

# Conjugate duality in stochastic optimization

Ari-Pekka Perkkiö,  
Institute of Mathematics,  
Aalto University

Ph.D. instructor/joint work with  
Teemu Pennanen,  
Institute of Mathematics,  
Aalto University

March 15th 2010

# Introduction

## Introduction

Introduction

Introduction

Introduction

Introduction

Stochastic Optimization

Stochastic Optimization

$\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality

Conjugate duality

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

We study convex stochastic optimization problems.

- ✓ Stochastic LP duality, linear quadratic control and calculus of variations
- ✓ Stochastic problems of Bolza, shadow price of information and optimal stopping
- ✓ Illiquid convex market models (Jouni&Kallal, Kabanov, Schachermayer, Guasoni, Pennanen)
- ✓ Super-hedging and pricing, utility maximization and optimal consumption

Convexity gives rise to dual optimization problems and dual characterisations of the objective functionals.

# Introduction

Introduction

Introduction

Introduction

Introduction

Introduction

Stochastic Optimization

Stochastic Optimization

$\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality

Conjugate duality

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

**Example** (Super-hedging in a liquid market).

$$\begin{aligned} & \inf x_0^0, \\ & \text{s.t. } \left\{ \begin{array}{l} C_T \leq x_0^0 + \int_T x_t \cdot dS_t \text{ a.s.} \end{array} \right. \end{aligned}$$

where  $x_0^0$  is the initial wealth,  $C_T$  is a claim,  $x$  is a predictable process (portfolio of risky assets) and  $S$  is a price process. The infimum is over initial wealths  $x_0$  and predictable processes  $x$ .

# Introduction

Introduction

Introduction

Introduction

Introduction

Introduction

Stochastic Optimization

Stochastic Optimization

$\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality

Conjugate duality

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

**Example** (Super-hedging in a liquid market).

$$\begin{aligned} & \inf x_0^0, \\ & \text{s.t. } \left\{ \begin{array}{l} C_T \leq x_0^0 + \int_T x_t \cdot dS_t \text{ a.s.} \end{array} \right. \end{aligned}$$

where  $x_0^0$  is the initial wealth,  $C_T$  is a claim,  $x$  is a predictable process (portfolio of risky assets) and  $S$  is a price process. The infimum is over initial wealths  $x_0$  and predictable processes  $x$ .

The dual problem is

$$\sup_{Q \in \mathcal{M}} \mathbb{E}^Q[C_T],$$

where the supremum is over martingale measures.

# Introduction

**Example** (Kabanov's model). Consider a set

$$\{x \in BV \mid (dx/|dx|)_t \in C(\omega, t) \forall t\}$$

where  $C(\omega, t) \subset \mathbb{R}^d$  is a convex cone for all  $(\omega, t)$ .  $C(\omega, t)$  is the set of self-financing trades in the market at time  $t$ . A predictable process of bounded variation is *self-financing* if

$$(dx(\omega)/|dx(\omega)|)_t \in C(\omega, t) \forall t \text{ a.s.}$$

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Introduction

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- inf  $\mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

**Example** (Kabanov's model). Consider a set

$$\{x \in BV \mid (dx/|dx|)_t \in C(\omega, t) \forall t\}$$

where  $C(\omega, t) \subset \mathbb{R}^d$  is a convex cone for all  $(\omega, t)$ .  $C(\omega, t)$  is the set of self-financing trades in the market at time  $t$ . A predictable process of bounded variation is **self-financing** if

$$(dx(\omega)/|dx(\omega)|)_t \in C(\omega, t) \forall t \text{ a.s.}$$

**Example** (Linear case).

$$C(\omega, t) = \{(x^0, x^1) \in \mathbb{R}^2 \mid x^0 + x^1 \cdot S_t(\omega) \leq 0\},$$

where  $S$  is the price process of a risky asset,  $x^0$  refers to a bank account and  $x^1$  to the risky asset. Portfolio is self-financing if all trades of the risky assets are financed using the bank account. The inequality allows a free disposal of money or assets.

