



Aalto-yliopisto  
Teknillinen korkeakoulu

# Stochastic Programs Without Duality Gaps

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# Outline

**Introduction**

**Stochastic Optimization**

**Examples**

**Convex duality**

**More general settings**



# Introduction

- We study dynamic stochastic optimization problems which arise in many applications in operations research and mathematical finance.

Pennanen. *Convex duality in stochastic programming and mathematical finance*. Mathematics of Operations Research (to appear)

Pennanen, Perkkiö. *Stochastic Programs Without Duality Gaps*. 2011 (submitted)

Pennanen, Perkkiö. *Convex duality in stochastic optimization over processes of bounded variation*. (manuscript)

Pennanen, Perkkiö. *Stochastic problems of Bolza over predictable processes of bounded variation*. (manuscript)



## Normal integrands

- Let  $(\Omega, \mathcal{F}, \mathbb{F}, P)$  be a complete filtered probability space.
- A function  $f : \mathbb{R}^n \times \Omega \rightarrow \overline{\mathbb{R}}$  is a **normal integrand** if  $f(\cdot, \omega)$  is lower semicontinuous for all  $\omega$ , and  $f$  is  $\mathcal{B}(\mathbb{R}^n) \otimes \mathcal{F}$ -measurable.
- The **integral functional**  $I_f : L^0(\Omega; \mathbb{R}^n) \rightarrow \overline{\mathbb{R}}$  defined by

$$I_f(x) = \mathbb{E}f(x(\omega), \omega)$$

is well-defined, where the expectation is defined as  $+\infty$  unless the positive part is integrable.

- If  $f(\cdot, \omega)$  is convex,  $I_f$  is convex.
- Rockafellar, *Integral functionals, normal integrands and measurable selections*, 1976

# Normal integrands

Examples of normal integrands

- **Carathéodory function:**  $f(\cdot, \omega)$  is continuous almost surely and  $f(x, \cdot)$  is measurable for all  $x$ .
- **Indicator function:**

$$f(x, \omega) = \begin{cases} 0 & \text{if } x \in D(\omega) \\ +\infty & \text{otherwise} \end{cases}$$

of a closed measurable set  $D$ .

- $f = f_1 + f_2$  where  $f_1$  and  $f_2$  are normal integrands.

# Stochastic optimization

For given integers  $n_t$ , let

$$\mathcal{N} = \{(x_t)_{t=0}^T \mid x_t \in L^0(\Omega, \mathcal{F}_t, P; \mathbb{R}^{n_t})\}$$

be the space of **strategies**. We consider the problem

$$\underset{x \in \mathcal{N}}{\text{minimize}} \quad \mathbb{E}[f(x(\omega), u(\omega), \omega)],$$

where  $f : \mathbb{R}^n \times \mathbb{R}^m \times \Omega \rightarrow \overline{\mathbb{R}}$  is a convex normal integrand,  $n = n_0 + \dots + n_T$  and  $u \in L^0(\Omega, \mathcal{F}, \mathbb{R}^m)$ .

The aim is to study the **value function**

$$\varphi(u) = \inf_{x \in \mathcal{N}} I_f(x, u).$$

## Example: Superhedging in liquid markets

Let  $S = (S_t)_{t=0}^T$  be an  $\mathbb{R}^d$ -valued adapted stochastic process,  $n_t = d$ ,  $m = 1$  and

$$f(x, u, \omega) = \begin{cases} 0 & \text{if } \sum_{t=0}^{T-1} x_t \cdot \Delta S_t(\omega) \geq u, \\ +\infty & \text{otherwise.} \end{cases}$$

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Let  $u \in L^0(\Omega, \mathcal{F}, P; \mathbb{R})$ . Now

$$\varphi(u) = \inf_{x \in \mathcal{N}} I_f(x, u) = \begin{cases} 0 & \text{if } u \in \mathcal{C} \\ \infty & \text{otherwise,} \end{cases}$$

where  $\mathcal{C} = \{u \mid \exists x \in \mathcal{N} : \sum_{t=0}^{T-1} x_t \cdot \Delta S_t \geq u \text{ } P\text{-a.s.}\}$  is the set of **claim processes** which can be **hedged** with zero cost.



