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Central limit theorems for coherent law-invariant risk measures

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Significance of coherent law-invariant risk measures

- Statistical functionals $\rho : \mathbb{F} \rightarrow \mathbb{R}$ of special type meeting requirements to serve as building blocks in quantitative risk management.
- Bringing viewpoints of insurance and finance together:

| | insurance | finance |
|----------------------|---|--|
| subject | premium calculation | risk assessment |
| object | contracts | financial risks |
| desirable properties | <ul style="list-style-type: none">• law-invariant• safety loadings• risk aversion | <ul style="list-style-type: none">• monotone w.r.t. stochastic order• minimal capital requirement• diversification• scale invariant |
| | derived Bäuerle/Müller (2006) | axiomatic Artzner et al. (1999) |

Estimation of risk measures

risk measure $\rho : \mathbb{F} \rightarrow \mathbb{R}$

- **Practical problems to calculate $\rho(F)$:**

- F unknown but data set x_1, \dots, x_n available
- Calculation of $\rho(F)$ for complicated F very difficult in practice.

↓

- **Canonical plug-in estimation:**

$$x_1, \dots, x_n \rightarrow F \approx \hat{F}_n \rightarrow \rho(F) \approx \rho(\hat{F}_n)$$

\hat{F}_n as easy as possible, **canonical choice:** $\hat{F}_n \hat{=}$ empirical d.f. w.r.t. (x_1, \dots, x_n) .

Subject of the talk: Limit distribution of $\rho(\hat{F}_n)$?

- 1 Motivation
- 2 A representation result for coherent law-invariant risk measures
- 3 The estimation method
- 4 Main result
- 5 Final remarks

Axiomatic approach by Artzner et al. (1999)

- $\mathbb{F}_{\mathcal{X}} =: \{F_X \mid X \in \mathcal{X}\}$ set of distribution functions on \mathbb{R} **loss distribution functions**
 - $(\Omega, \mathcal{F}, \mathbb{P})$ atomless probability space,
 - $\mathcal{L}^\infty(\Omega, \mathcal{F}, \mathbb{P}) \subseteq \mathcal{X} \subseteq \mathcal{L}^1(\Omega, \mathcal{F}, \mathbb{P})$ vector subspace, **financial risks**
 - \mathcal{X} **Stonean vector lattice**

$$X \wedge Y, X \vee Y, (X - c)^+, X \wedge c \in \mathcal{X} \text{ for } X, Y \in \mathcal{X}, c \in \mathbb{R}.$$

- $\rho : \mathbb{F}_{\mathcal{X}} \rightarrow \mathbb{R}$ **coherent law-invariant risk measure**, if

- **monotonicity**

$$\rho(F_{X_1}) \leq \rho(F_{X_2}) \text{ for } F_{X_1} \leq F_{X_2}.$$

- **cash invariance**

$$\rho(F_{X+c}) = \rho(F_X) - c \text{ for } X \in \mathcal{X} \text{ and } c \in \mathbb{R}.$$

- **sublinearity**

$$\rho(F_{\lambda_1 X_1 + \lambda_2 X_2}) \leq \lambda_1 \rho(F_{X_1}) + \lambda_2 \rho(F_{X_2}) \text{ for } X_1, X_2 \in \mathcal{X} \text{ and } \lambda_1, \lambda_2 \geq 0.$$

Examples

- Concave distortion risk measure:

$$\rho_{\psi}(F_X) := \rho(F_X) = \int_{-\infty}^0 \psi(F_X(x)) dx - \int_0^{\infty} [1 - \psi(F_X(x))] dx = \int (-X) d\psi(\mathbb{P}),$$

for some concave $\psi : [0, 1] \rightarrow [0, 1]$ which is a **distortion function**, i.e. nondecreasing with $\psi(0) = 0, \psi(1) = 1$.

- Average Value at Risk (Tail Value at Risk, Expected Shortfall): $\alpha \in]0, 1[$

$$\rho(F_X) =: AV@R_{\alpha}(F_X) := \frac{1}{\alpha} \int_0^{\alpha} q_{F_X}(1 - \beta) d\beta = \frac{1}{\alpha} \int_0^{\alpha} V@R_{\beta}(X) d\beta$$

$$AV@R_{\alpha} = \rho_{\psi} \text{ with } \psi(t) = \frac{1}{\alpha} (\alpha \wedge t)$$

Examples (not concave distortion risk measures)

- $\rho = \sup_{\psi \in \Psi} \rho_{\psi}$, where Ψ some set of concave distortion functions.

