \mathcal{B}_F(\mathcal{F}_{d+1})$  and thus every random closed set C in  $\mathbb{R}^{d+1}$  can alternatively be conceived as a measurable map C:  $(\Omega, \mathcal{A}) \to (\mathcal{F}_{d+1}, \mathcal{B}_F(\mathcal{F}_{d+1})).$ 

Now suppose that Z is a normal integrand. By definition and our last conclusion this means that  $\phi \circ Z : (\Omega, \mathcal{A}) \to (\mathcal{F}_{d+1}, \mathcal{B}_F(\mathcal{F}_{d+1}))$  is measurable. But  $\phi$  maps into  $\mathcal{E}$ , whence it follows from (27) that  $\phi \circ Z : (\Omega, \mathcal{A}) \to (\mathcal{E}, \mathcal{B}_{\delta}(\mathcal{E}))$  is measurable. By Proposition 2.3  $\phi^{-1}$  is a continuous map and therefore it is  $\mathcal{B}_{\delta}(\mathcal{E}) - \mathcal{B}_e(S)$  measurable. Since  $Z = \phi^{-1} \circ (\phi \circ Z)$  we can infer that Z is  $\mathcal{A} - \mathcal{B}_e(S)$  measurable as a composition of measurable maps.

For the other direction notice that by Proposition 2.3  $\phi$  is continuous and hence

$$\phi$$
 is  $\mathcal{B}_e(S) - \mathcal{B}_\delta(\mathcal{E})$  measurable. (28)

Let  $\mathbf{B} \in \mathcal{B}_F(\mathcal{F}_{d+1})$  be an arbitrary Borel-set. It has the inverse image  $\phi^{-1}(\mathbf{B}) = \phi^{-1}(\mathcal{E} \cap \mathbf{B})$ , where  $\mathcal{E} \cap \mathbf{B} \in \mathcal{B}_{\delta}(\mathcal{E})$  by the equalities in (27). Thus  $\phi^{-1}(\mathbf{B}) \in \phi^{-1}(\mathcal{B}_{\delta}(\mathcal{E})) \subseteq \mathcal{B}_e(S)$  by (28), whence  $(\phi \circ Z)^{-1}(\mathbf{B}) = Z^{-1}(\phi^{-1}(\mathbf{B})) \in \mathcal{A}$  for Z is  $\mathcal{A} - \mathcal{B}_e(S)$  measurable by assumption. This shows that  $\phi \circ Z$  is a random closed set and hereby Z is a normal integrand.

**Corollary 2.6.** Fix a subspace U of (S, e) and assume that Z is a normal integrand on  $(\Omega, \mathcal{A})$  with trajectories in U. Then  $Z : (\Omega, \mathcal{A}) \to (U, \mathcal{B}_e(U))$  is measurable.

Proof. Let  $B \in \mathcal{B}_e(U)$ . Since  $\mathcal{B}_e(U) = U \cap \mathcal{B}_e(S)$  by Lemma 1.6 in Kallenberg [11] it follows that  $B = U \cap \tilde{B}$  for some  $\tilde{B} \in \mathcal{B}_e(S)$ . We thus can infer that

$$Z^{-1}(B) = Z^{-1}(U) \cap Z^{-1}(\tilde{B}) = \Omega \cap Z^{-1}(\tilde{B}) = Z^{-1}(\tilde{B}) \in \mathcal{A}$$

by Lemma 2.5.

Notice: If Z is a stochastic process with trajectories in  $S_0$ , then it is a normal integrand by Lemma 2.4, whence by Corollary 2.6 it is  $\mathcal{A} - \mathcal{B}_e(S_0)$  measurable, which in turn is equivalent to  $\mathcal{A} - \mathcal{B}_e(S)$ -measurability. Therefore, given a sequence  $(Z_n)$  of stochastic processes with values in  $S_0$ , the measurability requirement in the definition of distributional convergence  $Z_n \xrightarrow{\mathcal{D}} Z$  in  $(S_0, e)$  and in (S, e) is fulfilled.

The following lemma gives an equivalent description for convergence in the Vietoristopology  $\tau_V$  and in the upper-Vietoris topology  $\tau_{uV}$ .

**Lemma 2.7.** Let F and  $F_n, n \in \mathbb{N}$ , be closed subsets of  $\mathbb{R}^d$ .

(1) The following statements (a) and (b) are equivalent:

- (a)  $F_n \to F$  in  $(\mathcal{F}_d, \tau_V)$ .
- (b) The miss-criterion (b1) and the hit-criterion (b2) are satisfied, where
  - (b1) For every  $H \in \mathcal{F}_d$  with  $F \cap H = \emptyset$  there exists a natural number  $n_0$  such that  $F_n \cap H = \emptyset$  for all  $n \ge n_0$ ,
  - (b2) For every  $G \in \mathcal{G}_d$  with  $F \cap G \neq \emptyset$  there exists a natural number  $n_1$  such that  $F_n \cap G \neq \emptyset$  for all  $n \ge n_1$ .
- (2)  $F_n \to F$  in  $(\mathcal{F}_d, \tau_{uV})$  if and only if the miss-criterion (b1) holds.

Proof. Both equivalences follow immediately from the definitions of the respective topologies upon noticing that for checking convergence it suffices to consider subbase-neighbourhoods.  $\hfill\square$ 

With the help of Lemma 2.7 we prove continuity of the map A. This plays a fundamental role in our paper. Here we deal with the superset  $U_0 := \{f \in S : f \text{ convex and proper}\} \supseteq S_0$ .

