

6th General AMaMeF and Banach Center Conference
Convex risk measures for càdlàg processes on Orlicz hearts

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Basic setting

Consider an incomplete financial market.

Let the interest rate be given by $\mathbf{0}$.

Time horizon: $T \in (0, \infty)$

Underlying space $(\Omega, \mathcal{F}, \mathbf{P}; \mathbf{F} = \{\mathcal{F}_t\}_{t \in [0, T]})$

\mathbf{F} is supposed to satisfy the usual condition, that is,

\mathbf{F} is right-continuous, $\mathcal{F}_T = \mathcal{F}$ and \mathcal{F}_0 contains all null sets of \mathcal{F} .

Young function and Orlicz space

$\Phi : \mathbf{R} \rightarrow [0, \infty]$ is called a **Young function**, if it is an even convex function with $\Phi(0) = 0$, $\Phi(x) \rightarrow \infty$ as $x \rightarrow \infty$ and $\Phi(x) < \infty$ for x in a neighborhood of 0 . Let Ψ be the conjugate function of Φ . $\Psi(y) := \sup_x \{xy - \Phi(x)\}$.

Let \mathcal{L}^0 be the set of all measurable functions on (Ω, \mathcal{F}) .

- ① **Orlicz space:** $L^\Phi := \{X \in \mathcal{L}^0 \mid E[\Phi(c|X|)] < \infty \text{ for some } c > 0\}$.
- ② **Orlicz heart:** $M^\Phi := \{X \in \mathcal{L}^0 \mid E[\Phi(c|X|)] < \infty \text{ for any } c > 0\}$.
- ③ **Luxemburg norm:** $\|X\|_\Phi := \inf \left\{ \lambda > 0 \mid E \left[\Phi \left(\left| \frac{X}{\lambda} \right| \right) \right] \leq 1 \right\}$,
- ④ **Orlicz norm:** $\|X\|_\Psi^* := \sup \{E[XY] \mid Y \in L^\Psi, \|Y\|_\Psi \leq 1\}$.

Henceforth, a Young function Φ is fixed.

For simplicity, Φ is assumed to be \mathbf{R} -valued, that is, continuous, and to satisfy $\Phi(x) > 0$ for any $x \in \mathbf{R} \setminus \{0\}$.

Examples

① $\Phi(x) = |x|^p/p, \infty > p > 1.$

In this case, $L^\Phi = M^\Phi = L^p$ and $\Psi(x) = |x|^q/q,$
where $1/p + 1/q = 1.$

② $\Phi(x) = e^{|x|} - |x| - 1$ or $\Phi(x) = e^{|x|} - 1.$

$L^\infty \subset M^\Phi \subset L^\Phi \subset L^p.$

Shortfall risk measure

Let $\mathcal{X} = M^\Phi$.

\mathcal{U} : a convex cone set, including $\mathbf{0}$, of attainable claims with zero initial endowment.

Let I be a nondecreasing convex function defined by

$$I(\mathbf{x}) = \begin{cases} \mathbf{0}, & \text{if } \mathbf{x} \leq \mathbf{0}, \\ \Phi(\mathbf{x}), & \text{otherwise.} \end{cases}$$

Shortfall risk with the **loss function** I for sellers is expressed by $E[I(-\mathbf{x} - \mathbf{U} + \mathbf{X})]$ when the price of a claim \mathbf{X} and the hedging strategy are given by $\mathbf{x} \in \mathbf{R}$ and $\mathbf{U} \in \mathcal{U}$, respectively.

← orlicz space

Shortfall risk measure 2

Fix $\delta > 0$. Define

$$\mathcal{A}^0 := \{X \in M^\Phi \mid E[I(-X)] \leq \delta\}.$$

For any $X \in M^\Phi$, define

$$\rho_I(X) := \inf\{x \in \mathbf{R} \mid \exists U \in \mathcal{U} \text{ s.t. } x + U + X \in \mathcal{A}^0\}.$$

$$\rho_I(-X) := \inf\{x \in \mathbf{R} \mid \exists U \in \mathcal{U} \text{ s.t. } x + U - X \in \mathcal{A}^0\}.$$

Proposition 1 (A, 2011)

Assume $\rho_I(\mathbf{0}) > -\infty$.

