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Dual solutions in convex stochastic optimization

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Abstract

This paper studies duality and optimality conditions for general convex stochastic optimization problems. The main result gives sufficient conditions for the absence of a duality gap and the existence of dual solutions in a locally convex space of random variables. It implies, in particular, the necessity of scenario-wise optimality conditions that are behind many fundamental results in operations research, stochastic optimal control and financial mathematics. Our analysis builds on the theory of Fréchet spaces of random variables whose topological dual can be identified with the direct sum of another space of random variables and a space of singular functionals. The results are illustrated by deriving sufficient and necessary optimality conditions for several more specific problem classes. We obtain significant extensions to earlier models e.g. on stochastic optimal control, portfolio optimization and mathematical programming.

Keywords. Stochastic programming, convexity, duality, optimality conditions

AMS subject classification codes. 46N10, 90C15, 90C46, 93E20, 91G80

1 Introduction

Given a probability space (Ω, \mathcal{F}, P) with a filtration $(\mathcal{F}_t)_{t=0}^T$ (an increasing sequence of sub- σ -algebras of \mathcal{F}), consider the problem

$$\text{minimize} \quad Ef(x, \bar{u}) \quad \text{over } x \in \mathcal{N} \quad (SP)$$

where \mathcal{N} is a linear space of stochastic processes $x = (x_t)_{t=0}^T$ adapted to $(\mathcal{F}_t)_{t=0}^T$ (i.e., x_t is \mathcal{F}_t -measurable) and \bar{u} is a \mathbb{R}^m -valued random variable. We assume that x_t takes values in a Euclidean space \mathbb{R}^{n_t} so the process $x = (x_t)_{t=0}^T$ takes values in \mathbb{R}^n where $n := n_0 + \dots + n_T$. The objective is defined on the space

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$L^0(\Omega, \mathcal{F}, P; \mathbb{R}^n \times \mathbb{R}^m)$ of \mathcal{F} -measurable $\mathbb{R}^n \times \mathbb{R}^m$ -valued random variables (x, u) by

$$Ef(x, u) := \int_{\Omega} f(x(\omega), u(\omega), \omega) dP(\omega),$$

where f is a *convex normal integrand* on $\mathbb{R}^n \times \mathbb{R}^m \times \Omega$, i.e. $f(\cdot, \omega)$ is a closed convex function for every $\omega \in \Omega$ and $\omega \mapsto \text{epi } f(\cdot, \omega)$ is an \mathcal{F} -measurable set-valued mapping; see [27, Chapter 14]. Here and in what follows, we define the integral of an extended real-valued random variable as $+\infty$ unless its positive part is integrable. By [27, Proposition 14.28], the function $\omega \mapsto f(x(\omega), \bar{u}(\omega), \omega)$ is measurable for any measurable x and \bar{u} so the integral functional Ef is a well-defined convex function on $L^0(\Omega, \mathcal{F}, P; \mathbb{R}^n \times \mathbb{R}^m)$.

Problems of the form (SP) unify and extend various more specific models in operations research, engineering and economics and they have been extensively studied since they first appeared in [23]; see e.g. [25, 26, 18, 31, 11, 4, 3] and the references there. Recently, the theory has been extended to beyond optimization problems to stochastic variational inequalities in [28]. Like [23, 18, 11, 3], this paper studies (SP) through convex duality. We follow the recent approach of [15], which yields explicit dual problems and allows for general adapted strategies x without any integrability restrictions. This is important e.g. in common models of financial mathematics where it may be impossible to find optimal trading strategies in an integrable space of stochastic processes; see [15] and its references.

The purpose of [15] was to lay out the general duality framework and to give optimality conditions in terms of dual solutions. However, it left open the question of the existence of dual solutions. The present paper gives sufficient conditions for the existence. Much like [23, 25], our arguments combine general functional analytic techniques with the special structure of subdifferentials of convex integral functionals originally developed in [20]. Our existence results cover more general problem classes than those studied in [23, 25] in terms of both the dualizing parameterizations and the underlying function spaces.

The analysis of [15] is based on applying the conjugate duality framework of [21] to the parametric optimization problem

$$\begin{aligned} & \text{minimize} && Ef(x, \bar{u}) && \text{over } x \in \mathcal{X} \\ & \text{subject to} && x - z \in \mathcal{X}_a \end{aligned} \tag{SP_{\mathcal{X}}}$$

where $\mathcal{X} \subset L^0(\Omega, \mathcal{F}, P; \mathbb{R}^n)$ is a locally convex vector space of random variables,

$$\mathcal{X}_a := \mathcal{X} \cap \mathcal{N}$$

and z and \bar{u} are the parameters. We assume that $z \in \mathcal{X}$ and that \bar{u} belongs to another locally convex space $\mathcal{U} \subset L^0(\Omega, \mathcal{F}, P; \mathbb{R}^m)$ of random variables. Following [15], we apply the conjugate duality framework to the extended real-valued convex function F on $\mathcal{X} \times \mathcal{X} \times \mathcal{U}$ defined by

$$F(x, z, u) := Ef(x, u) + \delta_{\mathcal{N}}(x - z),$$

where $\delta_{\mathcal{N}}$ is the *indicator function* of \mathcal{N} taking the value 0 on \mathcal{N} and $+\infty$ elsewhere. The associated optimum value function is denoted by

$$\varphi(z, u) := \inf_{x \in \mathcal{X}} \{Ef(x, u) \mid x - z \in \mathcal{N}\}.$$

We assume that \mathcal{X} and \mathcal{U} are in separating duality with spaces of random variables $\mathcal{V} \subset L^0(\Omega, \mathcal{F}, P; \mathbb{R}^n)$ and $\mathcal{Y} \subset L^0(\Omega, \mathcal{F}, P; \mathbb{R}^m)$, respectively. The bilinear forms are the usual ones, i.e.

$$\langle x, v \rangle := E[x \cdot v] \quad \forall x \in \mathcal{X}, v \in \mathcal{V}$$

and

$$\langle u, y \rangle := E[u \cdot y] \quad \forall u \in \mathcal{U}, y \in \mathcal{Y}.$$

It is assumed that all the spaces are *decomposable* and *solid*. Decomposability of \mathcal{X} means that

$$1_A x + 1_{\Omega \setminus A} x' \in \mathcal{X}$$

for every $x \in \mathcal{X}$, $x' \in L^\infty(\Omega, \mathcal{F}, P; \mathbb{R}^n)$ and $A \in \mathcal{F}$. Here $1_A : \Omega \rightarrow \mathbb{R}$ is the function taking the value 1 on A and 0 elsewhere. Solidity means that if $\bar{x} \in \mathcal{X}$ and $x \in L^0$ are such that $|x^i| \leq |\bar{x}^i|$ almost surely for every $i = 1, \dots, m$, then $x \in \mathcal{X}$. Similarly for \mathcal{U} , \mathcal{V} and \mathcal{Y} . Here and in what follows, we will simply write L^0 instead of $L^0(\Omega, \mathcal{F}, P; \mathbb{R}^n)$, when the underlying spaces are clear from the context.

As soon as $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, the conjugate of F can be expressed as

$$F^*(v, p, y) = Ef^*(v + p, y) + \delta_{\mathcal{X}_a^\perp}(p);$$

see [15, Lemma 13]. It follows that the dual problem becomes

$$\text{maximize} \quad \langle \bar{u}, y \rangle - Ef^*(p, y) \quad \text{over} \quad (p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y}. \quad (D)$$

It was shown in [15] that if the optimum value function φ is subdifferentiable at $(0, \bar{u})$ then there is no duality gap and a feasible $x \in \mathcal{N}$ solves (SP) if and only if there exists a dual feasible $(p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y}$ such that the scenario-wise optimality condition

$$(p, y) \in \partial f(x, \bar{u}) \text{ a.s.}$$

holds. The variable p can be interpreted as a “shadow price of information” introduced in [33, 22]; see also [1, 6]. It should be noted that the above applies to the original problem (SP) which does not directly fit the conjugate duality framework of [21] as the space \mathcal{N} is not locally convex. Allowing for general adapted strategies is key to obtaining the existence of primal solutions and the lower semicontinuity of the optimum value function φ in applications where the feasible set may be unbounded; see [15, Section 5]. An example is the celebrated “fundamental theorem of asset pricing” in financial mathematics [5, 30].

This paper gives sufficient conditions for subdifferentiability of the optimum value function φ of $(SP_{\mathcal{X}})$. A simple sufficient condition is that φ be Mackey-continuous at $(0, \bar{u})$ since that implies the existence of a subgradient in $\mathcal{V} \times \mathcal{Y}$; see

Theorem 57 in the Appendix. In some applications this is easy to establish e.g. by showing that φ has a Mackey continuous upper bound; see Corollary 51 in the Appendix as well as the discussion after Theorem 47. In others, however, the Mackey continuity fails. This is typical in problems with pointwise constraints such as those ones studied in [15] and in Section 5 below.

We follow the main arguments from [23, 14] and first establish, in Section 3, (relative) continuity of the value function φ with respect to a Fréchet topology which may be strictly stronger than the Mackey topology. Strengthening the topology makes it possible to establish the continuity in many applications. However, continuity with respect to such a topology does not, in general, give subgradients in the Mackey-dual $\mathcal{V} \times \mathcal{Y}$ of $\mathcal{X} \times \mathcal{U}$ but in a strictly larger dual space. Under an additional condition that extends classical “relatively complete recourse” condition from [23], we will show, in Section 4, that the strong subdifferential of φ at $(0, \bar{u})$ intersects the space $\mathcal{V} \times \mathcal{Y}$ thus giving a subgradient with respect to the original dual pairing. As noted earlier, this implies the absence of a duality gap and the existence of solutions to the dual problem (D).

Our main result extends those of [23, 14] in two important ways. First, we include the parameter $u \in \mathcal{U}$ which allows for dualization of different formats of stochastic optimization problems in a unified manner much like the conjugate duality framework of [21] unifies more specific problem formats. Second, we go beyond spaces of essentially bounded strategies and parameters. Allowing for more general Fréchet spaces \mathcal{X} and \mathcal{U} , widens the scope of applicability of the theory. The extensions are illustrated in Section 5 by deriving optimality conditions in mathematical programming, stochastic optimal control and portfolio optimization in a generality not seen before. In particular, we extend the main results of [25] by allowing equality constraints and unbounded feasible sets. Such extensions were obtained recently in [8] for Banach space valued strategies in a single period model under slightly more restrictive conditions. In stochastic optimal control, we find necessary and sufficient optimality conditions under far more general conditions that have been given before. In particular, we allow for both state and control constraints and unbounded feasible sets which are encountered e.g. in the classical models of linear-quadratic control. Stochastic control problems were treated via duality also in [29] but the analysis and the conclusions were quite different from classical functional analytic convex duality used here. In the case of portfolio optimization, we extend earlier results by the inclusion of portfolio constraints and statically held derivative portfolios.

The strategy outlined above is based on endowing the spaces \mathcal{X} and \mathcal{U} with topologies under which they become Fréchet spaces and under which their topological duals can be expressed as direct sums of \mathcal{V} and \mathcal{Y} , respectively, with linear spaces of certain singular functionals. Section 2 reviews the basic theory of such spaces and extends earlier results on conjugates of integral functionals with respect to the associated dual pairings.

2 Integral functionals under the strong topology

This section studies convex integral functionals with respect to a “strong topology” on \mathcal{U} which may be strictly stronger than the Mackey topology $\tau(\mathcal{U}, \mathcal{Y})$. The stronger the topology, the easier it is to establish continuity of a functional. By Theorem 57 in the Appendix, continuity implies subdifferentiability which, as we have seen, is closely connected to existence of dual solutions; see Section 6.3. Particularly convenient situation is where \mathcal{U} is a Fréchet space under the strong topology since then, lower semicontinuous convex functions on \mathcal{U} are continuous throughout the *core* of their domain; see Theorem 50 in the Appendix. Recall that a *Fréchet space* is a complete metrizable locally convex vector space. In particular, Banach spaces are Fréchet spaces.

Given a convex set $C \subset \mathcal{U}$, its core is the set

$$\text{core } C = \{u \in C \mid \text{pos}(C - u) = \mathcal{U}\},$$

where

$$\text{pos}(C - u) := \bigcup_{\alpha > 0} \alpha(C - u)$$

is the *positive hull* of $C - u$. More generally, Theorem 56 in the Appendix says that if a convex function g on a Fréchet space is the infimal projection of a closed convex function (as is the case with the optimum value function φ of (SP)) and if the affine hull of $\text{dom } g$ is closed, then g is relatively continuous throughout the relative core of its domain. By Theorem 57 in the Appendix, relative continuity is sufficient for subdifferentiability. Recall that the *relative core* of a set C is its core with respect to its affine hull. By Lemma 49 in the Appendix,

$$\text{rcore } C = \{u \in C \mid \text{pos}(C - u) \text{ is linear}\}$$

and, for every $u \in \text{core } C$, the positive hull $\text{pos}(C - u)$ is the linear translation of $\text{aff } C$. In many applications, one encounters convex functions that fail to be continuous throughout the whole space but, nevertheless, are relatively continuous.

If the strong topology is strictly stronger than the Mackey topology $\tau(\mathcal{U}, \mathcal{Y})$, then the corresponding topological dual of \mathcal{U} is strictly larger than \mathcal{Y} . The extra elements in the dual space cannot be represented by measurable functions but, often, they have properties that can be employed in the duality theory of integral functionals and, as we will see, in establishing the existence of solutions to the dual problem (D) .

2.1 Continuity of integral functionals

We will continue with the assumption that \mathcal{U} is a solid decomposable space of random variables. From now on, we will also assume that \mathcal{U} is endowed with a topology s at least as strong as the Mackey topology $\tau(\mathcal{U}, \mathcal{Y})$ such that s makes \mathcal{U} Fréchet space. This holds e.g. for the usual Lebesgue and Orlicz spaces. The

details and further examples can be found e.g. in [17]. We will call s the *strong topology*.

We say that a convex function g is *relatively continuous* at a point $u \in \text{dom } g$ if it is continuous at u in the relative topology of $\text{aff dom } g$. The following is a direct consequence of Corollary 55 in the Appendix.

Theorem 1. *Assume that Eh is proper and strongly lsc and that $\text{aff dom } Eh$ is strongly closed. Then Eh is strongly relatively continuous throughout $\text{rcore dom } Eh$.*

Recall that the *Köthe dual* or *associate space* of \mathcal{U} is the linear space

$$\mathcal{U}' := \{y \in L^0(\Omega, \mathcal{F}, P; \mathbb{R}^m) \mid u \cdot y \in L^1 \quad \forall u \in \mathcal{U}\};$$

see, e.g., [2, 34]. We denote by \mathcal{U}^* the topological dual of \mathcal{U} under the strong topology. That is, \mathcal{U}^* is the linear space of all strongly continuous linear functionals on \mathcal{U} .

Lemma 2. *We have $\mathcal{U}' \subseteq \mathcal{U}^*$ in the sense that, for every $y \in \mathcal{U}'$, the linear functional*

$$u \mapsto E[u \cdot y]$$

is strongly continuous on \mathcal{U} .

Proof. Let $y \in L^0$ be such that $u \cdot y \in L^1$ for all $u \in \mathcal{U}$. Given $u \in \mathcal{U}$, the solidity of \mathcal{U} implies that the random vector $u' \in L^0$ defined by $u'_i := |u_i| \text{sign}(y_i)$ belongs to \mathcal{U} . Since $u' \cdot y = \sum_{i=1}^m |u_i| |y_i|$, the function

$$p_y(u) := E\left[\sum_{i=1}^m |u_i| |y_i|\right]$$

is thus finite on \mathcal{U} . The above expression defines an extended real-valued convex function on L^1 . By Fatou's lemma, p_y is L^1 -lsc on L^1 so, by [13, Lemma 6], it is $\sigma(\mathcal{U}, \mathcal{Y})$ -lsc on \mathcal{U} . By Theorem 1, p_y is strongly continuous on \mathcal{U} . Since the function $u \mapsto E[u \cdot y]$ is majorized by p_y , it is strongly continuous by Theorem 50 in the Appendix. \square

Given a measurable mapping $C : \Omega \rightrightarrows \mathbb{R}^m$ and a space \mathcal{U} of measurable functions with values in \mathbb{R}^m , we will use the notation

$$\mathcal{U}(C) := \{u \in \mathcal{U} \mid u \in C \text{ a.s.}\}.$$

The following gives a simple criterion for strong continuity of an integral functional on an Orlicz space.

Example 3. *Let $\Phi : \mathbb{R} \times \Omega \rightarrow \overline{\mathbb{R}}$ be a convex normal integrand such that, for almost every $\omega \in \Omega$, the function $\Phi(\cdot, \omega)$ is nonconstant, symmetric, vanishes at the origin, has $\text{dom } \Phi(\cdot, \omega) \neq \{0\}$ and $\Phi(a) \in L^1$ for some constant $a > 0$. The Musielak-Orlicz space*

$$L^\Phi := \{u \in L^0(\Omega, \mathcal{F}, P) \mid \exists \alpha > 0 : E\Phi(|u|/\alpha) < \infty\}$$

is a solid decomposable Banach space; see [17].

Let h be a convex normal integrand such that Eh is proper and lsc on L^Φ and there exist constants $\epsilon, M > 0$ with

$$h(\epsilon u) \leq M\Phi(|u|) + M \quad \forall u \in \mathbb{R}^n \text{ a.s.}$$

Then $0 \in \text{core dom } Eh$ and Eh is continuous throughout $\text{core dom } Eh$. Defining

$$\tilde{h}(u, \omega) := h(u, \omega) + \delta_{H(\omega)}(u),$$

where H is a random linear set, we have $0 \in \text{rcore dom } E\tilde{h}$,

$$\text{aff dom } E\tilde{h} = L^\Phi(H) = L^\Phi(\text{aff dom } \tilde{h})$$

and $E\tilde{h}$ is strongly relatively continuous throughout $\text{rcore dom } E\tilde{h}$.

Proof. The continuity on $\text{core dom } Eh$ follows from Theorem 1. Given $u \in L^\Phi$, there exists an $\alpha > 0$ such that $E\Phi(|u|/\alpha) \leq 1$ so the assumption gives the existence of an $\epsilon > 0$ such that $Eh(\epsilon u) \leq 2M$. Thus, $0 \in \text{core dom } Eh$.

