Convex integral functionals of càdlàg processes

Ari-Pekka Perkkiö∗ Erick Treviño-Aguilar†

December 11, 2018

Abstract
This article characterizes conjugates and subdifferentials of convex integral functionals over linear spaces of càdlàg stochastic processes. The approach is based on new measurability results on the Skorokhod space and new interchange rules of integral functionals that are developed in the article. The main results provide a general approach to apply convex duality in a variety of optimization problems ranging from optimal stopping to singular stochastic control and mathematical finance.

Keywords. càdlàg stochastic processes; convex conjugate; integral functional; normal integrand, set-valued analysis;

AMS subject classification codes. 46N10, 60G07

1 Introduction

We fix a complete stochastic base (Ω, ℱ, (ℱₜ)ₜ≥₀, P) satisfying the usual hypotheses with a terminal time T > 0 which we allow to be +∞. For a space D of adapted processes with càdlàg paths (the french abbreviation for right continuous with left limits), we study conjugates of convex integral functionals

\[ F(y) = E \int_{[0,T]} h(y) dµ + \delta_D(S)(y) \]

for a convex normal integrand h : Ω × [0, T] × ℝᵈ → ℝ, a nonnegative random measure µ and hard constraints D(S). Here D(S) is a subset of D consisting of almost sure selections of the image closure S of dom hₜ(ω) = \{x | hₜ(x, ω) < +∞\} and δ_D(S)(y) takes the value zero if y belongs to D(S) and +∞ otherwise.

For a large class of Banach spaces of adapted càdlàg processes, the dual space can be identified with pairs of random measures under the bilinear form

\[ \langle y, (u, ū) \rangle = E \left[ \int ydu + \int y_-dū \right] , \]

∗Department of Mathematics, Ludwig Maximilians Universität München, Theresienstr. 39, 80333 München, Germany.
†Department of Economics and Finance, University of Guanajuato, Lascuráin de Retana 5, 36000, Guanajuato, México.
where \( y_ - \) is the left continuous version of \( y \); see [2, 13, 24]. This leads us to analyze a larger class of functionals

\[
\hat{F}(y) = E \left[ \int_{[0,T]} h(y)d\mu + \int_{[0,T]} \tilde{h}(y_-)d\tilde{\mu} \right] + \delta_{D(S)}(y) + \delta_{\tilde{D}(\tilde{S})}(y_-),
\]

(2)

where \( \tilde{h} \) is another convex normal integrand, \( \tilde{\mu} \) another non negative random measure, \( D \) a space of càglàd (the French abbreviation for left continuous with right limits) stochastic processes and \( \tilde{S} = \text{cl dom}(\tilde{h}) \).

The main results of the paper, Theorems 13 and 14, characterize the conjugates and subdifferentials of \( F \) and \( \hat{F} \). Our motivation arises from various optimization problems in stochastic control and mathematical finance. Applications to stochastic singular control are already presented in [22] which dealt with integral functionals on regular processes (processes that are optional projections of continuous processes). Examples in Section 8 illustrate how our results allow to model, e.g., bid-ask spreads and currency markets in mathematical finance while an application of our results to optimal stopping is given in [23]. Applications to partial hedging of American options will be presented in a forthcoming article by the authors. Further applications to finance and singular stochastic control going beyond [22] will be given elsewhere.

In Section 2, we recall basic facts from convex analysis and the theory of integral functionals. Since we work later on with the Skorokhod space where the functionals \( F \) and \( \hat{F} \) are not lower semicontinuous, we also present an extension of “an interchange rule between minimization and integration” to jointly measurable integrands on a non necessarily topological vector Suslin space. Another new interchange rule is provided in Section 3 for integral functionals on the space \( D \) of càdlàg functions. Such results go back to the seminal paper of [28] in decomposable spaces of \( \mathbb{R}^d \)-valued measurable functions. Extensions to Suslin-valued functions are studied in [33] and to non-necessarily decomposable spaces, e.g., in [7, 21, 30]. We build on their results.

Section 4 addresses the complication that measurable selection theorems behind the interchange rules require Suslin spaces, which \( D \) under the supremum norm is not. However, \( D \) becomes a Suslin space under the Skorokhod topology. A drawback is that the hard constraints are no longer closed in this topology. The main result of this section, which gives graph measurability of a set-valued mapping on \( D \), and hence that of the hard constraints, has a novel method of the proof since we can not use standard characterizations of measurability based on closed-valuedness.

Before presenting the main results in Section 6, we develop, in Section 5, the crucial interchange rule on a space of càdlàg processes. Stochastic settings for interchange rules and convex duality has been recently developed in [22] and [21] on spaces of regular processes and processes of bounded variation. Our interchange rule can be seen as an extension of one in [22] from the class of regular processes to classes of càdlàg processes.

One of the main assumptions in our main results is a sort of Michael representation (see [32]) of a set-valued mapping \( S \) consisting of càdlàg selections. We
analyze this condition in terms of standard continuity properties of set-valued mappings in Section 7. This section is of independent interest in set-valued analysis. In Section 8, we demonstrate how these results lead to straight-forward applications of the main theorems.

2 Convex conjugates and normal integrands

In this section we recall some fundamentals from convex analysis and the theory of normal integrands. We also give a minor extension of an interchange rule for integral functionals on decomposable spaces that we will need in the sequel.

When $X$ is in separating duality with another linear space $V$, the conjugate of an extended real-valued convex function $g$ on $X$ is the extended real-valued function $g^*$ on $V$ defined by

$$g^*(v) = \sup_{x \in X} \{ \langle x, v \rangle - g(x) \}.$$ 

A vector $v \in V$ is a subgradient of $g$ at $x$ if

$$g(x') \geq g(x) + \langle x' - x, v \rangle \quad \forall x' \in X.$$ 

The subdifferential $\partial g(x)$ is the set of all subgradients of $g$ at $x$. We often use the property that $v \in \partial g(x)$ if and only if

$$g(x) + g^*(v) = \langle x, v \rangle.$$ 

The recession function of a closed proper convex function $g$ is defined by

$$g^\infty(x) = \sup_{\alpha > 0} \frac{g(\bar{x} + \alpha x) - g(\bar{x})}{\alpha},$$

where the supremum is independent of the choice of $\bar{x} \in \text{dom } g = \{ x \in X \mid g(x) < \infty \}$; see [27, Corollary 3C]. By [27, Corollary 3D],

$$\sigma_{\text{dom } g^*} = g^\infty,$$ 

where $\sigma_C := \delta^*_C$ is the support function of $C$.

Let $X$ be a topological space equipped with its Borel $\sigma$-algebra $\mathcal{B}(X)$ and $(\Xi, \mathcal{A}, m)$ a measure space. A set-valued mapping $S : \Xi \to X$ is measurable if the inverse image $S^{-1}(O) := \{ \xi \in \Xi \mid S(\xi) \cap O \neq \emptyset \}$ of every open set $O$ is $\mathcal{A}$-measurable. A function $h : \Xi \times X \to \mathbb{R}$ is a normal integrand on $X$ if its epigraphical mapping

$$\text{epi } h(\xi) := \{ (x, \alpha) \in X \times \mathbb{R} \mid h(\xi, x) \leq \alpha \}$$

is closed-valued and measurable. When this mapping is also convex-valued, $h$ is a convex normal integrand. A general treatment of normal integrands on $\mathbb{R}^d$ can be found from [32, Chapter 14] while integrands on a Suslin space are studied in
In particular, a normal integrand $h$ is jointly measurable so that the integral functional
\[
\int h(w(\xi), \xi) dm
\]
is well-defined for any measurable $w : \Xi \to X$. Throughout the article, an integral is defined as $+\infty$ unless the positive part is integrable.

When $\mathcal{A}$ is $m$-complete, every jointly measurable $h : \Xi \times X \to \mathbb{R}$, that is lower semicontinuous in the second argument almost everywhere, is a normal integrand whenever $X$ and $X^*$ are Suslin locally convex spaces; see e.g., [8, Lemma VII-1] where this is actually taken as the definition of a normal integrand. Later on, we will work with non-complete $\sigma$-algebras (the predictable and optional $\sigma$-algebras) so we use the definition given in terms of the epigraphical mapping.

For a normal integrand $h$, $h^{\infty}$ is the integrand defined $\xi$-wise as the recession function of $h(\xi, \cdot)$. The conjugate integrand $h^*$ is defined $\xi$-wise as the conjugate of $h(\xi, \cdot)$. When $X$ and $X^*$ are Suslin and locally convex spaces, $h^*$ is a convex normal integrand due to [8, Corollary VII-2]. For our purposes, $h^{\infty}$ is a convex normal integrand when $X = \mathbb{R}^d$, see [32, Exercise 14.54]. Given a measurable function $w$, we denote by $\partial h(w(\xi), \xi)$. The set-valued mapping $\xi \mapsto \partial h(w(\xi), \xi)$.

Interchange rules for integral functionals on decomposable spaces go back to [28] for normal integrands on $\mathbb{R}^d$. Theorem 1 below is an extension of the main theorem in [33] that was formulated for a locally convex Suslin topological vector space $X$. However, our formulation is closer to [32, Theorem 14.60] stated for $X = \mathbb{R}^d$. We need this extension later on since the space of càdlàg functions equipped with the Skorokhod topology is not a topological vector space. The proof is almost identical, but we give it here for completeness.

**Theorem 1.** Assume that $(\Xi, \mathcal{A}, m)$ is a complete $\sigma$-finite measure space, and $X$ is a Suslin space. Let $h : \Xi \times X \to \mathbb{R}$ be $\mathcal{A} \otimes \mathcal{B}(X)$-measurable. Then, for $X \subset L^0(\Xi; X)$ decomposable
\[
\inf_{x \in X} \int h(x) dm = \int \inf_{x \in X} h(x) dm
\]
as soon as the left side is less than $+\infty$.

**Proof.** We note first that
\[
p(\xi) := \inf_{x \in X} h(\xi, x)
\]
is measurable. Indeed, for the projection $\prod : \Xi \times X \to \Xi$ and any $b \in \mathbb{R}$, we have $p^{-1}((\infty, b)) = \prod h^{-1}((-\infty, b))$, where the right side is measurable due to [12, III.44-45], since $\mathcal{A}$ is complete.

Let $\alpha = \int p dm$. There exists $\beta \in L^0(\Xi)$ such that $\beta > p$ and $\int \beta dm < \alpha$. Indeed, by $\sigma$-finiteness, there exists strictly positive $\beta \in L^1(\Xi, m)$ so that, for small enough $\epsilon$, we may choose $\beta := \epsilon \beta + p$. The set
\[
A := \{ (\xi, x) \in \Xi \times X \mid h(\xi, x) < \beta(\xi) \}
\]
is $\mathcal{A} \otimes \mathcal{B}(X)$-measurable and $\Pi A = \Xi$, so there exists $x \in L^0(X)$ such that $h(x) < \beta$ by [12, III.44-45].

By [6, Theorem 7.4.3], the law of $m \circ x^{-1}$ is Radon on $X$, so there exists a nondecreasing sequence $(\Xi_v)$, $\bigcup \Xi_v = \Xi$ such that $\text{cl } x(\Xi_v)$ are compact and $\int 1_{\Xi_v} x dm < 1/\nu$. Let now $\bar{x} \in \mathcal{X}$ be such that $\int h(\bar{x}) dm < \infty$, and define

$$\bar{x}^\nu = 1_{\Xi_v} x + 1_{\Xi_v} \bar{x}.$$  

By construction, $\bar{x}^\nu \in \mathcal{X}$, and $h(x^\nu) \leq \max\{h(x), h(\bar{x})\}$ for all $\nu$, so, by Fatou’s lemma, $\int h(x^\nu) dm < \alpha$ for $\nu$ large enough.