ρ_I is an \mathbf{R} -valued convex risk measure on M^Φ satisfying the following:

$$\rho_I(X) = \max_{Q \in \mathcal{P}^\Psi} \left\{ E_Q[-X] - \inf_{\lambda > 0} \frac{1}{\lambda} \left\{ \delta + E \left[\Psi \left(\lambda \frac{dQ}{dP} \right) \right] \right\} \right\},$$

where

$$\begin{aligned} \mathcal{M}^\Psi &:= \{Q \ll P \mid dQ/dP \in L^\Psi, E_Q[U] \leq 0 \text{ for any } U \in \mathcal{U}\}, \\ \mathcal{A}^1 &:= \{X^1 \in M^\Phi \mid \exists U \in \mathcal{U} \text{ s.t. } X^1 + U \geq 0\}. \end{aligned}$$

◀ Theorem 3

◀ Corollary

Motivation

Shortfall risk measure for American claims.

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In this talk, we consider convex risk measures for unbounded processes in the Orlicz heart framework.

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Convex risk measure on processes

Let \mathcal{X} be the set of all adapted càdlàg processes.

$$\mathbf{X}^* := \sup_{t \in [0, T]} |\mathbf{X}_t| \text{ for any } \mathbf{X} \in \mathcal{X}$$

$$\mathcal{R}^\Phi := \{\mathbf{X} \in \mathcal{X} | \mathbf{X}^* \in M^\Phi\}.$$

← orlicz space

Convex risk measure on processes

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◀ orlicz space

Define an order \leq between two processes in \mathcal{X} as follows:

for any $\mathbf{X}^1, \mathbf{X}^2 \in \mathcal{X}$,

$$\mathbf{X}^1 \leq \mathbf{X}^2 \stackrel{\text{def.}}{\iff} \mathbf{X}_t^1 \leq \mathbf{X}_t^2 \text{ a.s. for any } t \in [0, T].$$

Remark that the following hold:

$$\begin{aligned} \mathbf{X}^1 \leq \mathbf{X}^2 &\iff P(\mathbf{X}_t^1 \leq \mathbf{X}_t^2 \text{ for any } t \in [0, T]) = 1 \\ &\iff \mathbf{X}_\tau^1 \leq \mathbf{X}_\tau^2 \text{ a.s. for any } \tau \in \mathcal{T} \\ &\implies (\mathbf{X}^1)^* \leq (\mathbf{X}^2)^* \text{ a.s.,} \end{aligned}$$

where \mathcal{T} is the set of all stopping times on $[0, T]$.

Convex risk measure on processes 2

Defining a norm $\|\cdot\|_{\mathcal{R}^\Phi}$ on \mathcal{R}^Φ as

$$\|X\|_{\mathcal{R}^\Phi} := \|X^*\|_\Phi$$

for any $X \in \mathcal{R}^\Phi$.

we can see that $(\mathcal{R}^\Phi, \|\cdot\|_{\mathcal{R}^\Phi})$ is a Banach lattice with order \preceq .

Remark that \mathcal{R}^Φ is not always order continuous.

← orlicz space

Convex risk measure on \mathcal{R}^Φ

Definition

A functional $\rho : \mathcal{R}^\Phi \rightarrow (-\infty, +\infty]$ is called a convex risk measure on \mathcal{R}^Φ , if it satisfies the following four conditions:

- ① **Properness:** $\rho(\mathbf{0}) \in \mathbf{R}$,
- ② **Monotonicity:** $\rho(\mathbf{X}) \geq \rho(\mathbf{Y})$ for any $\mathbf{X}, \mathbf{Y} \in \mathcal{R}^\Phi$ with $\mathbf{X} \leq \mathbf{Y}$,
- ③ **Translation invariance:** $\rho(\mathbf{X} + \mathbf{m}) = \rho(\mathbf{X}) - \mathbf{m}$ for $\mathbf{X} \in \mathcal{R}^\Phi$ and $\mathbf{m} \in \mathbf{R}$,
- ④ **Convexity:** $\rho(\lambda \mathbf{X} + (1 - \lambda) \mathbf{Y}) \leq \lambda \rho(\mathbf{X}) + (1 - \lambda) \rho(\mathbf{Y})$ for any $\mathbf{X}, \mathbf{Y} \in \mathcal{R}^\Phi$ and $\lambda \in [0, 1]$.

Moreover, if a convex risk measure ρ satisfies

- ⑤ **Positive homogeneity:** $\rho(\lambda \mathbf{X}) = \lambda \rho(\mathbf{X})$ for any $\lambda \geq \mathbf{0}$,

then ρ is called a coherent risk measure.

Let $(\mathcal{R}^\Phi)'$ be the dual space of \mathcal{R}^Φ , that is, the space of all continuous linear functionals on \mathcal{R}^Φ .

$$(\mathcal{R}^\Phi)'_+ := \{\mathbf{J} \in (\mathcal{R}^\Phi)' \mid \mathbf{J}(\mathbf{X}) \geq \mathbf{0} \text{ for any } \mathbf{X} \geq \mathbf{0}\}.$$

$$(\mathcal{R}^\Phi)'_1 := \{\mathbf{J} \in (\mathcal{R}^\Phi)'_+ \mid \mathbf{J}(\mathbf{1}) = \mathbf{1}\}.$$

$(\mathcal{R}^\Phi)'_1$ is nonempty.

Let $(\mathcal{R}^\Phi)'$ be the dual space of \mathcal{R}^Φ , that is, the space of all continuous linear functionals on \mathcal{R}^Φ .

$$(\mathcal{R}^\Phi)'_+ := \{J \in (\mathcal{R}^\Phi)' \mid J(X) \geq 0 \text{ for any } X \geq 0\}.$$

$$(\mathcal{R}^\Phi)'_1 := \{J \in (\mathcal{R}^\Phi)'_+ \mid J(1) = 1\}.$$

$(\mathcal{R}^\Phi)'_1$ is nonempty.

By the **extended Namioka-Klee theorem**, any convex risk measure ρ defined on \mathcal{R}^Φ is represented as

$$\rho(X) = \max_{J \in (\mathcal{R}^\Phi)'_1} \{J(-X) - a_\rho(J)\} \text{ for } X \in \text{int}(\{\rho < \infty\}), \quad (1)$$

where a_ρ is defined as

$$a_\rho(J) := \sup_{X \in \mathcal{R}^\Phi} \{J(-X) - \rho(X)\}, \quad (2)$$

which is called the penalty function of ρ .

Let ρ be a $(-\infty, \infty]$ -valued functional on \mathcal{R}^Φ . ρ admits the representation

$$\rho(X) = \sup_{J \in (\mathcal{R}^\Phi)'_1} \{J(-X) - a_\rho(J)\}, \quad (3)$$

if and only if ρ is a $\sigma(\mathcal{R}^\Phi, (\mathcal{R}^\Phi)')$ -l.s.c. convex risk measure.

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Assumptions

Let Φ' be the left derivative of Φ , I the right-continuous inverse of Φ' . We moreover define

$$q_\Phi := \inf_{x>0} \frac{x\Phi'(x)}{\Phi(x)}, \quad p'_\Psi := \frac{q_\Phi}{q_\Phi - 1}, \quad \text{and} \quad p_\Psi := \sup_{y>0} \frac{yI(y)}{\Psi(y)}.$$

Note that $p'_\Psi \leq p_\Psi$.

Assumption 1

- ① $p_\Psi < \infty$, which implies that Ψ is moderate growth function.
- ② $\lim_{x \rightarrow \infty} \frac{\Phi(x)}{x} = +\infty$.