The claims concerning \tilde{h} follow from the fact that $E\tilde{h} = Eh + \delta_{L^\Phi(H)}$, where

$$L^\Phi(H) := \{u \in L^\Phi \mid u \in H \text{ a.s.}\}$$

is strongly closed, since it is $\sigma(\mathcal{U}, \mathcal{Y})$ -closed, by [15, Corollary 7]. Indeed, by Lemma 48 in the Appendix,

$$\text{pos}(\text{dom } E\tilde{h}) = \text{pos}(\text{dom } Eh \cap L^\Phi(H)) = \text{pos}(\text{dom } Eh) \cap L^\Phi(H) = L^\Phi(H).$$

It is clear that $\text{aff dom } \tilde{h} = H$. □

Taking a small enough space and a strong enough topology, it may be possible to establish relative continuity of Eh throughout $\text{rcore dom } Eh$ even when the integrand involves nonaffine pointwise constraints. The set-valued mappings

$$\omega \mapsto \text{dom } h(\cdot, \omega), \quad \omega \mapsto \text{aff dom } h(\cdot, \omega)$$

are measurable by [27, Proposition 14.28 and Exercise 14.12].

The following is obtained by modifying arguments in the proof of [20, Theorem 2] which required, in particular, that $\text{aff dom } h = \mathbb{R}^n$ almost surely; see also [14, Theorem 4].

Example 4. Assume that $Eh : L^\infty \rightarrow \overline{\mathbb{R}}$ is lsc and finite on

$$\mathcal{D} := \{u \in L^\infty(\text{dom } h) \mid \exists r > 0 : \mathbb{B}_r(u) \cap \text{aff dom } h \subseteq \text{dom } h \text{ a.s.}\}.$$

If $\mathcal{D} \neq \emptyset$, then $\mathcal{D} \subset \text{rcore dom } Eh$,

$$\text{aff dom } Eh = L^\infty(\text{aff dom } h)$$

and Eh is strongly relatively continuous throughout $\text{rcore dom } Eh$.

Proof. Let $u \in \mathcal{D}$. There is an $r > 0$ such that $\mathbb{B}_r(u) \cap \text{aff dom } h \subseteq \text{dom } h$ almost surely. Given $u' \in L^\infty(\text{aff dom } h)$, there is a $\lambda > 0$ such that

$$\mathbb{B}_{r/2}(\lambda(u' - u) + u) \subset \mathbb{B}_r(u).$$

Since $\lambda(u' - u) + u \in \text{aff dom } h$, we get $\lambda(u' - u) + u \in \mathcal{D}$ and thus, $L^\infty(\text{aff dom } h - u) \subseteq \text{pos}(\mathcal{D} - u)$. By assumption,

$$\mathcal{D} \subseteq \text{dom } Eh \subseteq L^\infty(\text{dom } h) \subseteq L^\infty(\text{aff dom } h),$$

so

$$\text{pos}(\mathcal{D} - u) = \text{pos}(\text{dom } Eh - u) = \text{pos}(L^\infty(\text{dom } h) - u) = L^\infty(\text{aff dom } h) - u.$$

Since $L^\infty(\text{aff dom } h)$ is affine, the first two claims follow from Lemma 49 in the Appendix. Since $L^\infty(\text{aff dom } h)$ is closed, the last claim follows from Theorem 1. \square

The following gives a simple example where $\text{pos dom } Eh$ is linear but not closed. In other words, we get $0 \in \text{rcore dom } Eh$ but $\text{aff dom } Eh$ is not closed. Corollary 55 thus implies that the function Eh is not relatively continuous on $\text{rcore dom } Eh$.

Counterexample 5. Let $\mathcal{U} = L^\infty$, $h(u, \omega) = \delta_{S(\omega)}(u)$, where $S(\omega) := \{u \mid |u| \leq \eta(\omega)\}$ for strictly positive $\eta \in L^\infty$ such that $1/\eta \notin L^\infty$. Then

$$\text{pos dom } Eh = \{u \in L^\infty \mid \exists \lambda > 0 : |u| \leq \lambda \eta\}$$

is linear but not closed in L^∞ . Indeed, defining $u^\nu := \min\{\sqrt{\eta}, \nu\eta\}$, we have $u^\nu \in \nu \text{ dom } Eh \subset \text{pos dom } Eh$ for all ν and $u^\nu \rightarrow \sqrt{\eta}$ in L^∞ but, since $1/\eta \notin L^\infty$, we have $\sqrt{\eta} \notin \nu \text{ dom } Eh$ for all ν .

In problems with nonlinear pointwise constraints, the strong relative core tends to be empty except in the strong topology of L^∞ .

Example 6. Let S be a closed convex random set and assume that the probability space is atomless and that, for every $w \in L^\infty$ and $(A^\nu)_{\nu=1}^\infty \subset \mathcal{F}$ with $P(A^\nu) \searrow 0$, $1_{A^\nu} w \rightarrow 0$ in the strong topology of \mathcal{U} . Then $u \in \text{rcore } \mathcal{U}(S)$ with $\text{aff } \mathcal{U}(S)$ closed if and only if $u \in S$ almost surely and S is affine-valued.

Proof. Sufficiency is clear. To prove necessity, let $u \in \text{rcore } \mathcal{U}(S)$, $\text{aff } \mathcal{U}(S)$ be closed and $S_u := S - u$. By Remark 54, the set $\mathcal{U}(S_u)$ is a neighborhood of the origin in $\text{aff } \mathcal{U}(S_u)$. Assume, for a contradiction, that S_u is not linear-valued. Then $\bar{A} := \{\omega \mid S_u(\cdot, \omega) \neq \text{aff } S_u(\omega)\}$ is not a null set. Let $w \in L^0$ be such that $w \in S_u \setminus \text{rint } S_u$ on \bar{A} . There exists constants $r < R$ such that $A := \{\omega \in \bar{A} \mid r < |w(\omega)| < R\}$ is not a null set. Now $1_A w \in \mathcal{U}(S)$, so $1_A 2w \in \text{aff } \mathcal{U}(S)$ but $1_A 2w \notin S_u$ almost surely. Since $\mathcal{U}(S - u)$ is a neighborhood of the origin in $\text{aff } \mathcal{U}(S - u)$, there exists, by the assumption in the statement, a non-null $A' \subseteq A$ such that $2w 1_{A'} \in \mathcal{U}(S - u)$. Thus, $2w 1_{A'} \in S_u$ almost surely, which is a contradiction. \square

The topological assumption in Example 6 holds, in particular, if any sequence $(u^\nu)_{\nu=1}^\infty$ in L^∞ with $|u^\nu| \searrow 0$ converges strongly to zero in \mathcal{U} . This condition fails in the strong topology of L^∞ but holds e.g. in L^p spaces with $p > 1$ or, more generally, in Orlicz spaces associated with finite Young functions Φ . These and more examples can be found in [17].

2.2 The strong dual

As in [15], we assume that \mathcal{U} and \mathcal{Y} are solid decomposable spaces of random variables in separating duality under the bilinear form $E[u \cdot y]$. We will also assume that \mathcal{U} is Fréchet under a given strong topology and that the corresponding topological dual \mathcal{U}^* can be expressed as

$$\mathcal{U}^* = \mathcal{Y} \oplus \mathcal{Y}^s$$

in the sense that, for every $y \in \mathcal{U}^*$, there exist unique $y^c \in \mathcal{Y}$ and $y^s \in \mathcal{Y}^s$ such that

$$\langle u, y \rangle = E[u \cdot y^c] + \langle u, y^s \rangle$$

for all $u \in \mathcal{U}$. Here and in what follows, \mathcal{Y}^s denotes the elements $y \in \mathcal{U}^*$ that are *singular* in the sense that, for every $u \in \mathcal{U}$, there exists a decreasing sequence $(A^\nu)_{\nu=1}^\infty \subset \mathcal{F}$ such that $P(A^\nu) \searrow 0$ and

$$\langle 1_{\Omega \setminus A^\nu} u, y \rangle = 0 \quad \forall \nu = 1, 2, \dots$$

By Lemma 2, \mathcal{Y} is necessarily the Köthe dual of \mathcal{U} . Most familiar Banach spaces of random variables have a topological dual of the above form; see [17].

The following extends Lemma 3 of [15] to the strong dual of \mathcal{U} .

Lemma 7. *Given a σ -algebra $\mathcal{G} \subset \mathcal{F}$ with $E^{\mathcal{G}}\mathcal{U} \subset \mathcal{U}$, the mapping $E^{\mathcal{G}} : \mathcal{U} \rightarrow \mathcal{U}$ is both strongly and $(\sigma(\mathcal{U}, \mathcal{Y}), \sigma(\mathcal{U}, \mathcal{Y}))$ -continuous. Moreover, the adjoint $(E^{\mathcal{G}})^* : \mathcal{U}^* \rightarrow \mathcal{U}^*$ is given by $(E^{\mathcal{G}})^*y = E^{\mathcal{G}}y$ for every $y \in \mathcal{Y}$. In particular, $E^{\mathcal{G}}\mathcal{Y} \subset \mathcal{Y}$.*

Proof. By Lemma 2, \mathcal{Y} is the Köthe dual of \mathcal{U} . Thus, by [15, Lemma 3], $E^{\mathcal{G}}$ is continuous with respect to $\sigma(\mathcal{U}, \mathcal{Y})$ -topology. In particular, $E^{\mathcal{G}}$ has $\sigma(\mathcal{U}, \mathcal{Y}) \times \sigma(\mathcal{U}, \mathcal{Y})$ -closed graph. Thus the graph is strongly closed as well. Since \mathcal{U} is Fréchet, the strong continuity follows from the closed graph theorem. Given $y \in \mathcal{Y}$, [15, Lemma 3] gives $E^{\mathcal{G}}y \in \mathcal{Y}$ and

$$\langle u, (E^{\mathcal{G}})^*y \rangle = \langle E^{\mathcal{G}}u, y \rangle = \langle u, E^{\mathcal{G}}y \rangle$$

for every $u \in \mathcal{U}$. Since \mathcal{U} separates points in \mathcal{U}^* , we have $(E^{\mathcal{G}})^*y = E^{\mathcal{G}}y$. \square

From now on, we will assume that \mathcal{X} has the same properties as \mathcal{U} . That is, we assume that there is a topology under which \mathcal{X} is a Fréchet space whose topological dual \mathcal{X}^* can be expressed as

$$\mathcal{X}^* = \mathcal{V} \oplus \mathcal{V}^s$$

in the sense that, for every $v \in \mathcal{X}^*$, there exist unique $v^c \in \mathcal{V}$ and $v^s \in \mathcal{V}^s$ such that, for all $x \in \mathcal{X}$

$$\langle x, v \rangle = E[x \cdot v^c] + \langle x, v^s \rangle.$$

Here \mathcal{V}^s is the set of singular elements of \mathcal{X}^* .

Lemma 8. *Let A be a random matrix such that $Ax \in \mathcal{U}$ for all $x \in \mathcal{X}$. Then $A^*y \in \mathcal{V}$ for all $y \in \mathcal{Y}$ and the linear mapping $\mathcal{A} : \mathcal{X} \rightarrow \mathcal{U}$ defined pointwise by*

$$\mathcal{A}x = Ax \quad a.s.$$

is both strongly and $(\sigma(\mathcal{X}, \mathcal{V}), \sigma(\mathcal{U}, \mathcal{Y}))$ -continuous. Moreover, the adjoint $\mathcal{A}^ : \mathcal{U}^* \rightarrow \mathcal{X}^*$ is given by $\mathcal{A}^*y = A^*y$ for every $y \in \mathcal{Y}$.*

Proof. By Lemma 2, \mathcal{V} is the Köthe dual of \mathcal{X} so the $(\sigma(\mathcal{X}, \mathcal{V}), \sigma(\mathcal{U}, \mathcal{Y}))$ -continuity follows from [15, Lemma 4]. In particular, $\text{gph } \mathcal{A}$ is weakly closed. Weak closedness implies strong closedness. Since \mathcal{U} and \mathcal{X} are Fréchet, the strong continuity follows from the closed graph theorem. Given $y \in \mathcal{Y}$, [15, Lemma 4] gives $A^*y \in \mathcal{V}$ and

$$\langle x, \mathcal{A}^*y \rangle = \langle Ax, y \rangle = \langle x, A^*y \rangle$$

for every $x \in \mathcal{X}$. Since \mathcal{X} separates points in \mathcal{X}^* , we have $\mathcal{A}^*y = A^*y$. \square

The following two examples are from [17].

Example 9 (Lebesgue spaces). *Let $\mathcal{U} = L^p$ and $\mathcal{Y} = L^q$ be the usual Lebesgue spaces, where p and q are conjugate exponents, and let the strong topology be the L^p -norm topology. Then \mathcal{U} is Banach and its dual can be expressed as*

$$\mathcal{U}^* = \mathcal{Y} \oplus \mathcal{Y}^s.$$

If $p < \infty$, then $\mathcal{Y}^s = \{0\}$. If $p = \infty$, then $\mathcal{U}^ \neq \mathcal{Y}$, in general, so that $\mathcal{Y}^s \neq \{0\}$.*

Example 10 (Musielak-Orlicz spaces). *Let $\mathcal{U} = L^\Phi$ and $\mathcal{Y} = L^{\Phi^*}$ be as in Example 3 and let the strong topology be the topology generated by the norm*

$$\|u\|_{L^\Phi} := \inf_{\alpha \in \mathbb{R}_+} \{\alpha \mid E\Phi(u/\alpha) \leq 1\}$$

Then \mathcal{U} is Banach and its dual can be expressed as

$$\mathcal{U}^* = \mathcal{Y} \oplus \mathcal{Y}^s.$$

If $\text{dom } E\Phi$ is a cone, then $\mathcal{Y}^s = \{0\}$. We have $\mathcal{Y}^s \neq \{0\}$ if L^∞ is not dense in L^Φ or if P is atomless and $\Phi(a, \cdot) \notin L^1$ for some $a > 0$.

If $\Phi(a, \cdot) \in L^1$ for all $a > 0$, then the Morse heart

$$M^\Phi := \{u \in L^0 \mid E\Phi(\alpha u) < \infty \forall \alpha \in \mathbb{R}_+\}$$

is Banach under the relative topology of L^Φ and L^{Φ^} is its strong dual.*

Much like Musielak-Orlicz spaces are often chosen so that a given objective is continuous (see Example 3), the following example shows how to design the space \mathcal{U} so that pointwise linear constraints define a continuous surjection to \mathcal{U} . Given a random matrix $A \in L^0(\mathbb{R}^{m \times n})$ with $\text{rge } A = \mathbb{R}^m$ almost surely, its scenario-wise Moore-Penrose inverse

$$A^\dagger := A^*(A^*A)^{-1}$$

is measurable.

Example 11 (Range spaces). *Let $A \in L^0(\mathbb{R}^{m \times n})$,*

$$\begin{aligned}\mathcal{U} &:= \{Ax \mid x \in \mathcal{X}\}, \\ \mathcal{Y} &:= \{y \in L^0 \mid A^*y \in \mathcal{V}\}\end{aligned}$$

and assume that $L^\infty \subset \mathcal{U}$ and $L^\infty \subset \mathcal{Y}$. We endow \mathcal{U} with the “final topology” induced by A , i.e. the strongest locally convex topology under which the linear mapping $x \mapsto Ax$ is continuous. Then \mathcal{U} is a decomposable Fréchet space. If \mathcal{X} is such that

$$\mathcal{X} = \{x \in L^0 \mid |x| \in \mathcal{X}_0\},$$

where \mathcal{X}_0 is a decomposable solid space of scalar random variables, then

$$\mathcal{U}^* = \mathcal{Y} \oplus \mathcal{Y}^s,$$

where \mathcal{Y}^s is the set of singular elements of \mathcal{U}^* . If, in addition, $|A||A^\dagger| \in L^\infty$ then \mathcal{U} is solid and

$$\mathcal{U} = \{A_1x \mid x \in \mathcal{X}\} \times \cdots \times \{A_mx \mid x \in \mathcal{X}\},$$

where A_i is the i -th row of A .

Proof. We apply Example 66 in the Appendix to the linear mapping $\mathcal{A} : \mathcal{X} \rightarrow L^0(\mathbb{R}^m)$ defined pointwise by $(\mathcal{A}x)(\omega) := A(\omega)x(\omega)$. Clearly, $\mathcal{U} = \text{rge } \mathcal{A}$. We have $\ker \mathcal{A} = \{x \in \mathcal{X} \mid Ax = 0\}$ which is weakly closed, by [15, Corollary 7], and thus, strongly closed. By Example 66, \mathcal{U} is then Fréchet and, for every $u^* \in \mathcal{U}^*$, there is a unique $v \in (\ker \mathcal{A})^\perp$ such that

$$\langle \mathcal{A}x, u^* \rangle = \langle x, v \rangle \quad \forall x \in \mathcal{X}.$$

Every $u \in \text{rge } \mathcal{A}$ can be expressed as $u = \mathcal{A}(A^\dagger u)$, where $A^\dagger u$ is an element of \mathcal{X} . Indeed, $u = Ax$ for some $x \in \mathcal{X}$ so

$$A^\dagger u = A^*(A^*A)^{-1}Ax$$

which is the scenario-wise Euclidean projection of $x(\omega)$ on $\text{rge } A^*$. Thus, $|A^\dagger Ax| \leq |x|$ almost surely. Since $\mathcal{X} = \{x \in L^0 \mid |x| \in \mathcal{X}_0\}$, we get $A^\dagger u \in \mathcal{X}$ for all

$u \in \text{rge } \mathcal{A}$. It follows that, for every $u^* \in (\text{rge } \mathcal{A})^*$, there is a $v \in (\ker \mathcal{A})^\perp$ such that

$$\begin{aligned} \langle u, u^* \rangle &= \langle A^\dagger u, v \rangle \\ &= E[A^\dagger u \cdot v^c] + \langle A^\dagger u, v^s \rangle \\ &= E[u \cdot (A^\dagger)^* v^c] + \langle A^\dagger u, v^s \rangle \end{aligned}$$

for all $u \in \text{rge } \mathcal{A}$. Here $(A^\dagger)^* v^c \in \mathcal{Y}$, since $A^*(A^\dagger)^* v^c = v^c \in \mathcal{V}$. Moreover, $u \mapsto \langle A^\dagger u, v^s \rangle$ is singular, which follows directly from the definition. Thus every continuous linear functional can be expressed by an element of $\mathcal{Y} \oplus \mathcal{Y}^s$.

