

# Law invariant convex risk measures on $\mathbb{R}^d$

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# The classical one-dimensional case

We extend the following theorems on risk measures, which are well-known for the one-dimensional case, to the  $d$ -dimensional case.

We work on a standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ .

## Reminder

A function  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow \mathbb{R}$  is called *law invariant* if  $X \sim Y$ , i. e.  $\text{law}(X) = \text{law}(Y)$ , implies that  $\varrho(X) = \varrho(Y)$ .

For  $F \in L^1_+(\Omega, \mathcal{F}, \mathbb{P})$  normalized by  $\mathbb{E}[F] = 1$ , we define the law invariant risk measure  $\varrho_F : L^\infty(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow \mathbb{R}$  by

$$\varrho_F(X) = \sup_{\tilde{X} \sim X} \mathbb{E}[-\tilde{X}F] = \sup_{\tilde{F} \sim F} \mathbb{E}[-X\tilde{F}].$$

The measures  $\varrho_F$  have the following co-monotonicity property.

## Definition

A risk measure  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow \mathbb{R}$  is called co-monotone if, for co-monotone random variables  $X, Y$ , we have

$$\varrho(X + Y) = \varrho(X) + \varrho(Y).$$

Recall that  $X$  is co-monotone to  $Y$  iff

$[X(\omega) - X(\omega')] \cdot [Y(\omega) - Y(\omega')] \geq 0$ , for  $\mathbb{P} \otimes \mathbb{P}$  almost all  $(\omega, \omega') \in \Omega \times \Omega$ . We write  $X \sim_c Y$ .

## Theorem A (Kusuoka, 2001)

For a law invariant convex risk measure  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow \mathbb{R}$   
t.f.a.e.

- (i)  $\varrho$  is co-monotone.
- (ii) There is  $F \in L^1(\Omega, \mathcal{F}, \mathbb{P})$  and  $0 \leq \alpha \leq 1$  s.t.  
$$\begin{aligned}\varrho(X) &= \alpha \varrho_F(X) + (1 - \alpha) \text{ess sup}(-X) \\ &=: \alpha \varrho_F(X) + (1 - \alpha) \varrho^\infty(X).\end{aligned}$$
- (iii)  $\varrho$  is *strongly coherent*, i.e.

$$\varrho(X) + \varrho(Y) = \sup_{\tilde{X} \sim X, \tilde{Y} \sim Y} \varrho(\tilde{X} + \tilde{Y}).$$

The equivalence (i)  $\Leftrightarrow$  (ii) is due to Kusuoka.

Property (iii) is easily seen to be equivalent to (i) in the one-dimensional setting.

While it is – a priori – not clear how to extend the notion (i) of *co-monotonicity* to the vector-valued setting, the notion (iii) of *strong coherence* extends to the vector-valued case on an obvious way. This is the reason why Ekeland-Galichon-Henry (2009) introduced this notion (in the vector valued setting).

Here is the second theorem which we state for the one-dimensional case, and later extend to the  $d$ -dimensional one.

## Theorem B (Kusuoka, 2001)

Let  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow \mathbb{R}$  be a law invariant convex risk measure. Then there is a function  $v : [0, 1] \times \mathcal{P} \rightarrow [0, \infty]$  such that

$$\varrho(X) = \max_{(\alpha, F) \in [0, 1] \times \mathcal{P}} \{ \alpha \varrho_F(X) + (1 - \alpha) \varrho^\infty(X) - v(\alpha, F) \}.$$

## Definition

A convex risk measure on  $\mathbb{R}^d$  is a function

$\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d) \rightarrow \mathbb{R}$  s.t.

(i) (normalization)  $\varrho(0) = 0,$ ,

(ii) (monotonicity)  $X \geq Y \Rightarrow \varrho(X) \leq \varrho(Y),$

(iii) (cash invariance)  $\varrho(X + m\mathbb{1}) = \varrho(X) - m,$  for  $m \in \mathbb{R},$

(iv) (convexity)

$\varrho(\alpha X + (1 - \alpha)Y) \leq \alpha\varrho(X) + (1 - \alpha)\varrho(Y),$  for  $0 < \alpha < 1.$

We call  $\varrho$  *coherent* if, in addition,

(v) (positive homogeneity)  $\varrho(\lambda X) = \lambda\varrho(X),$  for  $\lambda \geq 0.$



On the standard probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  we now denote by  $\mathcal{P}$  the subset of  $L^1_+(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$  of normalized functions  $F$  taking their values in  $\mathbb{R}_+^d$ ,

$$\mathcal{P} = \left\{ F = (F_i)_{i=1}^d \in L^1(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d) : F_i \geq 0, \mathbb{E} \left[ \sum_{i=1}^d F_i \right] = 1 \right\}.$$

## Definition (Rüschendorf)

For given  $F \in \mathcal{P}$  we define  $\varrho_F$ , the *maximal correlation risk measure in the direction  $F$* , by

$$\varrho_F(X) = \sup_{\tilde{X} \sim X} \mathbb{E} \left[ -(\tilde{X}, F) \right] = \sup_{\tilde{F} \sim F} \mathbb{E} \left[ -(X, \tilde{F}) \right].$$

Note that  $\varrho_F$  only depends on the law of  $F$ .