Examples

We introduce Φ s satisfying Assumption 1:

① $\Phi(x) = |x|^p / p, p > 1.$

In this case, $p_\Psi = p / (p - 1)$ and $L^\Phi = M^\Phi.$

② $\Phi(x) = e^{|x|} - |x| - 1.$

We can say easily $p'_\Psi \leq p_\Psi \leq 2.$

In this case, M^Φ is a proper subset of $L^\Phi.$

← orlicz space

We enumerate functions which do not satisfy Assumption 1:

① $\Phi(x) = |x|.$

The second assumption is not satisfied.

② $\Phi(x) = e^{|x|} - 1.$

We can see that $p'_\Psi = +\infty.$

③ $\Phi(x) = (|x| + 1) \log(|x| + 1) - |x|.$

④ $\Phi(x) = |x| - \log(|x| + 1).$

As for the last two functions, we have $q_\Phi = 1$, that is, $p_\Psi = \infty.$

The space \mathcal{D}

$$\mathcal{D} := \left\{ (D^-, D^+) \mid \begin{array}{l} D^\pm \text{ are right-continuous, integrable variation processes,} \\ V := \int_{(0, T]} |dD_t^-| + \int_0^T |dD_t^+| \in M^\Psi, \\ D^- \text{ is a predictable process with } D_0^- = \mathbf{0}, \\ D^+ \text{ is a purely discontinuous optional process} \end{array} \right\}.$$

◀ randomized stopping time

Theorem 1

The following theorem as an extension of Theorem VII.65 in Dellacherie and Meyer.

- ① For any $(\mathbf{D}^-, \mathbf{D}^+) \in \mathcal{D}$, defining a functional \mathbf{J} as

$$\mathbf{J}(X) := E \left[\int_{(0, T]} X_{t-} d\mathbf{D}_t^- + \int_0^T X_t d\mathbf{D}_t^+ \right] \text{ for any } X \in \mathcal{R}^\Phi, \quad (4)$$

we have $\mathbf{J} \in (\mathcal{R}^\Phi)'$. Moreover, $\|\mathbf{J}\| \leq 2\|\mathbf{V}\|_\Psi$, where
 $\|\mathbf{J}\| := \sup_{X \in \mathcal{R}^\Phi, \|X\|_{\mathcal{R}^\Phi} \leq 1} |\mathbf{J}(X)|$.

- ② For any $\mathbf{J} \in (\mathcal{R}^\Phi)'$, there exists $(\mathbf{D}^-, \mathbf{D}^+) \in \mathcal{D}$ uniquely, which satisfies (4). Furthermore, we have $\|\mathbf{V}\|_\Psi \leq p'_\Psi \|\mathbf{J}\|$.

Theorem 2

Any $\mathcal{J} \in (\mathcal{R}^\Phi)'_1$ is represented as

$$\mathcal{J}(X) = E \left[\int_{(0,T]} X_{t-} L_{t-} dK_t^- + \int_0^T X_t L_t dK_t^+ \right] \text{ for any } X \in \mathcal{R}^\Phi, \quad (5)$$

where processes L , K^- and K^+ satisfy the following conditions:

- ① L is a nonnegative local martingale with $L_0 = 1$.
- ② Both K^\pm are right-continuous nondecreasing processes. K^- is a predictable process with $K_0^- = 0$, and K^+ is a purely discontinuous optional process. Denoting $K := K^- + K^+$, $0 \leq K \leq 1$ holds.

- ③ $L_t = 1 + \int_{(0,t]} 1_{\{K_{s-} < 1\}} dL_s$ and $K_t = \int_0^t 1_{\{L_s > 0\}} dK_s$.

In addition, we can take K so that $K_T = 1$, and a triplet (L, K^-, K^+) with $K_T = 1$ is essentially unique.

Canonical density

By Theorem 2, each $\mathbf{J} \in (\mathcal{R}^\Phi)'_1$ is corresponding to a triplet $(\mathbf{L}, \mathbf{K}^-, \mathbf{K}^+)$ with $\mathbf{K}_T = \mathbf{1}$ satisfying conditions 1–3 in Theorem 2.