Conversely, let $(y^c, y^s) \in \mathcal{Y} \oplus \mathcal{Y}^s$. By definition of the final topology, $u \mapsto E[u \cdot y^c]$ is continuous on \mathcal{U} if and only if $x \mapsto E[(Ax) \cdot y^c]$ is continuous on \mathcal{X} . This holds by definition of \mathcal{Y} . The singular component y^s is continuous by definition.

Assume now that $|A||A^\dagger| \in L^\infty$. Let $u \in \mathcal{U}$ and $u' \in L^0(\mathbb{R}^m)$ with $|u'| \leq |u|$. There exists $x \in \mathcal{X}$ with $Ax = u$, so

$$|A^\dagger u'| \leq |A^\dagger||u| \leq |A^\dagger||A||x|.$$

The right side belongs to \mathcal{X}_0 , so $A^\dagger u' \in \mathcal{X}$ by assumption. Thus $A^\dagger u' \in \mathcal{X}$, so $u' = AA^\dagger u' \in \mathcal{U}$ which shows that \mathcal{U} is solid. The last claim follows from the fact that solid spaces are Cartesian products. \square

2.3 Conjugates of integral functionals

Let \mathcal{U} and \mathcal{U}^* be as in Section 2.2 and let h be a convex normal integrand on $\mathbb{R}^m \times \Omega$. This section gives expressions for the conjugate

$$(Eh)^*(y) := \sup_{u \in \mathcal{U}} \{\langle u, y \rangle - Eh(u)\}$$

and the subdifferential

$$\partial Eh(u) := \{y \in \mathcal{U}^* \mid Eh(u) + (Eh)^*(y) = \langle u, y \rangle\}$$

of the integral functional Eh with respect to the duality pairing of \mathcal{U} and \mathcal{U}^* .

The main result of this section makes use of the following simple observation.

Lemma 12. *We have*

$$\limsup_{\nu \rightarrow \infty} Eh(1_{A^\nu} u + 1_{\Omega \setminus A^\nu} \bar{u}) \leq Eh(\bar{u})$$

for all $u, \bar{u} \in \text{dom } Eh$ and decreasing $(A^\nu)_{\nu=1}^\infty \subset \mathcal{F}$ with $P(A^\nu) \searrow 0$.

Proof. For any $\bar{u}, u \in \text{dom } Eh$ and $A \in \mathcal{F}$, the right side in

$$h(1_A u + 1_{\Omega \setminus A} \bar{u}) \leq \max\{h(u), h(\bar{u})\}$$

is integrable, so the claim follows from Fatou's lemma. \square

The following extends [14, Theorem 10] which in turn refines [20, Corollary 1B]. Both [20] and [14] studied integral functionals on the space L^∞ of essentially bounded random variables. The following extends [14, Theorem 10] to general Fréchet spaces that satisfy the assumptions Section 2.2. It also yields [9, Theorem 2.6] in the case of Orlicz spaces of \mathbb{R}^n -valued random variables.

Lemma 13. *Let Eh be proper. If $\bar{u} \in \mathcal{U}$, $y \in \mathcal{U}^*$ and $\epsilon \geq 0$ are such that*

$$Eh(u) \geq Eh(\bar{u}) + \langle u - \bar{u}, y \rangle - \epsilon \quad \forall u \in \mathcal{U}, \quad (1)$$

then

$$Eh(u) \geq Eh(\bar{u}) + \langle u - \bar{u}, y^c \rangle - \epsilon \quad \forall u \in \mathcal{U}, \quad (2)$$

and

$$0 \geq \langle u - \bar{u}, y^s \rangle - \epsilon \quad \forall u \in \text{dom } Eh. \quad (3)$$

Proof. Let $u \in \text{dom } Eh$ and let $(A^\nu)_{\nu=1}^\infty \subset \mathcal{F}$ be the decreasing sequence of sets in the characterization of the singular component y^s of y . Let $u^\nu := \mathbb{1}_{A^\nu} \bar{u} + \mathbb{1}_{\Omega \setminus A^\nu} u$. Lemma 12 and (1) give

$$Eh(u) \geq \limsup Eh(u^\nu) \geq Eh(\bar{u}) + \limsup \langle u^\nu - \bar{u}, y \rangle - \epsilon.$$

Since $u^\nu - \bar{u} = \mathbb{1}_{\Omega \setminus A^\nu} (u - \bar{u})$, we have

$$\langle u^\nu - \bar{u}, y \rangle = \langle u^\nu - \bar{u}, y^c \rangle \rightarrow \langle u - \bar{u}, y^c \rangle$$

so (2) holds.

Now let $u^\nu := \mathbb{1}_{A^\nu} u + \mathbb{1}_{\Omega \setminus A^\nu} \bar{u}$ so that

$$Eh(\bar{u}) \geq \limsup Eh(u^\nu) \geq Eh(\bar{u}) + \limsup \langle u^\nu - \bar{u}, y \rangle - \epsilon.$$

Since $u^\nu - \bar{u} = \mathbb{1}_{A^\nu} (u - \bar{u})$, we have

$$\langle u^\nu - \bar{u}, y \rangle = \langle u^\nu - \bar{u}, y^c \rangle + \langle u^\nu - \bar{u}, y^s \rangle \rightarrow \langle u - \bar{u}, y^s \rangle$$

so $\langle u - \bar{u}, y^s \rangle - \epsilon \leq 0$. Since $u \in \text{dom } Eh$ was arbitrary, (3) holds. \square

Recall that $\mathcal{U}^* = \mathcal{Y} \oplus \mathcal{Y}^s$. We now come to the main result of this section which states that, at each $y = y^c + y^s \in \mathcal{U}^*$, the conjugate of Eh equals the sum of the integral functional of h^* evaluated at y^c and the *support function*

$$\sigma_{\text{dom } Eh}(y) := \sup_{u \in \text{dom } Eh} \langle u, y \rangle$$

of the domain of Eh ; see Lemma 58 in the Appendix. The following extends the main results of [20] on integral functionals on L^∞ to more general Fréchet spaces that satisfy the assumptions of Section 2.2.

Theorem 14. *If Eh is proper, then*

$$1. (Eh)^*(y) = Eh^*(y^c) + \sigma_{\text{dom } Eh}(y^s) \quad \forall y \in \mathcal{U}^*,$$

2. Eh is strongly lsc if and only if it is $\sigma(\mathcal{U}, \mathcal{Y})$ -lsc,
3. $y \in \partial Eh(\bar{u})$ if and only if $y^c \in \partial Eh(\bar{u})$ and $y^s \in N_{\text{dom } Eh}(\bar{u})$.

Proof. We have

$$\begin{aligned}
(Eh)^*(y) &= \sup_u \{\langle u, y \rangle - Eh(u)\} \\
&= \sup_u \{\langle u, y^c \rangle - Eh(u) + \langle u, y^s \rangle - \delta_{\text{dom } Eh}(u)\} \\
&\leq Eh^*(y^c) + \sigma_{\text{dom } Eh}(y^s)
\end{aligned}$$

while for every $y \in \text{dom}(Eh)^*$ and $\epsilon > 0$, there exists $\bar{u} \in \text{dom } Eh$ such that (1) holds. By Lemma 13 and Fenchel's inequality,

$$(Eh)^*(y^c) + \sigma_{\text{dom } Eh}(y^s) \leq \langle \bar{u}, y^c \rangle - Eh(\bar{u}) + \epsilon + \langle \bar{u}, y^s \rangle + \epsilon \leq (Eh)^*(y) + 2\epsilon.$$

This gives the expression for the conjugate since $\epsilon > 0$ was arbitrary.

It is clear that if Eh is $\sigma(\mathcal{U}, \mathcal{Y})$ -lsc it is strongly lsc. When Eh is proper and strongly lsc, the biconjugate theorem says that it has a proper conjugate. The first claim then implies that Eh^* is proper on \mathcal{Y} . The second claim thus follows from [17, Corollary 6].

When $\epsilon = 0$, Lemma 13 says that if $y \in \partial Eh(\bar{u})$ then $y^c \in \partial Eh(\bar{u})$ and $y^s \in N_{\text{dom } Eh}(\bar{u})$. The converse implication follows simply by adding (2) and (3) together. \square

Theorem 14 implies, in particular, that strong subdifferentiability of Eh implies subdifferentiability with respect to the pairing of \mathcal{U} with \mathcal{Y} .

Corollary 15. *Assume that Eh is closed proper and finite on \mathcal{U} . Then Eh is continuous both strongly and in $\tau(\mathcal{U}, \mathcal{Y})$.*

Proof. Strong continuity follows from Theorem 52. By Bourbaki-Alaoglu, level sets of $(Eh)^*$ are $\sigma(\mathcal{U}^*, \mathcal{U})$ -compact. By Theorem 14, the finiteness of Eh implies that all the level sets of $(Eh)^*$ are contained in \mathcal{Y} , so they are $\sigma(\mathcal{Y}, \mathcal{U})$ -compact. By the converse of Bourbaki-Alaoglu, Eh is $\tau(\mathcal{U}, \mathcal{Y})$ -continuous. \square

3 The strong dual problem

From now on, we will assume that both \mathcal{X} and \mathcal{U} satisfy the assumptions of Section 2.2. More precisely, \mathcal{X} and \mathcal{U} are endowed with a *strong* topology under which they are Fréchet, the dual of \mathcal{U}^* can be identified with $\mathcal{Y} \oplus \mathcal{Y}^s$ and, similarly, the topological dual of \mathcal{X}^* can be identified with $\mathcal{V} \oplus \mathcal{V}^s$ in the sense that, for each $v \in \mathcal{X}^*$, there exist unique $v^c \in \mathcal{V}$ and $v^s \in \mathcal{V}^s$ such that

$$\langle x, v \rangle = E[x \cdot v^c] + \langle x, v^s \rangle$$

for all $x \in \mathcal{X}$. Here, \mathcal{V}^s denotes the singular elements of \mathcal{X}^* .

Much like in [15], we apply the general conjugate duality framework of [21] to $(SP_{\mathcal{X}})$ with

$$F(x, z, u) := Ef(x, u) + \delta_{\mathcal{N}}(x - z),$$

but this time, with respect to the pairings of \mathcal{X} and \mathcal{U} with \mathcal{X}^* and \mathcal{U}^* , respectively. This gives the *strong dual problem*

$$\text{maximize } \langle \bar{u}, y \rangle - \varphi^*(p, y) \quad \text{over } (p, y) \in \mathcal{X}^* \times \mathcal{U}^*, \quad (D_s)$$

where φ^* is the conjugate of φ with respect the strong pairings. As long as the optimum value of $(SP_{\mathcal{X}})$ is finite, the subgradients of the optimum value function φ in the space $\mathcal{X}^* \times \mathcal{U}^*$ are solutions of (D_s) ; see Theorem 61 in the Appendix. Establishing the existence of a subgradient in the strong dual may be significantly easier than to establish the existence in the smaller space $\mathcal{V} \times \mathcal{Y} \subset \mathcal{X}^* \times \mathcal{U}^*$. Indeed, since \mathcal{X} and \mathcal{U} are Fréchet, Theorem 64 in the Appendix gives general conditions for the subdifferentiability. The special structure of the conjugates given in Section 2.1 will then allow us to show in Section 4 that the existence of solutions to (D_s) implies the existence of solutions of the original dual problem (D) and, moreover, that the optimum values are equal.

We will denote the orthogonal complement of \mathcal{X}_a in the strong dual of \mathcal{X} by

$$\mathcal{X}_a^\circ := \{p \in \mathcal{X}^* \mid \langle x, p \rangle = 0 \quad \forall x \in \mathcal{X}_a\}.$$

Clearly, $\mathcal{X}_a^\perp \subseteq \mathcal{X}_a^\circ$ in the sense that $p^c \in \mathcal{X}_a^\perp$ if and only if $(p^c, 0) \in \mathcal{X}_a^\circ$. The following lemma gives expressions for the conjugates of F and the optimum value function

$$\varphi(z, u) := \inf_{x \in \mathcal{X}} \{Ef(x, u) \mid x - z \in \mathcal{N}\}$$

with respect to the strong pairings.

Lemma 16. *If $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, then*

$$F^*(v, p, y) = Ef^*((v + p)^c, y^c) + \sigma_{\text{dom } Ef}((v + p)^s, y^s) + \delta_{\mathcal{X}_a^\circ}(p)$$

and, in particular,

$$\varphi^*(p, y) = Ef^*(p^c, y^c) + \sigma_{\text{dom } Ef}(p^s, y^s) + \delta_{\mathcal{X}_a^\circ}(p).$$

If, in addition, $\text{dom } Ef^* \cap (\mathcal{V} \times \mathcal{Y}) \neq \emptyset$, then F is proper and strongly lsc.

Proof. By Theorem 14,

$$\begin{aligned} F^*(v, p, y) &= \sup_{x \in \mathcal{X}, z \in \mathcal{X}, u \in \mathcal{U}} \{\langle x, v \rangle + \langle z, p \rangle + \langle u, y \rangle - Ef(x, u) \mid x - z \in \mathcal{X}_a\} \\ &= \sup_{x \in \mathcal{X}, z' \in \mathcal{X}, u \in \mathcal{U}} \{\langle x, v + p \rangle + \langle u, y \rangle - Ef(x, u) - \langle z', p \rangle \mid z' \in \mathcal{X}_a\} \\ &= Ef^*((v + p)^c, y^c) + \sigma_{\text{dom } Ef}((v + p)^s, y^s) + \delta_{\mathcal{X}_a^\circ}(p). \end{aligned}$$

When $\text{dom } Ef^* \neq \emptyset$, [15, Lemma 13] says that Ef is $\sigma(\mathcal{X}, \mathcal{V}) \times \sigma(\mathcal{U}, \mathcal{Y})$ -lsc and proper. The last claim thus follows from the fact that F is the sum of lsc functions. Clearly, weak lower semicontinuity implies strong semicontinuity and F is proper when Ef is proper. \square

As an immediate corollary, we get the following.

Theorem 17. *If $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, the strong dual problem (D_s) can be written as*

$$\text{maximize } \langle \bar{u}, y \rangle - Ef^*(p^c, y^c) - \sigma_{\text{dom } Ef}(p^s, y^s) \text{ over } (p, y) \in \mathcal{X}_a^\circ \times \mathcal{U}^*. \quad (D_s)$$

Restricting the dual variables (p, y) to $\mathcal{V} \times \mathcal{Y}$, problem (D_s) reduces to (D) , so

$$\inf (D_s) \leq \inf (D).$$

In particular, if there is no duality gap between $(SP_{\mathcal{X}})$ and (D) , the same holds between $(SP_{\mathcal{X}})$ and (D_s) . Clearly, the two dual problems coincide if $\mathcal{Y}^s = \{0\}$ and $\mathcal{X}^s = \{0\}$ as happens e.g. when $\mathcal{U} = \mathcal{X} = L^p$ with $p < \infty$; see Example 9. The problems coincide also in the special case where $\text{dom } Ef = \mathcal{X} \times \mathcal{U}$ since then, $\sigma_{\text{dom } Ef} = \delta_{\{(0,0)\}}$. Section 4 below gives more general conditions under which the optimum values of (D) and (D_s) coincide and one has a solution if and only if the other one does.

The following gives sufficient conditions for subdifferentiability of φ with respect to $\mathcal{X}^* \times \mathcal{U}^*$ and thus, by [21, Theorem 16], for the absence of a duality gap between $(SP_{\mathcal{X}})$ and (D_s) and for the existence of solutions for the latter.

Theorem 18. *If Ef is proper and strongly closed and*

$$\text{pos}(\text{dom } \varphi - (0, \bar{u}))$$

is linear and closed, then the strong dual problem (D_s) has a solution and

$$\inf (SP_{\mathcal{X}}) = \sup (D_s).$$

Proof. Follows directly from [21, Theorem 16] and Theorem 64 in the Appendix. \square

The set $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and strongly closed if $(0, \bar{u})$ belongs to the strong interior of $\text{dom } \varphi$. A simple extension of this is to require that the affine hull of $\text{dom } \varphi$ is strongly closed and $(0, \bar{u})$ belongs to the corresponding relative interior of $\text{dom } \varphi$. The following lemma will be useful in providing more general conditions.

Lemma 19. *If $\bar{x} \in \mathcal{X}_a$ is feasible, then*

$$\text{pos}(\text{dom } \varphi - (0, \bar{u})) = \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) - \mathcal{X}_a \times \{0\}$$

Proof. By definition,

$$\text{dom } \varphi = \text{dom } Ef - \mathcal{X}_a \times \{0\}$$

so

$$\text{dom } \varphi - (0, \bar{u}) = \text{dom } Ef - (\bar{x}, \bar{u}) - \mathcal{X}_a \times \{0\}$$

for any $\bar{x} \in \mathcal{X}_a$. If $\bar{x} \in \mathcal{X}_a$ is feasible, then $(\bar{x}, \bar{u}) \in \text{dom } Ef$ and

$$\text{pos}(\text{dom } \varphi - (0, \bar{u})) = \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) - \mathcal{X}_a \times \{0\},$$

by Lemma 48 in the Appendix. \square

We say that an $\bar{x} \in \mathcal{X}_a$ is *strictly feasible* if

$$\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$$

is linear and closed. If such a point exists, we say that the problem $(SP_{\mathcal{X}})$ is *strictly feasible*. This happens, in particular, if there exists an $\bar{x} \in \mathcal{X}_a$ such that $(\bar{x}, \bar{u}) \in \text{int } \text{dom } Ef$. This corresponds to the classical *Slater condition* in optimization theory. The condition implies that $\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) = \mathcal{X} \times \mathcal{U}$ and thus, by Lemma 19, $\text{pos}(\text{dom } \varphi - \bar{u}) = \mathcal{X} \times \mathcal{U}$, so φ is strongly subdifferentiable, by Theorem 18. More generally, if there exists a strictly feasible \bar{x} such that $\mathcal{X}_a \times \{0\} \subset \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$, we have $\text{pos}(\text{dom } \varphi - \bar{u}) = \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$ and, by Theorem 18 again, φ is subdifferentiable.

In general, Lemma 19 reduces the condition in Theorem 18 to the question on the closedness of the sum of two sets. Under strict feasibility, both sets are linear and closed. Example 4 gives the following sufficient condition for strict feasibility when $\mathcal{X} = L^\infty$ and $\mathcal{U} = L^\infty$.