**Theorem 2.8.** Let u be the usual metric on  $\mathbb{R}_+$  and  $e \times u$  be the product-metric on  $S \times \mathbb{R}_+$ . Then:

(1)  $A: (U_0 \times \mathbb{R}_+, e \times u) \to (\mathcal{F}_d, \tau_{uV})$  is continuous on  $U \times \mathbb{R}_+$ , where

 $U = \{ f \in S : f \text{ convex, proper and level-bounded} \}.$ 

(2)  $A: (U_0 \times \mathbb{R}_+, e \times u) \to (\mathcal{F}_d, \tau_V)$  is continuous on  $U^* \times \{0\}$ , where

 $U^* = \{f \in S : f \text{ convex, proper with unique minimizer}\} \subseteq U.$ 

Proof. (1) Let  $(f,r) \in U \times \mathbb{R}_+$  and  $(f_n, r_n)_{n \in \mathbb{N}}$  be a sequence in  $U_0 \times \mathbb{R}_+$  with  $(f_n, r_n) \to_{e \times d} (f, r)$ . Convergence by components and Proposition 2.3 yield that  $f_n \to_{epi} f$  and  $r_n \to r$ . By Exercise 7.32(c) in Rockafellar and Wets [19] the sequence  $(f_n)_{n \in \mathbb{N}}$  is eventually level-bounded, that means there exists some  $n_0 \in \mathbb{N}$  such that for every  $n \geq n_0$  we have that:

$$\forall \ \alpha \in \mathbb{R} \ \exists \ K = K_{\alpha} \in \mathcal{K}_d \text{ such that } \{f_n \le \alpha\} \subseteq K.$$
<sup>(29)</sup>

Notice that  $K = K_{\alpha}$  does not depend on n. Lemma 4.3 in the appendix ensures that

$$A(f_n, r_n) \neq \emptyset \quad \text{for all } n \ge n_0. \tag{30}$$

Furthermore Theorem 7.33 in Rockafellar and Wets [19] guarantees the convergence

$$\inf_{t \in \mathbb{R}^d} f_n(t) \to \inf_{t \in \mathbb{R}^d} f(t) \in \mathbb{R}.$$
(31)

Let  $t \in A(f_n, r_n)$ . Then  $f_n(t) \leq \inf_{t \in \mathbb{R}^d} f_n(t) + r_n$  and by (31) there exists an integer  $n_1 \in \mathbb{N}$  such that  $\inf_{t \in \mathbb{R}^d} f_n(t) \leq \inf_{t \in \mathbb{R}^d} f(t) + 1$  for all  $n \geq n_1$ . Moreover, since  $r_n \to r$  we have that  $r_n \leq r+1$  for all  $n \geq n_2$  for some  $n_2 \in \mathbb{N}$ . Thus  $f_n(t) \leq \inf_{t \in \mathbb{R}^d} f(t) + 2 + r =: \alpha \in \mathbb{R}$  for all  $n \geq n_3 := n_1 \lor n_2 \in \mathbb{N}$ . Conclude that  $A(f_n, r_n) \subseteq \{f_n \leq \alpha\}$  for all  $n \geq n_3$ . With K and  $n_0$  as in (29) plus  $n_4 := n_0 \lor n_3 \in \mathbb{N}$  we obtain that

$$A(f_n, r_n) \subseteq K \quad \forall \ n \ge n_4. \tag{32}$$

In order to verify the miss-criterion (b1) of Lemma 2.7 let us consider an arbitrary closed set  $H \in \mathcal{F}_d$  with  $A(f, r) \cap H = \emptyset$ . Then a fortiori

$$A(f,r) \cap H \cap K = \emptyset. \tag{33}$$

We shall show that

$$A(f_n, r_n) \cap H \cap K = \emptyset \quad \forall \ n \ge n_5 \text{ for some } n_5 \in \mathbb{N}.$$
(34)

Assume that (34) is not true, i. e., there exists a subsequence  $(n_j)_{j\in\mathbb{N}}\subseteq\mathbb{N}$  such that  $A(f_{n_j}, r_{n_j})\cap H\cap K\neq\emptyset$  for all  $j\in\mathbb{N}$ . Then one can find a sequence  $(x_{n_j})_{j\in\mathbb{N}}\subseteq H\cap K$  such that  $x_{n_j}\in A(f_{n_j}, r_{n_j})$  for each  $j\in\mathbb{N}$ . Since  $H\cap K$  is compact, the sequence  $(x_{n_j})_{j\in\mathbb{N}}$  has a subsequence  $(x_{n_{j_l}})_{l\in\mathbb{N}}$  with  $x_{n_{j_l}}\to x\in H\cap K$  as  $l\to\infty$ . For notational convenience we take  $x_{n_j}\to x, j\to\infty$  for granted. It follows that

$$x \in A(f, r). \tag{35}$$

Indeed, assume that  $x \notin A(f, r)$ , i.e.,  $f(x) > \inf_{t \in \mathbb{R}^d} f(t) + r$ , whence

$$f(x) > f(y) + r$$
 for some  $y \in \mathbb{R}^d$ . (36)

Recall that  $f_n \to_{epi} f$ . Thus by (24) and (25) there exists a sequence  $(y_n)$  with  $y_n \to y$  such that  $f(y) = \lim_{n\to\infty} f_n(y_n)$ . Conclude from (36) that

$$f(x) > \lim_{n \to \infty} f_n(y_n) + r = \liminf_{j \to \infty} f_{n_j}(y_{n_j}) + r.$$
(37)

Now  $x_{n_j} \in A(f_{n_j}, r_{n_j})$  entails  $f_{n_j}(y_{n_j}) \ge f_{n_j}(x_{n_j}) - r_{n_j}$  and so

$$\liminf_{j \to \infty} f_{n_j}(y_{n_j}) \ge \liminf_{j \to \infty} f_{n_j}(x_{n_j}) - r \ge f(x) - r, \tag{38}$$

where the last inequality holds by (24), because  $x_{n_j} \to x$  and by Proposition 2.3 the subsequence  $(f_{n_j})_{j\in\mathbb{N}}$  epi-converges to f as well. Combining (37) and (38) leads to f(x) > (f(x) - r) + r = f(x), a contradiction. Thus relation (35) is true and since  $x \in H \cap K$  we arrive at  $A(f,r) \cap H \cap K \neq \emptyset$ , which is a contradiction to (33). This shows (34).