## Example: Variance optimal hedging

Let  $S = (S_t)_{t=0}^T$  be an  $\mathbb{R}^d$ -valued  $\mathbb{F}$ -adapted stochastic process,  $u \in L^0(\Omega, \mathcal{F}, P; \mathbb{R})$  and

$$f(x, u, \omega) = \left( V_0 + \sum_{t=0}^{T-1} z_t \cdot \Delta S_{t+1}(\omega) - u \right)^2,$$

where  $x_0 = (z_0, V_0)$ ,  $x_t = z_t$  for  $t = 1, \dots, T$ .

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$$\varphi(u) = \inf_{x \in \mathcal{N}} E(V_0 + \sum_{t=0}^{T-1} z_t \cdot \Delta S_{t+1} - u)^2,$$

where  $V_0$  is as an initial value of a self-financing trading strategy and  $z_t$  is the portfolio of risky assets held over period  $[t, t + 1]$ .

Föllmer and Schied, *Stochastic Finance*, Section 10.3

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# Examples

- Superhedging in illiquid markets
- Optimal consumption in illiquid markets
- Liquidation problems in some market impact models.

More examples:

Pennanen. *Convex duality in stochastic programming and mathematical finance*. Mathematics of Operations Research (to appear)



# Convex duality

- We want to derive dual expressions for

$$\varphi(u) = \inf_{x \in \mathcal{N}} l_f(x, u).$$

- Let  $\mathcal{U} = L^p(\Omega; \mathbb{R}^m)$  and  $\mathcal{Y} = L^q(\Omega; \mathbb{R}^m)$  where  $1/p + 1/q = 1$ ,  $1 \leq p \leq \infty$ , and

$$\langle u, y \rangle = \mathbb{E}u(\omega) \cdot y(\omega).$$

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- The **conjugate** of  $\varphi$  on  $\mathcal{Y}$  is the convex function defined by

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

- If  $\varphi$  is convex, then  $\varphi^{**} = \text{cl } \varphi$ , where

$$\text{cl } \varphi = \begin{cases} \text{lsc } \varphi & \text{if } (\text{lsc } \varphi)(u) > -\infty \forall u \in \mathcal{U}, \\ -\infty & \text{otherwise.} \end{cases}$$

# Convex duality

- The **Lagrangian** associated with  $f$  is the extended real-valued function on  $\mathcal{N} \times \mathcal{Y}$  defined by

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

The Lagrangian is convex in  $x$  and concave in  $y$ .

- The **dual objective** is a concave function defined by

$$g(y) = \inf_{x \in \mathcal{N}} L(x, y).$$

# Convex duality

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$$g(y) = \inf_{x \in \mathcal{N}} L(x, y).$$

- We have  $g = -\varphi^*$  which gives the **dual representation**

$$(\text{cl } \varphi)(u) = \sup_{y \in \mathcal{Y}} \{\langle u, y \rangle + g(y)\},$$

# Convex duality

## Why convex duality?

- Optimality conditions can be given in terms of the dual representation
- Duality techniques are used in numerical optimization
- Dual variables have interpretations: martingale measures, consistent price systems, shadow prices of information
  - Why martingale measures in mathematical finance?





# Convex duality

Our framework differs from the duality framework of

R.T. Rockafellar, *Conjugate duality and optimization*, SIAM, 1974

by allowing a larger space of strategies (strategies are not restricted to be in a LCTVS)

- Our extension allows to use certain techniques from mathematical finance to close the duality gap in some situations where topological arguments fail.
- We do not put emphasis on the existence of dual solutions in general. The mere absence of duality gap is behind some fundamental results in mathematical finance.

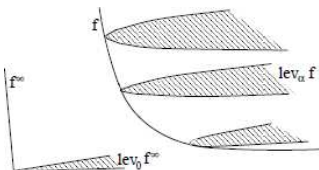
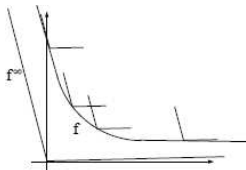


## Duality result

Let  $h : \mathbb{R}^n \times \Omega \rightarrow \overline{\mathbb{R}}$  be a convex normal integrand and assume  $\text{dom } h(\cdot, \omega)$  is nonempty, the **recession function** of  $h$  is

$$h^\infty(x, \omega) = \sup_{\lambda > 0} \frac{h(\lambda x + \bar{x}, \omega) - h(\bar{x}, \omega)}{\lambda},$$

which is independent of the choice of  $\bar{x} \in \text{dom } h(\cdot, \omega)$ .