Example 20. Let $\mathcal{X} = L^\infty$ and $\mathcal{U} = L^\infty$ and assume that Ef is finite on the set

$$\mathcal{D} := \{(x, u) \in (\mathcal{X} \times \mathcal{U})(\text{dom } f) \mid \exists r > 0 : \mathbb{B}_r(x, u) \cap \text{aff } \text{dom } f \subseteq \text{dom } f \text{ P-a.e.}\}.$$

If $\bar{x} \in \mathcal{X}_a$ is such that $(\bar{x}, \bar{u}) \in \mathcal{D}$, then \bar{x} is strictly feasible and

$$\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid (x, u) \in \text{aff } \text{dom } f - (\bar{x}, \bar{u}) \text{ a.s.}\}.$$

Closedness of sums of closed linear subspaces is a well-studied question in functional analysis for which various sufficient as well as necessary conditions have been given. In many applications, of stochastic optimization, the problem has additional structure that can be used to establish the closedness of the sum in Lemma 19. The following lemma gives a sufficient condition that may look awkward at first glance but turns out to be convenient in many applications.

Lemma 21. Let $\bar{x} \in \mathcal{X}_a$ be strictly feasible and let $\mathcal{L} := \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$. If there exists a strongly continuous linear idempotent mapping π on $\mathcal{X} \times \mathcal{U}$ such that $\pi\mathcal{L} \subseteq \mathcal{L}$ and

$$\mathcal{L} - \mathcal{X}_a \times \{0\} = \mathcal{L} + \text{rge } \pi,$$

then $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed. The equality holds, in particular, if

$$\mathcal{X}_a \times \{0\} \subseteq \text{rge } \pi \subseteq \mathcal{L} - \mathcal{X}_a \times \{0\}.$$

Proof. Strict feasibility means that there exists $\bar{x} \in \mathcal{X}_a$ such that $\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) = \mathcal{L}$ and that \mathcal{L} is closed. By Lemma 19,

$$\text{pos}(\text{dom } \varphi - (0, \bar{u})) = \mathcal{L} - \mathcal{X}_a \times \{0\},$$

so the first claim follows from Lemma 67 in the Appendix. The additional inclusions give

$$\mathcal{L} - \mathcal{X}_a \times \{0\} \subseteq \mathcal{L} + \text{rge } \pi \subseteq \mathcal{L} - \mathcal{X}_a \times \{0\},$$

which proves the second claim. \square

In some applications, it may be difficult to find a mapping π that satisfies the assumptions of Lemma 21. In such situations, the formulation in Example 22 below may be more convenient.

Example 22. *Let*

$$f(x, u, \omega) = g(x, A(\omega)x + u, \omega),$$

where g is a convex normal integrand on $\mathbb{R}^n \times \mathbb{R}^m \times \Omega$ and A is a random matrix with $A\mathcal{X} \subseteq \mathcal{U}$. Assume that $\bar{x} \in \mathcal{X}_a$ is such that

$$\mathcal{K} := \text{pos}(\text{dom } Eg - (\bar{x}, A\bar{x} + \bar{u}))$$

is linear and closed. Then \bar{x} is strictly feasible. If there exists a strongly continuous linear idempotent mapping π' on $\mathcal{X} \times \mathcal{U}$ such that $\pi'\mathcal{K} \subseteq \mathcal{K}$ and

$$\mathcal{K} - (I, A)\mathcal{X}_a = \mathcal{K} + \text{rge } \pi',$$

then $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed. The equality holds, in particular, if

$$(I, A)\mathcal{X}_a \subseteq \text{rge } \pi' \subseteq \mathcal{K} - (I, A)\mathcal{X}_a.$$

Proof. We have $\text{dom } Ef = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid (x, Ax + u) \in \text{dom } Eg\} = \bar{A}^{-1} \text{dom } g$, where

$$\bar{A} = \begin{bmatrix} I & 0 \\ A & I \end{bmatrix}.$$

By Lemma 48 in the Appendix,

$$\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) = \bar{A}^{-1} \text{pos}(\text{dom } Eg - \bar{A}(\bar{x}, \bar{u})) = \bar{A}^{-1}\mathcal{K}.$$

Since \mathcal{K} is linear and closed, by assumption, \bar{x} is strictly feasible.

To complete the proof, we apply Lemma 21 with the idempotent mapping $\pi = \bar{A}^{-1}\pi'\bar{A}$. We have $\mathcal{L} = \bar{A}^{-1}\mathcal{K}$ so $\pi'\mathcal{K} \subseteq \mathcal{K}$ implies $\pi\mathcal{L} \subseteq \mathcal{L}$. Moreover,

$$\begin{aligned} \mathcal{L} - \mathcal{X}_a \times \{0\} &= \{(x, u) \mid (x, Ax + u) \in \mathcal{K}\} - \mathcal{X}_a \times \{0\} \\ &= \{(x, u) \mid (x, Ax + u) \in \mathcal{K} - (I, A)\mathcal{X}_a\} \\ &= \bar{A}^{-1}(\mathcal{K} - (I, A)\mathcal{X}_a) \end{aligned}$$

and $\text{rge } \pi = \bar{A}^{-1} \text{rge } \pi'$, so the equality $\mathcal{K} - (I, A)\mathcal{X}_a = \mathcal{K} - \text{rge } \pi'$ implies the equality in Lemma 21. The additional inclusions give

$$\mathcal{K} - (I, A)\mathcal{X}_a \subseteq \mathcal{K} + \text{rge } \pi' \subseteq \mathcal{K} - (I, A)\mathcal{X}_a,$$

which proves the last claim. \square

4 Existence of dual solutions

The set of solutions of (D) is the intersection of the solution set of (D_s) with $\mathcal{V} \times \mathcal{Y}$. In general, it may happen that (D) does not admit solutions even if (D_s) does and there is no duality gap; see [24, Example 2]. This section gives conditions under which, for each point $(p, y) \in \mathcal{X}^* \times \mathcal{U}^*$, there is a dual feasible $(\tilde{p}, \tilde{y}) \in \mathcal{V} \times \mathcal{Y}$ that achieves a dual objective value at least as good as (p, y) . Combined with the existence results for strong dual solutions in the previous section, we then obtain existence results for (D) .

Let \mathcal{X} and \mathcal{U} be as in Section 3. We will assume, in addition, that $E_t \mathcal{X} \subseteq \mathcal{X}$ for all t . By Lemma 7, this implies that the conditional expectations E_t are both strongly and weakly continuous and that, when restricted to \mathcal{V} , the adjoints $E_t^* : \mathcal{X}^* \rightarrow \mathcal{X}^*$ coincide with the conditional expectations E_t . By [15, Lemma 3] and Lemma 2, we also have $E_t \mathcal{V} \subseteq \mathcal{V}$. Most familiar spaces of random variables satisfy the above assumptions.

We will denote the projection of the set $\text{dom } Ef \subset \mathcal{X} \times \mathcal{U}$ on \mathcal{U} by

$$\text{dom}_u Ef := \{u \in \mathcal{U} \mid \exists x \in \mathcal{X} : (x, u) \in \text{dom } Ef\}.$$

Assumption 23.

1. $\text{pos}(\text{dom}_u Ef - \bar{u})$ is linear and closed,
2. for every t and $z \in \mathcal{X}$ with $(z, \bar{u}) \in \text{dom } Ef$, there exists $\bar{z} \in \mathcal{X}$ with $(\bar{z}, \bar{u}) \in \text{dom } Ef$ and $\bar{z}^t = E_t z^t$.

Assumption 23 clearly holds if $\text{dom } Ef = \mathcal{X} \times \mathcal{U}$. Note that, in this case, the support function of $\text{dom } Ef$ is the indicator of the origin so feasible solutions in (D_s) have $(p^s, u^s) = (0, 0)$. More interesting sufficient conditions for Assumption 23 are given in Lemma 28 and Example 30 below.

First, however, we will show that under Assumption 23, one can restrict dual variables to $\mathcal{V} \times \mathcal{Y}$ without worsening the objective value in the strong dual problem (D_s) . The argument is based on a recursive application of the following somewhat technical lemma, the proof of which is an application of conjugate duality; see the appendix.

Lemma 24. *Assumption 23 implies that, for every t and $(p, y) \in \mathcal{X}^* \times \mathcal{U}^*$ with $p_{t'} = 0$ for $t' > t$, there exists a $\hat{y} \in \mathcal{U}^*$ such that*

$$\sigma_{\text{dom } Ef}(E_t^* p, \hat{y}) - \langle \bar{u}, \hat{y} \rangle \leq \sigma_{\text{dom } Ef}(p, y) - \langle \bar{u}, y \rangle.$$

Proof. Define a convex function g_t on \mathcal{X} by

$$g_t(z) := \inf_{\tilde{z} \in \mathcal{L}_t} Ef(z + \tilde{z}, \bar{u}),$$

where $\mathcal{L}_t := \{z \in \mathcal{X} \mid z^t = 0\}$. Assumption 23.2 means that

$$\delta_{\text{dom } g_t}(E_t z) \leq \delta_{\text{dom } g_t}(z) \quad \forall z \in \mathcal{X}.$$

By Lemmas 59 and 58 in the Appendix, this implies

$$(g_t^*)^\infty(E_t^*p) \leq (g_t^*)^\infty(p) \quad \forall p \in \mathcal{X}^*, \quad (4)$$

A direct calculation gives

$$g_t^* = (Ef(\cdot, \bar{u}))^* + \delta_{\mathcal{L}_t^\perp}(\cdot)$$

where $\mathcal{L}_t^\perp = \{p \in \mathcal{X}^* \mid p_{t'} = 0 \forall t' > t\}$. Since the recession function of the sum is the sum of the recession functions, we get by Lemma 58 in the Appendix that

$$(g_t^*)^\infty = \delta_{\text{dom } Ef(\cdot, \bar{u})^*} + \delta_{\mathcal{L}_t^\perp}(\cdot),$$

By Theorem 65 in the Appendix, Assumption 23.1 implies

$$\delta_{\text{dom } Ef(\cdot, \bar{u})^*}(p) = \inf_{y \in \mathcal{U}^*} \{\sigma_{\text{dom } Ef}(p, y) - \langle \bar{u}, y \rangle\}$$

where the infimum is attained for every $p \in \mathcal{X}^*$. Combining this with the expression for $(g_t^*)^\infty$ and with (4) proves the claim. \square

The following is the main result of this section. Its proof is based on Lemmas 16 and 24, The argument extends that of [14, Theorem 3] which in turn simplifies the arguments of [23].

Theorem 25. *Under Assumption 23, there exists, for every $(p, y) \in \mathcal{X}_a^\circ \times \mathcal{U}^*$, a dual feasible $(\tilde{p}, \tilde{y}) \in \mathcal{X}_a^\perp \times \mathcal{Y}$ with*

$$\varphi^*(\tilde{p}, \tilde{y}) - \langle \bar{u}, \tilde{y} \rangle \leq \varphi^*(p, y) - \langle \bar{u}, y \rangle. \quad (5)$$

Proof. Let $(p, y) \in \mathcal{X}_a^\circ \times \mathcal{U}^*$. We will show that if $p_{t'}^s = 0$ for $t' > t$, then there exists $(\tilde{p}, \tilde{y}) \in \mathcal{X}_a^\circ \times \mathcal{U}^*$ satisfying (5) and $\tilde{p}_{t'}^s = 0$ for $t' \geq t$. Since $p_{t'}^s = 0$ for $t' > T$, we then get, by induction, a $(\tilde{p}, \tilde{y}) \in \mathcal{X}_a^\perp \times \mathcal{U}^*$ satisfying (5).

By Lemma 16,

$$\varphi^*(p, y) - \langle \bar{u}, y \rangle = Ef^*(p^c, y^c) - \langle \bar{u}, y^c \rangle + \sigma_{\text{dom } Ef}(p^s, y^s) - \langle \bar{u}, y^s \rangle.$$

Assume that $p_{t'}^s = 0$ for $t' > t$. By Lemma 24, there is a $\hat{y} \in \mathcal{U}^*$ such that

$$\varphi^*(p, y) - \langle \bar{u}, y \rangle \geq Ef^*(p^c, y^c) - \langle \bar{u}, y^c \rangle + \sigma_{\text{dom } Ef}(E_t^*p^s, \hat{y}) - \langle \bar{u}, \hat{y} \rangle.$$

Applying Fenchel's inequality to Ef and the indicator of $\text{dom } Ef$, now gives

$$\begin{aligned} \varphi^*(p, y) - \langle \bar{u}, y \rangle &\geq \sup_{x \in \mathcal{X}, u \in \mathcal{U}} \{\langle x, p^c + E_t^*p^s \rangle + \langle u, y^c + \hat{y} \rangle - Ef(x, u)\} - \langle \bar{u}, y^c + \hat{y} \rangle \\ &= Ef^*(\tilde{p}^c, \tilde{y}^c) + \sigma_{\text{dom } Ef}(\tilde{p}^s, \tilde{y}^s) - \langle \bar{u}, \tilde{y} \rangle, \end{aligned}$$

where $\tilde{p} = p^c + E_t^*p^s$, $\tilde{y} = y^c + \hat{y}$ and the last equality holds by Theorem 14. Since, $p \in \mathcal{X}_a^\circ$, we have $E_t^*p_t = 0$ so $E_t^*p_t^s = -E_t^*p_t^c = -E_t p_t^c$ and thus, $\tilde{p}_{t'}^s = 0$

for every $t' \geq t$ as desired. It is easily checked that we still have $\tilde{p} \in \mathcal{X}_a^\circ$ so, by Theorem 16 again,

$$\varphi^*(\tilde{p}, \tilde{y}) - \langle \bar{u}, \tilde{y} \rangle \leq \varphi^*(p, y) - \langle \bar{u}, y \rangle.$$

Repeating the above procedure, we will get a $(\tilde{p}, \tilde{y}) \in \mathcal{X}_a^\perp \times \mathcal{U}^*$ satisfying (5) as claimed. Since

$$\sigma_{\text{dom } Ef}(0, y) - \langle \bar{u}, y \rangle \geq 0 \quad \forall y \in \mathcal{U}^*,$$

we can also drop the singular part of \tilde{y} with out increasing the value of $\varphi^*(\tilde{p}, \tilde{y}) - \langle \bar{u}, \tilde{y} \rangle$. \square

It is clear that the optimum value of the strong dual minorizes that of the weak dual. Theorem 25 can thus be reformulated as follows.

Theorem 26. *Under Assumption 23, the optimum values of (D_s) and (D) coincide and one has a solution if the other does.*

Combined with the results of Section 3, the above gives a two-step strategy for establishing the existence of solutions to (D) . First, one verifies subdifferentiability of the optimum value function φ with respect to the pairing of $\mathcal{X} \times \mathcal{U}$ with $\mathcal{X}^* \times \mathcal{U}^*$ and second, one checks whether Assumption 23 is satisfied. For strong subdifferentiability, we can use the sufficient conditions given in Section 3. Combining this with the optimality conditions from [15] gives the following.

Theorem 27. *Assume that $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed and that Assumption 23 holds. Then*

$$\inf(SP) = \inf(SP) = \sup(D)$$

and the dual optimum is attained. Moreover, a primal feasible x solves (SP) if and only if there exists a dual feasible (p, y) with $(p, y) \in \partial f(x, \bar{u})$ almost surely.

Proof. By Theorems 18 and 26, the assumptions imply the first two claims. The last claim follows from [15, Corollary 27]. \square

We end this section with some sufficient conditions for Assumption 23. Under strict feasibility, Assumption 23.1 can be stated as follows.

Lemma 28. *Let $\bar{x} \in \mathcal{X}_a$ be strictly feasible. Assumption 23.1 holds if and only if the projection of $\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$ to \mathcal{U} is strongly closed.*

Proof. Assumption 23.1 holds if and only if $\text{pos}(\text{dom}_u Ef - \bar{u})$ is linear and strongly closed. Strict feasibility means that there exists an $\bar{x} \in \mathcal{X}_a$ such that $\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))$ is linear and strongly closed. We have

$$\text{dom}_u Ef - \bar{u} = \{u \in \mathcal{U} \mid \exists x \in \mathcal{X} : (x, u) \in \text{dom } Ef - (\bar{x}, \bar{u})\}$$

so, by Lemma 48 in the Appendix,

$$\text{pos}(\text{dom}_u Ef - \bar{u}) = \{u \in \mathcal{U} \mid \exists x \in \mathcal{X} : (x, u) \in \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))\}.$$

This proves the claim. \square

Example 29. Consider Example 22 and assume that $\bar{x} \in \mathcal{X}_a$ is such that $(\bar{x}, A\bar{x} + \bar{u}) \in \text{core}_s \text{ dom } Eg$. If $\{u \in \mathcal{U} \mid (0, u) \in \mathcal{K} - \text{gph } A\}$ is strongly closed, then Assumption 23.1 holds.

Proof. We have

$$\text{dom } Ef = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid (x, Ax + u) \in \text{dom } Eg\}$$

so

$$\text{pos}(\text{dom } Ef - (\bar{x}, \bar{u})) = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid (x, Ax + u) \in \text{pos}(Eg - (\bar{x}, A\bar{x}, \bar{u}))\}.$$

The projection of this set to \mathcal{U} can be written as

$$\begin{aligned} & \{u \in \mathcal{U} \mid \exists x \in \mathcal{X} : (x, u) \in \text{pos}(\text{dom } Ef - (\bar{x}, \bar{u}))\} \\ &= \{u \in \mathcal{U} \mid \exists x \in \mathcal{X} : (x, Ax + u) \in \mathcal{K}\} \\ &= \{u \in \mathcal{U} \mid (0, u) \in \mathcal{K} - \text{gph } A\} \end{aligned}$$

so the claim follows from Lemma 28. \square

We say that a set-valued mapping $S : \Omega \rightrightarrows \mathbb{R}^{n_0} \times \dots \times \mathbb{R}^{n_T}$ is $(\mathcal{F}_t)_{t=0}^T$ -adapted if, for each $t = 0, \dots, T$, the projection

$$S^t(\omega) := \{x^t \in \mathbb{R}^{n_t} \mid x \in S(\omega)\}$$

is \mathcal{F}_t -measurable. If $S(\omega) = \{s(\omega)\}$ for a stochastic process s , then S is adapted if and only if the process s is adapted.

Our condition in the following example extends the "bounded recourse condition" of [26] from problems of Bolza to the general setting. If S is essentially bounded, then the condition becomes the "relatively complete recourse" condition from [23].