### 3 Integral functionals of càdlàg functions

This section gives an interchange rule for integral functionals on the space of càdlàg functions. Our main results build on the interchange rule.

The space $D$ of $\mathbb{R}^d$-valued càdlàg functions on $[0,T]$ is Banach for the norm

$$\|y\|_\infty = \sup_{t \in [0,T]} |y_t|.$$  

We denote $y_{0-} := 0$ and note that $\lim_{t \to T} y_t$ exists also in the case $T = \infty$. The dual of $D$ can be identified with $\hat{M} := M \times \hat{M}$ under the bilinear form

$$\langle y, \hat{u} \rangle := \int y d\mu + \int y_- d\tilde{u},$$  

where $\hat{u} = (u, \tilde{u}) \in \hat{M}$, $M$ is the space of $\mathbb{R}^d$-valued Radon measures on $[0,T]$, $\hat{M} \subset M$ is the space of purely atomic measures, and $y_-$ denotes the left continuous version of $y$; this is a deterministic special case of [13, Theorem VII.65 and Remark VII.4 (a)].

Given a nonnegative Radon measure $\mu$ on $[0,T]$ and a convex normal integrand $h : [0,T] \times \mathbb{R}^d \to \mathbb{R}$ on $\mathbb{R}^d$, the associated integral functional on the space of measurable $\mathbb{R}^d$-valued functions is

$$I_h(y) := \int h(y) d\mu := \int_{[0,T]} h_t(y_t) d\mu_t.$$  

For the domain mapping $\text{dom } h_t := \{ y \in \mathbb{R}^d \mid h_t(y) < \infty \}$,

$$S_t := \{ y \in \mathbb{R}^d \mid y \in \text{cl } \text{dom } h_t \}$$

is its image closure and

$$D(S) := \{ y \in D \mid y_t \in S_t \ \forall t \}$$

is the set of càdlàg selections of $S$.

Theorem 2 below is an interchange rule for integral functionals on the non-decomposable space $D$. The assumptions of the theorem are analogous to those of [25, Theorem 5] that was formulated on continuous functions.
**Assumption 1.** We have

\[ S_t = \text{cl}\{y_t \mid y \in D(S)\} \quad \forall \; t, \]

\[ D(S) = \text{cl}(\text{dom} \; I_h \cap D(S)), \]

where the latter closure is with respect to the supremum norm.

**Theorem 2.** Under Assumption 1,

\[ \inf_{y \in D(S)} I_h(y) = \int \inf_{y \in \mathbb{R}^d} h(y) d\mu \]

as soon as the left side is less than \(+\infty\).

**Proof.** We have \( D(S) = \text{cl}(\text{dom} \; I_h \cap D(S)) \) and \( D(S) \) is PCU-stable in the sense of [7], so, by [7, Theorem 1],

\[ \inf_{y \in D(S)} I_h(y) = \int \inf_{y \in \Gamma_t} h_t(y) d\mu_t, \]

where \( \Gamma \) is the essential supremum of \( D(S) \), i.e., the smallest (up to a \( \mu \)-null set) closed-valued mapping for which every \( y \in D(S) \) is a selection of \( \Gamma \) \( \mu \)-almost everywhere. We have \( S_t \subseteq \Gamma_t \) \( \mu \)-almost everywhere by (4), so the infimum over \( \Gamma_t \) can be taken instead over all of \( \mathbb{R}^d \). \( \square \)

The above result has a variant on the space \( D_\ell \) of càglàd functions (the French abbreviation of left continuous with right limits), which we equip likewise with the supremum norm. Given another nonnegative Radon measure \( \tilde{\mu} \) on \([0,T]\) and convex normal integrand \( \tilde{h} \), we define likewise

\[ I_{\tilde{h}}(y) := \int \tilde{h}(y) d\tilde{\mu} := \int \tilde{h}_t(y_t) d\tilde{\mu}_t \]

on the space of measurable \( \mathbb{R}^d \)-valued functions, \( \tilde{S}_t := \text{cl} \text{dom} \tilde{h}_t \) and

\[ D_\ell(\tilde{S}) := \{y \in D_\ell \mid y_t \in \tilde{S}_t \; \forall t\}. \]

We use the convention \( \tilde{S}_{0-} = \{0\} \).

**Assumption 2.** We have

\[ \tilde{S}_t = \text{cl}\{y_t \mid y \in D_\ell(\tilde{S})\} \quad \forall \; t, \]

\[ D_\ell(\tilde{S}) = \text{cl}(\text{dom} \; I_{\tilde{h}} \cap D_\ell(\tilde{S})), \]

where the latter closure is with respect to the supremum norm.

**Theorem 3.** Under Assumption 2

\[ \inf_{y \in D_\ell(\tilde{S})} I_{\tilde{h}}(y) = \int \inf_{y \in \mathbb{R}^d} \tilde{h}(y) d\tilde{\mu} \]

as soon as the left side is less than \(+\infty\).
Next we state the deterministic special case of our main result, Theorem 13 below. That is, we give a pointwise expression for the conjugate of
\[ \hat{f}(y) = I_h(y) + \delta_{D(S)}(y) + I_{\hat{h}}(y) + \delta_{D_l(S)}(y) \]
and for its subdifferentials. This result can be used to extend duality results in deterministic singular control using the machinery in, e.g., [31, 20].

It will turn out that the conjugate can be expressed in terms of \( J^{S}_{h} \), where for a normal integrand \( h \), the functional \( J^{S}_{h} : M \to \mathbb{R} \) is defined by
\[
J^{S}_{h}(\theta) = \int h(\frac{d\theta^a/d\mu}{d\theta}) d\mu + \int h^\infty(\frac{d\theta^s/d|\theta^s|}{d|\theta^s|}) d|\theta^s|.
\] (5)

Here \( \theta^a \) and \( \theta^s \) are the absolutely continuous and the singular part, respectively, of \( \theta \) with respect to \( \mu \) and \( |\theta^s| \) is the total variation of \( \theta^s \). The formula (3) makes it evident that the properties of the domain mapping \( S \) play an important role for the validity of the representation in terms of \( J^{S}_{h} \). For integral functionals on continuous functions, such conjugate formulas go back to [30]; see also [25] for necessity and sufficiency. We use the notation \( \partial^{S}_{h} := \partial \delta_{S} \).

**Theorem 4.** Assume that \( \tilde{\mu} \) is purely atomic and, for every \( \epsilon > 0 \), \( y \in \text{dom} \ I_h \cap D(S) \) and \( \tilde{y} \in \text{dom} \ I_{\tilde{h}} \cap D_l(\tilde{S}) \) and \( (u, \tilde{u}) \in \hat{M} \), there exists \( \tilde{y} \in D(S) \) and \( \hat{y} \in D_l(\hat{S}) \) with
1. \( I_h(\tilde{y}) + \int \tilde{y} du \leq I_h(y) + \int y du + \epsilon \) and \( \tilde{y}_{-} \in \text{dom} \ I_h \cap D_l(\tilde{S}) \),
2. \( I_{\tilde{h}}(\tilde{y}) + \int \tilde{y} \tilde{d}u \leq I_{\tilde{h}}(\tilde{y}) + \int \tilde{y} \tilde{d}u + \epsilon \) and \( \hat{y}_{+} \in \text{dom} \ I_{\tilde{h}} \cap D(S) \).

Then, under Assumptions 1 and 2, \( \hat{f} \) is a proper lower semicontinuous convex function on \( D \),
\[ \hat{f}^{\ast}(u, \tilde{u}) = J^{h}_{\ast}(u) + J^{\tilde{h}}_{\ast}(\tilde{u}) \]
and \( (u, \tilde{u}) \in \partial \hat{f}(y) \) if and only if
\[
\frac{du/d\mu}{d\theta} \in \partial h(y) \quad \mu-a.e.,
\frac{du/|u^s|}{d\theta} \in \partial^{S}_{h}(y) \quad |u^s|-a.e.,
\frac{d\tilde{u}/d\tilde{\mu}}{d\theta} \in \partial \tilde{h}(y_{-}) \quad \mu-a.e.,
\frac{d\tilde{u}/|\tilde{u}^s|}{d\theta} \in \partial^{S}_{\tilde{h}}(y_{-}) \quad |\tilde{u}^s|-a.e.
\]

4 Graph measurability of integral functionals

In this section we endow \( \Xi = \Omega \times [0, T] \) with the product \( \sigma \)-algebra \( \mathcal{A} := \mathcal{F} \otimes \mathcal{B}([0, T]) \). From now on, we fix a convex normal integrand \( h : \Xi \times \mathbb{R}^d \to \mathbb{R} \).

For a measurable closed convex-valued mapping \( \Gamma : \Omega \times [0, T] \rightrightarrows \mathbb{R}^d \) we define \( D(\Gamma) : \Omega \rightrightarrows D \) by
\[ D(\Gamma)(\omega) := \{ y \in D \mid y_t \in \Gamma_t(\omega) \ \forall \ t \}. \]
Similarly, we define, pathwise, \( I_h \) for any \( \mathbb{R}^d \)-valued measurable function \( w \) on \([0, T]\) and \( J_h \) for any \( u \in M \) as
\[
I_h(\omega, w) := I_{h(\omega, \cdot)}(w),
\]
\[
J_h(\omega, u) := J_{h(\omega, \cdot)}(u).
\]

We will consider c\'adl\'ag processes as \( D \)-valued random variables where the measurability is understood with respect to the Borel-\( \sigma \)-algebra \( \mathcal{B}(D) \) generated by the Skorokhod topology \( \tau_S \). The reason is, that \( \tau_S \) is Suslin whereas the topology \( \tau_D \) generated by the supremum norm is not (since, e.g., \( \tau_D \) is not separable). However, \( \mathcal{B}(D) \) coincides with the Borel-\( \sigma \)-algebra generated by the continuous linear functionals on \( \tau_D \); see [26, Theorem 3]. It is also generated by point-evaluations; see [4, Theorem 12.5]. For the definition of the Skorokhod topology and its basic properties, we refer to [16] and [15].

**Lemma 5.** The function \( I_h : \Omega \times D \to \mathbb{R} \) is \( \mathcal{F} \otimes \mathcal{B}(D) \)-measurable.

**Proof.** By [15, Theorem 15.12], the sequential convergence in the Skorohod topology implies pointwise convergence outside a countable set. Thus, by [22, Theorem 20], \( I_h : \Omega \times D \to \mathbb{R} \) is \( \mathcal{F} \otimes \mathcal{B}(D) \)-measurable. \( \square \)

**Lemma 6.** For a random time \( \tau \), the mappings \((\omega, y) \mapsto y_{\tau(\omega)} \) and \((\omega, y) \mapsto y_{\tau(\omega)-} \) are \( \mathcal{F} \otimes \mathcal{B}(D) \)-measurable.

**Proof.** We assume that \( 0 < \tau < T \) almost surely, the extension to general \( \tau \) is evident. Defining \( g^{\nu} : \Omega \times D \to \mathbb{R} \) by
\[
g^{\nu}(\omega, y) := \frac{1}{\nu} \int_{\tau(\omega)}^{\tau(\omega)+1/\nu} y_s ds,
\]
g\( ^{\nu} \cdot y \) is measurable and \( g^{\nu}(\omega, \cdot) \) is continuous. Here the continuity follows from the dominated convergence and the facts that convergence in the Skorokhod topology implies uniform boundedness and pointwise convergence outside a countable set. By [1, Lemma 4.51], \( g^{\nu} \) is jointly measurable, so we get the first claim from \( \lim_{\nu \to \infty} g^{\nu}(\omega, y) = y_{\tau(\omega)} \). The second claim can be proved similarly. \( \square \)

The next theorem is of independent interest of the Skorokhod topology and it will be crucial for the main results of the paper. In general, \( D(\Gamma) \) in the theorem is not closed-valued in the Skorokhod topology, since sequences therein need not converge pointwise. By \( \tau \), we denote the topology on \( \mathbb{R} \) generated by the right-open intervals \( \{(s, t) \mid s < t\} \).

