

Multivariate Risk Measures

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Zurich, May 5, 2011

Monivariate risk measures

A *monivariate risk measure* (1-rm) is a function $\rho : L^\infty(\Omega, \mathcal{F}, P) \rightarrow \infty$ such that:

- $\rho(0) = 0$
- $X \geq Y \implies \rho(X) \leq \rho(Y)$
- $\rho(X + m) = \rho(X) - m$ for $m \in \mathbb{R}$

It is

- *convex* if ρ is a convex function
- *coherent* if it is convex and positively homogeneous:

$$\begin{aligned}\rho(X + Y) &\leq \rho(X) + \rho(Y) \\ \rho(\lambda X) &= \lambda \rho(X)\end{aligned}$$

Law-invariant 1-r.m.

We shall write $X \sim Y$ to mean that X and Y have the same law. ρ is law-invariant if $X \sim Y$ implies that $\rho(X) = \rho(Y)$

Definition

A 1-rm is strongly coherent if it is convex, law-invariant and:

$$\rho(X) + \rho(Y) = \sup \left\{ \rho(X + \tilde{Y}) \mid Y \sim \tilde{Y} \right\}$$

There are two fundamental examples of s.c. 1-r.m.

- 1 Let $F \in L^1$ be a probability density, so that $F \geq 0$ and $E[F] = 1$. Define

$$\rho_F := \sup \left\{ \mathbb{E} \left[-F\tilde{X} \right] \mid \tilde{X} \sim X \right\}$$

- 2 Set:

$$\rho_\infty(X) := \text{ess sup } -X$$

Both ρ_F and ρ_∞ are strongly coherent.

Kusuoka (2001) has shown that all strongly coherent risk measures can be built from ρ_F and ρ_∞

Theorem

ρ is a strongly coherent 1-r.m. if and only if there is a probability density F and a number s with $0 \leq s \leq 1$ such that:

$$\forall X \in L^\infty, \quad \rho(X) = s\rho_\infty(X) + (1-s)\rho_F(X)$$

A d -dimensional risk measure (d -r.m.) is a function $\rho : L^\infty(\Omega, \mathcal{F}, P; R^d) \rightarrow R$ such that:

- $\rho(0) = 0$
- $X \geq Y \implies \rho(X) \leq \rho(Y)$
- $\rho(X + me) = \rho(X) - m$ for $m \in R$ and $e = (1, \dots, 1)$

It is

- *convex* if ρ is a convex function
- *coherent* if it is convex and positively homogeneous:

$$\begin{aligned}\rho(X + Y) &\leq \rho(X) + \rho(Y) \\ \rho(\lambda X) &= \lambda \rho(X)\end{aligned}$$

ρ is law-invariant if $X \sim Y$ implies that $\rho(X) = \rho(Y)$

Definition

(Galichon, Henry) A d-rm is *strongly coherent* if it is convex, law-invariant and:

$$\rho(X) + \rho(Y) = \sup \left\{ \rho(X + \tilde{Y}) \mid Y \sim \tilde{Y} \right\}$$

Examples

There are two fundamental examples of s.c. 1-r.m.

- 1 Let $F = (F_1, \dots, F_d) \in L^1(R^d)$ satisfy $F_i \geq 0, 1 \leq i \leq d$, and $\sum_i E[F_i] = 1$. Define

$$\rho_F := \sup \left\{ \mathbb{E} \left[-F\tilde{X} \right] \mid \tilde{X} \sim X \right\} = \sup \left\{ \sum_i \mathbb{E} \left[-F_i \tilde{X}_i \right] \mid \tilde{X} \sim X \right\}$$

- 2 Denote by S^d the unit simplex in R^d and let $\xi \in S^d$, so that $\xi_i \geq 0$ and $\sum \xi_i = 1$. Define

$$\rho_\xi(X) := \text{ess sup} -X\xi = \text{ess sup} - \sum X_i \xi_i$$

- 3 Let μ be a probability on S^d . Set:

$$\rho_\mu(X) := \int_{S^d} \rho_\xi(X) d\mu(\xi)$$

ρ_F and ρ_μ are strongly coherent d-r.m.