Example 30 (bounded recourse condition). *If $\mathcal{X} = L^\infty$ then Assumption 23.2 holds if, for all r large enough, the mapping*

$$S_r(\omega) := \{x \in \mathbb{B}_r \mid (x, \bar{u}(\omega)) \in \text{dom } f(\omega)\}$$

is adapted and if $(x, \bar{u}) \in \text{dom } Ef$ for all $x \in L^0(S_r)$.

If the set-valued mapping $S(\omega) := \{x \in \mathbb{R}^n \mid (x, \bar{u}(\omega)) \in \text{dom } f(\omega)\}$ is essentially bounded, the above clearly holds if S is adapted. Note, however, that a set-valued mapping S can be adapted while its truncations S_r are not. This happens e.g. when $S(\omega) = \{x \in \mathbb{R}^n \mid \xi(\omega)x_0 \leq x_1\}$ with $\xi \in L^0 \setminus L^\infty$.

Proof. Let $z \in \mathcal{X}$ such that $(z, \bar{u}) \in \text{dom } Ef$ and let $r \geq \|z\|_{L^\infty}$. We have $z \in S_r$ and thus, $z^t \in S_r^t$ almost surely. When S_r is adapted, Jensen's inequality, gives $E_t z^t \in S_r^t$ almost surely. Applying the measurable selection theorem to

$$\omega \mapsto \{x \in \mathbb{R}^n \mid x^t = [E_t z^t](\omega), x \in S_r(\omega)\},$$

gives the existence of a $\bar{z} \in L^0$ such that $\bar{z}^t = E_t z^t$ and $\bar{z} \in S_r$ almost surely. The last condition now gives $(\bar{z}, \bar{u}) \in \text{dom } Ef$. \square

As noted in [23], any problem with essentially bounded domain can be reduced to a problem satisfying relatively complete recourse. The following gives the details.

Remark 31 (Induced constraints). *Assume that $S(\omega) := \text{dom } f(\cdot, \bar{u}(\omega), \omega)$ is closed and essentially bounded and that $(x, \bar{u}) \in \text{dom } Ef$ when $(x, \bar{u}) \in \text{dom } f$ almost surely. The projections $S^t(\omega)$ of $S(\omega)$ are then closed and essentially bounded so, by [32, Theorem 1.8], there exist \mathcal{F}_t -measurable mappings D^t whose \mathcal{F}_t -measurable selections coincide with those of S^t . Clearly, every feasible $x \in \mathcal{X}_a$ has $x^t \in D^t$ almost surely. Thus, defining*

$$\tilde{f}(x, u, \omega) := f(x, u, \omega) + \sum_{t=0}^T \delta_{D^t(\omega)}(x^t),$$

we have $E\tilde{f}(x, \bar{u}) = Ef(x, \bar{u})$ for every $x \in \mathcal{X}_a$. Moreover, \tilde{f} satisfies the assumptions of Example 30. The random sets D^t can be thought of as “induced constraints” arising from the requirement of feasible future recourse. If S is adapted, we have $D^t = S^t$ and $f = \tilde{f}$.

5 Applications

This section applies the general results above to specific instances of $(SP_{\mathcal{X}})$. The dual problems and optimality conditions were derived in [15].

5.1 Mathematical programming

Consider the problem

$$\begin{aligned} & \text{minimize} && Ef_0(x) \quad \text{over } x \in \mathcal{N}, \\ & \text{subject to} && f_j(x) \leq 0 \quad j = 1, \dots, l \text{ a.s.}, \\ & && f_j(x) = 0 \quad j = l + 1, \dots, m \text{ a.s.} \end{aligned} \tag{MP}$$

where f_j are convex normal integrands with f_j affine for $j > l$. This extends the problem formulation from [25] by relaxing the boundedness assumptions on the feasible set and by including the equality constraints.

Problem (MP) fits the general duality framework with $\bar{u} = 0$ and

$$f(x, u, \omega) = \begin{cases} f_0(x, \omega) & \text{if } x \in \text{dom } H, H(x) + u \in K, \\ +\infty & \text{otherwise,} \end{cases}$$

where $K = \mathbb{R}_-^l \times \{0\}$ and H is the K -convex random function defined by

$$\text{dom } H(\cdot, \omega) = \bigcap_{j=1}^m \text{dom } f_j(\cdot, \omega) \quad \text{and} \quad H(x, \omega) = (f_i(x, \omega))_{i=1}^m.$$

It was shown in [15] that if $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, then the dual problem can be written as

$$\begin{aligned} & \text{maximize } E \inf_{x \in \mathbb{R}^n} \{f_0(x) + y \cdot H(x) - x \cdot p\} \text{ over } (p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y} \\ & \text{subject to } y \in K^* \text{ a.s.} \end{aligned} \quad (D_{MP})$$

To get more explicit expressions for f^* and the dual problem, additional structure is needed; see [15].

In order to state our assumptions, we first write the problem in the form

$$\begin{aligned} & \text{minimize } Ef_0(x) \text{ over } x \in \mathcal{N}, \\ & \text{subject to } F(x) \in \mathbb{R}_-^l \text{ a.s.,} \\ & Ax = b \text{ a.s.,} \end{aligned}$$

where F is an \mathbb{R}_-^l -convex random function and A is a random $n \times (m-l)$ -matrix. Without loss of generality, we may assume that $b = (b_t)_{t=0}^T$ and

$$A = \begin{bmatrix} A_{0,0} & 0 & \cdots & 0 \\ \vdots & & & \vdots \\ & & & 0 \\ A_{T,0} & \cdots & A_{T,T} \end{bmatrix},$$

where $b_t \in \mathbb{R}^{m_t^2}$ and $A_{s,t} : \mathbb{R}^{n_s} \rightarrow \mathbb{R}^{m_t^2}$ with $m_1^2 + \cdots + m_T^2 = m-l$. Similarly, we may assume that $F(x, \omega) = (F^0(x^0, \omega), \dots, F^T(x^T, \omega))$, where F^t is a $\mathbb{R}_-^{m_t^1}$ -convex random function on \mathbb{R}^{n_t} with $m_1^1 + \cdots + m_T^1 = l$.

Accordingly, we can write the normal integrand f as

$$\begin{aligned} f(x, u, \omega) &= \begin{cases} f_0(x, \omega) & \text{if } x \in \text{dom } H, H(x) + u \in K, \\ +\infty & \text{otherwise,} \end{cases} \\ &= \begin{cases} f_0(x, \omega) & \text{if } F(x, \omega) + (u_j)_{j=1}^l \in \mathbb{R}_-^l, A(\omega)x + (u_j)_{j=l+1}^m = b(\omega), \\ +\infty & \text{otherwise.} \end{cases} \end{aligned}$$

We assume that $\mathcal{X} = L^\infty$ and that $\mathcal{U} = L^\infty(\mathbb{R}^l) \times \mathcal{U}^e$, where \mathcal{U}^e is a space of \mathbb{R}^{m-l} -valued random variables satisfying the assumptions in Section 2.2. We denote $A^t := (A_{t,0}, \dots, A_{t,t})$. Since \mathcal{U}^e is solid, we can write it accordingly as $\mathcal{U}^e = \mathcal{U}_0^e \times \cdots \times \mathcal{U}_T^e$, where \mathcal{U}_t^e is a decomposable solid Fréchet space of $\mathbb{R}^{m_t^2}$ -valued random variables.

Assumption 32.

A Ef_0 is finite on \mathcal{X} and there exist $\bar{x} \in \mathcal{X}_a$ and $\epsilon > 0$ such that

$$\begin{aligned} f_j(\bar{x} + x) + \epsilon &\leq 0 \quad \forall x \in \mathbb{B}_\epsilon, \quad j = 1, \dots, l \text{ a.s.} \\ A(\bar{x}) &= b \text{ a.s.} \end{aligned}$$

B $A\mathcal{X} \subseteq \mathcal{U}^e$ and $A\mathcal{X}_a$ and $A\mathcal{X}$ are strongly closed in \mathcal{U}^e ,

C For every t , A^t , b_t and F^t are \mathcal{F}_t -measurable, and

$$\{x_t \in \mathcal{X}_t \mid F^t(x^{t-1}, x_t) \in \mathbb{R}_-^{m_t^1}, A^t(x^{t-1}, x_t) = b_t \text{ a.s.}\}$$

is nonempty for every \mathcal{F}_{t-1} -measurable $x^{t-1} \in \mathcal{X}^{t-1}$ such that $F^{t'}(x^{t'}) \in \mathbb{R}_-^{m_{t'}^1}$ and $A^{t'}x^{t'} = b_{t'}$ for all $t' < t$.

The following extends the main results of [25] relaxing the compactness assumptions made there and by allowing for affine equality constraints. The strategy of [25] was to first relax the nonanticipativity constraint by using a shadow price of information p and then to construct a dual variable y via measurable selection arguments. The following employs the general theory of Sections 3 and 4 to establish the existence of a dual optimal pair (p, y) directly.

Theorem 33. *Under Assumption 32, $\inf(MP) = \sup(D_{MP})$ and the optimum in (D_{MP}) is attained. In particular, a feasible $x \in \mathcal{N}$ solves (MP) if and only if there exists $(p, y) \in \mathcal{N}^\perp \times \mathcal{Y}$ feasible in (D_{MP}) such that*

$$\begin{aligned} p &\in \partial_x[f_0 + y \cdot H](x), \\ H(x) &\in K, \quad y \in K^*, \quad y \cdot H(x) = 0 \end{aligned}$$

almost surely.

Proof. Recall that, in (MP) , $\bar{u} = 0$. It was shown in [15, Section 6.1] that $(p, y) \in \partial f(x, \bar{u})$ can be written as the scenario-wise optimality conditions given here. Thus, by Theorem 27, it suffices to show that $\text{pos dom } \varphi$ is linear and closed and Assumption 23 holds. By Assumption 32.A,

$$\begin{aligned} \text{dom } Ef - (\bar{x}, 0) &= \{(z, u) \in \mathcal{X} \times \mathcal{U} \mid f_j(\bar{x} + z) + u_j \leq 0 \ j = 1, \dots, l \text{ a.s.}\} \\ &\quad \cap \{(z, u) \in \mathcal{X} \times \mathcal{U} \mid Az + (u_j)_{j=l+1}^m = 0\}, \end{aligned}$$

where the first set on the right has $(0, 0)$ in its strong interior. Thus, by Lemma 48 in the Appendix,

$$\text{pos}(\text{dom } Ef - (\bar{x}, 0)) = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid Ax + (u_j)_{j=l+1}^m = 0\}.$$

By Lemma 8, the first property in B implies that $A : \mathcal{X} \rightarrow \mathcal{U}$ is continuous so the problem is strictly feasible. By Lemma 19,

$$\begin{aligned} \text{pos dom } \varphi &= \text{pos}(\text{dom } Ef - (\bar{x}, 0)) - \mathcal{X}_a \times \{0\} \\ &= \{(z, u) \in \mathcal{X} \times \mathcal{U} \mid Az + (u_j)_{j=l+1}^m \in A\mathcal{X}_a\}, \end{aligned}$$

so $\text{pos dom } \varphi$ linear closed by B. Indeed, by Lemma 8, the first property in B implies that $A : \mathcal{X} \rightarrow \mathcal{U}$ is continuous.

The projection of $\text{pos}(\text{dom } Ef - (0, 0))$ to \mathcal{U} equals $L^\infty(\mathbb{R}^l) \times A\mathcal{X}$, which is strongly closed by the last property in B. Thus, by Lemma 28, Assumption 23.1

holds. As to Assumption 23.2, let $z \in \mathcal{X}$ with $(z, 0) \in \text{dom } Ef$. By Assumption 32.C, A^t , b_t and F^t are \mathcal{F}_t -measurable, so, by Jensen's inequality,

$$\begin{aligned} F^{t'}(E_t z^{t'}) &\in \mathbb{R}_-^{t'} \quad \forall t' \leq t, \\ A^{t'}(E_t z^{t'}) &= b_{t'} \quad \forall t' \leq t. \end{aligned}$$

Under Assumption 32.C, $E_t z^t$ can be extended to a \bar{z} such that $\bar{z}^t = E_t z^t$ and $(\bar{z}, 0) \in \text{dom } Ef$. \square

The following gives sufficient conditions for the closedness conditions in Assumption 32.B. Recall that, if A is a matrix with full row rank, its Moore–Penrose inverse is given by

$$A^\dagger = A^*(A^*A)^{-1}.$$

Remark 34. Assume that $A\mathcal{X} \subseteq \mathcal{U}^e$ and that, for every t , A^t is \mathcal{F}_t -measurable and

$$\mathcal{U}_t^e = A_{t,t}\mathcal{X}_t. \quad (6)$$

Then *B* holds. If there are no inequality constraints and b_t are \mathcal{F}_t -measurable, then *C* holds.

Condition (6) holds trivially if \mathcal{U}_t^e is the range space of the mapping $\mathcal{A}_{t,t} : \mathcal{X}_t \rightarrow L^0$ defined pointwise by $(\mathcal{A}_{t,t}x_t)(\omega) = A(\omega)x_t(\omega)$. By Example 11, such a space \mathcal{U}_t^e satisfies the assumptions of the present chapter (i.e. it is a solid Fréchet space and its dual is a direct sum of measurable functions and singular elements) provided $|A_{t,t}| |A_{t,t}^\dagger| \in L^\infty$ and

$$\mathcal{X}_t = \{x_t \in L^0 \mid |x_t| \in \mathcal{X}_t^0\}$$

for solid decomposable spaces of scalar-valued random variables \mathcal{X}^0 .

Proof. To verify *B*, it suffices to prove that $A\mathcal{X}_a = \mathcal{U}_a^e$ and $A\mathcal{X} = \mathcal{U}^e$. Here \mathcal{U}_a^e is the set of $u \in \mathcal{U}^e$ such that the component u_t of u corresponding to A^t is \mathcal{F}_t -measurable. The block-triangular structure and the \mathcal{F}_t -measurability of A^t imply $A\mathcal{X}_a \subseteq \mathcal{U}_a^e$. Given $u \in \mathcal{U}_a^e$, assume that there exists $x^{t-1} \in \mathcal{X}_a^{t-1}$ such that $A^{t'}x^{t'} + u_{t'} = 0$ for all $t' \leq t-1$. This holds trivially for $t=0$. Since $A\mathcal{X} \subseteq \mathcal{U}^e$, condition (6) gives the existence of an $x_t \in \mathcal{X}_t$ with $A^t x_t + u_t = 0$. Taking conditional expectations, $A^t E_t x_t + u_t = 0$, so there exists $x^t \in \mathcal{X}_a^t$ with $A^{t'} x^{t'} + u_{t'} = 0$ for all $t' \leq t$. By induction, $A\mathcal{X}_a = \mathcal{U}_a^e$. The equality, $A\mathcal{X} = \mathcal{U}^e$ is proved similarly. Assumption *A* implies $b \in \mathcal{U}^e$ so, in the absence of inequality constraints, the above argument also shows that *C* holds. \square

5.2 Optimal stopping

Let R be a real-valued adapted stochastic process with $R_t \in L^1$ for all t and consider the *optimal stopping problem*

$$\text{maximize } ER_\tau \quad \text{over } \tau \in \mathcal{T}, \quad (OS)$$

where \mathcal{T} is the set of *stopping times*, i.e. measurable functions $\tau : \Omega \rightarrow \{0, \dots, T+1\}$ such that $\{\omega \in \Omega \mid \tau(\omega) \leq t\} \in \mathcal{F}_t$ for each $t = 0, \dots, T$. Choosing $\tau = T+1$ is interpreted as not stopping at all. The problem

$$\begin{aligned} & \text{maximize} && E \sum_{t=0}^T R_t x_t \quad \text{over } x \in \mathcal{N}, \\ & \text{subject to} && x \geq 0, \sum_{t=0}^T x_t \leq 1 \quad \text{a.s.} \end{aligned} \tag{ROS}$$

is the convex relaxation of (OS) in sense that their optimal values coincide and the extreme points of the feasible set of (ROS) can be identified with \mathcal{T} ; see [16, Section 5.2].

Problem (ROS) fits the general duality framework with $n_t = 1$, $\mathcal{X} = L^\infty$, $\mathcal{V} = L^1$, $m = 1$, $\mathcal{U} = L^\infty$, $\mathcal{Y} = L^1$,

$$f(x, u, \omega) = \begin{cases} -\sum_{t=0}^T x_t R_t(\omega) & \text{if } x \geq 0 \text{ and } \sum_{t=0}^T x_t + u \leq 0, \\ +\infty & \text{otherwise} \end{cases}$$

and $\bar{u} = -1$. It was shown in [15, Section 6.2] that the dual of (ROS) can be written as

$$\begin{aligned} & \text{minimize} && Ey \quad \text{over } (p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y}_+ \\ & \text{subject to} && p_t + R_t \leq y \quad t = 0, \dots, T \text{ a.s.} \end{aligned} \tag{DOS}$$

As noted in [15, Section 6.2], a pair $(p, y) \in \mathcal{V} \times \mathcal{Y}$ solves (DOS) if and only if $p_t = y - E_t y$ and the process $y_t := E_t y$ solves the “reduced dual”

$$\begin{aligned} & \text{minimize} && Ey_0 \quad \text{over } y \in \mathcal{M}_+^\mathcal{Y} \\ & \text{subject to} && R_t \leq y_t \quad t = 0, \dots, T \text{ a.s.,} \end{aligned}$$

where $\mathcal{M}_+^\mathcal{Y}$ is the cone of nonnegative martingales y with $y_T \in \mathcal{Y}$.