# Introduction

**Example** (Optimal consumption in a convex market model).

$$\begin{aligned} & \sup \mathbb{E} \int_T U_t(\omega, dc), \\ \text{s.t. } & \begin{cases} (d(x(\omega) + c(\omega)) / |d(x(\omega) + c(\omega))|)_t \in C(\omega, t) \quad \forall t \text{ a.s.} \\ x_t(\omega) \in D(\omega, t) \quad \forall t \text{ a.s.} \end{cases} \end{aligned}$$

where  $U_t$  is an utility function for all  $t$  almost surely and  $D(\omega, t)$  is the set of allowed portfolio positions at time  $t$ . The supremum is over predictable processes of bounded variation  $x$  and  $c$ .  $x$  is the portfolio process, and  $c$  is the consumption process.

- Introduction
- Introduction
- Introduction
- Introduction**
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Introduction

**Example** (Optimal consumption in a convex market model). *A dual problem is*

$$\begin{aligned} & \inf \mathbb{E} - \int_T U_t^*(y_t) dt, \\ \text{s.t. } & \begin{cases} y_t(\omega) \in C^*(\omega, t) \quad \forall t \text{ a.s.} \\ (da(\omega)/|da(\omega)|)_t \in D^*(\omega, t) \quad \forall t \text{ a.s.} \end{cases} \end{aligned}$$

where  $U_t^*$  is the concave conjugate of the utility function,  $C^*(\omega, t)$  is the polar of  $C(\omega, t)$  ( $y$  is a consistent price system),  $D^*(\omega, t)$  is the polar of  $D(\omega, t)$ , and the supremum is over semimartingales with the canonical decomposition  $y = m + a$ .

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization



# Introduction

**Example** (Optimal consumption in a convex market model). *A dual problem is*

$$\begin{aligned} & \inf \mathbb{E} - \int_T U_t^*(y_t) dt, \\ \text{s.t. } & \begin{cases} y_t(\omega) \in C^*(\omega, t) \quad \forall t \text{ a.s.} \\ (da(\omega)/|da(\omega)|)_t \in D^*(\omega, t) \quad \forall t \text{ a.s.} \end{cases} \end{aligned}$$

where  $U_t^*$  is the concave conjugate of the utility function,  $C^*(\omega, t)$  is the polar of  $C(\omega, t)$  ( $y$  is a consistent price system),  $D^*(\omega, t)$  is the polar of  $D(\omega, t)$ , and the supremum is over semimartingales with the canonical decomposition  $y = m + a$ .

The aim is to formulate problems like this in a general framework and deduce the dual problems by general methods.

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Stochastic Optimization

Introduction  
Introduction  
Introduction  
Introduction  
Introduction

## Stochastic Optimization

Stochastic Optimization  
 $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality  
Conjugate duality  
Conjugate duality in  
stochastic optimization  
Conjugate duality in  
stochastic optimization  
Conjugate duality in  
stochastic optimization

Let  $(\Omega, \mathcal{F}, \mathbb{F}, P)$  be a complete filtered probability space, Let  $\mathcal{N}$  be the set of predictable processes of bounded variation. Let  $U$  be a separable Banach (or its dual).

# Stochastic Optimization

Introduction  
Introduction  
Introduction  
Introduction  
Introduction

## Stochastic Optimization

Stochastic Optimization  
 $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality

Conjugate duality

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Let  $(\Omega, \mathcal{F}, \mathbb{F}, P)$  be a complete filtered probability space, Let  $\mathcal{N}$  be the set of predictable processes of bounded variation. Let  $U$  be a separable Banach (or its dual). Define  $F : \mathcal{N} \times L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{+\infty\}$  by

$$F(x, u) = \mathbb{E}[f(\omega, x(\omega), u(\omega))],$$

where  $f$  is a normal-integrand. The *value function* is

$$\phi(u) = \inf_{x \in \mathcal{N}} F(x, u) = \inf_{x \in \mathcal{N}} \mathbb{E}[f(\omega, x(\omega), u(\omega))].$$