**Theorem 7.** Assume that \( \Gamma : \Omega \times [0, T] \Rightarrow \mathbb{R}^d \) is a \( \mathcal{F} \otimes \mathcal{B}([0, T]) \)-measurable closed convex-valued mapping. Then, \( D(\Gamma) : \Omega \Rightarrow D \) has a measurable graph.

**Proof.** Let \((t^n)_{n=1}^\infty \) be a dense sequence in \([0, T]\) containing \( T \). For each rational vector \( q \in \mathbb{Q}^{d+1} \), let \( H_q = \{ x \in \mathbb{R}^d \mid (q_1, \ldots, q_d) \cdot x \leq q_0 \} \) be the associated “rational” half-space. By [32, Theorem 14.3(i)], the sets \( A_q := \{(\omega, t) \mid \)
Γ_τ(ω) ⊂ H_q}, are measurable, so, by time-reversing the argument in the proof of [15][Theorem 4.2], each

\[ t'_q := \sup\{t \leq t', t \in A_q\} \]

is a random time. By the measurable selection theorem, for each \( t'_q \), there exists random times \( t''_q,j \in A_q \) such that \( t'_q - 1/j \leq t''_q,j \leq t'_q \). We redefine \( t''_q,j = t'_q \) on \( \{t_q \in A_q\} \). It suffices to show

\[ D(Γ) = \{y \in D \mid y_{t''_q,j} \in Γ_{t''_q} \forall q, \nu, j\}, \]

since the right side has a measurable graph, by Lemma 6. In the rest of the proof, we may argue pathwise, so we fix \( \omega \).

We show first that, for each \( q \), \( (t''_q,j)_{j,\nu=1}^{\infty} \) is \( \tau_r \)-dense in \( A_q \). Let \( t \in A_q \).

If there exists \( t'' \succ t \) such that \( (t, t'') \cap A_q = \emptyset \), then \( t = t'_q \). Otherwise, there is a subsequence \( t'' \searrow t \) such that \( (t, t'') \cap A_q \neq \emptyset \) for each \( \nu \). Thus \( t < t''_q \leq t'' \). For \( j(\nu') \) large enough, \( t \leq t'_q,j(\nu') \leq t''_q \leq t'' \), so we have eventually \( t''_q,j(\nu') \searrow t \).

Assume now that \( y \in D \) satisfies \( y_{t''_q,j} \in Γ_{t''_q,j} \), for each \( t''_q,j \). Fix \( t \) and choose \( q \) such that \( Γ_t \subset H_q \) (note that \( H_q = \mathbb{R}^d \) for \( q = 0 \)). Since \( t \in A_q \) and \( (t''_q,j)_{j,\nu=1}^{\infty} \) is \( \tau_r \)-dense in \( A_q \), there is a sequence \( (t''_q,j) \) converging to \( t \) in \( \tau_r \) such that \( t''_q,j \in A_q \).

Thus

\[ y_t = \lim y_{t''_q,j} \in H_q. \]

We obtained

\[ y_t \in \bigcap_q \{H_q \mid Γ_t \subset H_q\}, \]

where the right side equals \( Γ_t \), since every closed convex set is an intersection of rational half-spaces containing the set.

The following corollary extends [22, Lemma 5], where pathwise inner semi-continuity of \( S \) was assumed, with a completely different proof. The subspace \( C \subset D \) of continuous functions is closed in \( D \) and its relative topology w.r.t. the Skorokhod topology is generated by the supremum norm.

**Corollary 8.** Assume that \( Γ \) is a measurable closed convex-valued stochastic mapping. Then \( C(Γ) \) is measurable and closed convex-valued.

**Proof.** Since \( Γ \) is closed convex-valued, \( C(Γ) \) is closed convex-valued as well. By Theorem 7, \( D(Γ) \) has a measurable graph, so \( C(Γ) \) has a measurable graph. For complete \( F \), graph measurable and closed nonempty-valued mappings to Polish spaces are measurable [8, Theorem III.30]. If \( C(Γ) \) is not nonempty-valued, we may apply the above with \( Ω \) replaced by dom \( C(Γ) \), which is \( F \)-measurable by the projection theorem.

\[ \square \]
5 Integral functionals of càdlàg processes

We denote by $L^0_+(M)$ the space of nonnegative random Radon measures which are measurable in the sense of [13, VI.86] and by $L^1_+(M)$ elements $\mu \in L^0_+(M)$ such that the random variable $\mu(\cdot;[0,T])$ belongs to the space $L^1(P)$ of $P$ integrable functions. A random measure $\mu$ is optional if $E \int vd\mu = E\int v\,dM$ for every bounded measurable $v$ and $M$ is the optional projection of $v$. For any subset $A \subset M$ we denote by $L^1_+(A)$ the elements $\mu$ of $L^1_+(M)$ such that $\mu(\omega,1) \in A$ almost surely. We set $L^1(A) = L^1_+(A) - L^1_+(A)$.

A normal integrand $h : \Omega \times [0,T] \times \mathbb{R}^d \to \mathbb{R}$ on $\mathbb{R}^d$ is optional (resp. predictable) if its epigraphical mapping is optional (resp. predictable). In what follows, we use the qualifier a.s.e. (almost surely everywhere) for a property satisfied outside an evanescent set.

Let $\mathcal{D}$ be a linear subspace of adapted càdlàg processes.

**Assumption 3.** The convex normal integrand $h$ is optional and

1. $(t,x) \to h_t(x,\omega)$ satisfies the Assumption 1 almost surely,

2. there exists an optional process $x$ and nonnegative measurable process $\alpha$ with $E \int |x_t||y|d\mu < \infty$ for all $y \in \mathcal{D}$ and $E \int \alpha d\mu < \infty$ such that

$$h_t(y,\omega) \geq x_t(\omega) \cdot y - \alpha_t(\omega) \quad \text{a.s.e.}$$

Throughout, $L^0(D)$ denotes the space of càdlàg-valued random variables $y : \Omega \to D$ which are $(\mathcal{F},\mathcal{B}(D))$-measurable, $L^0(D,S)$ the elements of $L^0(D)$ that are almost surely selections of $S$, and $L^\infty(D)$ the elements $y \in L^0(D)$ such that $\|y\|_\infty \in L^\infty$. We set $\mathcal{D}(S) = \mathcal{D} \cap L^0(D,S)$ and $L^i(D,S) = L^i(D) \cap L^0(D,S)$ for $i \in \{0,\infty\}$. Let $\mathcal{D}^\infty$ be the space of bounded adapted càdlàg processes and $\mathcal{D}^\infty(S) := \mathcal{D}^\infty \cap L^0(D,S)$.

**Theorem 9.** Let $\mathcal{D}$ be a space of adapted càdlàg processes containing $\mathcal{D}^\infty$. Under Assumption 3,

$$\inf_{y \in \mathcal{D}} [EI_h(y) + \delta_{\mathcal{D}(S)}(y)] = E \left[ \inf_{x \in \mathbb{R}^d} h(x) d\mu \right]$$

as soon as the left side is less than $+\infty$.

**Proof.** Assume first that $EI_h$ is finite for some $y \in \mathcal{D}^\infty(S)$. By Lemma 5 $I_h$ is $\mathcal{F} \otimes \mathcal{B}(D)$-measurable while $\delta_{\mathcal{D}(S)}$ is $\mathcal{F} \otimes \mathcal{B}(D)$-measurable due to Theorem 7. Then, $I_h + \delta_{\mathcal{D}(S)}$ is $\mathcal{F} \otimes \mathcal{B}(D)$-measurable and

$$\inf_{y \in L^\infty(D)} E[I_h(y) + \delta_{\mathcal{D}(S)}(y)] = E \left[ \inf_{x \in \mathcal{D}} \left\{ \int h(x) d\mu + \delta_{\mathcal{D}(S)}(x) \right\} \right]$$

$$= E \left[ \inf_{x \in \mathcal{D}(S)} \int h(x) d\mu \right]$$

$$= E \left[ \int \inf_{x \in \mathbb{R}^d} h(x) d\mu \right],$$
where the first equality follows from Theorem 1, the second is clear, and the third follows from Theorem 2. On the other hand
\[
\inf_{y \in D^\infty(S)} E_I h(y) = \inf_{y \in L^\infty(D)} \left[ E_I h(y) + \delta_{D(S)}(y) \right]
\leq \inf_{y \in L^\infty(D,S)} E_I h(y)
\leq \inf_{y \in D^\infty(S)} E_I h(y),
\]
where in the first equality we used the fact that optional projections of bounded càdlàg processes are bounded and càdlàg [13, Theorem VI.47] and that optional projections of selections are again selections; see [18, Corollary 4]. In the first inequality we applied Jensen’s inequality Lemma 22 and Jensen’s inequality for optional set-valued mappings [18, Theorem 9].

Assume now that \( E_I h \) is finite for arbitrary \( \bar{y} \in \mathcal{D}(S) \). Then
\[
\bar{h}_t(y, \omega) := h_t(y + \bar{y}(\omega)),
\]
satisfies Assumption 3 and the corresponding integral functional is finite at the origin. Thus the result follows from the first part.

Next, we formulate a variant of the above result for adapted càglàd processes. A random measure \( \tilde{\mu} \in L^1_+(M) \) is predictable if \( E \int v d\mu = E \int p v d\mu \) for each bounded measurable process \( v \), where \( p v \) is the predictable projection of \( v \). Let \( \tilde{\mu} \in L^1_+(M) \) be a nonnegative predictable random Radon measure, \( \tilde{h} : \Omega \times [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R} \) a convex normal integrand on \( \mathbb{R}^d \). Let \( \mathcal{D}_t \) be a linear subspace of adapted càglàd processes.

**Assumption 4.** The convex normal integrand \( h \) is predictable and
1. \((t, x) \rightarrow h_t(x, \omega)\) satisfies the Assumption 2 almost surely,
2. there exists a predictable process \( x \) and nonnegative measurable process \( \alpha \) with \( E \int |x| |y| d\tilde{\mu} < \infty \) for all \( y \in \mathcal{D}_t \) and \( E \alpha d\tilde{\mu} < \infty \) such that
\[
\tilde{h}_t(y, \omega) \geq x_t(\omega) \cdot y - \alpha_t(\omega) \quad \text{a.s.e.}
\]

We denote the space of bounded adapted càglàd processes by \( \mathcal{D}_t^\infty \) and by \( \mathcal{D}_t(\tilde{S}) \) the elements of \( \mathcal{D}_t \) which are almost surely selections of \( \tilde{S} \).

**Theorem 10.** Let \( \mathcal{D}_t \) be a space of adapted càglàd processes containing \( \mathcal{D}_t^\infty \). Under Assumption 4,
\[
\inf_{y \in \mathcal{D}_t} \left[ EI_h(y) + \delta_{\mathcal{D}_t(\tilde{S})}(y) \right] = E \left[ \int_{x \in \mathbb{R}^d} h(x) d\tilde{\mu} \right]
\]
as soon as the left side is less than \( +\infty \).