Theorem 35. *We have $\inf(OS) = \sup(DOS)$ and the optimum in (DOS) is attained. In particular, a stopping time $\tau \in \mathcal{T}$ is optimal if and only if there exists $(p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y}$ such that $p_t + R_t \leq y$ for all t and $p_\tau + R_\tau = y$ almost surely. This is equivalent to the existence of a martingale y such that $R_t \leq y_t$ for all t and $R_\tau = y_\tau$ almost surely.*

Proof. Recall that in (ROS) , $\bar{u} = -1$. By [15, Theorem 38 and Example 39], it suffices to show that φ is subdifferentiable at $(0, \bar{u})$. Thus, by Theorem 27, it suffices to show that $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed and Assumption 23 holds. We have

$$\text{dom } Ef = \{(x, u) \in L^\infty \times L^\infty \mid x \geq 0, \sum_{t=0}^T x_t + u \leq 0\}$$

and $\text{dom } \varphi = L^\infty$. Thus $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is the whole space so it is linear and closed. We have $\text{dom}_u Ef = L^\infty$ so Assumption 23.1 holds. The mapping $S(\omega) := \text{dom } f(\cdot, \bar{u}(\omega), \omega) = \{x \in \mathbb{R}^n \mid x \geq 0, \sum_{t=0}^T x_t \leq 1\}$ is deterministic and, in particular, adapted so Assumption 23.2 holds by Example 30. \square

5.3 Optimal control

Consider the optimal control problem

$$\begin{aligned} & \text{minimize} && E \left[\sum_{t=0}^T L_t(X_t, U_t) \right] && \text{over } (X, U) \in \mathcal{N}, \\ & \text{subject to} && \Delta X_t = A_t X_{t-1} + B_t U_{t-1} + W_t && t = 1, \dots, T \end{aligned} \quad (OC)$$

where the *state* X and the *control* U are processes with values in \mathbb{R}^N and \mathbb{R}^M , respectively, A_t and B_t are \mathcal{F}_t -measurable random matrices, W_t is an \mathcal{F}_t -measurable random vector and the functions L_t are convex normal integrands. The linear constraints in (OC) are called the *system equations*.

The problem fits the general duality framework with $x = (X, U)$, $\bar{u} = (W_t)_{t=1}^T$ and

$$f(x, u, \omega) = \sum_{t=0}^T L_t(X_t, U_t, \omega) + \sum_{t=1}^T \delta_{\{0\}}(\Delta X_t - A_t(\omega)X_{t-1} - B_t(\omega)U_{t-1} - u_t).$$

As in [15, Section 6.3], we assume that

$$\begin{aligned} \mathcal{X}_t &= \mathcal{S} \times \mathcal{C}, & \mathcal{U}_t &= \mathcal{S} \\ \mathcal{V}_t &= \mathcal{S}' \times \mathcal{C}', & \mathcal{Y}_t &= \mathcal{S}' \end{aligned}$$

where \mathcal{S} and \mathcal{C} are solid decomposable spaces in separating duality with \mathcal{S}' and \mathcal{C}' , respectively. In order to apply the general theory of Sections 3 and 4, we assume, in addition, that $E_t \mathcal{S} \subseteq \mathcal{S}$, $E_t \mathcal{C} \subseteq \mathcal{C}$ for all t and that \mathcal{S} and \mathcal{C} are endowed with Fréchet topologies under which their topological duals can be expressed as

$$\mathcal{S}^* = \mathcal{S}' \oplus (\mathcal{S}')^s \quad \text{and} \quad \mathcal{C}^* = \mathcal{C}' \oplus (\mathcal{C}')^s,$$

where $(\mathcal{S}')^s$ and $(\mathcal{C}')^s$ are singular elements of \mathcal{S}^* and \mathcal{C}^* , respectively. These conditions simply mean that the spaces \mathcal{X} and \mathcal{U} satisfy the assumptions made in Sections 3 and 4.

It was shown in [15, Section 6.3] that if $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, then the dual of (OC) can be written as

$$\begin{aligned} & \text{maximize} && E \left[\sum_{t=1}^T W_t \cdot y_t - \sum_{t=0}^T L_t^*(p_t - (\Delta y_{t+1} + A_{t+1}^* y_{t+1}, B_{t+1}^* y_{t+1})) \right] \\ & \text{over} && (p, y) \in \mathcal{X}_a^\perp \times \mathcal{Y}. \end{aligned} \quad (DOC)$$

Assumption 36. For $t = 0, \dots, T$

A $A_t \mathcal{S} \subseteq \mathcal{S}$ and $B_t \mathcal{C} \subseteq \mathcal{C}$,

B EL_t is proper on $\mathcal{S} \times \mathcal{C}$, and there exists a feasible $(\bar{X}, \bar{U}) \in \mathcal{X}_a$ such that

$$\mathcal{K}_t := \text{pos}(\text{dom } EL_t - (\bar{X}_t, \bar{U}_t)) \quad t = 0, \dots, T$$

are linear and strongly closed.

- C (a) $E_t \mathcal{K}_{t'} \subset \mathcal{K}_{t'}$ for all $t' \leq t$
(b) for every $X_t \in \mathcal{S}$, there exists $U_t \in \mathcal{C}$ with $(X_t, U_t) \in \mathcal{K}_t$.

Clearly, Assumption 36.B and C hold if the functions EL_t are finite on $\mathcal{S} \times \mathcal{C}$. Assumption 36.A was used also in [15, Section 6.3] to derive a reduced dual problem. Part (a) of C holds automatically if each L_t is \mathcal{F}_t -measurable.

Let \mathcal{Y}_a be the adapted elements of \mathcal{Y} . By Lemma 7, ${}^a y \in \mathcal{Y}_a$ for every $y \in \mathcal{Y}$, where $({}^a y)_t := E_t y_t$ is the *adapted projection* of y . As noted in [15, Section 6.3], if Assumption 36.A holds, each L_t is \mathcal{F}_t -measurable and EL_t is proper on $\mathcal{S} \times \mathcal{C}$, then a pair (p, y) solves (D_{OC}) if and only if ${}^a y$ solves the reduced dual

$$\text{maximize} \quad E \left[\sum_{t=1}^T W_t \cdot y_t - \sum_{t=0}^T [L_t^*(-E_t(\Delta y_{t+1} + A_{t+1}^* y_{t+1}, E_t B_{t+1}^* y_{t+1}))] \right] \quad \text{over} \quad y \in \mathcal{Y}_a$$

and

$$p_t = (\Delta y_{t+1} + A_{t+1}^* y_{t+1}, B_{t+1}^* y_{t+1}) - E_t(\Delta y_{t+1} + A_{t+1}^* y_{t+1}, B_{t+1}^* y_{t+1}).$$

Lemma 37. *Under Assumption 36, $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed and Assumption 23.1 holds.*

Proof. This fits the format of Example 22 with $f(x, u) = g(x, Ax + u)$, where $g(x, u) := \sum L_t(x_t) + \delta_{\{0\}}(u)$, $A(x)_t = -\Delta X_t + A_t X_{t-1} + B_t U_{t-1}$ and $\pi'(x, u) := ({}^a x, {}^a u)$. By Assumption 36.B,

$$\mathcal{K} := \text{pos}(\text{dom } Eg - (\bar{x}, A\bar{x} + \bar{u})) = \{(x, u) \in \mathcal{X} \times \mathcal{U} \mid x_t \in \mathcal{K}_t \ \forall t, u = 0\}$$

is linear and strongly closed. By Assumption 36.C, $\pi' \mathcal{K} \subseteq \mathcal{K}$. By Theorem 18 and Example 22, it thus suffices to show that

$$(I, A)\mathcal{X}_a \subset \text{rge } \pi' \subseteq \mathcal{K} + (I, A)\mathcal{X}_a.$$

We have $\text{rge } \pi' = \mathcal{X}_a \times \mathcal{U}_a$. Since A_t and B_t are \mathcal{F}_t -measurable, the first inclusion is clear. As to the second, let $(x, u) \in \mathcal{X}_a \times \mathcal{U}_a$. We have $(x, u) = (I, A)x + (0, u - Ax)$, where $Ax \in \mathcal{U}_a$, so it suffices to show that $(0, u) \in \mathcal{K} + (I, A)\mathcal{X}_a$, i.e., there exists $x \in \mathcal{X}_a$ with $(X_t, U_t) \in \mathcal{K}_t$ and $\Delta X_t = A_t X_{t-1} + B_t U_{t-1} + u_t$ for all t .

Assume that $(X^{t-1}, U^{t-1}) \in \mathcal{X}_a^{t-1}$ with

$$\Delta X_{t'} = A_{t'} X_{t'-1} + B_{t'} U_{t'-1} + u_{t'} \quad \forall t' = 0 \dots t-1$$

and $(X_{t-1}, U_{t-1}) \in \mathcal{K}_{t-1}$. By Assumption 36.A, the \mathcal{F}_t -measurable X_t defined by

$$\Delta X_t := A_t X_{t-1} + B_t U_{t-1} + u_t$$

belongs to \mathcal{S} . By Assumption 36.C(b), there exists \tilde{U}_t such that $(X_t, \tilde{U}_t) \in \mathcal{K}_t$. Since X_t is \mathcal{F}_t -measurable, Assumption 36.C.(a) gives $(X_t, U_t) \in \mathcal{K}_t$, where $U_t := E_t \tilde{U}_t$. Thus the claim follows by induction.

To verify Assumption 23.1, we apply Example 29. It suffices to show that $\{u \in \mathcal{U} \mid (0, u) \in \mathcal{K} - \text{rge } A\} = \mathcal{U}$. This means that, for every $u \in \mathcal{U}$, there exists $x \in \mathcal{X}$ with $(X_t, U_t) \in \mathcal{K}_t$ and $\Delta X_t = A_t X_{t-1} + B_t U_{t-1} + u_t$ for all t . The existence of such x follows by repeating the arguments in the above paragraph. \square

Assumption 38. For all t ,

1. $E_t \text{ dom } EL_{t'} \subset \text{dom } EL_{t'}$ for all $t' \leq t$
2. for every $X_t \in \mathcal{S}$, there exists $U_t \in \mathcal{C}$ with $(X_t, U_t) \in \text{dom } EL_t$.

Assumption 38 implies Assumption 36.C. Note also that Assumption 38.1 holds automatically if each L_t is \mathcal{F}_t -measurable.

Theorem 39. *Under Assumptions 36 and 38, problem (OC) is feasible for any $W \in \mathcal{U}_a$, $\inf(OC) = \sup(D_{OC})$ and the optimum in (D_{OC}) is attained. In particular, a feasible (X, U) is optimal in (OC) if and only if there exists a dual feasible $(p, y) \in \mathcal{N}^\perp \times \mathcal{Y}$ such that*

$$p_t - (\Delta y_{t+1} + A_{t+1}^* y_{t+1}, B_{t+1}^* y_{t+1}) \in \partial L_t(X_t, U_t)$$

almost surely. If each L_t is \mathcal{F}_t -measurable, this is equivalent to the existence of a $y \in \mathcal{Y}_a$ feasible in the reduced dual such that

$$-E_t(\Delta y_{t+1} + A_{t+1}^* y_{t+1}, B_{t+1}^* y_{t+1}) \in \partial L_t(X_t, U_t)$$

almost surely.

Proof. Recall that in (OC), $\bar{u} = W$. By [15, Theorem 41 and Remark 45], it suffices to show that φ is subdifferentiable at $(0, \bar{u})$. Thus, by Theorem 27 and Lemma 37, it suffices to show that Assumption 23.2 holds and that $(0, \bar{u}) \in \text{dom } \varphi$ for all $\bar{u} \in \mathcal{U}_a$. We start with the latter.

Let $\bar{u} \in \mathcal{U}_a$ and assume that $(X^{t-1}, U^{t-1}) \in \mathcal{X}_a^{t-1}$ with

$$\Delta X_{t'} = A_{t'} X_{t'-1} + B_{t'} U_{t'-1} + \bar{u}_{t'} \quad \forall t' = 0 \dots t-1$$

and $(X_{t-1}, U_{t-1}) \in \text{dom } EL_{t-1}$. By Assumption 36, the \mathcal{F}_t -measurable X_t defined by

$$\Delta X_t = A_t X_{t-1} + B_t U_{t-1} + \bar{u}_t$$

belongs to \mathcal{S} . By Assumption 38.2, there exists \tilde{U}_t such that $(X_t, \tilde{U}_t) \in \text{dom } EL_t$. Since X_t is \mathcal{F}_t -measurable, Assumption 38.1 gives $(X_t, U_t) \in \text{dom } EL_t$, where $U_t := E_t \tilde{U}_t$. By induction, there exists $x \in \mathcal{X}_a$ such that $(x, \bar{u}) \in \text{dom } Ef$, so $(0, \bar{u}) \in \text{dom } \varphi$.

To verify Assumption 23.2, let $(z, \bar{u}) \in \text{dom } Ef$. For all $t' \leq t$, $E_t z_{t'} \in \text{dom } EL_{t'}$, by Assumption 38.1, and, by [15, Lemma 44],

$$E_t X_{t'} = E_t X_{t'-1} + A_{t'} E_t X_{t'-1} + B_{t'} E_t U_{t'-1} + \bar{u}_{t'}.$$

An induction argument similar to the above paragraph gives $\bar{z} \in \mathcal{X}$ with $(\bar{z}, \bar{u}) \in \text{dom } Ef$ and $\bar{z}^t = E_t z^t$, so Assumption 23.2 holds. \square

Example 40 (Bounded strategies). Assume that $\mathcal{S} = L^\infty$ and $\mathcal{C} = L^\infty$ are endowed with the usual norm-topologies. If EL_t is finite on

$$\mathcal{D}_t := \{(X, U) \in L^\infty \times L^\infty \mid \exists \epsilon > 0 : (X_t, U_t) + \epsilon \mathbb{B} \subset \text{dom } L_t \text{ P-a.s.}\}$$

and if there is a feasible point (\bar{X}, \bar{U}) such that $(\bar{X}_t, \bar{U}_t) \in \mathcal{D}_t$ for all t , then, by Example 4, Assumption 36.B holds.

5.4 Problems of Lagrange

Consider the problem

$$\text{minimize } E \sum_{t=0}^T K_t(x_t, \Delta x_t) \quad \text{over } x \in \mathcal{N}, \quad (L)$$

where x is a process of fixed dimension d , K_t are convex normal integrands and $x_{-1} := 0$. Problem (L) can be thought of as a discrete-time version of a problem studied in calculus of variations. Other problem formulations have $K_t(x_{t-1}, \Delta x_t)$ instead of $K_t(x_t, \Delta x_t)$ in the objective, or an additional term of the form $Ek(x_0, x_T)$ (see [26]), all of which fit the general format of stochastic optimization.

This fits the general duality framework with $\bar{u} = 0$ and

$$f(x, u, \omega) = \sum_{t=0}^T K_t(x_t, \Delta x_t + u_t, \omega).$$

As in [15, Section 6.4], we assume that

$$\mathcal{X}_t = \mathcal{S}, \quad \mathcal{Y}_t = \mathcal{S}', \quad \mathcal{U} = \mathcal{X}, \quad \mathcal{Y} = \mathcal{V},$$

where \mathcal{S} is solid decomposable space in separating duality with \mathcal{S}' . In addition, we assume that $E_t \mathcal{S} \subset \mathcal{S}$ for all t and that \mathcal{S} is endowed with a Fréchet topology under which its topological dual can be expressed as

$$\mathcal{S}^* = \mathcal{S}' \oplus (\mathcal{S}')^s,$$

where $(\mathcal{S}')^s$ are singular elements of \mathcal{S}^* . These conditions simply mean that the spaces \mathcal{X} and \mathcal{U} satisfy the assumptions made in Sections 3 and 4.

It was shown in [15, Section 6.4] that if $\text{dom } Ef \cap (\mathcal{X} \times \mathcal{U}) \neq \emptyset$, then the dual problem can be written as

$$\text{maximize } E \left[- \sum_{t=0}^T K_t^*(p_t + \Delta y_{t+1}, y_t) \right] \quad \text{over } y \in \mathcal{Y}, p \in \mathcal{X}_a^\perp \quad (D_L)$$

where $y_{T+1} := 0$.

As noted in [15, Section 6.4], if each K_t is \mathcal{F}_t -measurable and EK_t is proper on $\mathcal{S} \times \mathcal{S}$, then a pair (p, y) solves the dual problem if and only if ${}^a y$ solves the reduced dual problem

$$\text{maximize} \quad E\left[-\sum_{t=0}^T K_t^*(E_t \Delta y_{t+1}, y_t)\right] \quad \text{over } y \in \mathcal{Y}_a.$$

and $p_t = E_t y_{t+1} - y_{t+1}$.

Assumption 41.

A each EK_t is proper lsc on $\mathcal{S} \times \mathcal{S}$.

B there exists $\bar{x} \in \mathcal{X}_a$ such that, for all t ,

$$\mathcal{K}_t := \text{pos}(\text{dom } EK_t - (\bar{x}_t, \Delta \bar{x}_t))$$

is linear and strongly closed,

C for every t ,

(a) $E_t \mathcal{K}_{t'} \subset \mathcal{K}_{t'}$ for all $t' \leq t$,

(b) for every $u_t, x_{t-1} \in \mathcal{S}$, there exists $x_t \in \mathcal{S}$ such that $(x_t, \Delta x_t + u_t) \in \mathcal{K}_t$.

Lemma 42. *Under Assumption 41, $\text{pos dom } \varphi$ is linear and closed and Assumption 23.1 holds.*

Proof. This fits the format of Example 22 with $f(x, u) = g(x, Ax + u)$, where $g(x, u) := \sum_{t=0}^T K_t(x_t, u_t)$, $A(x)_t := \Delta x_t$ and $\pi'(x, u) := ({}^a x, {}^a u)$. We have

$$\mathcal{K} := \text{pos}(Eg - (\bar{x}, A\bar{x} + \bar{u})) = \prod_{t=0}^T \mathcal{K}_t,$$

which is linear and strongly closed, by Assumption 41.B. Assumption 41.C(a) implies $\pi' \mathcal{K} \subseteq \mathcal{K}$. By Theorem 18 and Example 22, it thus suffices to show that

$$(I, A)\mathcal{X}_a \subseteq \text{rge } \pi' \subseteq \mathcal{K} - (I, A)\mathcal{X}_a.$$

The first inclusion is clear.

The second inclusion means that, for every $(x, u) \in \text{rge } \pi' = \mathcal{X}_a \times \mathcal{X}_a$, there exists $\tilde{x} \in \mathcal{X}_a$ such that $(x_t + \tilde{x}_t, u_t + \Delta \tilde{x}_t) \in \mathcal{K}_t$ for all t . By change of variables, this means for every $u \in \mathcal{X}_a$, there exists $\tilde{x} \in \mathcal{X}_a$ such that $(\tilde{x}_t, u_t + \Delta \tilde{x}_t) \in \mathcal{K}_t$ for all t . Assume that there is $\tilde{x}^t \in \mathcal{X}_a^t$ such that $(x_{t'}, u_{t'} + \Delta \tilde{x}_{t'}) \in \mathcal{K}_{t'}$ for all $t' \leq t$. By Assumption 41.C(b), there exists $\tilde{x}_{t+1} \in \mathcal{S}$ such that $(\tilde{x}_{t+1}, u_{t+1} + \Delta \tilde{x}_{t+1}) \in \mathcal{K}_{t+1}$. By Assumption 41.C(a) and \mathcal{F}_{t+1} measurability of x_{t+1} and u_{t+1} , we can choose \tilde{x}_{t+1} as \mathcal{F}_{t+1} -measurable. Thus the second inclusion follows by induction on t .