Extension of Kusuoka's theorem

Schachermayer and IE have shown that all strongly coherent risk measured-r.m. can be built from ρ_F and ρ_∞

Theorem

ρ is a strongly coherent d -r.m. if and only if there is some $F \in L^1_+(R^d)$ with $\sum_i E[F_i] = 1$, a probability μ on S^d and a number s with $0 \leq s \leq 1$ such that:

$$\forall X \in L^\infty, \quad \rho(X) = s\rho_\mu(X) + (1-s)\rho_F(X)$$

This result builds on an earlier result by Galichon, Henry and IE, which we will explain in due course.

If $\rho : L^\infty(R^d) \rightarrow R$ is a convex and law-invariant, it is lower semi-continuous wrt $\sigma(L^\infty, L^1)$ (Jouini, Schachermayer, Touzi, 2006). It follows that ρ is coherent (homogeneous) if and only if there exists a closed, convex, law-invariant subset C of $L^1(R^d)$ such that.

$$\rho(X) = \sup_{F \in C} \langle F, X \rangle$$

We shall set $\rho = \rho_C$ to remember this fact. Denote by \mathcal{T} the set of measure-preserving maps τ from Ω into itself. Set:

$$K := \{(F, F \circ \tau) \mid F \in C, \tau \in \mathcal{T}\}$$

Since C is law-invariant, $K \subset C \times C$. Conversely:

Lemma

ρ is strongly coherent iff $C \times C$ is the closed convex hull of K

Suppose $\rho = \rho_C$ and $C \times C$ is the closed convex hull of K . Then:

$$\begin{aligned}
 \rho_C(X) + \rho_C(Y) &= \sup \{ -\langle F, X \rangle - \langle G, X \rangle \mid (F, G) \in C \times C \} \\
 &= \sup \{ -\langle F, X \rangle - \langle F \circ \tau, Y \rangle \mid F \in C, \tau \in \mathcal{T} \} \\
 &= \sup \{ -\langle F, X + Y \circ \tau^{-1} \rangle \mid F \in C, \tau \in \mathcal{T} \} \\
 &\leq \sup \{ -\langle F, X + \tilde{Y} \rangle \mid \tilde{Y} \sim Y \} = \rho_C(X + \tilde{Y})
 \end{aligned}$$

and the reverse inequality is always true

Theorem

Let $C \subset L^1(\mathbb{R}^d)$ be strongly coherent. Then there exists a number t , with $0 \leq t \leq 1$, and two closed convex law-invariant subsets C^r and C^s of $L^1(\mathbb{R}^d)$ such that:

- 1 $C = (1 - t) C^r + t C^s$
- 2 C^r is $\sigma(L^1, L^\infty)$ -compact
- 3 $\overline{C^s} \subset (L^\infty)^*$ is $\sigma((L^\infty)^*, L^\infty)$ -compact, and its extreme points are purely singular

The regular case

This is the case when $C^s = \emptyset$. In other words, $C = C^r$ is weakly compact in $L^1(R^d)$. For instance, $C \subset L^p(R^d)$, so that $\rho = \rho_C$ is continuous in the $L^p(R^d)$ norm. This case was dealt with in an earlier paper by Galichon, Henry and IE.

Since C is weakly compact, it is the closed convex hull of its set of strongly exposed points. Let $F \in C$ be such a point. We claim that $\rho = \rho_F$.