For  $n_6 := n_4 \lor n_5 \in \mathbb{N}$  observe that  $A(f_n, r_n) = A(f_n, r_n) \cap K$  by (32) for all  $n \ge n_6$ , whence (34) yields:

$$A(f_n, r_n) \cap H = A(f_n, r_n) \cap H \cap K = \emptyset \quad \forall \ n \ge n_6,$$

whence the miss-criterion (b1) in Lemma 2.7(2) is fulfilled and therefore  $A(f_n, r_n) \rightarrow A(f, r)$  in  $(\mathcal{F}_d, \tau_{uV})$ . This shows continuity of A at every point  $(f, r) \in U \times \mathbb{R}_+$ .

(2) Let  $(f_n, r_n) \to_{e \times d} (f, 0)$  with  $f \in U^*$ . It follows from Lemma 4.2 in the appendix that f is level-bounded and thus  $U^* \subseteq U$ . In particular, the missing-criterion (b1) is fulfilled by (1) above. Therefore it remains to show the hit-criterion (b2) in Lemma 2.7. For that purpose let  $G \in \mathcal{G}_d$  with  $A(f, 0) \cap G \neq \emptyset$ . We have to show that

$$A(f_n, r_n) \cap G \neq \emptyset \quad \text{for eventually all } n \in \mathbb{N}.$$
(39)

Assume that (39) does not hold, i. e., there exists a subsequence  $(n_j)_{j \in \mathbb{N}}$  of the natural numbers such that  $A(f_{n_j}, r_{n_j}) \cap G = \emptyset$  or equivalently  $A(f_{n_j}, r_{n_j}) \subseteq G^c$  for every  $j \in \mathbb{N}$ , where  $G^c := E \setminus G$  denotes the complement of G in E. From (32) we can deduce that there exist a compact set K and a  $j_0 \in \mathbb{N}$  such that  $A(f_{n_j}, r_{n_j}) \subseteq K$  for all  $j \geq j_0$  and consequently  $A(f_{n_j}, r_{n_j}) \subseteq G^c \cap K$  for all  $j \geq j_0$ . By (30) there exists a  $j_1 \in \mathbb{N}$  such that  $A(f_{n_j}, r_{n_j}) \neq \emptyset$  for all  $j \geq j_1$ . Put  $j_2 = j_0 \lor j_1 \in \mathbb{N}$ . Then for every  $j \geq j_2$  there exists some  $z_{n_j} \in A(f_{n_j}, r_{n_j}) \subseteq G^c \cap K$ . Since G is open  $G^c \cap K$  is compact, whence w.l.o.g. we may assume that  $z_{n_j} \to z \in G^c \cap K$  as  $j \to \infty$ . Now,  $z_{n_j} \in A(f_{n_j}, r_{n_j})$  means that  $f_{n_j}(z_{n_j}) \leq \inf_{s \in \mathbb{R}^d} f_{n_j}(s) + r_{n_j}$  for all  $j \geq j_2$ . From  $f_{n_j} \to_{epi} f$  it follows with (24) that

$$f(z) \le \liminf_{j \to \infty} f_{n_j}(z_{n_j}) \le \liminf_{j \to \infty} \inf_{s \in \mathbb{R}^d} f_{n_j}(s) + \liminf_{j \to \infty} r_{n_j} = \inf_{s \in \mathbb{R}^d} f(s),$$

where the last equality holds by (31) and  $r_n \to 0$ . Conclude that  $z \in A(f, 0)$ , where by  $f \in U^*$  the argmin-set  $A(f, 0) = \operatorname{Argmin}(f)$  is a singleton. Hence  $A(f, 0) = \{z\}$ . However, recall that  $A(f, 0) \cap G \neq \emptyset$ , which results in  $z \in G$  in contradiction to  $z \in G^c \cap K$ .

**Proposition 2.9.** Let Z be a stochastic process with trajectories in  $S_1$  and let  $\epsilon$  be a  $\mathbb{R}_+$ -valued random variable both defined on  $(\Omega, \mathcal{A})$ . Then  $A(Z, \epsilon) = A \circ (Z, \epsilon)$  is a  $\mathcal{A} - \mathcal{B}_{uV}$  measurable map from  $\Omega$  into  $\mathcal{F}_d$ .

Proof. By Lemma 2.4 Z is a normal integrand and therefore by Corollary 2.6 it is a  $\mathcal{A} - \mathcal{B}_e(S_1)$  measurable map from  $\Omega$  into  $S_1$ . Thus  $(Z, \epsilon) : (\Omega, \mathcal{A}) \to (S_1 \times \mathbb{R}_+, \mathcal{B}_e(S_1) \otimes \mathcal{B}_u(\mathbb{R}_+))$  is measurable. It follows from Proposition 2.3 that the subspace  $(S_1, e)$  is separable, and clearly  $(\mathbb{R}_+, u)$  is also separable. Consequently

$$\mathcal{B}_e(S_1) \otimes \mathcal{B}_u(\mathbb{R}_+) = \mathcal{B}_{e \times u}(S_1 \times \mathbb{R}_+).$$
(40)

By  $S_1 \subseteq U$  Theorem 2.8 ensures that the restriction  $A : (S_1 \times \mathbb{R}_+, e \times u) \to (\mathcal{F}_d, \tau_{uV})$ is continuous and consequently  $\mathcal{B}_{e \times u}(S_1 \times \mathbb{R}_+) - \mathcal{B}_{uV}$  measurable. The assertion now follows from (40), which shows that  $A \circ (Z, \epsilon)$  is a composition of measurable maps.

In view of  $\operatorname{Argmin}(Z) = A(Z, 0)$  we immediately obtain

**Corollary 2.10.** If Z is a stochastic process with trajectories in  $S_1$ , then  $\operatorname{Argmin}(Z)$  is  $\mathcal{A} - \mathcal{B}_{uV}$  measurable.

Next, we seek conditions under which a random closed set  $C : \Omega \to \mathcal{F}_d$  is actually  $\mathcal{A} - \mathcal{B}_{uV}$  measurable. An answer is given in

**Proposition 2.11.** If the random closed set  $C : \Omega \to \mathcal{F}_d$  is convex and bounded with  $int(C) \neq \emptyset$ , then it is  $\mathcal{A} - \mathcal{B}_{uV}$  measurable.