- Risk measures based on one-sided moments: [Fischer (2003)]

$$\rho(F_X) =: \rho_{p,a}(F_X) := -\mathbb{E}[X] + a \|(X - \mathbb{E}[X])^{-}\|_p \quad (a \in [0, 1], p \in [1, \infty]).$$

- Expectiles: [Newey/Powell (1987), Müller (2010)]

$$\rho_{\alpha}(F_X) := \operatorname{argmin}_{x \in \mathbb{R}} \left[(1 - \alpha) \|((-X) - x)^{-}\|_2^2 + \alpha \|((-X) - x)^{+}\|_2^2 \right] \quad (\alpha \in [1/2, 1])$$

Regular coherent law-invariant risk measures

$\rho : \mathbb{F}_{\mathcal{X}} \rightarrow \mathbb{R}$ coherent law-invariant risk measure

- **associated distortion function:** [K./Zähle (2010)]

$\psi_{\rho} : [0, 1] \rightarrow \mathbb{R}$, $t \mapsto \rho(F_{-X(t)})$, $X(t) \sim B(1, t)$, is a distortion function

- ρ **regular** if $\lim_{k \rightarrow \infty} \rho(F_{-(X-k)^+}) = 0$ for nonnegative $X \in \mathcal{X}$.

ρ **regular** $\Rightarrow \rho(F_X) = \sup_{m \in \mathbb{N}} \inf_{k \in \mathbb{N}} \rho(F_{[X+\wedge k] - [X-\wedge m]})$ for $X \in \mathcal{X}$ [K. (2007)]

- ρ **strictly regular** if ρ regular, and $\lim_{t \rightarrow 0^+} \psi_{\rho}(t) = 0$.

- **Prominent examples:** Exists Riesz-seminorm $\|\cdot\|$ on \mathcal{X}

- $\|\cdot\|$ **complete** $\Rightarrow \rho$ **regular**.

- $\|\cdot\|$ **complete and σ -order continuous** $\Rightarrow \rho$ **strictly regular**.

- $\|\cdot\|$ **Luxemburg seminorm** on Orlicz space or Orlicz heart w.r.t. **continuous Young function** $\Rightarrow \rho$ **is strictly regular**.

Representation by concave distortion risk measures

■ **2.1 Proposition:** [Belomestny/K. (2010)]

- $\rho : F_{\mathcal{X}} \rightarrow \mathbb{R}$ regular coherent law-invariant risk measure $\Rightarrow \rho = \sup_{\psi \in \Psi} \rho_{\psi}$,

where Ψ some set of concave distortion function.

- Representing Ψ compact w.r.t. uniform metric $\Leftarrow \rho$ strictly regular.

[Based on Kusuoka (2001) with Jouini/Schachermayer/Touzi (2006)]

■ **Representation typically unknown!** Partial information by ψ_{ρ}

$$\psi_{\rho} = \sup_{\psi \in \Psi} \psi \text{ and } \sup_{\psi \in \Psi} |\psi(t) - \psi(s)| \leq \psi_{\rho}(|t - s|) \text{ if } \rho = \sup_{\psi \in \Psi} \rho_{\psi}$$

The set up

- $\rho : \mathbb{F}_{\mathcal{X}} \rightarrow \mathbb{R}$ **strictly regular** coherent law-invariant risk measure.
- $(X_i)_i$ **strictly stationary** sequence of random variables with common distribution function $F \in \mathbb{F}_{\mathcal{X}}$.
- $F_n \hat{=}$ empirical distribution function based on (X_1, \dots, X_n)
- **Estimation:**

$$\rho_n(F) := \rho(F_n).$$

Examples

- **Concave distortion risk measure:** $\rho = \rho_\psi$

$$\rho_n(F) = - \int_0^1 q_{F_n}(t) \psi'(t) dt. \quad L\text{-Statistic.}$$

- **Average Value at Risk:** $\rho = AV @ R_\alpha$

$$\rho_n(F) = \frac{1}{\alpha} \frac{1}{n} \sum_{i=1}^n (q_{F_n}(\alpha) - X_i)^+ - q_{F_n}(\alpha).$$

- **Risk measures based on one-sided moments:** $\rho = \rho_{p,a}$

$$\rho_n(F) = -\frac{1}{n} \sum_{j=1}^n X_j + a \left(\sum_{i=1}^n \left(\left[\frac{1}{n} \sum_{j=1}^n X_j - X_i \right]^+ \right)^p \frac{\#\{k \mid X_k = X_i\}}{n} \right)^{1/p}.$$

- **Expectiles:** $\rho = \rho_\alpha$

$$\rho_n(F) = \operatorname{argmin}_{x \in \mathbb{R}} \sum_{i=1}^n \left[(1 - \alpha) \left((-X_i - x)^- \right)^2 + \alpha \left((-X_i - x)^+ \right)^2 \right].$$

Basic assumptions

(M) Data

$(X_i)_{i \in \mathbb{N}}$ strongly mixing with mixing coefficients $\alpha(i) \leq \bar{\alpha}_0 \exp(-\bar{\alpha}_1 i)$ for some $\bar{\alpha}_0, \bar{\alpha}_1 > 0$ (e.g. i.i.d. data, also general ARMA- or GARCH-processes).