**Proof.** Recalling that predictable projections of bounded càdlàg processes are bounded and càdlàg [13, Theorem VI.47], the theorem is proved similarly to Theorem 9. \( \square \)
The next assumption gives a necessary compatibility condition so that we get an interchange rule for
\[ \hat{F}(y) = E \left[ I_h(y) + I_h(y_-) \right] + \delta_{\mathcal{D}(S)}(y) + \delta_{\mathcal{D}_l(S)}(y_-). \]

**Assumption 5.** The predictable measure \( \hat{\mu} \) is purely atomic and, for every \( \epsilon > 0, y \in \text{dom } EI_h \cap \mathcal{D}(S), \) and \( \check{y} \in \text{dom } EI_h \cap \mathcal{D}_l(S) \), there exists \( \tilde{y} \in \mathcal{D}(S) \) and \( \check{y} \in \mathcal{D}_l(S) \) with
\[
\begin{align*}
1. & \quad EI_h(\tilde{y}) \leq EI_h(y) + \epsilon \text{ and } \tilde{y} \in \text{dom } EI_h \cap \mathcal{D}_l(S), \\
2. & \quad EI_h(\check{y}) \leq EI_h(y) + \epsilon \text{ and } \check{y} \in \text{dom } EI_h \cap \mathcal{D}(S).
\end{align*}
\]

A class of adapted càdlàg processes \( \mathcal{D} \) is called solid if \( y \in \mathcal{D} \) whenever \( |y| \leq |\check{y}| \) for some \( \check{y} \in \mathcal{D} \), and max-stable if \( \max\{|y_1|,|y_2|\} \in \mathcal{D} \) whenever \( y_1, y_2 \in \mathcal{D} \).

**Theorem 11.** Let \( \mathcal{D} \) be a solid max-stable space containing \( \mathcal{D}\infty \) and let \( \mathcal{D}_l = \{y_- \mid y \in \mathcal{D}\} \). Under Assumptions 3, 4 and 5,
\[
\inf_{y \in \mathcal{D}} \hat{F}(y) = E \left[ \int_{x \in \mathbb{R}^d} h(x)d\mu + \int_{x \in \mathbb{R}^d} \check{h}(x)d\hat{\mu} \right]
\]
as soon as the left side is less than \( +\infty \).

**Proof.** Let \( \epsilon > 0 \). By Theorems 9 and 10 and Assumption 5, there exist \( y, \check{y} \in \mathcal{D}(S) \) such that \( \check{y} \in \mathcal{D}_l(S), \) \( \check{y} \in \text{dom } EI_h \cap \mathcal{D}(S), \) \( y_- \in \text{dom } EI_h \cap \mathcal{D}_l(S) \) and
\[
EI_h(y) + EI_h(\check{y}_-) \leq E \left[ \int_{x \in \mathbb{R}^d} h_t(x)d\mu + \int_{x \in \mathbb{R}^d} \check{h}_t(x)d\hat{\mu} \right] + \epsilon. \tag{6}
\]

In the next construction we take \( T = \infty \), the case \( T < \infty \) being more simple. Since \( \hat{\mu} \) is purely atomic and predictable, the process \( w \) defined by
\[
w_t := \int_{[0,t]} d\hat{\mu}
\]
is predictable and purely discontinuous whose jump times belong to the set \( A := \{\Delta w \neq 0\} \). For each \( \nu = 1, 2, \ldots \) let \( A_{\nu} := \{\Delta w \geq \nu^{-1}\} \). For each \( n = 0, 1, \ldots \), the predictable set \( A_{\nu} \cap \{n,n+1\} \) is a union of graphs of the elements of an increasing sequence \( (\tau_{\nu,n,j})_{j=1}^{\infty} \) of predictable times with graphs in \( \{n,n+1\} \). Fix a decreasing sequence of positive real numbers converging to zero \( \{c_n\}_{n \in \mathbb{N}} \). Let \( \{\sigma_{\nu,n,j,k}\}_{k \in \mathbb{N}} \) be an announcing sequence for \( \tau_{\nu,n,j} \) as in Lemma 23 below in the interval \( \{n,n+1\} \) with sequence \( \{1/2^{\nu+k}\}_{k \in \mathbb{N}} \). We define a process \( y_{\nu}^{\nu,k} \) by
\[
y_{\nu}^{\nu,k}(\omega) = \begin{cases} 
\check{y}_t(\omega) & \text{if } (\omega,t) \in \bigcup_{n=0}^{\infty} \{n,n+1\} \cap \bigcup_{j=1}^{\infty} \{\sigma_{\nu,n,j,k}, \tau_{\nu,n,j}\}, \\
y_t(\omega) & \text{otherwise.}
\end{cases} \tag{7}
\]
Outside a null set, the process $y_{\nu,k}$ has the following properties

$$y_{\nu,k}$$ has càdlàg paths, \hspace{1cm} (8)

$$\lim_{k \to \infty} y_{\nu,k} = y, \text{ a.s.} \hspace{1cm} (9)$$

$$\lim_{k \to \infty} y_{\nu,k}^{-} = y_{-}, \text{ a.s.} \text{ stationarily on } A_{\nu}, \hspace{1cm} (10)$$

$$\lim_{k \to \infty} y_{\nu,k}^{-} = \tilde{y}_{-}, \text{ a.s.} \text{ stationarily on } A_{\nu}, \hspace{1cm} (11)$$

which we verify next.

Fix $\nu$, $n$ and $k$. Let $D_j := \{\sigma_{\nu,n,j,k} < \infty\}$ and $B_j := \{\tau_{\nu,n,j} = \infty\}$. We have $P(D_j \cap B_j) \leq \frac{\epsilon}{2^{j}}$ and so $C := \limsup_{j \to \infty} D_j \cap B_j$ is a null set due to Borel-Cantelli Lemma. Let $N$ be a null set such that $w_T(\omega) < \infty$ for $\omega \in \Omega/\mathbb{C} \cup \mathbb{N}$. Let $M$ be a null set such that on $\Omega/\mathbb{C} \cup \mathbb{N}$ the stopping times $\sigma_{\nu,n,j,k}$ converge to $\tau_{\nu,n,j}$. Let $p_{\nu,n,k}(\omega) := \left\{ t \in [n, n + 1] \mid (\omega, t) \in \bigcup_{j=1}^{\infty} [\sigma_{\nu,n,j,k}, \tau_{\nu,n,j}] \right\}$.

We get (8) once we show that $p_{\nu,n,k}(\cdot)$ is a finite union of semiopen sets $[a, b)$ on $\Omega/(\mathbb{C} \cup \mathbb{N} \cup M)$, and it is actually included on $(n, n + 1]$. For $\omega \in \Omega/(\mathbb{C} \cup \mathbb{N} \cup M)$ there exists $j_0$ such that for all $j \geq j_0$ we have $\omega \notin D_j \cap B_j$, since $\omega$ is not an element of $C$. There are two alternatives:

$$\sigma_{\nu,n,j,k}(\omega) = \infty \text{ or } \tau_{\nu,n,j}(\omega) < \infty.$$ 

The second alternative can only happen for a finite number of indexes $j \geq j_0$, since $\omega$ is not an element of $N$. In the first alternative we clearly have that the interval $[\sigma_{\nu,n,j,k}(\omega), \tau_{\nu,n,j}(\omega))$ is empty. Thus

$$p_{\nu,n,k}(\omega) = \bigcup_{j=1}^{\tilde{j}} [\sigma_{\nu,n,j,k}(\omega), \tau_{\nu,n,j}(\omega))$$

for some $\tilde{j}$. The set $p_{\nu,n,k}(\omega)$ is included on $(n, n + 1]$ since each $\sigma_{\nu,n,j,k}$ is strictly greater than $n$. Now (8) is established. Other properties of $p_{\nu,n,k}(\cdot)$ on $\Omega/(\mathbb{C} \cup \mathbb{N} \cup M)$ are $p_{\nu,n,k+1} \subset p_{\nu,n,k}$ and $\cap_{k \in \mathbb{N}} p_{\nu,n,k} = \emptyset$. Thus (9) holds.

Let

$$q_{\nu,n,k}(\omega) := \left\{ t \in (n, n + 1] \mid (\omega, t) \in (n, n + 1] \cap \bigcup_{j=1}^{\infty} [\sigma_{\nu,n,j,k}, \tau_{\nu,n,j}] \right\}.$$ 

Since $p_{\nu,n,k}$ is almost surely a finite union, we have, for $t \in (n, n + 1]$

$$y_{\nu,k}^{-}(\omega) = \begin{cases} \tilde{y}_{t-}(\omega) & \text{if } t \in q_{\nu,n,k}(\omega) \\ y_{t-}(\omega) & \text{otherwise.} \end{cases} \hspace{1cm} (12)$$
Since also $q^{ν,n,k+1} ⊂ q^{ν,n,k}$ and $\bigcap_{k=1}^{∞} q^{ν,n,k}$ is equal to $\bigcup_{j=1}^{∞} \pi^{ν,n,j}$ we get (10) and (11). Now (9) gives
\[
\lim_{k→∞} h(y^{ν,k}) = h(y), \quad P\text{-a.s. for all } t \in [0, T],
\]
and, for all $(ω, t) ∈ A$, (10) and (11) give
\[
\lim_{k→∞} ˜h(y^{ν,k}) = ˜h(˜y-1_{Aν}).
\]
By the definition of $Aν$ it follows that $\lim_{ν→∞} ˜h(y-1_{Aν} + ˜y-1_{Aν}) = ˜h(˜y-)$. We have, by (7) and (12),
\[
h(y^{ν,k}) + ˜h(y^{ν,k}) ≤ \max\{h(˜y), h(y)\} + \max\{ ˜h(˜y-), ˜h(y-)\},
\]
where the right side is integrable, by the existence of the lower bounds in Assumptions 3 and 4 and by the fact that $E[Ih(y) + I ˜h(˜y) + I ˜h(y-) + I ˜h(˜y-)] < ∞$.
By Fatou’s lemma,
\[
\limsup_{ν→∞} \limsup_{k→∞} \{ EIh(y^{ν,k}) + EI ˜h(y^{ν,k}) \} ≤ EIh(y) + EI ˜h(˜y-)
\]
which in combination with (6) finishes the proof. □

6 Conjugates of integral functionals

This section presents the main results of the article. That is, we characterize the convex conjugates and the subdifferentials of $F$ and $F$ defined respectively in (1) and (2). The results are based on the interchange rules developed in the previous sections. We start by specifying appropriate spaces $D$ and $M$, of adapted càdlàg processes and random measures in duality.

From now on, we assume that $D$ and $Dl$ satisfy the conditions in Theorem 11 (i.e. $D$ is a solid max-stable space containing $D∞$ and $Dl = \{y- \mid y ∈ D\}$), $M$ is a subspace of
\[
\{(u, ˜u) ∈ L^1(M) × L^1(M) \mid u \text{ optional}, ˜u \text{ predictable}\}
\]
containing $M∞ := \{(u, ˜u) ∈ L^∞(M) × L^∞(M) \mid u \text{ optional}, ˜u \text{ predictable}\}$ and that, for all $y ∈ D$ and $(u, ˜u) ∈ M$,
\[
E \left[ \int |y|d|u| + ∫ |y-|d| ˜u| \right] < ∞.
\]
In particular, the bilinear form
\[
(y, (u, ˜u)) := E \left[ ∫ ydu + ∫ y-d ˜d\right]
\]
is well-defined on $D × M$. We equip $D$ and $M$ with topologies compatible with this bilinear form. The next example shows how many familiar Banach spaces of adapted càdlàg processes together with their duals fit in our setting.
Example 1. For $p \in (1, \infty)$, let $D^p$ be the space of adapted càdlàg processes whose pathwise supremum belongs to $L^p$. When endowed with the norm $\|y\|_{D^p} = (E\|y\|^p)^{1/p}$, $D^p$ is Banach space whose dual may be identified with

$$\hat{\mathcal{M}}^q = \{(u, \tilde{u}) \in L^q(M) \times L^q(M) \mid u \text{ optional, } \tilde{u} \text{ predictable}\}$$

for $1/p + 1/q = 1$; see, e.g., [13]. These dual pairs evidently satisfy our assumptions as $D$ and $\hat{\mathcal{M}}$. In this setting, $L^p$ can be replaced by the Morse heart of an appropriate Orlicz space; see [2].