Proof. Consider the special indicator function

$$Z(\omega, t) := \delta_{C(\omega)}(t) := \begin{cases} 0, & t \in C(\omega) \\ \infty, & t \neq C(\omega) \end{cases}$$

Observe that for each fixed  $t \in \mathbb{R}^d$  and every  $\alpha \in \mathbb{R}$  the set  $\{\omega \in \Omega : Z(\omega, t) \leq \alpha\}$ is equal to  $\{\omega \in \Omega : t \in C(\omega)\}$ , if  $\alpha \geq 0$  and it is equal to  $\emptyset$ , if  $\alpha < 0$ . Recall that  $\mathcal{B}_F := \mathcal{B}_F(\mathcal{F}_d) = \sigma(\{\mathcal{M}(K) : K \in \mathcal{K}_d\})$ . Therefore  $\{\omega \in \Omega : t \in C(\omega)\} = \{\omega \in \Omega : C(\omega) \cap \{t\} \neq \emptyset\} \in \mathcal{A}$ , since  $\{t\} \in \mathcal{K}_d$ . This shows that Z is an integrand (stochastic process). Similarly, one sees that for each fixed  $\omega \in \Omega$  the level-set  $\{t \in \mathbb{R}^d : Z(\omega, t) \leq \alpha\}$ is equal to  $C(\omega)$  or  $\emptyset$  according as  $\alpha \geq 0$  or  $\alpha < 0$ . Consequently Z is level-bounded, because C is bounded by assumption. Furthermore,  $\operatorname{epi}(Z(\omega)) = C(\omega) \times [0, \infty)$  is a closed subset of  $\mathbb{R}^d \times \mathbb{R}$ , whence Z is lsc. It is easy to check that Z is also convex and proper with dom Z = C. To sum up, Z is a an integrand with trajectories in  $S_1$ . Thus Corollary 2.10 yields the assumption upon noticing that  $\operatorname{Argmin}(Z) = C$ .

Let  $D = \{t_i : i \in \mathbb{N}\}$  be a countable and dense subset of  $\mathbb{R}^d$ . We define the projectionmap  $\pi_D : S_0 \to \overline{\mathbb{R}}^\infty$  by  $\pi_D(f) := (f(t_i) : i \in \mathbb{N}), f \in S_0$ . Let  $\varrho$  be the metric of coordinatewise convergence on  $\overline{\mathbb{R}}^\infty$  or in other words  $\varrho$  is the product-metric pertaining to the metric on  $\overline{\mathbb{R}}$ . Further let  $R := \pi_D(S_0) \subseteq \overline{\mathbb{R}}^\infty$  be the range of  $\pi_D$ . We obtain:

**Theorem 2.12.** For every countable and dense subset D the corresponding projectionmap  $\pi_D : (S_0, e) \to (R, \varrho)$  is bijective and its inverse  $\pi_D^{-1} : (R, \varrho) \to (S_0, e)$  is continuous. Proof. Write  $\pi = \pi_D$  for short. For the first assertion it suffices to show that  $\pi$  is injective. So, assume  $\pi(f) = \pi(g)$ , that is f(s) = g(s) for all  $s \in D$ . We firstly show that

$$int(dom f) = int(dom g).$$
 (41)

For that purpose consider  $t \in \operatorname{int}(\operatorname{dom} f)$ . Since D lies dense in  $\mathbb{R}^d$  there exists a sequence  $(s_m)_{m\in\mathbb{N}}$  in D such that  $s_m \to t$ . Observe that  $\operatorname{int}(\operatorname{dom} f)$  is an open neighborhood of t, whence there exists a  $m_0 \in \mathbb{N}$  such that  $s_m \in \operatorname{int}(\operatorname{dom} f)$  for all  $m \ge m_0$ . Now,  $f \in S_0$  implies that f is finite on dom  $f \neq \emptyset$ , which is convex. Recall that the nonempty interior of a convex set in  $\mathbb{R}^d$  is convex as well, see Theorem 2.33 of Rockafellar and Wets [19]. Thus f is a finite convex function on the open and convex set  $O := \operatorname{int}(\operatorname{dom} f)$ . Since by assumption  $O \neq \emptyset$  Corollary 2.36 in Rockafellar and Wets [19] says that f is continuous on O. This makes us to infer that

$$\infty > f(t) = \lim_{m \to \infty} f(s_m) = \lim_{m \to \infty} g(s_m) = \liminf_{m \to \infty} g(s_m) \ge g(t),$$

where the last inequality holds because g is lsc. Conclude that  $g(t) < \infty$ , whence  $t \in \text{dom } g$ . This shows that  $\text{int}(\text{dom } f) \subseteq \text{dom } g$ , which in turn gives  $\text{int}(\text{dom } f) \subseteq \text{int}(\text{dom } g)$  for int(dom g) is the largest open set contained in dom g. Using the same arguments with f and g reversing their roles yields  $\text{int}(\text{dom } g) \subseteq \text{int}(\text{dom } f)$  and thus the equality (41).

For every  $t \in int(\text{dom } f)$  as above we obtain that

$$f(t) = \lim_{m \to \infty} f(s_m) = \lim_{m \to \infty} g(s_m) = g(t),$$

because f and g are continuous on  $O := \operatorname{int}(\operatorname{dom} f) = \operatorname{int}(\operatorname{dom} g)$ . This means that f and g coincide on O, which as nonempty set agrees with the relative interiors  $\operatorname{rint}(\operatorname{dom} f)$  and  $\operatorname{rint}(\operatorname{dom} g)$ . Thus Exercise 2.46(a) in Rockafellar and Wets [19] guarantees that f = g upon noticing that f and g are lsc. Consequently,  $\pi$  is injective.

For proving continuity of the inverse  $\pi^{-1}$  let  $(y_n)$  be a sequence in the range R with

$$y_n \to_{\varrho} y \in R$$
, that is  $\varrho(y_n, y) \to 0.$  (42)

Observe that  $y_n = \pi(f_n) = (f_n(t_i) : i \in \mathbb{N})$  and  $y = \pi(f) = (f(t_i) : i \in \mathbb{N})$  with  $f_n = \pi^{-1}(y_n)$  and  $f = \pi^{-1}(y)$  by the first part. Then by definition of  $\rho$  the convergence (42) means that

$$f_n(t_i) \to f(t_i)$$
 for all  $i \in \mathbb{N}$ .