By definition, there is some $X \in L^\infty(R^d)$ such that $\sup_{H \in C} -\langle H, X \rangle = -\langle F, X \rangle$ and if $H_n \in C$ is a maximizing sequence, then $H_n \rightarrow F$ strongly in $L^1(R^d)$. Since ρ is strongly coherent, we have:

$$\rho(X) + \rho(Y) = \sup \{ -\langle G, X \rangle - \langle G \circ \tau, Y \rangle \mid G \in C, \tau \in \mathcal{T} \}$$

Let (G_n, τ_n) be a maximizing sequence. Since we have $=$ instead of \geq , we must have $\rho(X) = \sup -\langle G_n, X \rangle$ and $\rho(Y) = \sup -\langle G_n \circ \tau_n, Y \rangle$ so $G_n \rightarrow F$ and:

$$\rho(Y) = \sup_{\tau} -\langle F \circ \tau, Y \rangle = \sup_{\tau} -\langle F, Y \circ \tau^{-1} \rangle = \sup_{Y \sim \tilde{Y}} -\langle F, \tilde{Y} \rangle$$

so $\rho = \rho_F$ as desired

The singular case 1

Let $\mathcal{G} = (A_i)$ be a finite partition of Ω . On each A_i we are given a point $x_i \in S^d$ and an element $\beta_i \in (L^\infty(R))^*$. We then build an element $\beta_{\mathcal{G}} \in (L^\infty(R^d))^*$ by the formula

$$\langle \beta_{\mathcal{G}}, X \rangle = \sum \langle \beta_i, (x_i \cdot X) \rangle$$

Denote by δ_i the Dirac mass at x_i . We associate with $\beta_{\mathcal{G}}$ the Borel measure $\mu_{\mathcal{G}}$ on S^d defined by:

$$\mu_{\mathcal{G}} = \sum \delta_i \beta_i(A_i)$$

Lemma

Suppose $\beta_{\mathcal{G}}$ is law-invariant. Then, for $X \in L^\infty(R^d)$ we have:

$$\sup_{\tau \in \mathcal{T}} -\langle \beta_{\mathcal{G}} \circ \tau, X \rangle = \int_{S^d} \sup - (x \cdot X) d\mu(x)$$

$$\begin{aligned}
\int_{S^d} \sup - (x \cdot X) d\mu(x) &= \int_{S^d} \sup - (x \cdot X \circ \tau^{-1}) d\mu(x) \\
&= \sum \sup - (x_i \cdot X \circ \tau^{-1}) \beta_i(A_i) \\
&\geq -\langle \beta_G \circ \tau, X \rangle
\end{aligned}$$

and the reverse inequality follows from the fact that β_G is purely singular. Taking points (ξ_i) and disjoint sets (B_i) with $\mathbb{P}(B_i) > 0$ such that

$$\text{ess sup } -x_i \cdot X = -x_i \cdot \xi_i \text{ and } B_i \subset \{|X - \xi_i| < \varepsilon\}$$

we can find a set N with $0 < \mathbb{P}(N) < \inf \mathbb{P}(B_i)$ such that $\beta_G \mathbf{1}_N = \beta_G$. We then find a measure-preserving transformation which maps $N \cap A_i$ into B_i and the result follows

The singular case: simple elements

Let $\beta \in (L^\infty(\mathbb{R}^d))^* \geq 0$ be a purely singular element. We define a finitely additive measure $|\beta|$ by:

$$|\beta|[A] = \langle \mathbf{e}\mathbf{1}_A, \beta \rangle$$

Let $\mathcal{G} = (A_i)$ be a finite partition of Ω . Set:

$$\begin{aligned}\beta_{\mathcal{G}} &= \sum \xi_i (|\beta|[G_i]) \\ \xi_i &= \frac{\langle \mathbf{e}_i \mathbf{1}_{G_i}, \beta \rangle}{\langle \mathbf{e} \mathbf{1}_{G_i}, \beta \rangle}\end{aligned}$$

We can approximate β by the $\beta_{\mathcal{G}}$ which have a simpler form. We associate with $\beta_{\mathcal{G}}$ the Borel measure:

$$\mu_{\mathcal{G}} = \sum |\beta|[G_i] \delta_{\xi_i}$$

and we show that the $\mu_{\mathcal{G}}$ converge to some μ as the $\beta_{\mathcal{G}}$ converge to β .