# Stochastic Optimization

Introduction  
Introduction  
Introduction  
Introduction  
Introduction

## Stochastic Optimization

Stochastic Optimization  
 $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

Conjugate duality

Conjugate duality

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Conjugate duality in  
stochastic optimization

Let  $(\Omega, \mathcal{F}, \mathbb{F}, P)$  be a complete filtered probability space, Let  $\mathcal{N}$  be the set of predictable processes of bounded variation. Let  $U$  be a separable Banach (or its dual). Define  $F : \mathcal{N} \times L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{+\infty\}$  by

$$F(x, u) = \mathbb{E}[f(\omega, x(\omega), u(\omega))],$$

where  $f$  is a normal-integrand. The *value function* is

$$\phi(u) = \inf_{x \in \mathcal{N}} F(x, u) = \inf_{x \in \mathcal{N}} \mathbb{E}[f(\omega, x(\omega), u(\omega))].$$

A function  $f : \Omega \times (X \times U) \rightarrow \mathbb{R} \cup \{+\infty\}$  is a **normal integrand** if the epigraph  $\text{epi } f \subset \Omega \times X \times U \times \mathbb{R}$  is measurable and  $\omega$ -sections are closed.

In particular  $\omega \mapsto f(\omega, x(\omega), u(\omega))$  is measurable when  $x \in L^0(\Omega; X)$  and  $u \in L^0(\Omega; U)$ , and for fixed  $\omega$ ,  $(x, u) \mapsto f(\omega, x, u)$  is lower semicontinuous. Moreover,  $F$  is convex if  $f(\omega, \cdot, \cdot)$  is convex.

# Stochastic Optimization $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization  $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

**Example** (Convex market models). Let  $U = BV$ , and

$$f(\omega, x, u) = k(\omega, x, u) + \delta_{\mathcal{D}(\omega)}(x) + \delta_{\mathcal{C}(\omega)}(dx + du),$$

where (and similarly for  $\delta_{\mathcal{C}(\omega)}$ )

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

# Stochastic Optimization $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization  $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

**Example** (Convex market models). Let  $U = BV$ , and

$$f(\omega, x, u) = k(\omega, x, u) + \delta_{\mathcal{D}(\omega)}(x) + \delta_{\mathcal{C}(\omega)}(dx + du),$$

where (and similarly for  $\delta_{\mathcal{C}(\omega)}$ )

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

and

$$\mathcal{C}(\omega) = \{x \in BV \mid (dx/|dx|)_t \in C(\omega, t) \forall t\},$$

$$\mathcal{D}(\omega) = \{x \in BV \mid x_t \in D(\omega, t) \forall t\},$$

and  $k$  is a normal integrand which gives the criterion one wants to minimize/maximize (e.g. utility).

# Stochastic Optimization $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization  $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

**Example** (Convex market models). Let  $U = BV$ , and

$$f(\omega, x, u) = k(\omega, x, u) + \delta_{\mathcal{D}(\omega)}(x) + \delta_{\mathcal{C}(\omega)}(dx + du),$$

where (and similarly for  $\delta_{\mathcal{C}(\omega)}$ )

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

and

$$\mathcal{C}(\omega) = \{x \in BV \mid (dx/|dx|)_t \in C(\omega, t) \forall t\},$$

$$\mathcal{D}(\omega) = \{x \in BV \mid x_t \in D(\omega, t) \forall t\},$$

and  $k$  is a normal integrand which gives the criterion one wants to minimize/maximize (e.g. utility). In this case  $u \in L^p(\Omega; U)$  can be interpreted as a claim process or a consumption process.

# Conjugate duality

Let  $Y$  be a separable Banach space and  $U$  be its dual space. The pairing  $\langle u, y \rangle = \mathbb{E}\langle u(\omega), y(\omega) \rangle$  is finite for all  $u \in L^p(\Omega; U)$  and  $y \in L^q(\Omega; Y)$ . We equip these spaces with weak topologies induced by the pairing.