Since  $D = \{t_i : i \in \mathbb{N}\}$  is a dense subset of  $\mathbb{R}^d$ , Theorem 7.17 of Rockafellar and Wets [19] yields that  $f_n \to_{epi} f$ , which by Proposition 2.3 is equivalent to  $f_n \to_e f$  and thus  $\pi^{-1}(y_n) \to_e \pi^{-1}(y)$ . This shows continuity of the inverse.  $\Box$ 

With our last theorem we can prove that for lsc and convex stochastic processes convergence of the finite dimensional distributions entails *epi-convergence in distribution*. More precisely we have

**Proposition 2.13.** Fix some countable and dense subset  $D = \{t_1, t_2, \ldots\}$  of  $\mathbb{R}^d$ . Let Z and  $Z_n, n \in \mathbb{N}$ , be integrands with trajectories in  $S_0$ .

If  $Z_n \xrightarrow{fd}_D Z$  then

$$Z_n \xrightarrow{\mathcal{D}} Z \quad \text{in} (S_0, e)$$

$$\tag{43}$$

and

$$Z_n \xrightarrow{\mathcal{D}} Z \quad \text{in } (S, e).$$
 (44)

Proof. Again let  $\pi = \pi_D$ . Since  $\overline{\mathbb{R}}$  is separable, the assumption  $Z_n \xrightarrow{fd}_D Z$  in combination with Theorem 3.29 in Kallenberg [11] yields that  $\pi(Z_n) \xrightarrow{\mathcal{D}} \pi(Z)$  in  $(\overline{\mathbb{R}}^{\infty}, \varrho)$ . By the Subspace-Lemma 3.26 in Kallenberg [11] this is equivalent to  $\pi(Z_n) \xrightarrow{\mathcal{D}} \pi(Z)$  in  $(R, \varrho)$ . By Theorem 2.12 the inverse  $\pi^{-1} : (R, \varrho) \to (S_0, e)$  is continuous, whence the Continuous Mapping Theorem ensures (43), because  $Z_n = \pi^{-1}(\pi(Z_n))$ . Another application of the Subspace-Lemma gives (44).

#### 3. PROOFS

In this section we prove our results in section 1. With the preparations made in section 2 the proofs reduce to a few lines.

Proof. (of Theorem 1.1) By Proposition 2.13 we have that  $Z_n \xrightarrow{\mathcal{D}} Z$  in (S, e). Since (S, e) and  $(\mathbb{R}_+, u)$  are separable Slutsky's theorem yields  $(Z_n, \epsilon_n) \xrightarrow{\mathcal{D}} (Z, c)$  in  $(S \times \mathbb{R}_+, e \times u)$ , which in particular entails  $(Z_n, \epsilon_n) \xrightarrow{\to} (Z, c)$  in  $(S \times \mathbb{R}_+, e \times u)$ . Theorem 2.8 (1) says that A is  $\tau_{uV}$ -continuous on  $U \times \mathbb{R}_+ \supseteq S_1 \times \mathbb{R}_+$ , whence the set of discontinuity-points  $D_A := \{(f, r) \in S \times \mathbb{R}_+ : A \text{ is not } \tau_{uV}$ -continuous at  $(f, r)\}$  of A is contained in  $(S \setminus S_1) \times \mathbb{R}_+$ . Consequently,  $\mathbb{P}_*((Z, c) \in D_A) \leq \mathbb{P}_*(Z \notin S_1) = \mathbb{P}(Z \notin S_1) = 0$  and the Continuous Mapping Theorem for  $\twoheadrightarrow$ , see Lemma 4.5 yields the desired result (3). The second part follows from Lemma 4.3.

Proof. (of Theorem 1.2) By Proposition 2.9 the random sets  $A(Z_n, \epsilon_n), n \in \mathbb{N}$ , and A(Z, c) are  $\mathcal{A} - \mathcal{B}_{uV}$  measurable. Therefore

$$\bigcap_{F \in \mathcal{F}'} \{ A(Z_n, \epsilon_n) \cap F \neq \emptyset \} = \{ A(Z_n, \epsilon_n) \in \bigcap_{F \in \mathcal{F}'} \mathcal{H}(F) \} \in \mathcal{A}$$

and

ł

$$\bigcap_{F \in \mathcal{F}'} \{ A(Z,c) \cap F \neq \emptyset \} = \{ A(Z,c) \in \bigcap_{F \in \mathcal{F}'} \mathcal{H}(F) \} \in \mathcal{A},$$

because  $\bigcap_{F \in \mathcal{F}'} \mathcal{H}(F)$  is  $\tau_{uV}$ -closed and in particular a Borel-set in  $\mathcal{B}_{uV}$ . Thus (5) follows from Theorem 1.1, because (4) reduces to (6), since  $\mathbb{P}^* = \mathbb{P} = \mathbb{P}_*$  on  $\mathcal{A}$ . Finally, by the Portmanteau-Theorem (6) is equivalent to (5). Again the second part is a consequence of Lemma 4.3

**Proof.** (of Corollary 1.3) First observe that by complementation the sequence  $(C_n)$  satisfies

Argmin-sets of convex stochastic processes

$$\limsup_{n \to \infty} \mathbb{P}^*(C_n \not\subseteq A(Z_n, \epsilon_n)) = 0.$$
(45)

Now, since  $\{C_n \cap F \neq \emptyset\} \cap \{C_n \subseteq A(Z_n, \epsilon_n)\} \subseteq \{A(Z_n, \epsilon_n) \cap F \neq \emptyset\}$ , a decomposition of the set  $\bigcap_{F \in \mathcal{F}'} \{C_n \cap F \neq \emptyset\}$  results in

$$\limsup_{n \to \infty} \mathbb{P}^* \Big( \bigcap_{F \in \mathcal{F}'} \{ C_n \cap F \neq \emptyset \} \Big)$$
  
$$\leq \limsup_{n \to \infty} \mathbb{P}^* \Big( \bigcap_{F \in \mathcal{F}'} \{ A(Z_n, \epsilon_n) \cap F \neq \emptyset \} \Big) + \limsup_{n \to \infty} \mathbb{P}^* (C_n \not\subseteq A(Z_n, \epsilon_n) \Big)$$