# Duality result

## Theorem

Assume that there is a  $y \in \mathcal{Y}$  and an  $m \in L^1(\Omega, \mathcal{F}, P)$  such that for  $P$ -almost every  $\omega$ ,

$$f(x, u, \omega) \geq u \cdot y(\omega) + m(\omega) \quad \forall (x, u) \in \mathbb{R}^n \times \mathbb{R}^m$$

and that  $\{x \in \mathcal{N} \mid f^\infty(x(\omega), 0, \omega) \leq 0 \text{ a.s.}\}$  is a linear space. Then

$$\varphi(u) = \inf_{x \in \mathcal{N}} I_f(x, u)$$

is lower semicontinuous on  $\mathcal{U}$  and the infimum is attained for every  $u \in \mathcal{U}$ .

## Example: Superhedging in liquid markets

For

$$f(x, u, \omega) = \begin{cases} 0 & \text{if } \sum_{t=0}^{T-1} x_t \cdot \Delta S_t(\omega) \geq u, \\ +\infty & \text{otherwise,} \end{cases}$$

we have

$$\{x \in \mathcal{N} \mid f^\infty(x(\omega), 0, \omega) \leq 0 \text{ a.s.}\} = \{x \in \mathcal{N} \mid \sum_{t=0}^{T-1} x_t \cdot \Delta S_t \geq 0\}.$$

Thus  $f$  satisfies the linearity condition of the theorem if and only if the price process  $S$  satisfies the *no-arbitrage* condition.

## Example: Variance optimal hedging

For

$$f(x, u, \omega) = \left( V_0 + \sum_{t=0}^{T-1} z_t \cdot \Delta S_{t+1}(\omega) - u \right)^2$$

the linearity condition holds, since

$$\{x \in \mathcal{N} \mid f^\infty(x(\omega), 0, \omega) \leq 0 \text{ a.s.}\} = \{x \in \mathcal{N} \mid V_0 + \sum_{t=0}^{T-1} z_t \cdot \Delta S_{t+1} = 0\}.$$

Thus the infimum in

$$\varphi(u) = \inf_{x \in \mathcal{N}} E \left( V_0 + \sum_{t=0}^{T-1} z_t \cdot \Delta S_{t+1} - u \right)^2$$

is attained for every  $u \in \mathcal{U}$ , and  $\varphi$  is lower semicontinuous.

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# Remarks on the duality result

- The proof uses a characterization of the linearity condition which is based on dynamic programming equations.
  - With certain assumptions the dynamic programming equations take the more familiar form of Bellman equations.
- $\varphi : \mathcal{U} \rightarrow \overline{\mathbb{R}}$ , where  $\mathcal{U}$  is more general space than  $L^p(\Omega; \mathbb{R}^n)$ 
  - Parameters and dual variables can belong to more general LCTVS than Banach spaces.
  - The presented duality result extends to this setting provided that  $\mathcal{U}$  is **decomposable**.

## More general settings

- Let  $f : X \times M \times \Omega \rightarrow \overline{\mathbb{R}}$ , where  $X$  is the space of functions of bounded variation on  $[0, T]$  and  $M$  is the space of Radon measures on  $[0, T]$ . Let  $\mathcal{N}$  be the predictable processes of bounded variation and consider

$$\underset{x \in \mathcal{N}}{\text{minimize}} \quad \mathbb{E}[f(x(\omega), u(\omega), \omega)],$$

where  $u \in L^0(\Omega, \mathcal{F}; M)$ .

- Continuous times illiquid markets
- Stochastic Lagrange variational problems

Pennanen, Perkkiö. *Convex duality in stochastic optimization over processes of bounded variation*. (manuscript)

Pennanen, Perkkiö. *Stochastic problems of Bolza over predictable processes of bounded variation*. (manuscript)