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality**
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization



# Conjugate duality

Let  $Y$  be a separable Banach space and  $U$  be its dual space. The pairing  $\langle u, y \rangle = \mathbb{E}\langle u(\omega), y(\omega) \rangle$  is finite for all  $u \in L^p(\Omega; U)$  and  $y \in L^q(\Omega; Y)$ . We equip these spaces with weak topologies induced by the pairing.

The *convex conjugate* of  $\phi : L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{\pm\infty\}$  is defined by

$$\phi^*(y) = \sup_{u \in L^p(\Omega; U)} \{\langle u, y \rangle - \phi(u)\},$$

which is a convex lower semicontinuous function on  $L^q(\Omega; Y)$ .

Introduction  
Introduction  
Introduction  
Introduction  
Introduction  
Stochastic Optimization  
Stochastic Optimization  
 $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$   
Conjugate duality  
Conjugate duality  
Conjugate duality in  
stochastic optimization  
Conjugate duality in  
stochastic optimization  
Conjugate duality in  
stochastic optimization

# Conjugate duality

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- inf  $\mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality**
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

Let  $Y$  be a separable Banach space and  $U$  be its dual space. The pairing  $\langle u, y \rangle = \mathbb{E}\langle u(\omega), y(\omega) \rangle$  is finite for all  $u \in L^p(\Omega; U)$  and  $y \in L^q(\Omega; Y)$ . We equip these spaces with weak topologies induced by the pairing.

The *convex conjugate* of  $\phi : L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{\pm\infty\}$  is defined by

$$\phi^*(y) = \sup_{u \in L^p(\Omega; U)} \{\langle u, y \rangle - \phi(u)\},$$

which is a convex lower semicontinuous function on  $L^q(\Omega; Y)$ .

The biconjugate satisfies  $\phi^{**} = \text{cl co } \phi$ , where

$$\text{cl } \phi = \begin{cases} -\infty & \text{if } \text{lsc } \phi(u) = -\infty \text{ for some } u, \\ \text{lsc } \phi & \text{otherwise.} \end{cases}$$

In particular, if  $\phi$  is convex and closed, then  $\phi = \phi^{**}$  (the *dual representation*).

# Conjugate duality

Recall the value function  $\phi : L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{\pm\infty\}$  was given by

$$\phi(u) = \inf_{x \in \mathcal{N}} \mathbb{E} f(x(\omega), u(\omega)),$$

which is a convex function on  $\mathcal{U}$  (if  $F$  is convex, which we assume).

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- inf  $\mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality**
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Conjugate duality

Recall the value function  $\phi : L^p(\Omega; U) \rightarrow \mathbb{R} \cup \{\pm\infty\}$  was given by

$$\phi(u) = \inf_{x \in \mathcal{N}} \mathbb{E} f(x(\omega), u(\omega)),$$

which is a convex function on  $\mathcal{U}$  (if  $F$  is convex, which we assume).

Define the *dual objective* by

$$g(y) = -\phi^*(y),$$

which is a concave upper semicontinuous function on  $\mathcal{Y}$ . If  $\phi$  is lower semicontinuous and proper, then  $\phi$  has the dual representation

$$\phi(u) = \sup_{y \in L^q(\Omega; Y)} \{\langle u, y \rangle + g(y)\}.$$

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- inf  $\mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality**
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Conjugate duality in stochastic optimization

$$\phi(u) = \inf_{x \in \mathcal{N}} \mathbb{E} f(\omega, x(\omega), u(\omega)).$$

- ✓ Calculating the dual objective  $g = -\phi^*$  is based on conjugacy of integral functionals and theory of normal-integrands. Adaptiveness constraints lead to stochastic analysis; also the dual problem may not be a pure integral functional anymore.