Here, by (45) the second summand vanishes and by Theorem 1.1 the first summand can be estimated as in (4). Thus we obtain

$$\limsup_{n \to \infty} \mathbb{P}^* \Big( \bigcap_{F \in \mathcal{F}'} \{ C_n \cap F \neq \emptyset \} \Big) \le \mathbb{P}_* \Big( \bigcap_{F \in \mathcal{F}'} \{ A(Z, c) \cap F \neq \emptyset \} \Big) \quad \text{for all } \mathcal{F}' \subseteq \mathcal{F}_d,$$
(46)

which by the Borel law Portmanteau Theorem gives the assertion (9). In case of measurable  $C_n$ 's we can argue analogously as in the above proof to conclude that (46) holds without the asteriks \*, which by the Portmanteau-Theorem results in (10).

Proof. (of Theorem 1.4) First notice that by Theorem 1.1 Argmin(Z) = A(Z, 0) is a.s. non-empty, whence Argmin $(Z) = \{\xi\}$  a.s. by (11).

From the proof of Theorem 1.1 we know that  $(Z_n, \epsilon_n) \to (Z, 0)$  in  $(S \times \mathbb{R}_+, e \times u)$ . Theorem 2.8 (2) yields that  $D_A := \{(f, r) \in S \times \mathbb{R}_+ : A \text{ is not } \tau_V \text{-continuous at } (f, r)\} \subseteq (S \times \mathbb{R}_+) \setminus (U^* \times \{0\}) = ((S \setminus U^*) \times \mathbb{R}_+) \cup (S \times (\mathbb{R}_+ \setminus \{0\}))$ . Thus it follows that  $\mathbb{P}_*((Z, 0) \in D_A) \leq \mathbb{P}_*(Z \notin U^*) = 0$  by (12) and so the CMT (Lemma 4.5) gives (13).  $\Box$ 

Proof. (of Theorem 1.5) The first part (12) follows from Theorem 1.4. From the proof of Theorem 1.1 we know that  $(Z_n, \epsilon_n) \xrightarrow{\mathcal{D}} (Z, 0)$  in  $(S \times \mathbb{R}_+, e \times u)$ , whence by the subspace-lemma

$$(Z_n, \epsilon_n) \xrightarrow{\mathcal{D}} (Z, 0) \text{ in } (S_1 \times \mathbb{R}_+, e \times u).$$
 (47)

Since  $S_1 \subseteq U$  and  $\tau_{uF} \subseteq \tau_{uV}$  it follows from Theorem 2.8(1) that  $A : (S_1 \times \mathbb{R}_+, e \times u) \to (\mathcal{F}_d, \tau_{uF})$  is continuous and herewith A is  $\mathcal{B}_{e \times u}(S_1 \times \mathbb{R}_+) - \mathcal{B}_{uF}$  measurable. Recall that  $\mathcal{B}_{uF} = \mathcal{B}_F$ , see the proof of Proposition 2.5. Therefore  $A : (S_1 \times \mathbb{R}_+, e \times u) \to (\mathcal{F}_d, \tau_F)$  is Borel-measurable. From  $\tau_F \subseteq \tau_V$  and Theorem 2.8(2) we can infer that A is  $\tau_F$ -continuous on  $U^* \times \{0\}$ . Thus the assertion (14) follows from (47) and the CMT.  $\Box$ 

Proof. (of Theorem 1.6)  $C_n := \{\xi_n\}$  is  $\mathcal{A} - \mathcal{B}_{uV}$  measurable by Lemma 4.6 in the appendix and so Corollary 1.3 yields the distributional convergence (15) of the singletons. By the Portmanteau-Theorem this is equivalent to

$$\limsup_{n \to \infty} \mathbb{P}\Big(\bigcap_{F \in \mathcal{F}'} \{\{\xi_n\} \cap F \neq \emptyset\}\Big) \le \mathbb{P}\Big(\bigcap_{F \in \mathcal{F}'} \{A(Z, c) \cap F \neq \emptyset\}\Big) \text{ for all } \mathcal{F}' \subseteq \mathcal{F}_d,$$

which with  $\mathcal{F}' = \{F\}$  simplifies to (16) since  $\{\{\xi_n\} \cap F \neq \emptyset\} = \{\xi_n \in F\}.$ 

Proof. (of Theorem 1.7) By Theorem 1.4 Argmin $(Z) = \{\xi\}$  a.s. From Corollary 1.3 with  $C_n := \{\xi_n\}$  and c = 0 we know that  $\{\xi_n\} \twoheadrightarrow^{\sim} A(Z, 0) = \operatorname{Argmin}(Z)$  in  $(\mathcal{F}_d, \tau_{uV})$ . Use (4) with  $\mathcal{F}' := \{F\}$  to infer that

$$\limsup_{n \to \infty} \mathbb{P}^*(\xi_n \in F) \le \mathbb{P}_*(\operatorname{Argmin}(Z) \cap F \neq \emptyset) = \mathbb{P}_*(\{\xi\} \cap F \neq \emptyset) = \mathbb{P}_*(\xi \in F) \ \forall F \in \mathcal{F}.$$

Since the  $\xi'_n$ 's and  $\xi$  are random variables it follows that  $\{\xi_n \in F\} \in \mathcal{A}$  for every  $n \in \mathbb{N}$  and  $\{\xi \in F\} \in \mathcal{A}$  as well. Consequently, the above inequalities hold without the asterisks leftmost and rightmost and therefore

$$\limsup_{n \to \infty} \mathbb{P}(\xi_n \in F) \le \mathbb{P}(\xi \in F) \; \forall F \in \mathcal{F},$$

which by the Portmanteau-Theorem gives  $\xi_n \xrightarrow{\mathcal{D}} \xi$  in  $\mathbb{R}^d$ .