Furthermore, $\mathbf{V}_J := \int_{(0,T]} \mathbf{L}_{t-} d\mathbf{K}_t^- + \int_0^T \mathbf{L}_t d\mathbf{K}_t^+$ is said the **canonical density** of \mathbf{J} .

◀ Theorem 3

◀ Optimal stopping

Randomized stopping times 1

Let \mathcal{R}^1 be the space \mathcal{R}^Φ corresponding to the case where $\Phi(\mathbf{x}) = |\mathbf{x}|$.

A positive linear continuous functional $\gamma \in (\mathcal{R}^1)'$ is called a **randomized stopping time**, if there exists a pair of adapted nondecreasing right-continuous processes $(\mathbf{A}^-, \mathbf{A}^+)$ such that \mathbf{A}^- is a predictable process with $\mathbf{A}_0^- = \mathbf{0}$, \mathbf{A}^+ is a purely discontinuous optional process, $\mathbf{A}_T^- + \mathbf{A}_T^+ = \mathbf{1}$, and

$$\gamma(\mathbf{X}) = E \left[\int_{(0,T]} \mathbf{X}_{t-} d\mathbf{A}_t^- + \int_0^T \mathbf{X}_t d\mathbf{A}_t^+ \right] \text{ for any } \mathbf{X} \in \mathcal{R}^1.$$

Set $\mathbf{X}_\gamma := \int_{(0,T]} \mathbf{X}_{t-} d\mathbf{A}_t^- + \int_0^T \mathbf{X}_t d\mathbf{A}_t^+$.

Denote by Γ the set of all randomized stopping times.

Randomized stopping times 2

Any stopping time $\tau \in \mathcal{T}$ is clearly a randomized stopping time, that is, $\mathcal{T} \subset \Gamma$.

Note that Γ forms a convex set.

$\Gamma = \{\mathbf{J} \in (\mathcal{R}^\Phi)'_1 \mid V_{\mathbf{J}} = \mathbf{1}\}$ holds.

Corollary

For any $\mathbf{J} \in (\mathcal{R}^\Phi)'_1$, there exist a nonnegative local martingale \mathbf{L} with $\mathbf{L}_0 = \mathbf{1}$, and a randomized stopping time $\gamma \in \Gamma$ such that

$$\mathbf{J}(\mathbf{X}) = E[(\mathbf{X}\mathbf{L})_\gamma] \text{ for any } \mathbf{X} \in \mathcal{R}^\Phi.$$

That is, each $\mathbf{J} \in (\mathcal{R}^\Phi)'_1$ has a corresponding pair (\mathbf{L}, γ) .

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Optimal stopping risk measure

$$\mathcal{E}(X) := \sup_{\tau \in \mathcal{T}} E[-X_\tau] \text{ for any } X \in \mathcal{R}^\Phi.$$

\mathcal{E} is rewritten as $\mathcal{E}(X) = \sup_{\gamma \in \Gamma} E[-X_\gamma]$.

Moreover, since \mathcal{E} is \mathbf{R} -valued, we can rewrite it as $\mathcal{E}(X) = \max_{\gamma \in \Gamma} E[-X_\gamma]$.

The penalty function $\mathbf{a}_\mathcal{E}$ is given as

$$\mathbf{a}_\mathcal{E}(\mathbf{J}) = \begin{cases} 0, & \text{if } V_{\mathbf{J}} = 1, \\ \infty, & \text{if } V_{\mathbf{J}} \neq 1, \end{cases}$$

where $V_{\mathbf{J}}$ is the canonical density of \mathbf{J} .