To verify Assumption 23.1, we apply Example 29. It suffices to show that $\{u \in \mathcal{U} \mid (0, u) \in \mathcal{K} - \text{rge } A\} = \mathcal{U}$. This means that, for every $u \in \mathcal{U}$, there exists $x \in \mathcal{X}$ with $(x_t, \Delta x_t + u_t) \in \mathcal{K}_t$ for all t . The existence of such x follows by repeating the induction argument in the above paragraph. \square

Assumption 43. For every t ,

1. $E_t \text{ dom } EK_{t'} \subset \text{ dom } EK_{t'}$ for all $t' \leq t$,
2. for every \mathcal{F}_t -measurable $x^t \in \mathcal{X}^t$ such that $(x_t, \Delta x_t) \in \text{ dom } EK_t$, there exists $x_{t+1} \in \mathcal{X}_{t+1}$ with $(x_{t+1}, \Delta x_{t+1}) \in \text{ dom } EK_{t+1}$.

When

$$\text{ dom } EK_t = \{(x_t, u_t) \in \mathcal{S} \times \mathcal{S} \mid (x_t, u_t) \in \text{ dom } K_t \text{ a.s.}\},$$

1 holds, by Jensen, as soon as $\text{ dom } K_t$ is \mathcal{F}_t -measurable. Moreover, 2 holds if $\text{ dom}_1 EK_t$ or $\text{ dom}_2 EK_t$ is the whole space.

Like in [15, Section 6.4], we will write the optimality conditions in terms of the associated *Hamiltonians*

$$H_t(x_t, y_t, \omega) := \inf_{u_t \in \mathbb{R}^d} \{K_t(x_t, u_t, \omega) - u_t \cdot y_t\}.$$

Theorem 44. *Under Assumptions 41 and 43, $\inf(L) = \sup(D_{OC})$ and the optimum in (D_L) is attained. In particular, a feasible x is optimal in (L) if and only if there exists a dual feasible $(p, y) \in \mathcal{N}^\perp \times \mathcal{Y}$ such that*

$$\begin{aligned} p_t + \Delta y_{t+1} &\in \partial_x H_t(x_t, y_t), \\ \Delta x_t &\in \partial_y [-H_t](x_t, y_t), \end{aligned}$$

almost surely. If, in addition, each K_t is \mathcal{F}_t measurable, this is equivalent to the existence of a $y \in \mathcal{Y}_a$ feasible in the reduced dual such that

$$\begin{aligned} E_t \Delta y_{t+1} &\in \partial_x H_t(x_t, y_t), \\ \Delta x_t &\in \partial_y [-H_t](x_t, y_t), \end{aligned}$$

almost surely.

Proof. Recall that in (L) , $\bar{u} = 0$. By [15, Theorem 49 and Remark 51], it suffices to show that φ is subdifferentiable at the origin. Thus, by Theorem 27 and Lemma 42, it suffices to show that Assumption 23.2 holds.

Let $z \in \mathcal{X}$ be such that $Ef(z, 0) < \infty$. Let $\bar{z}^t = E_t z^t$. By Assumption 43.1, $(\bar{z}_{t'}, \Delta \bar{z}_{t'}) = E_t(z_{t'}, \Delta z_{t'}) \in \text{ dom } EK_{t'}$ for all $t' \leq t$. By Assumption 43.2, there is an $\tilde{z}_{t+1} \in \mathcal{X}_{t+1}$ with $(\tilde{z}_{t+1}, \Delta \tilde{z}_{t+1}) \in \text{ dom } EK_{t+1}$. Let $\bar{z}_{t+1} = E_{t+1} \tilde{z}_{t+1}$. By Assumption 43.1, $(\bar{z}_{t+1}, \Delta \bar{z}_{t+1}) \in \text{ dom } EK_{t+1}$. Repeating until T , we find a $\bar{z} \in \mathcal{X}$ that satisfies the conditions in Assumption 23.2. \square

Example 45 (Bounded strategies). *Assume that $\mathcal{S} = L^\infty$ is endowed with the usual norm-topology. If EK_t is finite on*

$$\mathcal{D}_t := \{(x_t, u_t) \in L^\infty \times L^\infty \mid \exists \epsilon > 0 : (x_t, u_t) + \epsilon \mathbb{B} \subset \text{ dom } K_t \text{ } P\text{-a.s.}\}$$

and if there is a feasible \bar{x} such that $(\bar{x}_t, \Delta \bar{x}_t) \in \mathcal{D}_t$ for all t , then, by Example 4, Assumption 41.B holds.

5.5 Financial mathematics

This section analyzes the semi-static hedging problem from [15]. Let $s = (s_t)_{t=0}^T$ be an adapted \mathbb{R}^J -valued stochastic process describing the unit prices of a finite set J of perfectly liquid tradeable assets. We also assume that there is a finite set K of derivative assets that can be bought or sold at time $t = 0$ and that provide random payments $C^j \in L^0, j \in K$ at time $t = T$. We denote $C = (C^j)_{j \in K}$. The cost of buying a derivative portfolio $\bar{x} \in \mathbb{R}^K$ at the best available market prices is denoted by $S_{-1}(x_{-1})$. Such a function is convex and lsc; see e.g. [10, 12].

Consider the problem of finding a dynamic trading strategy $x = (x_t)_{t=0}^T$ in the liquid assets J and a static portfolio x_{-1} in the derivatives K so that their combined revenue provides the “best hedge” against the financial liability of delivering a random amount $c \in L^0$ of cash at time T . If we assume that cash (or another numeraire asset) is a perfectly liquid asset that can be lent and borrowed at zero interest rate, the problem can be written as

$$\begin{aligned} \text{minimize} \quad & EV \left(c - \sum_{t=0}^{T-1} x_t \cdot \Delta s_{t+1} - C \cdot x_{-1} + S_{-1}(x_{-1}) \right) \text{ over } x \in \mathcal{N}, x_{-1} \in \mathbb{R}^K, \\ \text{subject to} \quad & x_t \in D_t \quad t = 0, \dots, T-1 \text{ a.s.}, \end{aligned} \tag{SSH}$$

where V is a random “loss function” on \mathbb{R} and D_t is a random \mathcal{F}_t -measurable set describing possible portfolio constraints. More precisely, the function V is a convex normal integrand such that $V(\cdot, \omega)$ nondecreasing and nonconstant for all ω . We will assume $D_T = \{0\}$, which means that all positions have to be closed at the terminal date.

As soon as $c \in \mathcal{U}$, problem (SSH) fits the general duality framework with the time index running from -1 to $T-1$, $\mathcal{F}_{-1} = \{\Omega, \emptyset\}$, $\bar{u} = c$ and

$$f(x, u, \omega) = V \left(u - \sum_{t=0}^{T-1} x_t \cdot \Delta s_{t+1}(\omega) - C(\omega) \cdot x_{-1} + S_{-1}(x_{-1}), \omega \right) + \sum_{t=0}^{T-1} \delta_{D_t(\omega)}(x_t, \omega).$$

We will assume that $S_{-1}(0) = 0$ and $0 \in D_t$ almost surely for all t . It was shown in [15, Section 6.5] that as soon as EV is proper on \mathcal{U} , then the dual of (SSH) can be written as

$$\text{maximize}_{p \in \mathcal{X}_t^+, y \in \mathcal{Y}} \quad E \left[cy - V^*(y) - \sum_{t=0}^{T-1} \sigma_{D_t}(p_t + y \Delta s_{t+1}) - (y S_{-1})^*(p_{-1} + y C) \right]. \tag{DSSH}$$

We assume that $\mathcal{X} = L^\infty$ and that \mathcal{U} is a space of real-valued random variables satisfying the assumptions in Section 2.2.

Assumption 46.

- A $\Delta s_t^j, C_j \in \mathcal{U}$ for all $j \in J, \bar{j} \in K$ and $t = 0, \dots, T$,
- B EV is finite on \mathcal{U} , EV^* is proper on \mathcal{Y} ,

C there exists a feasible $\bar{x} \in \mathcal{X}_a$ and an $\epsilon > 0$ such that $\bar{x}_{-1} \in \text{int dom } S_{-1}$ and $\mathbb{B}_\epsilon(\bar{x}_t) \subseteq D_t$ almost surely,

D D_t is \mathcal{F}_t -measurable for $t = 0, \dots, T$ and $S_{-1}(z_{-1}) \in \mathcal{U}$ for all $z_{-1} \in L^\infty(\text{dom } S_{-1})$.

As noted in [15, Section 6.5], if $E_t \mathcal{U} \subseteq \mathcal{U}$ and $\Delta s_{t+1} \in \mathcal{U}$, then (p, y) solves (D_{SSH}) if and only if y solves the reduced dual problem

$$\text{maximize}_{y \in \mathcal{Y}} \quad E \left[cy - V^*(y) - \sum_{t=0}^{T-1} \sigma_{D_t}(E_t[y\Delta s_{t+1}]) - (E[y]S_{-1})^*(E[y\bar{c}]) \right].$$

and

$$p_{-1} := \frac{E[y\bar{c}]}{E[y]}y - y\bar{c} \quad \text{and} \quad p_t = E_t[y\Delta s_{t+1}] - y\Delta s_{t+1} \quad t = 0, \dots, T-1.$$

Theorem 47. *Under Assumption 46, $\inf(SSH) = \sup(D_{SSH})$ and the optimum in (D_{SSH}) is attained. In particular, a feasible $x \in \mathcal{N}$ solves (SSH) if and only if there exists $(p, y) \in \mathcal{N}^\perp \times \mathcal{Y}$ feasible in (D_{SSH}) such that*

$$\begin{aligned} y &\in \partial V(u - \sum_{t=0}^{T-1} x_t \Delta s_{t+1} - C \cdot x_{-1} + S_{-1}(x_{-1})), \\ p_t + y\Delta s_{t+1} &\in N_{D_t}(x_t) \quad t = 0, \dots, T, \\ p_{-1} + yC &\in \partial(yS_{-1})(x_{-1}) \end{aligned}$$

almost surely. This holds if and only if

$$\begin{aligned} y &\in \partial V(u - \sum_{t=0}^{T-1} x_t \Delta s_{t+1} - C \cdot x_{-1} + S_{-1}(x_{-1})), \\ E_t[y\Delta s_{t+1}] &\in N_{D_t}(x_t) \quad t = 0, \dots, T, \\ \frac{E[yC]}{E[y]}y &\in \partial(yS_{-1})(x_{-1}), \end{aligned}$$

where the fraction is interpreted as 0 if $E[y] = 0$.

Proof. It was shown in [15, Theorem 56 and Remark 58] that the scenario-wise optimality condition $(p, y) \in \partial f(x, \bar{u})$ can be written as the scenario-wise optimality conditions given here. Thus, by Theorem 27, it suffices to show that $\text{pos}(\text{dom } \varphi - (0, \bar{u}))$ is linear and closed and Assumption 23 holds.

We have

$$\varphi(z, u) \leq EV(u - \sum_{t=0}^T ((\bar{x}_t + z_t) \cdot \Delta s_{t+1}) - C \cdot (\bar{x}_{-1} + z_{-1}) + S_{-1}(\bar{x}_{-1} + z_{-1})) + E \sum_{t=0}^T \delta_{D_t}(\bar{x}_t + z_t).$$

By A, B and C, the right hand side is finite on a neighborhood of $(0, \bar{u})$, so $(0, \bar{u}) \in \text{int dom } \varphi$. Assumption 23.1 is clear since $\text{dom}_u Ef = \mathcal{U}$. As to Assumption 23.2, note first that $(z, \bar{u}) \in \text{dom } Ef$ implies $z_{-1} \in \text{dom } S_{-1}$ and $z_t \in D_t$

almost surely for $t = 0, \dots, T$. By D, there is a $u \in \mathcal{U}$ such that $S_{-1}(z_{-1}) \leq u$. By the conditional Jensen's inequality, $S_{-1}(E_t z_{-1}) \leq E_t u$ so $E_t z_{-1} \in \text{dom } S_{-1}$ and thus, $S_{-1}(E_t z_{-1}) \in \mathcal{U}$, by D. Applying Jensen's inequality (see e.g. [15, Theorem 8]) to the indicator functions of D_t shows that $E_t z_{t'} \in D_{t'}$ for $t' \leq t$. The process \bar{z} in Assumption 23.2 can thus be taken $\bar{z}_{t'} = E_t z_{t'}$ for $t' \leq t$ and $\bar{z}_{t'} = 0$ for $t' > t$. \square

If, in addition to Assumption 46, we assume that there are no portfolio constraints and that S_{-1} is finite on \mathbb{R}^K , the right side of the inequality in the above proof is, by Corollary 15, $\tau(\mathcal{X} \times \mathcal{U}, \mathcal{V} \times \mathcal{Y})$ -continuous under Assumption 46. Since a convex function that is bounded from above on an open set is continuous on the set, this implies Mackey continuity of φ and thus the existence of a subgradient in $\mathcal{V} \times \mathcal{Y}$. In that case, one can thus avoid going through the arguments in Sections 3 and 4. In the presence of portfolio constraints, the Mackey continuity fails but we still find a dual solution in $\mathcal{V} \times \mathcal{Y}$. Note also that, the above proof gives the existence of a solution in the strong dual (D_s) even without part D in Assumption 46. Strong dual solutions would exist even if, in part C, the finiteness of EV was weakened by only requiring that EV be strongly continuous at $\bar{u} - \sum_{t=0}^{T-1} \bar{x}_t \Delta s_{t+1} - C \cdot \bar{x}_{-1} + S_{-1}(\bar{x}_{-1})$; see Example 3.

6 Appendix

6.1 Relative core of a convex set

Let U be a vector space and $C \subset U$ convex. The *positive hull*

$$\text{pos } C := \bigcup_{\lambda > 0} \lambda C$$

of C is the intersection of all cones containing C . If $0 \in C$ and $0 < \lambda_1 < \lambda_2$, then by convexity, $\lambda_1 C \subseteq \lambda_2 C$.

Lemma 48. *Given a linear $A : X \rightarrow U$ and convex sets $C, C' \subset X$ and $D \subset U$, we have*

1. $\text{pos}(AC) = A \text{pos } C$,
2. $\text{pos}(A^{-1}D) = A^{-1} \text{pos } D$,
3. $\text{pos}(C \times D) = \text{pos } C \times \text{pos } D$ if $0 \in C$ and $0 \in D$,
4. $\text{pos}(C \cap C') = \text{pos } C \cap \text{pos } C'$ if $0 \in C$ and $0 \in C'$,
5. $\text{pos}(C + C') = \text{pos } C + \text{pos } C'$ if $0 \in C$ and $0 \in C'$,

Proof. The first two claims are clear. As to 3, since $0 \in C$ and $0 \in D$, we have $C \times D \subset \text{pos } C \times \text{pos } D$ so $\text{pos}(C \times D) \subset \text{pos } C \times \text{pos } D$. If $(x, u) \in \text{pos } C \times \text{pos } D$, we have $x \in \lambda_1 C$ and $u \in \lambda_2 D$ for some $\lambda_i > 0$. Since the sets contain the origins, we have $x \in \max\{\lambda_1, \lambda_2\}C$ and $u \in \max\{\lambda_1, \lambda_2\}D$,

so $(x, u) \in \text{pos}(C \times D)$. Defining $A : X \rightarrow X \times X$ by $Ax = (x, x)$, we have $C \cap C' = A^{-1}(C \times C')$, so 4 follows from 2 and 3. Defining $A : X \times X \rightarrow X$ by $A(x, x') = x + x'$, we have $A(C \times C') = C + C'$, so 5 follows from 1 and 3. \square

The *core* of a set $C \subset U$, denoted by $\text{core } C$, is the set of points $u \in C$ for which $\text{pos}(C - u) = U$. The *relative core* of C , denoted by $\text{rcore } C$, is the core of C relative to the affine hull of C . Recall that the *affine hull* $\text{aff } C$ of C is the smallest affine set containing C . A set C is *affine* if $\lambda u + (1 - \lambda)u' \in C$ for all $u, u' \in C$ and $\lambda \in \mathbb{R}$.

Lemma 49. *Given a convex set C ,*

1. $\text{rcore } C = \{u \in C \mid \text{pos}(C - u) = \text{aff}(C - u)\}$,
2. $\text{pos}(C - x) = \text{pos}(C - x')$ for every $x, x' \in \text{rcore } C$.

Proof. By definition, $\text{pos}(C - u) \subseteq \text{aff}(C - u)$ for any $u \in C$. The converse holds if and only if $u \in \text{rcore } C$. Since $\text{aff}(C - u)$ is independent of the choice of $u \in C$, the second claim follows from the first one. \square

6.2 Continuity of convex functions

Let U be a topological vector space.

Theorem 50. *A convex function which is bounded from above on an open set is either proper and continuous or identically $-\infty$ throughout the core of its domain.*

Proof. In the case of a proper convex function, the proof can be found e.g. in [7, Proposition I.2.5]. A simple line segment argument shows that, if a convex function equals $-\infty$ at some point, then it equals $-\infty$ throughout the core of its domain. \square

We say that a function g is *relatively continuous* at $u \in \text{dom } g$ if g is continuous at u relative to $\text{aff dom } g$. The following is a straightforward consequence of Theorem 50.

Corollary 51. *A convex function which is bounded from above on a relatively open set of $\text{aff dom } g$ is either proper and relatively continuous or identically $-\infty$ throughout $\text{rcore dom } g$.*

Even simpler conditions for continuity are available if the underlying space is *barreled* in the sense that $\text{core } C = \text{int } C$ for every closed convex set C . Fréchet spaces are barreled. The following implies that the same holds if, instead of being closed, C is the domain of a lsc convex function.