Proof. (of Corollary 1.8) By assumption  $\operatorname{Argmin}(Z) = \{\xi\}$  a.s. and thus  $Z \in S_1$  a.s. according to Lemma 4.2 and in particular (18) is fulfilled. Then the assertion follows from Theorem 1.7 with  $\epsilon_n := c_n$ .

Proof. (of Corollary 1.9) First notice that every convex and real-valued function  $f: \mathbb{R}^d \to \mathbb{R}$  is lsc (actually even continuous) and proper with  $\operatorname{dom}(f) = \mathbb{R}^d$ . Therefore, the processes Z and  $Z_n, n \in \mathbb{N}$ , especially have trajectories in  $S_0$ . Moreover, by the subspace-lemma (21) entails  $Z_n \xrightarrow{fd}_D Z$ , whence the proposition follows from Corollary 1.8 with  $c_n := 0$ .

Proof. (of Corollary 1.10)  $Z_n(t) = U'_n t + D_n(t)$  with  $D_n(t) := \frac{1}{2}t'V_n t + r_n(t)$ . Let  $t_1, \ldots, t_k \in \mathbb{R}^k$ . Then  $(U'_n t_1, \ldots, U'_n t_k) \xrightarrow{\mathcal{D}} (U't_1, \ldots, U't_k)$  by the Continuous Mapping Theorem. By continuity  $D_n(t) \xrightarrow{\mathbb{P}} D(t) = \frac{1}{2}t'Vt$  and stochastic convergence by components gives  $(D_n(t_1), \ldots, D_n(t_k)) \xrightarrow{\mathbb{P}} (D(t_1), \ldots, D(t_k))$ . Thus Slutsky's lemma yields the finite dimensional convergence (21), where  $Z(t) = U't + \frac{1}{2}t'Vt$ . Since Z has unique minimizer  $-V^{-1}U$  the statement follows from Corollary 1.9.

### 4. APPENDIX

**Lemma 4.1.** For every  $f : \mathbb{R}^d \to \overline{\mathbb{R}}$  lsc and every real  $r \ge 0$  we have that

$$A(f,r) = \{t \in \mathbb{R}^d : f(t) \le \inf_{s \in \mathbb{R}^d} f(s) + r\}$$

is a closed subset of  $\mathbb{R}^d$ .

Proof. If  $\inf_{s\in\mathbb{R}^d} f(s) = +\infty$  then  $A(f,r) = \mathbb{R}^d \in \mathcal{F}(\mathbb{R}^d)$  and if  $\inf_{s\in\mathbb{R}^d} f(s) = -\infty$ then  $A(f,r) = \{f \leq -\infty\}$ , which is closed, because for each sequence  $t_n \to t \in \mathbb{R}^d$  with  $f(t_n) \leq -\infty$  it follows by lower-semicontinuity of f that  $f(t) \leq \liminf_{n\to\infty} f(t_n) \leq -\infty$ , whence  $t \in \{f \leq -\infty\}$ . Finally, assume that  $\inf_{s\in\mathbb{R}^d} f(s) \in \mathbb{R}$ . Then  $\alpha := \inf_{s\in\mathbb{R}^d} f(s) + r \in \mathbb{R}$  and  $A(f,r) = \{f \leq \alpha\} \in \mathcal{F}(\mathbb{R}^d)$ , since f lsc means that  $\{f > \alpha\}$  is open for each real  $\alpha$ .

**Lemma 4.2.** Let  $f : \mathbb{R}^d \to \overline{\mathbb{R}}$  be lsc, convex and proper. Then  $\operatorname{Argmin}(f)$  is non-empty and bounded if and only if f is level-bounded.

Proof. The if-part follows from Theorem 1.9 in Rockafellar and Wets [19]. For the other direction first observe that by  $\operatorname{Argmin}(f) \neq \emptyset$  there exists  $t_0 \in \mathbb{R}^d$  such that  $f(t_0) = \inf_{\in \mathbb{R}^d} f(t)$ . Since f is proper  $f(t) > -\infty$  for all  $t \in \mathbb{R}^d$  and so in particular  $f(t_0) > -\infty$ . Moreover, there exists  $s \in \mathbb{R}^d$  such that  $f(s) < \infty$ , whence  $f(t_0) \leq f(s) < \infty$ . Consequently,  $\alpha_0 := f(t_0) \in \mathbb{R}$ . It follows that  $\{f \leq \alpha_0\} = \{f = \alpha_0\} = \operatorname{Argmin}(f)$ . Thus by assumption on  $\operatorname{Argmin}(f)$  the level-set  $\{f \leq \alpha_0\}$  is non-empty and bounded and hence compact, because  $\{f \leq \alpha_0\}$  is closed by lower-semicontinuity of f. Now, the assertion that f is level-bounded follows from Proposition 2.3.1 of Bertsekas [2], Convex Analysis and Optimization.

**Lemma 4.3.** If f is lsc, convex, proper and level-bounded, then A(f, r) is non-empty and compact for every real  $r \ge 0$ .

Proof. Conclude from  $\operatorname{Argmin}(f) = A(f, 0) \subseteq A(f, r)$  and Lemma 4.2 that  $A(f, r) \neq \emptyset$  for all real  $r \geq 0$ . As in the proof of Lemma 4.2 we see that  $A(f, r) = \{f \leq \alpha_0 + r\}$ , where  $\alpha_0 = \inf_{t \in \mathbb{R}^d} f(t) \in \mathbb{R}$  and another application of Proposition 2.3.1 of Bertsekas [2] yields that  $\{f \leq \alpha_0 + r\}$  is compact as desired.  $\Box$ 

**Lemma 4.4.** Let  $f : \mathbb{R}^d \to \overline{\mathbb{R}}$  be convex. Then the following statements are equivalent:

- (1) f is proper and  $int(dom f) \neq \emptyset$ .
- (2) f is finite on some nonempty open set

Proof. If (1) holds then f is finite on domf and in particular on  $\operatorname{int}(\operatorname{dom} f) \neq \emptyset$ . For the reverse let  $G \neq \emptyset$  be open such that  $\infty < f(x) < \infty$  for all  $x \in G$ . Then  $G \subseteq$ domf and thus  $G \subseteq \operatorname{int}(\operatorname{dom} f)$ , whence  $\operatorname{int}(\operatorname{dom} f) \neq \emptyset$ . Next, assume that f is not proper. By Exercise 2.5 in Rockafellar and Wets [19] it follows that  $f(x) = -\infty$  for all  $x \in \operatorname{int}(\operatorname{dom} f) \supset G$ , which contradicts  $f > -\infty$  on G.