Let $D^1$ be the space of adapted càdlàg processes of class $(D)$. When endowed with the norm $\sup_{\tau \in T} E|y_\tau|$, $D^1$ is a Banach space whose dual can be identified with $\hat{\mathcal{M}}^\infty$; see, e.g., [24]. This dual pair satisfies our assumptions as well. This setting extends to appropriate spaces with the so called “Choquet property” [24].

We denote $\hat{\mathcal{M}} = \{u \mid (u, 0) \in \hat{\mathcal{M}}\}$. Recall the definition of $J_{h^\ast}$ in (5).

Lemma 12. We have property 2 in Assumption 3 if and only if $EJ_{h^\ast}(u)$ is finite for some $u \in M$. In this case, for every $y \in D$

$$\sup_{u \in M} \left\{ E \int y du - EJ_{h^\ast}(u) \right\} = F(y), \quad (13)$$

and in particular, $F$ is lower semicontinuous on $D$. We have an analogous result for $EJ_{h^\ast}(\cdot)$ with respect to Property 2 in Assumption 4 and the class $\hat{\mathcal{M}} = \{\tilde{u} \mid (0, \tilde{u}) \in \hat{\mathcal{M}}\}$.

Proof. We only prove the claims for $EJ_{h^\ast}(u)$, since for $EJ_{h^\ast}(u)$ the proof is similar. Assume that Property 2 in Assumption 3 holds. The measure $u$ defined by $du = x d\mu$ is an element of $\mathcal{M}$ and $EJ_{h^\ast}(u) = E \int h^\ast(x) d\mu \leq E \int x d\mu < \infty$. Conversely, let $u \in \mathcal{M}$ be such that $EJ_{h^\ast}(u)$ is finite. Then Property 2 in Assumption 3 holds with $\alpha = (h^\ast(\frac{du}{d\mu}))^+$ and $x = \frac{du}{d\mu}$.

Take $y \in D$. Assume first that $EJ_{h^\ast}$ is finite at the origin. Let $\theta$ be the optional measure on the optional $\sigma$-algebra $\mathcal{O}$ given by $\theta(A) = E \int 1_A d\mu$. Then,

$$\sup_{u \in M} \left\{ E \int y du - EJ_{h^\ast}(u) \right\} \geq \sup_{w \in L^\infty(\Xi, \mathcal{O}, \theta)} \left\{ \int y wd\theta - \int h^\ast(w) d\theta \right\}$$

$$= \int h(y) d\theta$$

$$= EI_h(y), \quad (14)$$

where the first equality follows from [32, Theorem 14.60], since the process identically equal to zero belongs to $L^\infty(\Xi, \mathcal{O}, \theta)$. For a stopping time $\tau$, $h_\tau$ is a normal integrand; see the discussion after Theorem 2 in [18]. Assume that $\theta(\{\tau\}) = 0$. By [32, Theorem 14.60] again, and by (3), we get

$$\sup_{u \in \hat{\mathcal{M}}} \left\{ E \int y du - EJ_{h^\ast}(u) \right\} \geq \sup_{\eta \in L^\infty(\mathcal{F}_\tau)} E[\eta \cdot (h^\ast(\eta))] = E\delta_{S_\tau}(y_\tau). \quad (15)$$
As a consequence, for \( y \) such that the left hand side of (13) is finite, we get \( y \in \mathcal{D}(S) \) by (14) and (15). Conversely, if \( y \) is not an element of \( \mathcal{D}(S) \) then (15) yields that the left hand side of (13) is infinite. Indeed, the set \( \{(\omega, t) \mid S_t(\omega) \cap \{y\} = \emptyset\} \) is optional and we conclude with the optional section theorem.

We have shown that

\[
\sup_{u \in \mathcal{M}} \left\{ E \int y du - EJ_h^*(u) \right\} \geq F(y)
\]

while the opposite inequality follows from Fenchel’s inequality. If \( EJ_h^* \) is not finite at the origin but at \( \bar{u} \), we apply the above to \( u \to EJ_h^*(u + \bar{u}) \).

Assumption 6. \( \hat{F} \) is finite at some point. The predictable measure \( \hat{\mu} \) is purely atomic and, for every \( \epsilon > 0 \), \( y \in \text{dom} \, EI_h \cap \mathcal{D}(S) \) and \( \tilde{y} \in \text{dom} \, EI_{\tilde{h}} \cap \mathcal{D}(\tilde{S}) \) and \((u, \tilde{u}) \in \mathcal{M}\), there exists \( \tilde{y} \in \mathcal{D}(S) \) and \( \hat{y} \in \mathcal{D}(\tilde{S}) \) with

1. \( EI_h(\tilde{y}) + E \int \tilde{y} du \leq EI_h(y) + E \int y du + \epsilon \) and \( \tilde{y} \in \text{dom} \, EI_h \cap \mathcal{D}(\tilde{S}) \),
2. \( EI_{\tilde{h}}(\tilde{y}) + E \int \tilde{y} d\tilde{u} \leq EI_{\tilde{h}}(\hat{y}) + E \int \hat{y} d\tilde{u} + \epsilon \) and \( \tilde{y} \in \text{dom} \, EI_{\tilde{h}} \cap \mathcal{D}(S) \).

The following is our first main result.

**Theorem 13.** Under Assumptions 3, 4 and 6, \( \hat{F} \) is a proper lower semicontinuous convex function on \( \mathcal{D} \),

\[
\hat{F}^*(u, \tilde{u}) = E \left[ J_h^*(u) + J_{\tilde{h}}^*(\tilde{u}) \right]
\]

and \((u, \tilde{u}) \in \partial \hat{F}(y) \) if and only if almost surely,

\[
\begin{align*}
\frac{du}{d\mu} &\in \partial h(y) \quad \mu\text{-a.e.,} \\
\frac{du}{d|u^a|} &\in \partial^a h(y) \quad |u^a|\text{-a.e.,} \\
\frac{d\tilde{u}}{d\tilde{\mu}} &\in \partial h(y_-) \quad \mu\text{-a.e.,} \\
\frac{d\tilde{u}}{d|\tilde{u}^a|} &\in \partial^a h(y_-) \quad |\tilde{u}^a|\text{-a.e.}
\end{align*}
\]

**Proof.** We note first that \( F \) is proper and lsc. Indeed, applying Lemma 12 with \( \mathcal{M} \) and \( \hat{\mathcal{M}} \), we see that \( \hat{F} \) is lsc and it never takes the value \(-\infty\), while finitess at some point is assumed explicitly in Assumption 6. We have

\[
\hat{F}^*(u, \tilde{u}) = \sup_{y \in \mathcal{D}} \{ (y, (u, \tilde{u})) - \hat{F}(y) \}
\]

\[
= -\inf_{y \in \mathcal{D}} \left\{ E \left[ I_h(y) + I_{\tilde{h}}(y_-) - \int y du - \int y_- d\tilde{u} \right] + \delta_{\mathcal{D}(S)}(y) + \delta_{\mathcal{D}(\tilde{S})}(y_-) \right\}.
\]

Let \( \tilde{u} = \tilde{u}^a + \tilde{u}^s \) be the Lebesgue decomposition of \( \tilde{u} \) with respect to \( \tilde{\mu} \) and \( \tilde{A} \) be a predictable set such that

\[
E \int 1_{\tilde{A}^c \cap \tilde{B}} d\tilde{\mu} = E \int 1_{\tilde{A} \cap \tilde{B}} d|\tilde{u}^a| = 0
\]
for any predictable set \( B \); see [15, Theorem 5.15] and [10, Theorem 2.1]. Defining
\[
\tilde{\mu} := |\tilde{u}^*| + \mu
\]
we have
\[
\tilde{h}(y) := \begin{cases} 
\tilde{h}(y) - y \cdot \frac{du^*}{d\tilde{\mu}} & \text{on } \tilde{A}, \\
\delta_{\tilde{S}}(y) - y \cdot \frac{du^*}{d\tilde{\mu}} & \text{on } \tilde{A}^c,
\end{cases}
\]
and \( \tilde{S}_t := \text{cl dom } \tilde{h}_t = \tilde{S}_t \), we have
\[
E \left[ I_{\tilde{h}}(y,-) - \int y_- d\tilde{\mu} \right] + \delta_{\tilde{D}(\tilde{S})}(y,-) = E \int \tilde{h}(y,-) d\tilde{\mu} + \delta_{\tilde{D}(\tilde{S})}(y,-).
\]
Likewise, let \( u = u^a + u^s \) be the Lebesgue decomposition of \( u \) with respect to \( \mu \) and \( A \) an an optional set such that
\[
E \int 1_{A^c \cap B} d\mu = E \int 1_{A \cap B} |u^s| = 0
\]
for any optional set \( B \). Defining \( \hat{\mu} := |u^s| + \mu 
\]
we have
\[
\hat{h}(y) := \begin{cases} 
\hat{h}(y) - y \cdot \frac{du^s}{d\hat{\mu}} & \text{on } A, \\
\delta_{\hat{S}}(y) - y \cdot \frac{du^s}{d\hat{\mu}} & \text{on } A^c,
\end{cases}
\]
and \( \hat{S}_t := \text{cl dom } \hat{h}_t = S_t \), we have that
\[
E \left[ I_{\hat{h}}(y) - \int y d\mu \right] + \delta_{D(S)}(y) = E \int \hat{h}(y) d\hat{\mu} + \delta_{D(S)}(y).
\]
Recalling (3), we have
\[
\inf_{y \in \mathbb{R}^d} \hat{h}_t(y, \omega) = \begin{cases} 
-\hat{h}_t^*((d\hat{u}/d\hat{\mu})_t(\omega), \omega) & \text{if } (\omega, t) \in \tilde{A}, \\
-(\hat{h}_t^*)^\infty((d\hat{u}/d\hat{\mu}^*)_t(\omega), \omega) & \text{otherwise}
\end{cases}
\]
and similarly for \( \tilde{h} \). It is straightforward (although slightly tedious) to verify the assumptions in Theorem 11 for \( \hat{h} \) and \( \hat{h} \), which then gives the conjugate formula.