◀ Canonical density

Greatest coherent risk measure

- ① $\mathbb{M}_1(\mathbf{X}) := \sup_{Q \in \mathcal{P}^\Psi} E_Q[\sup_{t \in [0, T]} (-X_t)],$
- ② $\mathbb{M}_2(\mathbf{X}) := \sup_{J \in (\mathcal{R}^\Phi)'_1} J(-\mathbf{X}),$
- ③ $\mathbb{M}_3(\mathbf{X}) := \sup_{Q \in \mathcal{P}^\Psi, \tau \in \mathcal{T}} E_Q[-X_\tau] = \sup_{Q \in \mathcal{P}^\Psi, \gamma \in \Gamma} E_Q[-X_\gamma] = \sup_{J \in \mathcal{J}} J(-\mathbf{X}),$

where \mathcal{P}^Ψ is the set of all probability measures $Q \ll P$ with $dQ/dP \in M^\Psi$.

Note that $\mathbb{M}_1 \geq \mathbb{M}_2 \geq \mathbb{M}_3$ holds true, and

\mathbb{M}_2 is the greatest $\sigma(\mathcal{R}^\Phi, (\mathcal{R}^\Phi)')$ -l.s.c. coherent risk measure.

Let us see $\mathbb{M}_1 \geq \mathbb{M}_2$.

For any $J \in (\mathcal{R}^\Phi)'_1$, there exists a $Q^J \in \mathcal{P}^\Psi$ with $dQ^J/dP = V_J$.

Thus, we have $J(-\mathbf{X}) \leq E_{Q^J}[\sup_{t \in [0, T]} (-X_t)] \leq \mathbb{M}_1(\mathbf{X})$, namely,

$\mathbb{M}_1(\mathbf{X}) \geq \sup_{J \in (\mathcal{R}^\Phi)'_1} J(-\mathbf{X}).$

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We presume an investor selling an American option H . She is supposed to control her shortfall risk at the moment when H is exercised. If she selects $C \in \mathcal{C}$ as her hedging strategy and H is exercised at a stopping time τ , her final cash-flow is $x + C_\tau - H_\tau$, and its shortfall is $(-x - C_\tau + H_\tau) \vee 0$.

Denoting her loss function by I , her shortfall risk is described by $E[I(-x - C_\tau + H_\tau)]$.

We assume that $I(x) = 0$ if $x \leq 0$, and $I(x) = \Phi(x)$ if $x > 0$.

Given the limitation of shortfall risk which she can endure as $\delta > 0$, the least price for H which she can accept is given by

$$\inf\{x \in \mathbf{R} \mid \text{there exists } C \in \mathcal{C} \text{ such that } x + C - H \in \mathcal{A}^0\}, \quad (6)$$

where $\mathcal{A}^0 := \{X \in \mathcal{R}^\Phi \mid \sup_{\tau \in \mathcal{T}} E[I(-X_\tau)] \leq \delta\}$.

Regarding (6) as a functional, denoted by ρ_I , it is defined as

$$\rho_I(\mathbf{X}) := \inf\{\mathbf{x} \in \mathbf{R} \mid \text{there exists } \mathbf{C} \in \mathcal{C} \text{ such that } \mathbf{x} + \mathbf{C} + \mathbf{X} \in \mathcal{A}^0\}$$

for any $\mathbf{X} \in \mathcal{R}^\Phi$, and call it shortfall risk measure. Then, we have $\rho_I(-\mathbf{H}) = (6)$.

Proposition 4

Assuming $\rho_I(\mathbf{0}) > -\infty$, ρ_I is an \mathbf{R} -valued convex risk measure on \mathcal{R}^Φ .

Representations of ρ_I

Henceforth, we assume that $\rho_I(\mathbf{0}) > -\infty$.

Since ρ_I is \mathbf{R} -valued, Namioka-Klee theorem implies that ρ_I is represented as

$$\rho_I(X) = \max_{J \in (\mathcal{R}^\Phi)'_1} \{J(-X) - a_{\rho_I}(J)\},$$

where a_{ρ_I} is the penalty function of ρ_I defined in (2).

We focus on estimating the penalty function a_{ρ_I} . First of all, we illustrate another description of a_{ρ_I} .