## Proposition (Rüschendorf)

A coherent (resp. convex) risk measure  $\varrho : L_d^\infty \rightarrow \mathbb{R}$  is *law invariant* if and only if it can be represented as

$$\varrho(X) = \sup_{F \in C} \varrho_F(X)$$

$$\text{resp. } \varrho(X) = \sup_{F \in C} \{\varrho_F(x) - v(F)\}$$

where  $C$  is a subset of  $\mathcal{P}$  and  $v : C \rightarrow \mathbb{R}_+$  a non-negative function defined on  $C$ .

Apart from the risk measures  $\rho_F$ , where  $F \in \mathcal{P}$ , a second type of risk measures will play a special role. It generalizes the maximal loss measure from the one- to the  $d$ -dimensional case.

## Definition

For  $\xi \in S^d$ , where

$$S^d := \left\{ \xi \in \mathbb{R}^d : \xi_i \geq 0, \sum_{i=1}^d \xi_i = 1 \right\},$$

we define the *maximal loss measure in the direction*  $\xi$  by

$$\varrho_{\xi}^{\infty}(X) = \text{ess sup} \left\{ - \sum_{i=1}^d \xi_i X_i \right\}.$$

More generally, for a probability measure  $\mu$  on  $S^d$  we may define

$$\varrho_{\mu}^{\infty}(X) = \int_{S^d} \varrho_{\xi}^{\infty}(X) d\mu(\xi).$$

## Theorem (Ekeland, Galichon, Henry, 2009)

Assume that  $\varrho$  is a convex, law invariant risk measure on  $\mathbb{R}^d$  which extends continuously from  $L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$  to  $L^p(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ , for some  $p < \infty$ . Then there is a function  $v : \mathcal{P} \rightarrow [0, \infty]$  such that

$$\varrho(X) = \max_{F \in \mathcal{P}} \{\varrho_F(X) - v(F)\}$$

## Theorem (Ekeland, S., 2011)

Assume that  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d) \rightarrow \mathbb{R}$  is a convex, law invariant risk measure on  $\mathbb{R}^d$ . Then there is a function  $v : [0, 1] \times \mathcal{P} \times \mathcal{M}_+^1(S^d) \rightarrow [0, \infty]$  such that

$$\varrho(X) = \max_{(\alpha, F, \mu) \in [0, 1] \times \mathcal{P} \times \mathcal{M}_+^1(S^d)} \{\alpha \varrho_F(X) + (1 - \alpha) \varrho_\mu^\infty(X) - v(\alpha, F, \mu)\}$$

The law invariant risk measure  $\varrho$  is coherent if and only if  $v$  can be chosen to take only values in  $\{0, \infty\}$ .

### Definition (Ekeland, Galichon, Henry, 2009)

A convex, law invariant risk measure  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d) \rightarrow \mathbb{R}$  is *strongly coherent* if

$$\varrho(X) + \varrho(Y) = \sup \left\{ \varrho(\tilde{X} + \tilde{Y}) : \tilde{X} \sim X, \tilde{Y} \sim Y \right\}.$$

### Theorem (Ekeland, Galichon, Henry, 2009)

Let  $\varrho$  be a convex, law invariant risk measure which extends continuously from  $L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$  to  $L^p(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ , for some  $p < \infty$ . Then  $\varrho$  is strongly coherent if and only if there is some  $F \in \mathcal{P}$  such that

$$\varrho(X) = \alpha \varrho_F(X).$$

### Theorem (Ekeland, S., 2011)

A convex, law invariant risk measure  $\varrho : L^\infty(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d) \rightarrow \mathbb{R}$  is strongly coherent if and only if there is some  $F \in \mathcal{P}, \mu \in \mathcal{M}_+(S^d)$  and  $\alpha \in [0, 1]$  such that

$$\varrho(X) = \alpha \varrho_F(X) + (1 - \alpha) \varrho_\mu^\infty(X).$$