To prove the subgradient formula, let \( y \in \text{dom } \hat{F} \) and \( (u, \tilde{u}) \in \hat{M} \). By Fenchel’s inequality, almost surely,
\[
\hat{h}(y) + \hat{h}^*(du/d\hat{\mu}) \geq y \cdot (du/d\hat{\mu}) \quad \mu \text{-a.e.},
\]
\[
(\hat{h}^*)^\infty(du/d|u^s|) \geq y \cdot (du/d|u^s|) \quad |u^s| \text{-a.e.},
\]
\[
\tilde{h}(y_-) + \tilde{h}^*(d\tilde{u}/d\tilde{\mu}) \geq y_- \cdot (d\tilde{u}/d\tilde{\mu}) \quad \tilde{\mu} \text{-a.e.},
\]
\[
(\tilde{h}^*)^\infty(d\tilde{u}/d|\tilde{u}^*|) \geq y_- \cdot (d\tilde{u}/d|\tilde{u}^*|) \quad |\tilde{u}^*| \text{-a.e.}
\]
We have \( (u, \tilde{u}) \in \partial \hat{F}(y) \) if and only if \( \hat{F}(y) + \hat{F}^*(u, \tilde{u}) = \langle y, (u, \tilde{u}) \rangle \) which by the conjugate formula, is equivalent to having the above inequalities satisfied as equalities which in turn is equivalent to the stated pointwise subdifferential conditions. \( \square \)
Theorem 14. Assume that $EI_h$ is finite on $\mathcal{D}(S)$, Assumption 3 holds and that

$$\hat{S}_t(\omega) := \text{cl}\{y_{t-} \mid y \in \mathcal{D}(S(\omega))\}$$

defines a predictable set-valued mapping. Then

$$F^*(u, \tilde{u}) = EJ_{h^*}(u) + EJ_{\sigma_{\tilde{u}}}(\tilde{u}).$$

Moreover, $F^*$ is the lower semicontinuous hull of $(u, \tilde{u}) \mapsto EJ_{h^*}(u) + \delta_{\{0\}}(\tilde{u})$.

Proof. Let $\tilde{h} = \delta_{\tilde{S}}$ so that $F = \tilde{F}$. Assumption 4 holds, so by Theorem 13, it suffices to show that Assumption 6 is satisfied. Indeed, the last claim then holds as well, by Lemma 12 and the biconjugate theorem.

If $y \in \mathcal{D}(S)$, then $y_- \in \mathcal{D}(\hat{S})$, and property 1 in Assumption 6 holds. To show property 2, take $\tilde{y} \in \mathcal{D}(\hat{S})$, $\epsilon > 0$ and $(u, \tilde{u}) \in \hat{M}$. Choose $y \in \mathcal{D}(S)$.

Since $\tilde{u}$ is purely atomic and predictable, the process $w$ defined by

$$w_t := \int_{[0,t]} (|\tilde{y}| + |y_+|)d|\tilde{u}|$$

is predictable increasing and purely discontinuous whose jump times belong to the set $A := \{\Delta w_t \neq 0\}$. Defining $A_\nu := \{\Delta w_t \geq 1/\nu\}$ and fixing $\nu$ large enough, we get

$$E\left[\int_{A_\nu^c} (|\tilde{y}| + |y_-|)d|\tilde{u}|\right] < \epsilon/2.$$

Here $A_\nu$ is supported on a union of graphs of an increasing disjoint sequence $(\tau^j)$ of predictable times. Take $\alpha > 0$ such that $E[||\tilde{u}|(A)||] < \alpha$. Let

$$\hat{S}_t(\omega) = \{x \in \mathbb{R}^d \mid |x| \leq |y_t(\omega)| + |\tilde{y}_{t+}(\omega)| + 1\}$$

$$\Gamma^j(\omega) = \{z \in \mathcal{D} \mid z_{\tau^j(\omega)_-} \in \tilde{y}_{\tau^j(\omega)}(\omega) + \frac{\epsilon}{\alpha 2^{j+1}B}\}.$$

We have that $D(S)$ and $D(\hat{S})$ are graph-measurable by Theorem 7. Let $T_j : \Omega \times \mathcal{B}(\mathcal{D}) \to \mathbb{R}$ be defined by $T_j(\omega, z) = z_{\tau^j(\omega)_-} - \tilde{y}_{\tau^j(\omega)}(\omega)$. Then $T_j$ is $\mathcal{F} \otimes \mathcal{B}(\mathcal{D})$-measurable by Lemma 6. Hence, $\Gamma^j$ is graph-measurable since $(\text{gph} \ \Gamma^j) = T_j^{-1}(\frac{\epsilon}{\alpha 2^{j+1}B})$. As a consequence, we see the graph-measurability of the mapping

$$\Gamma(\omega) := D(S) \cap D(\hat{S}) \cap \bigcap_j \Gamma^j.$$

Now we check $\Gamma$ is nonempty-valued. We fix $\omega$. For all $j$ there exists $z^j \in D(S(\omega))$ with $z^j_{\tau^j(\omega)_-} \in \tilde{y}_{\tau^j(\omega)}(\omega) + \frac{\epsilon}{\alpha 2^{j+1}B}$ by the definition of $\hat{S}$. For $\delta > 0$ let

$$z := \sum_j z^j 1_{[\tau^j(\omega)-\delta, \tau^j(\omega))]} + y(\omega)(1 - 1_{\bigcup_j [\tau^j(\omega)-\delta, \tau^j(\omega))]}).$$
The series defining $z$ is a finite sum since $A_\nu$ is $\omega$-wise finite. Thus, $z$ is a càdlàg function and it is clear that it is a selection of $\hat{S}(\omega)$. We choose $\delta$ in such a way that $z_\nu^j \in z_{\tau^j(\omega)}^j + \frac{\gamma_j}{\tau^j-\delta}B$ and $\hat{y}_\nu(\omega) \in \hat{y}_{\tau^j(\omega)}^j(\omega) + \frac{\gamma_j}{\tau^j-\delta}B$ for $t \in [\tau^j(\omega)-\delta, \tau^j(\omega)]$. Then $z \in D(\hat{S}(\omega))$. It is also clear that $z \in \Gamma^j(\omega)$. Thus, $\Gamma(\omega)$ is nonempty.

By [8, Theorem III.22], there exists a càdlàg selection $z$ of $\Gamma$ which seen as a process is measurable although possibly non-adapted. The bound

$$|z| \leq |y(\omega)| + |\hat{y}_+ (\omega)| + 1$$

implies that $\sigma z$ exists and belongs to $D$. The process $z$ satisfies $(\sigma z)_- = p(z_-)$, by [24, Lemma 4]. Moreover, by [18, Corollary 4], $\sigma z \in D(S)$. Let $(\sigma^j, \nu)$ be an announcing sequence for $\tau^j$, where we may assume that $\sigma^j+1, \nu \geq \tau^j$ for every $j$ and $\nu$. Defining

$$\hat{y}^\nu = \sum_j \sigma z_\nu^{\sigma^j, \nu, \tau^j} + y_\nu(\bigcup_j [\sigma^j, \nu, \tau^j))^{ev},$$

we have that $\hat{y}^\nu \in D(S)$ and hence $\hat{y}^\nu \in \text{dom} EI_h$. For $\nu$ large enough,

$$E \int \hat{y}^\nu d\bar{u} \leq E \int \hat{y} d\bar{u} + \epsilon$$

which shows property 2 in Assumption 6.

The following result is an immediate corollary of Theorem 14.

**Corollary 15.** Assume that $S$ is an optional set-valued mapping with

$$S_t(\omega) = \text{cl}\{y_t \mid y \in D(S)(\omega)\}$$

and that $\bar{S}$ defined by

$$\bar{S}_t(\omega) := \text{cl}\{y_t^- \mid y \in D(S)(\omega)\}$$

is predictable. Then $D(S)$ is closed and

$$\sigma_{D(S)}(u, \bar{u}) = EJ_{\sigma_S}(u) + EJ_{\sigma_{\bar{S}}}(\bar{u})$$

as soon as $D(S) \neq \emptyset$. Moreover, $\sigma_{D(S)}$ is the lower semicontinuous hull of

$$(u, \bar{u}) \rightarrow EJ_\sigma(u) + \delta_{[0]}(\bar{u}).$$

## 7 Càdlàg selections of set-valued mappings

One of the conditions in our main results above is

$$S_t = \text{cl}\{y_t \mid y \in D(S)\}$$
which is a sort of Michael representation (see [32]) of $S$ consisting of càdlàg selections. In this section we analyze this condition in terms of standard continuity properties of $S_t$ as a function of $t$. These results are of independent interest in set-valued analysis. Our results also lead to sufficient conditions for the predictability of $\tilde{S}$ in Theorem 14.

Recall that a function is right-continuous (càd) in the usual sense if and only if it is continuous with respect to the topology $\tau_r$ generated by the right-open intervals $\{(s, t) \mid s < t\}$. A set-valued mapping $\Gamma : [0, T] \rightrightarrows \mathbb{R}^d$ is said to be right-inner semicontinuous (right-isc) if $\Gamma^{-1}(O)$ is $\tau_r$-open for any open $O \subseteq \mathbb{R}^d$. Left-inner semicontinuity (left-isc) is defined analogously using the topology $\tau_l$ generated by the left-open intervals $\{(s, t] \mid s < t\}$.

The mapping $\Gamma$ is said to be right-outer semicontinuous (right-osc) if its graph is closed in the product topology of $\tau_r$ and the usual topology on $\mathbb{R}^d$. The mapping $\Gamma$ is right-continuous (càd) if it is both right-isc and right-osc. Left-outer semicontinuous (left-osc) and left-continuous (càg) mappings are defined analogously. We say that $\Gamma$ has limits from the left (làg) if, for all $t$,

$$\liminf_{s \uparrow t} \Gamma_s = \limsup_{s \uparrow t} \Gamma_s,$$

where the limits are in the sense of [32, Section 5.B] and are taken along strictly increasing sequences. Having limits from the right (làd) is defined analogously. A mapping $\Gamma$ is càdlàg (resp. càglàd) if it is both càd and làg (both càg and làd). Recall that a convex-valued $\Gamma$ is solid if $\text{int } \Gamma_t \neq \emptyset$ for all $t$. For any mapping $\Gamma$ we let $\bar{\Gamma}_0 := \{0\}$ and for $t > 0$

$$\bar{\Gamma}_t := \liminf_{s \uparrow t} \Gamma_s.$$

In the following theorem, the distance of $x$ to $\Gamma_t$ is defined, as usual, by

$$d(x, \Gamma_t) = \inf_{x' \in \Gamma_t} d(x, x'),$$

where the distance of two points is given by the euclidean metric.

**Theorem 16.** Let $\Gamma : [0, T] \rightrightarrows \mathbb{R}^d$ be a càdlàg nonempty convex-valued mapping. For every $x \in \mathbb{R}^d$, the function $y$ defined by

$$y_t = \arg\min_{x' \in \Gamma_t} d(x, x')$$

satisfies $y \in D(\Gamma)$ and

$$y_{t-} = \arg\min_{x' \in \Gamma_t} d(x, x').$$

In particular,

$$\Gamma_t = \text{cl}\{y_t \mid y \in D(\Gamma)\}$$

and $\bar{\Gamma}$ is càglàd nonempty convex-valued with

$$\bar{\Gamma}_t = \text{cl}\{y_{t-} \mid y \in D(\Gamma)\}.$$
Proof. By strict convexity of the distance mapping, the argmin in the definition of $y$ is single-valued. By [32, Proposition 4.9], $y$ is càdlàg. On the other hand, for every strictly increasing $t' \nearrow t$, $\Gamma_{t'} \to \bar{\Gamma}_t$, so $y$ is càdlàg, by [32, Proposition 4.9] again.

Next we show

$$y_t^- = \arg\min_{x' \in \Gamma_t} d(x, x').$$

Since $\Gamma$ is càdlàg, we get $y_t^- \in D_t(\bar{\Gamma})$, so the inequality $d(x, \bar{\Gamma}_t) \leq d(x, y_t^-)$ is trivial.