(J) Jumps of distribution

F has a finite set $D(F)$ s.t. $F \llcorner_{q_F(0), q_F(1)} \setminus D(F)$ is continuously differentiable with strictly positive derivative.

(I) Tails of distribution

F fulfills the following integrability condition:

$$\int_{\mathbb{R}} F(x)^{-1/2-2\delta} (1-F(x))^{1/2-\delta} \psi_{\rho}(\lambda F(x)^{1+\delta}) dx < \infty$$

for some $\lambda \in]0, 1/2[$, $\delta \in [0, 1/2[$, where $\delta = 0$ iff i.i.d. data **“loss” tail matters!**

Basic idea

$\rho : \mathbb{F}^{\mathcal{X}} \rightarrow \mathbb{R}$ **strictly regular** coherent law-invariant risk measure

■ $\rho = \sup_{\psi \in \Psi} \rho_{\psi}$, Ψ compact w.r.t. uniform metric (**repres. possibly unknown!**).

■ **Reduction to convergence of stochastic processes:**

■ (T) \rightarrow stochastic processes $(\rho_{\psi}(F_n))_{\psi \in \Psi}$, $(\rho_{\psi}(F))_{\psi \in \Psi}$ **Borel random elements in path space**

$UCB(\Psi) := \{f : \Psi \rightarrow \mathbb{R} \mid f \text{ uniformly continuous w.r.t. uniform metric}\}$
endowed with sup norm

■ Under (M), (J), (I) **convergence in law in $UCB(\Psi)$**

$$(\sqrt{n}[\rho_{\psi}(F_n) - \rho_{\psi}(F)])_{\psi \in \Psi} \xrightarrow{L} G$$

\Rightarrow application of functional Delta method for sup functionals

$$\sqrt{n}[\rho_n(F) - \rho(F)] = \sqrt{n} \left[\sup_{\psi} \rho_{\psi}(F_n) - \sup_{\psi} \rho_{\psi}(F) \right] \xrightarrow{L} \sup_{\psi, \rho(F) = \rho_{\psi}(F)} G(\psi)$$

Ingredients of proof

- Approximation lemma for Borel random elements in separable metric spaces
← compactness of Ψ , (T) [Billingsley (1968)]
- Donsker's functional theorem ← (M) [Ben Hariz (2005)]
- Bounds for empirical distribution function ← (M) [Puri/Tran (1980)]
- Functional Delta method for quantile transforms ← (J) [van der Vaart (1998)]

■ 4.1 Theorem: [Belomestny/K. (2010)]

Under **(M)**, **(J)**, **(I)**, we may find

- a compact set $S(\rho(F))$ of continuous, concave distortion functions,
- some centered Gaussian process $(G(\psi))_{\psi \in S(\rho(F))}$ with continuous paths w.r.t. uniform metric, and

$$\begin{aligned} \mathbb{E}[G(\psi_1)G(\psi_2)] &= \int_{\mathbb{R}^2} \psi_1'(F(x))\psi_2'(F(y))\{F(x \wedge y) - F(x)F(y) \\ &\quad + 2 \sum_{k=1}^{\infty} [\mathbb{P}(\{X_1 \leq x, X_{k+1} \leq y\}) - F(x)F(y)]\} dx dy \end{aligned}$$

such that $(\sqrt{n}[\rho_n(F) - \rho(F)])_{n \in \mathbb{N}}$ converges in law to $\max_{\psi \in S(\rho(F))} G(\psi)$.

- Moreover, $\sup_{\psi \in S(\rho(F))} G(\psi) = G(Z)$ for some Borel-random element Z of $S(\rho(F))$

if $\mathbb{E}[G(\psi_1) - G(\psi_2)]^2 \neq 0$ for any two different $\psi_1, \psi_2 \in S(\rho(F))$.

In particular $\sup_{\psi \in S(\rho(F))} G(\psi) = G(\psi_\rho)$ if ρ concave distortion risk measure.

Limit random variable $\sup_{\psi \in S(\rho(F))} G(\psi)$ in main result with $\sigma := \sup_{\psi \in S(\rho(F))} \mathbb{E}[G(\psi)^2]$

How to utilize for confidence intervals?

- Large deviation principle for suprema of Gaussian processes:

$$\mathbb{P}(\{ \sup_{\psi \in S(\rho(F))} G(\psi) < z \}) \geq \Phi(z/\sigma)$$

- Idea for one-sided confidence interval:

Find estimator $\hat{\sigma}_n \xrightarrow{P} \bar{\sigma} \geq \sigma$, and choose suitable quantile λ of $N(0, 1)$.

$$\text{confidence interval: }] -\infty, \rho_n(F) - \frac{\lambda \hat{\sigma}_n}{\sqrt{n}} [$$

- In case of i.i.d. data simple ad hoc methods for $\hat{\sigma}_n$ available, but in general not!
→ future work (subsampling? use of associated distortion function might help!)