Theorem 52. [21, Corollary 8B] *In a barreled space, a lsc convex function is either proper and continuous or identically $-\infty$ throughout the core of its domain.*

The *relative interior* $\text{rint } C$ of a set C is the interior of C with respect to $\text{aff } C$. If C is a closed subset of a topological vector space U and $\text{aff } C$ is barreled, then the earlier argument applied on (the linear translation of) $\text{aff } C$ gives

$$\text{rint } C = \text{rcore } C.$$

The following is an immediate corollary of Theorem 52. We say that an affine set is *barreled* if its translation to the origin is barreled. Again, a closed affine set in a metrizable space is barreled.

Corollary 53. *Let g be a lsc convex function on a topological vector space. If $\text{aff dom } g$ is barreled, then g is either proper and relatively continuous or identically $-\infty$ throughout $\text{rcore dom } g$.*

Remark 54. *Closed subspaces of barreled spaces need not be barreled but closed subspaces of Fréchet spaces are Fréchet, and thus, barreled.*

Corollary 55. *A proper lsc convex function g with $\text{rcore dom } g \neq \emptyset$ on a Fréchet space is relatively continuous throughout $\text{rcore dom } g$ if and only if $\text{aff dom } g$ is closed.*

Proof. A closed subspace of a Fréchet space is Fréchet and, in particular, barreled. Thus, by Corollary 51, the closedness of $\text{aff dom } g$ implies relative continuity. To prove the converse, let $(u^\nu)_{\nu=\infty}$ be a sequence in $\text{aff dom } g$ converging to a \bar{u} . The sequence is bounded so, for every neighborhood N of the origin, there is an $\alpha > 0$ such that $(u^\nu)_{\nu=\infty} \subset \alpha N$. Relative continuity of g gives the existence of an open set N , a $u_0 \in \text{rcore dom } g$ and an $\beta > g(u_0)$ such that $N \cap \text{aff dom } g \subset (\text{lev}_\beta g - u_0)$. It follows that there is a constant $\alpha > 0$ such that $g(\alpha u^\nu) \leq \beta$ for all ν . The lower semicontinuity of g gives $g(\alpha \bar{u}) \leq \beta$ so $\bar{u} \in \text{aff dom } g$. \square

The following extends Corollary 53 by relaxing the lower semicontinuity assumption.

Theorem 56. *[35, Theorem 2.7.1.(vi)] Let X be a Fréchet space, F a lsc convex function on $X \times U$ and*

$$\varphi(u) := \inf_{x \in X} F(x, u).$$

If $\text{aff dom } \varphi$ is barreled, then φ is either proper and continuous or identically $-\infty$ throughout $\text{rcore dom } \varphi$.

6.3 Conjugates and subgradients

Let U and Y be vector spaces in *separating duality* under a bilinear form

$$(u, y) \mapsto \langle u, y \rangle.$$

The *conjugate* of $g : U \rightarrow \overline{\mathbb{R}}$ is the extended real-valued convex function on \mathcal{Y} defined by

$$g^*(y) = \sup_{u \in U} \{\langle u, y \rangle - g(u)\}.$$

The conjugate of the *indicator function*

$$\delta_C(u) := \begin{cases} 0 & \text{if } u \in C, \\ +\infty & \text{otherwise} \end{cases}$$

of a set $C \subset U$ is the *support function*

$$\sigma_C(y) := \sup_{u \in C} \langle u, y \rangle$$

of C .

By definition, a function and its conjugate satisfy the *Fenchel inequality*

$$g(u) + g^*(y) \geq \langle u, y \rangle$$

for all $u \in U$ and $y \in Y$. When $g(u)$ is finite, the inequality holds as an equality if and only if

$$g(u') \geq g(u) + \langle u' - u, y \rangle \quad \forall u' \in U.$$

We then say that y is a *subgradient* of g at u . The set of subgradients of g at u is known as the *subdifferential* of g at u and denoted by $\partial g(u)$. The subdifferential is defined as the empty set unless $g(u)$ is finite.

Theorem 57. [35, Theorem 2.4.12] *If g is relatively continuous and finite at $u \in \text{dom } g$, then $\partial g(u) \neq \emptyset$.*

Given a proper convex function g , its *recession function* is the convex positively homogeneous function given by

$$g^\infty(u) = \lim_{\alpha \nearrow \infty} \frac{g(\bar{u} + \alpha u) - g(\bar{u})}{\alpha},$$

where $\bar{u} \in \text{dom } g$; see e.g. [19].

Lemma 58. [19, Corollary 3C] *If g^* is proper, then*

$$(g^*)^\infty = \sigma_{\text{dom } g}.$$

The following was used in the proof of Lemma 24.

Lemma 59. *Let π be a continuous linear mapping on U and $g : U \rightarrow \overline{\mathbb{R}}$ such that $g \circ \pi \leq g$. Then $g^* \circ \pi^* \leq g^*$. If π is idempotent and g is subdifferentiable at $u \in \text{rge } \pi$ in the relative topology of $\text{rge } \pi$, then g is subdifferentiable at u .*

Proof. The first claim follows from

$$\begin{aligned} g^*(\pi^* y) &= \sup\{\langle u, \pi^* y \rangle - g(u)\} \\ &\leq \sup\{\langle \pi u, y \rangle - g(\pi u)\} \\ &\leq \sup\{\langle u, y \rangle - g(u)\} \\ &= g^*(y). \end{aligned}$$

Assume now that g is subdifferentiable at $u \in \text{rge } \pi$ relative to $\text{rge } \pi$. By Hahn-Banach, linear functionals on $\text{rge } \pi$ can be expressed by elements of Y . Thus, there is $y \in Y$ such that

$$g(u') \geq g(\pi u') \geq g(u) + \langle \pi u' - u, y \rangle = g(u) + \langle u' - u, \pi^* y \rangle \quad \forall u' \in U,$$

where the last equality holds when π is idempotent. \square

6.4 Duality in optimization

This appendix gives a brief summary of the conjugate duality framework of [21]. We include an extra linear perturbation in the primal objective which allows us to formulate certain results in a more convenient form. It does not, however, interfere with the original arguments from [21] as we see below.

Let X and U be in separating duality with V and Y , respectively. Given a convex function F on $X \times U$, the primal problem is

$$\text{minimize } F(x, u) - \langle x, v \rangle \quad \text{over } x \in X.$$

and the dual problem is

$$\text{maximize } \langle u, y \rangle - F^*(v, y) \quad \text{over } y \in Y.$$

The primal value function is

$$\varphi_v(u) := \inf_{x \in X} \{F(x, u) - \langle x, v \rangle\}$$

and the dual value function

$$\gamma_u(v) := \inf_{y \in Y} \{F^*(v, y) - \langle u, y \rangle\}.$$

By Fenchel's inequality,

$$F(x, u) + F^*(v, y) \geq \langle x, v \rangle + \langle u, y \rangle$$

for all $(x, u) \in X \times U$ and $(v, y) \in V \times Y$, so

$$\varphi_v(u) \geq -\gamma_u(v) \quad \forall u \in U, v \in V.$$

If $\varphi_v(u) > -\gamma_u(v)$, a *duality gap* is said to exist.

By definition, $\varphi_v^*(y) = F^*(v, y)$, so the dual problem can be written as

$$\text{maximize } \langle u, y \rangle - \varphi_v^*(y) \quad \text{over } y \in Y.$$

The properties of conjugates and subgradients from the previous section thus imply that the absence of a duality gap and existence of dual solutions come down to closedness and subdifferentiability of φ_v at u ; see Theorems 60 and 61 below.

The *Lagrangian* associated with F is the convex-concave function on $X \times Y$ given by

$$L(x, y) := \inf_{u \in U} \{F(x, u) - \langle u, y \rangle\}.$$

Clearly, the conjugate of F can be expressed as

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

The Lagrangian saddle-point problem is to find a saddle-value and/or a saddle-point of the convex-concave function

$$L_{v,u}(x, y) := L(x, y) - \langle x, v \rangle + \langle u, y \rangle.$$

We have

$$\langle u, y \rangle - F^*(v, y) = \inf_x L_{v,u}(x, y)$$

so the dual problem is equivalent to the maximization half of the Lagrangian minimax problem. When F is closed in u , the biconjugate theorem gives

$$F(x, u) - \langle x, v \rangle = \sup_y L_{v,u}(x, y)$$

so the primal problem is the minimization half of the minimax problem.

Theorem 60. *The following are equivalent,*

1. *There is no duality gap.*
2. *φ_v is closed at u .*

If F is closed in u , the above are equivalent to

3. *$L_{v,u}$ has a saddle-value.*

If F is closed, the above are equivalent to

4. *γ_u is closed at v .*

Proof. We have $\varphi_v^*(y) = \sup_{v,y} \{\langle x, v \rangle + \langle u, y \rangle - F(x, u)\} = F^*(v, y)$, so

$$\varphi_v^{**}(u) = \sup_y \{\langle u, y \rangle - F^*(v, y)\} = -\gamma_u(v),$$

and the equivalence of 1 and 2 follows from the biconjugate theorem. When F is closed in u , 1 is equivalent to 3 by the remarks before the statement. When F is closed, $F = F^{**}$ by the biconjugate theorem, so 4 is equivalent to 1, by symmetry. \square

Theorem 61. *If $\varphi_v(u) < \infty$, then the following are equivalent,*

1. *There is no duality gap and y solves the dual.*
2. *Either $\varphi_v(u) = -\infty$ or $y \in \partial\varphi_v(u)$.*

If F is closed in u , the above are equivalent to

$$3. \inf_x \sup_y L_{v,u}(x, y) = \inf_x L_{v,u}(x, y).$$

If F is closed, the above are equivalent to

$$4. \gamma_u \text{ is closed at } v \text{ and } y \text{ solves the dual.}$$

Proof. Condition 2 means that either $\varphi_v(u) = -\infty$ or $\varphi_v(u) + \varphi_v^*(y) = \langle u, y \rangle$. The equivalence of 1 and 2 thus follows from the biconjugate theorem. When F is closed in u , the equivalence of 1 and 3 follows from the equivalence of 1 and 3 in Theorem 60. When F is closed, the equivalence of 1 and 4 follows from the equivalence of 1 and 4 in Theorem 60. \square

Theorem 62. *The following are equivalent,*

1. *There is no duality gap, x solves the primal, y solves the dual and both problems are feasible.*
2. *$y \in \partial\varphi_v(u)$ and x solves the primal.*
3. *$(v, y) \in \partial F(x, u)$.*

If F is closed in u , the above are equivalent to

$$4. v \in \partial_x L(x, y) \text{ and } u \in \partial_y [-L](x, y).$$

If F is closed, the above are equivalent to

5. *$x \in \partial\gamma_u(v)$ and y solves the dual.*
6. *$(x, u) \in \partial F^*(v, y)$.*

Proof. The equivalence of 1 and 2 follows from Theorem 61. Condition 2 means that

$$F(x', u') - \langle x', v \rangle \geq F(x, u) - \langle x, v \rangle + \langle u' - u, y \rangle \quad \forall x', u'$$

which is 3. It is clear from the discussion just before Theorem 60 that, when F is closed in u , 1 means that (x, y) is a saddle-point of $L_{v,u}$, which is 4. When F is closed, 3 is equivalent to 6, by the biconjugate theorem. The equivalence of 5 and 6 follows like that of 2 and 3. \square

Theorem 63. *Assume that $v \in V$ is such that φ_v is either relatively continuous at $u \in U$ or $\varphi_v(u) = -\infty$. Then*

$$\gamma_u(v) := \inf_{y \in Y} \{F^*(v, y) - \langle u, y \rangle\}$$

is closed at v and

$$\inf_x \{F(x, u) - \langle x, v \rangle\} = \sup_y \{\langle u, y \rangle - F^*(v, y)\},$$

where the supremum is attained. If F is proper lsc and the first assumption holds for all $v \in V$, then γ_u is proper lsc, and

$$\gamma_u^\infty(v) = \inf_{y \in Y} \{(F^*)^\infty(v, y) - \langle u, y \rangle\}, \quad (7)$$

where the infimum is attained.

Proof. If $\varphi_v(u)$ is finite then, by Theorem 57, there exists $y \in \partial\varphi_v(u)$ which gives the second claim. The claims in the first paragraph follow now from Theorem 61.

Under the additional assumption, γ_u is proper lsc, so, by the biconjugate theorem, $F(\cdot, u)^* = \gamma_u$. Since F is proper lsc, there exists $(\bar{v}, \bar{y}) \in \text{dom } F^*$. By Fenchel,

$$F(x, u) + F^*(\bar{v}, \bar{y}) \geq \tilde{F}(x, u),$$

where $\tilde{F}(x, u) := \delta_{\text{dom } F}(x, u) + \langle x, \bar{v} \rangle + \langle u, \bar{y} \rangle$. Thus,

$$\varphi_v(u) + F^*(\bar{v}, \bar{y}) \geq \tilde{\varphi}_v(u),$$

where

$$\tilde{\varphi}_v(u) := \inf_x \{\tilde{F}(x, u) - \langle x, v \rangle\}.$$

We have $\text{dom } \tilde{\varphi}_v = \text{dom } \varphi_v$ and, by assumption, $u \in \text{dom } \varphi_v$. When $\tilde{\varphi}_v(u)$ is finite, the above inequality implies that $\varphi_v(u)$ is finite as well, and then continuity of φ_v at u implies that of $\tilde{\varphi}_v$. Thus, by the first part,

$$\sup_x \{\langle x, v \rangle - \tilde{F}(x, u)\} = \inf_{y \in Y} \{\tilde{F}^*(v, y) - \langle u, y \rangle\},$$

where the infimum is attained. Since $F(\cdot, u)^* = \gamma_u$, Lemma 58 implies $\tilde{F}(\cdot, u)^*(v) = \gamma_u^\infty(v - \bar{v}) - \langle u, \bar{y} \rangle$ and $\tilde{F}^*(v, y) = (F^*)^\infty(v - \bar{v}, y - \bar{y})$. Substituting into above,

$$\begin{aligned} \gamma_u^\infty(v - \bar{v}) - \langle u, \bar{y} \rangle &= \inf_{y \in Y} \{(F^*)^\infty(v - \bar{v}, y - \bar{y}) - \langle u, y \rangle\} \\ &= -\langle u, \bar{y} \rangle + \inf_{y \in Y} \{(F^*)^\infty(v - \bar{v}, y) - \langle u, y \rangle\}, \end{aligned}$$

which proves the claim. \square

Note that $\text{dom } \varphi_v$ does not depend on v . We denote this set by $\text{dom } \varphi$. Theorem 56 gives the following.

Theorem 64. *If X and U are Fréchet, F is lsc and $\text{aff dom } \varphi$ is closed, then for every $v \in V$, φ_v is either proper and continuous or identically $-\infty$ throughout $\text{core dom } \varphi$.*

The conditions of Theorem 64 clearly hold if $0 \in \text{core dom } \varphi$. This, in turn, generalizes the classical Slater-type conditions which require $0 \in \text{int dom } \varphi$. The primal problem is said to be *strictly feasible* if there is an $\bar{x} \in X$ such that

$\text{pos}(\text{dom } F - (\bar{x}, \bar{u}))$ is linear and closed. Since $\text{dom } \varphi$ is the projection of $\text{dom } F$ to U , Lemma 48 in the Appendix gives

$$\text{pos}(\text{dom } \varphi - \bar{u}) = \{u \mid \exists x : (x, u) \in \text{pos}(\text{dom } F - (\bar{x}, \bar{u}))\}.$$

Thus, the assumptions of Theorem 64 are satisfied if the primal is strictly feasible and the projection of the closed linear subspace $\text{aff dom } F$ is closed in U . The closedness is a classical question in functional analysis and various conditions have been given. We refer the reader to Section 3 for examples in the context of stochastic optimization.

Combining Theorems 63 and 64 gives the following.

Theorem 65. *If X and U are Fréchet, F is lsc and $\text{aff dom } \varphi$ is closed, then for every $v \in V$ and $u \in \text{rcore dom } \varphi$, the infimum in the definition of γ_u is attained, γ_u is lsc proper and*

$$\gamma_u^\infty(v) = \inf_{y \in Y} \{(F^*)^\infty(v, y) - \langle u, y \rangle\},$$

where the infimum is attained.

Proof. By Theorem 64, φ_v is relatively continuous for every $v \in V$. The claims thus follow from Theorem 63 and recalling that, by Theorem 61, the subgradients of φ_v at u are exactly the minimizers in the definition of γ_u . \square

The following was used in Example 11.

Example 66. *Let A be a linear mapping from a Fréchet X space to a vector space U such that $\ker A$ is closed. Then $\text{rge } A$ is a Fréchet space with respect to the final topology induced by A . Moreover, the topological dual of $\text{rge } A$ can be identified with $(\ker A)^\perp$ in the sense that for every $u^* \in (\text{rge } A)^*$ there is a unique $v \in (\ker A)^\perp$ such that*

$$\langle Ax, u^* \rangle = \langle x, v \rangle$$

for all $x \in X$.

Proof. The quotient space $X/\ker A$ and $\text{rge } A$ are in one-to-one correspondence under the linear mapping i that sends an $[x] \in X/\ker A$ to Ax . Indeed, $i^{-1}(u) = \{x \in X \mid Ax = u\}$. The final topology on $\text{rge } A$ is topologically isomorphic under i with the quotient space topology on $X/\ker A$. If $\ker A$ is closed, then $X/\ker A$ is Hausdorff and $X/\ker A$ is Fréchet and so too is $\text{rge } A$.

Continuous linear functionals l on $\text{rge } A$ can be expressed as $l(u) = l'(i^{-1}(u))$, where l' is a continuous linear functional on $X/\ker A$. The continuous linear functionals $l' \in (X/\ker A)^*$ are given by $l'([x]) = \langle x, v \rangle$, where $v \in (\ker A)^\perp$. \square

The following was used in the proof of Lemma 21.

Lemma 67. *Let $L \subset X$ and $A : X \rightarrow U$ a continuous linear mapping such that AL is closed. Then $\ker A + L$ is closed. In particular, if L is closed and linear and π a continuous linear idempotent mapping on U such that $\pi L \subseteq L$, then $\text{rge } \pi + L$ is closed.*

Proof. We have

$$\begin{aligned}\ker A + L &= \{z + x' \mid Az = 0, x' \in L\} \\ &= \{x \mid A(x - x') = 0, x' \in L\} \\ &= \{x \in X \mid Ax \in AL\} \\ &= A^{-1}(AL),\end{aligned}$$

which proves the first claim. The second claim follows from the first by choosing $A = I - \pi$. Indeed, $AL = L - \pi L = L$ is closed and, since π is idempotent, $\ker A = \text{rge } \pi$. \square

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