For the other direction, assume for a contradiction the existence of $\bar{\Gamma}$ is right-isc, there is, for every $u$ be small enough so that $y_t^- \in \Gamma_u$ for $u \in [\bar{t}, t + \delta]$ since $\Gamma$ is right-isc and solid. Indeed, let $\bar{y}_i$ be a finite set of points in int $\Gamma_{\bar{t}}$ such that $y_i^\nu$ belongs to the interior of the convex hull $\text{co}\{\bar{y}_i\}$. Let $\epsilon > 0$ be small enough so that $y_i^\nu \in \text{co}\{v_i^\nu\}$ whenever, for every $i$, $v_i^\nu \in (\bar{y}_i + \epsilon \mathbb{B})$. Since $\Gamma$ is right-isc, there is, for every $i$, $u_i^\nu > \bar{t}$ such that $\Gamma_u \cap (\bar{y}_i + \epsilon \mathbb{B}) \neq \emptyset$ for every $u \in [\bar{t}, u_i^\nu]$. Denoting $\bar{u} = \min u_i^\nu$, we have, by convexity of $\Gamma$, that $y_i^\nu \in \Gamma_u$ for every $u \in [\bar{t}, \bar{u}]$.

Now assume $\bar{t} > 0$ and take $y_i^\nu \in \text{int} \bar{\Gamma}_{\bar{t}}$. We now show the existence of $s < \bar{t}$ such that

$$y_i^s \in \Gamma_u$$

for every $u \in (s, \bar{t}]$. Assume for a contradiction the existence of $t'' \nearrow \bar{t}$ such that $y_i^s \notin \Gamma_{t''}$. Let $\bar{y}_i \in \text{int} \bar{\Gamma}_{\bar{t}}$ be $d + 1$ points and $\epsilon > 0$ such that for any points $\bar{y}_i + \epsilon \mathbb{B}$, $\bar{y}_i \in \Gamma_{\bar{t}}$ and $y_i^s \in \text{co}\{\bar{y}_i\}$. By the definition of $\bar{\Gamma}$ as a left-limit, there exists $\nu_0 \in \mathbb{N}$ such that for all $\nu > \nu_0$ there exists $y_i^\nu \in \Gamma_{t''}$ with $y_i^\nu \in \bar{y}_i + \epsilon \mathbb{B}$ for $i = 1, \ldots, d + 1$. Then, $y_i^s \in \text{co}\{y_i^\nu\}$ and this last set is included in $\Gamma_{t''}$ by convexity. Then, $y_i^s \in \Gamma_{t''}$, a contradiction.

Theorem 17. Let $\Gamma : [0, T] \to \mathbb{R}^d$ be a closed convex-valued solid mapping such that $\bar{\Gamma}$ is solid. We have

$$\Gamma_t = \text{cl}\{y_t^- \mid y \in D(\Gamma)\}$$

if and only if $\Gamma$ is right-isc. In this case, $\bar{\Gamma}$ is left- isc and

$$\bar{\Gamma}_t = \text{cl}\{y_t^- \mid y \in D(\Gamma)\}.$$
We have proved that for any $t \in [0, T]$ there exists $\delta > 0$ and $y', y'' \in \mathbb{R}^d$ such that $y_u = y_1^{t_1} u + y_1^{t_2} u$ is a càdlàg selection of $\Gamma$ in the interval $[t - \delta, t + \delta] \cap [0, T]$. Now we can paste together these local selections using partitions of unity as in [19, Theorem 3.2’]. We have proved that $D(\Gamma) \neq \emptyset$. Now $\Gamma_t = \{ y_t \mid y \in D(\Gamma) \}$ is an easy consequence to (16) and $D(\Gamma) \neq \emptyset$. We have shown the sufficiency.

To prove $\Gamma_t = \{ y_t \mid y \in D(\Gamma) \}$, the inclusion $\supseteq$ is clear. Now take $y' \in \text{int} \Gamma_t$ and $s < t$ as in (17) and $y \in D(\Gamma)$. Defining

$$
\begin{cases}
y' & t \in [s, t) \\
y_t & \text{otherwise},
\end{cases}
$$

we get the inclusion $\subseteq$. From the representation $\Gamma_t = \{ y_t \mid y \in D(\Gamma) \}$ it follows that $\Gamma$ is left-isc.

The following preparatory lemma characterizes preimages of $\Gamma_t$. For a set $A$, $\text{cl}A$ denotes the limit points from the left of $A$. Thus, for $t \in \text{cl}A$ and $s < t$, $(s, t) \cap A \neq \emptyset$.

**Lemma 18.** For any closed set $A \subset \mathbb{R}^d$,

$$
\bar{\Gamma}^{-1}(A) = \bigcup_{N \in \mathbb{N}} \bigcap_{n \in \mathbb{N}} \bigcup_{a \in \mathbb{Q}^d \cap NB} \left[ \text{cl}C_{\nu, n, a}/\text{cl}C^c_{\nu, n, a} \right],
$$

where $C_{\nu, n, a} := \Gamma^{-1}(A \cap (a + n^{-1}B) + \nu^{-1}B)$.

**Proof.** To prove the inclusion $\supseteq$, let

$$
t \in \bigcap_{N \in \mathbb{N}} \bigcap_{n \in \mathbb{N}} \bigcup_{a \in \mathbb{Q}^d \cap NB} \left[ \text{cl}C_{\nu, n, a}/\text{cl}C^c_{\nu, n, a} \right]
$$

for some fixed $N$. For all $\nu, n$ there exists $a^{\nu, n} \in \mathbb{Q}^d \cap NB$ such that $t \in \text{cl}C_{\nu, n, a^{\nu, n}}/\text{cl}C^c_{\nu, n, a^{\nu, n}}$. By compactness, there exists a subsequence $n_j \to \infty$ and $a^* \in \mathbb{R}^d$ such that $a_j := a^{n_j, n_j} \to a^*$. There exists $s_j < t$ such that $(s_j, t] \subset C_{a_j, j, a_j}$ since $t \in \text{cl}C_{a_j, j, a_j}/\text{cl}C^c_{a_j, j, a_j}$. As a consequence, for any sequence $t_k \not\to t$, there exists $y_k \in \Gamma_{t_k}$ with $y_k \in A \cap (a_{jk} + n^{-1}_{jk}B) + n^{-1}_{jk}B$ for an appropriate subsequence $j_k \to \infty$. Then $y_k \to a^*$ and $a^* \in A = A$. Thus $t \in \bar{\Gamma}^{-1}(A)$.

Now we show $\subseteq$. Take $t > 0$ in $\bar{\Gamma}^{-1}(A)$ and $a^* \in A \cap \bar{\Gamma}_t \cap NB$. Fix $n$ and $\nu$. There exists $a \in \mathbb{Q}^d \cap NB$ such that $a^* \in A \cap (a + n^{-1}B)$. Then $t \in \bar{\Gamma}^{-1}(A \cap (a + n^{-1}B))$ and this yields the existence of $s < t$ such that $(s, t) \subset C_{\nu, n, a}$. Thus $t \in \text{cl}C_{\nu, n, a}/\text{cl}C^c_{\nu, n, a}$. 

We finish the section with the stochastic setting. For $\Gamma : \Omega \times [0, T] \to \mathbb{R}^d$, $\bar{\Gamma}$ is defined pathwise as above, and for $C \subset \Omega \times [0, T]$, $\text{cl} C$ denotes the pathwise limit points from the left.
Lemma 19. Let $A$ be a $\mathcal{F} \otimes \mathcal{B}([0,T])$-measurable set. Then $\bar{\text{cl}}A$ is measurable.

Proof. The process $D_s(\omega) = \inf\{t > s \mid (t, \omega) \in A\}$ is measurable; see the proof of [12, Theorem IV.32]. Since

$$
\bar{\text{cl}}A = \bigcap_{\nu \in \mathbb{N}} \{(s, \omega) \mid D_{(s-\frac{1}{2})\nu}(\omega) < s\},
$$

$\bar{\text{cl}}A$ is thus measurable. $\Box$

Lemma 20. Let $\Gamma : \Omega \times [0,T] \rightarrow \mathbb{R}^d$ be a measurable mapping. Then $\bar{\Gamma}$ is measurable. If $\Gamma$ is progressively measurable, then $\bar{\Gamma}$ is predictable.

Proof. For a closed set $A \subset \mathbb{R}^d$, Lemma 18 gives

$$
\bar{\Gamma}^{-1}(A) = \bigcup_{N \in \mathbb{N}} \bigcap_{n \in \mathbb{N}} \bigcap_{a \in \mathbb{Q}^d \cap NB} \left[\bar{\text{cl}}C_{\nu,n,a}/\bar{\text{cl}}C_{\nu,n,a}\right],
$$

where the sets on the right hand side are measurable by Lemma 19. Hence, $\bar{\Gamma}^{-1}(A)$ is a measurable set. Then, $\bar{\Gamma}$ is a measurable mapping by [32, Theorem 14.3 (b)]. When $\Gamma$ is progressively measurable, the sets on the right are predictable by [12, Theorem IV.89]. $\Box$

Combining results of this section, we get sufficient conditions for the predictability assumption in Theorem 14.

Theorem 21. Let $h$ be an optional normal integrand and $S_t(\omega) = \text{cl} \text{dom} h_t(\omega)$. Then the set-valued mapping defined by

$$
\tilde{S}_t(\omega) = \text{cl}\{y_t - \mid y \in D(S)(\omega)\}
$$

is predictable and coincides with $\bar{S}$ under either of the following conditions:

1. $S$ is càdlàg,
2. $S$ and $\tilde{S}$ are solid.

Proof. By Theorems 16 and 17, $\bar{S} = \tilde{S}$ under either condition, so the claim follows from Lemma 20. $\Box$

8 Applications

In this section we demonstrate how Corollary 15 together with Theorem 21 lead to well-known models in mathematical finance. The article [23] applies our results to optimal stopping while partial hedging of American options will be studied in a forthcoming article by the authors. Further applications to finance and to singular stochastic control will be presented elsewhere. We assume throughout the section the conditions of Theorem 14 for $D$. That is, $D$, is a solid max-stable space of adapted càdlàg processes of class $(D)$ containing $D^\infty$. 23
For a stochastic process $b$, 
\[ \bar{b}_t := \limsup_{s \uparrow t} b_s \]
is its \textit{left upper semicontinuous regularization}. A process is \textit{left-usc} if $b \geq \bar{b}$. By [12, Theorem IV.90], $\bar{b}$ is predictable whenever $b$ is optional. For a càdlàg $b$, $\bar{b}$ is càglàd and $\bar{b} = b_-$. Likewise, $b$ is said to be \textit{right-use} if $b \geq \underline{b}$, where 
\[ \underline{b}_t := \limsup_{s \downarrow t} b_s \]
is the \textit{right-upper semicontinuous regularization} of $b$. These regularizations appear in the context of optimal stopping as properties of reward processes, e.g., in [5, 14], or more recently in and [3, 23].

\textbf{Example 2.} Let $b$ be optional and right-usc dominated by some $\tilde{y} \in \mathcal{D}$ and 
\[ S_b(\omega) := \{ y \in \mathbb{R} | y \geq b_t(\omega) \}. \]
Then $\mathcal{D}(S)$ is nonempty closed convex with 
\[ \sigma_{\mathcal{D}(S)}(u, \tilde{u}) = E \left[ \int bdu + \int \tilde{b}d\tilde{u} \right] + \delta_{\mathcal{M}_-}(u, \tilde{u}). \]
Moreover, $\sigma_{\mathcal{D}(S)}$ is the lower semicontinuous hull of 
\[ (u, \tilde{u}) \mapsto E \int bdu + \delta_{\mathcal{M}_- \times \{0\}}(u, \tilde{u}). \]

\textbf{Proof.} Let 
\[ \tilde{S}_b(\omega) := \{ y \in \mathbb{R} | y \geq \tilde{b}_t(\omega) \}. \]
The result follows from Corollary 15 once we verify that $\tilde{S}_b = \text{cl}\{y_t- | y \in D(S)\}$. In the following argument, $\omega$ is fixed.