**Lemma 4.5.** (CMT for  $\to^{\sim}$ ) Let  $(X_1, \mathcal{O}_1)$  and  $(X_2, \mathcal{O}_2)$  be topological spaces and let  $h : X_1 \to X_2$  be a mapping with pertaining set  $D_h := \{x \in X_1 : h \text{ is not continuous at } x\}$  of all discontinuity-points of h. For mappings  $Y_n : (\Omega, \mathcal{A}) \to X_1$  and  $Y : (\Omega, \mathcal{A}) \to X_1$  assume that

$$Y_n \twoheadrightarrow^{\sim} Y$$
 in  $(X_1, \mathcal{O}_1)$ .

If  $\mathbb{P}_*(Y \in D_h) = 0$ , then

 $h(Y_n) \twoheadrightarrow^{\sim} h(Y)$  in  $(X_2, \mathcal{O}_2)$ .

Proof. Let F be closed in  $(X_2, \mathcal{O}_2)$ . Check that

$$cl_1(h^{-1}(F)) \subseteq h^{-1}(F) \cup D_h, \tag{48}$$

where  $cl_1(A)$  denotes the closure of  $A \subset (X_1, \mathcal{O}_1)$ . It follows that

$$\limsup_{n \to \infty} \mathbb{P}^*(h(Y_n) \in F) = \limsup_{n \to \infty} \mathbb{P}^*(Y_n \in h^{-1}(F)) \le \limsup_{n \to \infty} \mathbb{P}^*(Y_n \in \operatorname{cl}_1(h^{-1}(F)))$$
$$\le \mathbb{P}_*(Y \in \operatorname{cl}_1(h^{-1}(F))) \le \mathbb{P}_*(Y \in h^{-1}(F)) + \mathbb{P}_*(Y \in D_h) = \mathbb{P}_*(h(Y) \in F).$$

Here, in the second row the first inequality follows from the Borel-Portmanteau-Theorem, the second inequality from (48) and the subsequent equality from the requirement  $\mathbb{P}_*(Y \in D_h) = 0$ . Another application of the Borel-Portmanteau-Theorem yields the assertion.

**Lemma 4.6.** Let  $\xi_1, \ldots, \xi_n$  be finitely many random variables defined on a measurable space  $(\Omega, \mathcal{A})$  with values in  $(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$ . Then  $C := \{\xi_1, \ldots, \xi_n\}$  is  $\mathcal{A} - \mathcal{B}_{uV}$ -measurable.

Proof. Clearly, C maps into  $\mathcal{F}_d$ . Let  $\mathbf{O} \in \tau_{uV}$ . Since  $\mathcal{S}_{uV}$  is actually a base for  $\tau_{uV}$ , there exists a family  $(F_i)_{i\in I} \subseteq \mathcal{F}_d$  with some index-set  $I \neq \emptyset$  such that  $\mathbf{O} = \bigcup_{i\in I} \mathcal{M}(F_i)$ . If  $\pi_l : (\mathbb{R}^d)^n \to \mathbb{R}^d$  with  $l \in \{1, \ldots, n\}$  denotes the *l*-th projection, i.e.,  $\pi_l(x_1, \ldots, x_n) = x_l$  for  $(x_1, \ldots, x_n) \in (\mathbb{R}^d)^n$ , then  $\{C \in \mathbf{O}\} = \{(\xi_1, \ldots, \xi_n) \in V\}$ , where  $V = \bigcup_{i\in I} \bigcap_{l=1}^n \pi_l^{-1}(F_i^c)$  is open in  $(\mathbb{R}^d)^n$ . In particular,  $V \in \mathcal{B}((\mathbb{R}^d)^n)$ . Now,  $\mathcal{B}((\mathbb{R}^d)^n) = (\mathcal{B}(\mathbb{R}^d))^n$ , whence  $\{(\xi_1, \ldots, \xi_n) \in V\} \in \mathcal{A}$ . Thus  $\{C \in \mathbf{O}\} \in \mathcal{A}$  for every open  $\mathbf{O} \in \tau_{uV}$ , which yields that C is  $\mathcal{A} - \mathcal{B}_{uV}$ -measurable.  $\Box$ 

Proof. (of Lemma 2.1) Let  $x \in \mathbb{R}^d$ . In case 1 assume that  $f(x) = -\infty$ . Then  $(x, \alpha) \in \operatorname{epi}(f) = \operatorname{epi}(g)$  for every  $\alpha \in \mathbb{R}$ , and so  $g(x) \leq \alpha$  for every  $\alpha \in \mathbb{R}$ , which means that  $g(x) = -\infty = f(x)$ . In case 2 let  $-\infty < f(x) < \infty$ . Then  $(x, f(x)) \in \operatorname{epi}(f) = \operatorname{epi}(g)$ , whence  $(\star) g(x) \leq f(x) < \infty$ . Assume that  $g(x) = -\infty$ . Then as in case 1 (exchange f for g) it followed that  $f(x) = -\infty$ , a contradiction. Therefore  $g(x) \in \mathbb{R}$  and consequently  $(x, g(x)) \in \operatorname{epi}(g) = \operatorname{epi}(f)$  resulting in  $f(x) \leq g(x)$  and by  $(\star)$  we obtain that f(x) = g(x). Finally, let  $f(x) = \infty$ . Assume that  $g(x) < \infty$ . Then either  $g(x) = -\infty$  and as in case 1 it followed that  $f(x) = -\infty$  (contradiction) or  $-\infty < g(x) < \infty$  and as in case 2 it followed that  $f(x) < \infty$  (contradiction). Consequently,  $g(x) = \infty = f(x)$ .

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