Lemma

We have $a_{\rho_I}(J) = \sup_{X \in \tilde{\mathcal{A}}} J(-X)$ for any $J \in (\mathcal{R}^\Phi)'_1$, where

$\tilde{\mathcal{A}} := \{X \in \mathcal{R}^\Phi \mid \text{there exists } C \in C \text{ such that } X + C \geq Y \text{ for some } Y \in \mathcal{A}^0\}$.

Denoting $\mathcal{A}^1 := \{\mathbf{X} \in \mathcal{X} \mid \text{there exists } \mathbf{C} \in \mathcal{C} \text{ such that } \mathbf{C} + \mathbf{X} \succeq \mathbf{0}\}$, we have $\tilde{\mathcal{A}} := \{\mathbf{X}^0 + \mathbf{X}^1 \mid \mathbf{X}^0 \in \mathcal{A}^0, \mathbf{X}^1 \in \mathcal{A}^1\}$. Thus, we can rewrite $\mathbf{a}_{\rho_l}(\mathbf{J})$ as

$$\mathbf{a}_{\rho_l}(\mathbf{J}) = \sup_{\mathbf{X}^0 \in \mathcal{A}^0} \mathbf{J}(-\mathbf{X}^0) + \sup_{\mathbf{X}^1 \in \mathcal{A}^1} \mathbf{J}(-\mathbf{X}^1). \quad (7)$$

Since \mathcal{C} is cone, we have, for any $\mathbf{J} \in (\mathcal{R}^\Phi)'_1$,

$$\sup_{\mathbf{X} \in \mathcal{A}^1} \mathbf{J}(-\mathbf{X}) = \begin{cases} \mathbf{0}, & \text{if } \mathbf{J} \in (\mathcal{R}^\Phi)'_{\mathcal{C}} \\ +\infty, & \text{otherwise,} \end{cases}$$

where $(\mathcal{R}^\Phi)'_{\mathcal{C}} = \{\mathbf{J} \in (\mathcal{R}^\Phi)'_1 \mid \mathbf{J}(\mathbf{C}) \leq \mathbf{0} \text{ for any } \mathbf{C} \in \mathcal{C}\}$. We can rewrite as

$$\rho_l(\mathbf{X}) = \max_{\mathbf{J} \in (\mathcal{R}^\Phi)'_{\mathcal{C}}} \left\{ \mathbf{J}(-\mathbf{X}) - \sup_{\mathbf{X}^0 \in \mathcal{A}^0} \mathbf{J}(-\mathbf{X}^0) \right\}.$$

Hence, we estimate only the first term of the RHS of (7).

Theorem 3

For any $\mathbf{J} \in (\mathcal{R}^\Phi)'_1$, we have

$$\inf_{\lambda > 0} \frac{1}{\lambda} \{ \delta + \mathbf{E}[I^*(\lambda L^*)] \} \geq \sup_{X \in \mathcal{A}^0} \mathbf{J}(-X) \geq \inf_{\lambda > 0} \frac{1}{\lambda} \{ \delta + \mathbf{E}[I^*(\lambda V_J)] \} \quad (8)$$

where I^* is the conjugate function of I , L is the corresponding local martingale to \mathbf{J} in the sense of Theorem 2, and V_J is the canonical density of \mathbf{J} .

◀ Prop1

◀ canonical density

◀ Theorem 2

Corollary

The shortfall risk measure ρ_I is evaluated as follows:

$$\begin{aligned} & \sup_{J \in (\mathcal{R}^\Phi)'_C} \left\{ J(-X) - \inf_{\lambda > 0} \frac{1}{\lambda} \{ \delta + E[I^*(\lambda L^*)] \} \right\} \\ & \leq \rho_I(X) \leq \sup_{J \in (\mathcal{R}^\Phi)'_C} \left\{ J(-X) - \inf_{\lambda > 0} \frac{1}{\lambda} \{ \delta + E[I^*(\lambda V_J)] \} \right\}. \end{aligned}$$

◀ Prop1

Thank you for your attention!