If $y \in D(S)$, then $y_- = \tilde{y} \geq \bar{b}$, so $\tilde{S}_b \supseteq \text{cl}\{y_t- | y \in D(S)\}$. To prove the converse, it suffices to show that $\tilde{y} := \tilde{b}_t \in \text{cl}\{y_t- | y \in D(S)\}$ for a given $\tilde{t} \in (0, T]$. Fix $\epsilon > 0$. Since $\bar{b}$ is left-usc, the set $\{s | \bar{b}_s < \tilde{y} + \epsilon\}$ contains a $\tau_\epsilon$-neighborhood of $\tilde{t}$. Thus $b_s < \tilde{y} + \epsilon$ for all $s \in [u, \tilde{t})$ for some $u < \tilde{t}$. We have $z \geq b$ and $z_{t-} = \tilde{b}_t + \epsilon$, where 
\[ z_t := (\tilde{y} + \epsilon)[1_{[u, \tilde{t})}] + \tilde{y}_t[1_{[u, \tilde{t})}]c(t). \]
Since $\epsilon > 0$ was arbitrary, $\tilde{b}_t \in \text{cl}\{y_t- | y \in D(S)\}$. \hfill \Box

Let $a$ be a stochastic process and $a_t := \lim\inf_{s \uparrow t} a_s$. By construction, $a$ is left-lsc. By [12, Theorem IV.90], $a$ is predictable whenever $a$ is predictable. In the next example, the set-valued mapping $S$ describes \textit{bid-ask spreads} in proportional transaction cost models; see [9] and references therein. Our example allows for general bid and ask prices given by right-usc and right-lsc processes.
Example 3. [Bid-ask spreads] Let \( b \) be optional and right-usc and \( a \) be optional and right-lsc such that there exists \( \bar{y} \in \mathcal{D} \) with \( b < \bar{y} < a \) and \( \bar{b} < \bar{y} < a \) a.s.e. Let

\[
S_t(\omega) := \{ y \in \mathbb{R} \mid b_t(\omega) \leq y \leq a_t(\omega) \}
\]

Then \( \mathcal{D}(S) \) is nonempty closed convex with

\[
\sigma_{\mathcal{D}(S)}(u, \tilde{u}) = \mathbb{E} \left[ \int adu^+ - \int bdu^- + \int \tilde{adu}^- - \int \tilde{bdu}^- \right].
\]

Moreover, \( \sigma_{\mathcal{D}(S)} \) is the lower semicontinuous hull of

\[
(u, \tilde{u}) \mapsto \mathbb{E} \left[ \int adu^+ - \int bdu^- \right] + \delta_{\{0\}}(\tilde{u}).
\]

Proof. Let

\[
\tilde{S}_t(\omega) := \{ y \in \mathbb{R} \mid \bar{b}_t(\omega) \leq y \leq \bar{a}_t(\omega) \}.
\]

The result follows from Corollary 15 provided that \( \tilde{S}_t = \text{cl} \{ \bar{y}_t \mid y \in \mathcal{D}(S) \} \). To this end, one may proceed as in Example 2 using \( \bar{y} \) as a bound.

In the next example, \( \mathcal{C} \) describes the set of self-financing portfolio processes in a currency market of \( d \) different currencies, and the martingales \( y \in \mathcal{D} \) with \( y \in \mathcal{D}(S) \) and \( y^- \in \mathcal{D}(\tilde{S}) \) are consistent price systems. We refer to [17] for a detailed treatment of the subject. The claims in the example follow immediately from Theorem 21 and Corollary 15.

Example 4 (Currency markets). Assume that \( S \) is an optional right-lsc solid convex cone-valued mapping and that \( \tilde{S} \) is solid. Then \( \mathcal{D}(S) \) is a closed convex cone and its polar cone is

\[
\mathcal{C} = \{ (u, \tilde{u}) \in \mathcal{M} \mid (du/d|u|)_t \in S^*_t, (d\tilde{u}/d|\tilde{u}|)_t \in \tilde{S}^*_t \},
\]

where \( S^*_t = \{ x \mid x \cdot y \leq 0 \ \forall \ y \in S_t(\omega) \} \).

We finish this section by showing that the assumptions in the above currency market model are satisfied by the general model in [17, Section 3.6.6].

Example 5 (Campi-Schachermayer model). Let \( G \) (resp. \( \tilde{G} \)) be an optional (resp. predictable) closed convex cone-valued mapping. We assume

- “Efficient friction”: \( G_t \cap (-G_t) = \{0\} \) and \( \tilde{G}_t \cap (-\tilde{G}_t) = \{0\} \) for all \( t \);
- “Regularity hypotheses”: \( G_{t,t^+} = G_t \) and \( G_{t-,t} = \tilde{G}_t \), where

\[
G_{s,t}(\omega) := \text{cl cone} \{ G_r(\omega) \mid r \in [s,t) \},
\]

\[
G_{s,t}(\omega) := \bigcap_{\epsilon > 0} G_{s,t+\epsilon}(\omega),
\]

\[
G_{s-,t}(\omega) := \bigcap_{\epsilon > 0} G_{s-\epsilon,t}(\omega),
\]

and the last is defined as \( G_{0,t}(\omega) \) for \( s = 0 \).
Then $S = G^*$ is right-isc solid-valued and $\tilde{S} = G^*$ is solid.

Proof. The regularity assumptions $G_{t, t+} = G_t$ and $G_{t-, t} = \tilde{G}_t$ imply that $G$ is right-osc and that $\tilde{G}$ is left-osc; we show the latter, the former being simpler. Let $t', y' \in \mathcal{G}_{t'}$ with $y' \to y$. We need to show $y \in \mathcal{G}_t$ to which end we may assume that $t' < t$ for each $\nu$, otherwise the claim is trivial. For any $\epsilon > 0$, $t' \in (t - \epsilon, t)$ for $\nu$ large enough. Since $y_{t'} \in \mathcal{G}_{t'} = G_{t'}$, there are sequences $t'_k \uparrow t'$ and $y_{t'_k} \to y_{t'}$ with $t'_k > t - \epsilon$ and $y_{t'_k} \in G_{t'_k}$. By diagonalization, there exists $k(\nu)$ such that $t'_{k(\nu)} \uparrow t$ and $y_{t'_{k(\nu)}} \to y$. Thus $y \in G_{t-\epsilon, t}$ and, since $\epsilon > 0$ was arbitrary, $y \in G_{t-\epsilon, t}$.

By [32, Corollary 11.35], $G^*$ and $\tilde{G}^*$ are thus right- and left-isc, respectively. While $G_{t-, t} = \tilde{G}_t$ then means that $\tilde{G}^* = G_t^*$. By [29, Corollary 13.4.2], a convex cone $K$ is solid if and only if $K \cap (-K) = \{0\}$, so the efficient friction assumption implies that $G^*$ and $\tilde{G}^*$ are solid-valued. 

9 Appendix

The following extends [21, Lemma 4.2] that was formulated for bounded $w$.

Lemma 22 (Jensen’s inequality). Assume that $h$ is an optional convex normal integrand, $\mu$ is an optional nonnegative random measure,

$$h(x) \geq x \cdot v - \alpha$$

for some optional $v$ and nonnegative $\alpha$ such that $\int |v|d\mu$ and $\int \alpha d\mu$ are integrable, and that $w$ is a raw measurable process with $E \int |w||v|d\mu < \infty$. If $w$ has an optional projection, then

$$EI_h(w) \geq EI_h(\circ w).$$

If $h$, $\mu$ and $v$ are predictable and $w$ has a predictable projection, then

$$EI_h(w) \geq EI_h(p w).$$

Proof. Let $\hat{\mu} \ll \mu$ be defined by $d\hat{\mu}/d\mu = \beta := \circ(1/(1 + \int d\mu))$. Then $\hat{\mu}(A) = E \int 1_A d\hat{\mu}$ defines an optional bounded measure on $\Omega \times [0, T]$. Moreover, $EI_h(w) = E \int \hat{h}(w)d\hat{\mu}$, where $\hat{h}(w) = h(w)/\beta$ is an optional convex normal integrand. We have

$$\hat{h}^*(v) = h^*(\beta v)/\beta,$$

so the lower bound implies that $E \int \hat{h}^*(v/\beta)d\hat{\mu}$ is finite. Thus we may apply the interchange of integration and minimization on $(\Omega \times [0, T], \mathcal{O}, \hat{\eta})$ and on
\( (\Omega \times [0,T], \mathcal{F} \otimes B([0,T]), \hat{\eta}) \) (see [32, Theorem 14.60]) to get

\[
EI_h(\omega w) = E \int \hat{h}(\omega w) d\hat{\mu} \\
= \sup_{v \in L^1(\Omega \times [0,T], \mathcal{O}, \hat{\eta})} E \int [\omega w \cdot v - \hat{h}^*(v)] d\hat{\mu} \\
\leq \sup_{v \in L^1(\Omega \times [0,T], \mathcal{F} \otimes B([0,T]), \hat{\eta})} E \int [w \cdot v - \hat{h}^*(v)] d\hat{\mu} \\
= E \int \hat{h}(w) d\hat{\mu} \\
= EI_h(w).
\]

The predictable case is proved similarly.

The next lemma was used in the proof of Theorem 11.

**Lemma 23.** Let \( \tau \) be a predictable time such that \( [\tau] \subset (m, m + 1] \) for a given \( m \in \{0, 1, \ldots\} \). Let \( \{\epsilon_n\}_{n \in \mathbb{N}} \) be a decreasing sequence of positive real numbers converging to zero. Then, there exists a nondecreasing sequence of stopping times \( \{\sigma^n\}_{n \in \mathbb{N}} \) converging to \( \tau \) with \( [\sigma^n] \subset (m, m + 1) \) for every \( n \), and with the following properties: \( \{\tau < \infty\} \subset \{\sigma^n < \tau\} \) and

\[
|P(\{\sigma^n = \infty\}) - P(\{\tau = \infty\})| \leq \epsilon_n.
\]

**Proof.** By [11, Theorem IV.11], there exists an announcing sequence \( \{\tau^n\}_{n \in \mathbb{N}} \) of \( \tau \) with graphs on \( (m, m + 1) \) and \( \{\tau < \infty\} \subset \{\tau^n < \tau\} \). Note that

\[
\{\tau = \infty\} = \bigcup_{n \in \mathbb{N}} \{\tau^n > m + 1\}.
\]

Then, for any sequence \( \{\epsilon_n\}_{n \in \mathbb{N}} \) decreasing to zero, we might assume, by taking a subsequence if necessary, that

\[
|P(\{\tau^n > m + 1\}) - P(\{\tau = \infty\})| < \epsilon_n.
\]

For \( n \in \mathbb{N} \), let \( \sigma^n \) be the stopping time \( \tau^n_{\{\tau^n \leq m + 1\}} \). The sequence \( \{\sigma^n\}_{n \in \mathbb{N}} \) is non-decreasing and converges to \( \tau \). Moreover

\[
P(\sigma^n = \infty) = P(\tau^n > m + 1) \geq P(\tau = \infty) - \epsilon_n,
\]

which proves the claim. 

**Acknowledgements** Erick Treviño gratefully acknowledges financial support from Alexander von Humboldt Foundation while visiting the Technical University of Berlin.
References