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization**
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Conjugate duality in stochastic optimization

$$\phi(u) = \inf_{x \in \mathcal{N}} \mathbb{E} f(\omega, x(\omega), u(\omega)).$$

- ✓ Calculating the dual objective  $g = -\phi^*$  is based on conjugacy of integral functionals and theory of normal-integrands. Adaptiveness constraints lead to stochastic analysis; also the dual problem may not be a pure integral functional anymore.
- ✓ In convex analysis there exists a lot of results for the lower semicontinuity of  $\phi$ . These are based on LCTVS structure of the strategy space and compactness type arguments.

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization**
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

# Conjugate duality in stochastic optimization

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization**
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

$$\phi(u) = \inf_{x \in \mathcal{N}} \mathbb{E} f(\omega, x(\omega), u(\omega)).$$

- ✓ Calculating the dual objective  $g = -\phi^*$  is based on conjugacy of integral functionals and theory of normal-integrands. Adaptiveness constraints lead to stochastic analysis; also the dual problem may not be a pure integral functional anymore.
- ✓ In convex analysis there exists a lot of results for the lower semicontinuity of  $\phi$ . These are based on LCTVS structure of the strategy space and compactness type arguments.
- ✓  $\mathcal{N}$  is not LCTVS. In mathematical finance there exists results for lower semicontinuity of  $\phi$  in this case, but only when  $f(\omega, x, u)$  is an indicator function or of some other very restrictive form.

# Conjugate duality in stochastic optimization

One of our contributions has been to extend the arguments used in convex analysis and mathematical finance to obtain lower semicontinuity of  $\phi$  in more general cases.

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization



# Conjugate duality in stochastic optimization

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

One of our contributions has been to extend the arguments used in convex analysis and mathematical finance to obtain lower semicontinuity of  $\phi$  in more general cases.

**Example.** Assume there exists  $(v, y)$  such that  $\mathbb{E}f^*(\omega, v(\omega), y(\omega)) < \infty$ ,  $v$  is a martingale, and

$$\{x \in X \mid \exists u \in B(\omega), f(\omega, x, u(\omega)) - \langle x, v(\omega) \rangle \leq \beta(\omega)\}$$

is compact almost surely for some  $\beta \in L^0(\Omega; \mathbb{R})$ , where  $B(\omega)$  is a neighborhood of the origin almost surely. Then the value function  $\phi$  is lower semicontinuous at the origin.

# Conjugate duality in stochastic optimization

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization
- $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization
- Conjugate duality in stochastic optimization

One of our contributions has been to extend the arguments used in convex analysis and mathematical finance to obtain lower semicontinuity of  $\phi$  in more general cases.

**Example.** Assume there exists  $(v, y)$  such that  $\mathbb{E}f^*(\omega, v(\omega), y(\omega)) < \infty$ ,  $v$  is a martingale, and

$$\{x \in X \mid \exists u \in B(\omega), f(\omega, x, u(\omega)) - \langle x, v(\omega) \rangle \leq \beta(\omega)\}$$

is compact almost surely for some  $\beta \in L^0(\Omega; \mathbb{R})$ , where  $B(\omega)$  is a neighborhood of the origin almost surely. Then the value function  $\phi$  is lower semicontinuous at the origin.

**Remark.** The path spaces  $X, U, Y$  can be generalized to Souslin LCTVS, and perturbation space  $L^p(\Omega; U)$  can be generalized to LCTVS. Banach space structure shown in the slides was just an example.

# Conjugate duality in stochastic optimization

- Introduction
- Introduction
- Introduction
- Introduction
- Introduction
- Stochastic Optimization
- Stochastic Optimization  
 $\inf \mathbb{E}[f(\omega, x(\omega), u(\omega))]$
- Conjugate duality
- Conjugate duality
- Conjugate duality in  
stochastic optimization
- Conjugate duality in  
stochastic optimization
- Conjugate duality in  
stochastic optimization

And back to introduction: many convex stochastic optimization problems are covered by this duality framework.

- ✓ Stochastic LP duality, linear quadratic control and calculus of variations
- ✓ Stochastic problems of Bolza, shadow price of information and optimal stopping
- ✓ Illiquid convex market models
- ✓ Super-hedging and pricing, utility maximization and optimal consumption