

Statistical inference for risk measures

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DAA-Workshop für junge Mathematiker
Schloss Reisenburg
September 4–6, 2014

Based on joint work with

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1. Introduction

What is a risk measure?

Let X be a risky position, e.g.

1. the total claim of an insurance collective in the next period,
2. the shortfall of the equity of an insurer at a reporting date
(= “cash value” of liabilities – “cash value” of assets).

A monetary risk measure is a “suitable” map $\rho : \mathcal{X} \rightarrow \mathbb{R}$ which assigns to the position $X \in \mathcal{X}$ the minimal amount of capital $\rho(X)$ which, if added to the position and invested in a risk-free manner, makes the position “acceptable”.

A position $Y \in \mathcal{X}$ is “acceptable” with respect to ρ if $\rho(Y) \leq 0$.

$$\rho(X - \rho(X)) = \rho(X) - \rho(X) = 0 \quad (\rightarrow \text{axiom (2)}).$$

1. Introduction

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A monetary risk measure is a “suitable” map $\rho : \mathcal{X} \rightarrow \mathbb{R}$ which assigns to the position $X \in \mathcal{X}$ the minimal amount of capital $\rho(X)$ which, if added to the position and invested in a risk-free manner, makes the position “acceptable”. E.g.

1. $\rho(X)$ = premium for the whole collective
2. $\rho(X)$ = solvency capital requirement (SCR) of insurer

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be atomless, and $\mathcal{X} \subset L^0(\Omega, \mathcal{F}, \mathbb{P})$ be a vector space containing the constants. Let $\rho : \mathcal{X} \rightarrow \mathbb{R}$ be a map, and consider the following conditions:

- (1) monotonicity: $\rho(X_1) \leq \rho(X_2)$ for all $X_1, X_2 \in \mathcal{X}$ with $X_1 \leq X_2$.
- (2) cash additivity: $\rho(X + m) = \rho(X) + m$ for all $X \in \mathcal{X}$ and $m \in \mathbb{R}$.
- (3) subadditivity: $\rho(X_1 + X_2) \leq \rho(X_1) + \rho(X_2)$ for all $X_1, X_2 \in \mathcal{X}$.
- (4) positive homogeneity: $\rho(\lambda X) = \lambda \rho(X)$ for all $X \in \mathcal{X}$ and $\lambda \geq 0$.

ρ is a **monetary risk measure** if (1)–(2) hold.

ρ is a **coherent risk measure** if (1)–(4) hold.

ρ is **law-invariant** if $\rho(X_1) = \rho(X_2)$ whenever $\mathbb{P}_{X_1} = \mathbb{P}_{X_2}$.

Example 1 The **expectation**

$$\mathbb{E}[X] := \int X d\mathbb{P} = \int x dF_X(x)$$

is a law-invariant **coherent** risk measure on $\mathcal{X} = L^1(\Omega, \mathcal{F}, \mathbb{P})$.
It has the representation

$$\mathbb{E}[X] = - \int_{-\infty}^0 F_X(x) dx + \int_0^{\infty} (1 - F_X(x)) dx.$$

Example 2 The **Value at Risk** at level $\alpha \in (0, 1)$

$$\text{V@R}_\alpha(X) := F_X^{\leftarrow}(1 - \alpha) = \inf\{x \in \mathbb{R} : F_X(x) \geq 1 - \alpha\}$$

is a law-invariant and positively homogeneous monetary risk measure on $\mathcal{X} = L^0(\Omega, \mathcal{F}, \mathbb{P})$. But it is not subadditive, hence **not coherent**.

Example 3 The **Average Value at Risk** at level $\alpha \in (0, 1)$

$$\text{AV@R}_\alpha(X) := \frac{1}{\alpha} \int_0^\alpha \text{V@R}_s(X) ds$$

is a law-invariant **coherent** risk measure on $\mathcal{X} = L^1(\Omega, \mathcal{F}, \mathbb{P})$.
If F_X is continuous at $\text{V@R}_\alpha(X)$, then

$$\text{AV@R}_\alpha(X) = \mathbb{E}[X | X \geq \text{V@R}_\alpha(X)].$$

Example 4 Let g be a convex distortion function, i.e. a convex nondecreasing function $g : [0, 1] \rightarrow [0, 1]$ with $g(0) = 0$ and $g(1) = 1$. The **distortion risk measure** associated with g

$$\rho_g(X) := - \int_{-\infty}^0 g(F_X(x)) dx + \int_0^{\infty} (1 - g(F_X(x))) dx$$

is a law-invariant **coherent** risk measure on $\mathcal{X} = \mathcal{X}_g := \{\dots\}$.
For right-continuous g , we have the representations

$$\rho_g(X) = \int_0^1 \text{V@R}_s(X) dg(s) = \int_{-\infty}^{\infty} x d(g \circ F_X)(x).$$

If specifically $g(t) = \max\{(t - (1 - \alpha))/\alpha; 0\}$, then $\rho_g = \text{AV@R}_\alpha$.

Example 4 Let g be a convex distortion function, i.e. a convex nondecreasing function $g : [0, 1] \rightarrow [0, 1]$ with $g(0) = 0$ and $g(1) = 1$. The **distortion risk measure** associated with g

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Distortion risk measures associated with convex distortion functions are the building blocks of rather general law-invariant coherent risk measures

...

Theorem Let ρ be a law-invariant coherent risk measure on $\mathcal{X} = L^p(\Omega, \mathcal{F}, \mathbb{P})$ for some $p \in [1, \infty]$. Then there is some set \mathcal{G}_ρ of continuous convex distortion functions such that

$$\rho(X) = \sup_{g \in \mathcal{G}_\rho} \rho_g(X) \quad \text{for all } X \in \mathcal{X}$$

“Kusuoka representation”.

Kusuoka (2001)

Föllmer/Schied (2004)

Krätschmer/H. Z. (2011)

Belomestny/Krätschmer (2012)

For law-invariant ρ and $\mathcal{M}(\mathcal{X}) := \{\mathbb{P}_X : X \in \mathcal{X}\}$, we may define a map

$$\mathcal{R}_\rho : \mathcal{M}(\mathcal{X}) \longrightarrow \mathbb{R} \quad \text{by} \quad \mathcal{R}_\rho(\mathfrak{m}) := \rho(X_{\mathfrak{m}}),$$

where $X_{\mathfrak{m}} \in \mathcal{X}$ has law \mathfrak{m} . We call \mathcal{R}_ρ risk functional associated with ρ .

For law-invariant ρ and $\mathbb{F}(\mathcal{X}) := \{F_X : X \in \mathcal{X}\}$, we may define a map

$$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R} \quad \text{by} \quad \mathcal{R}_\rho(F) := \rho(X_F),$$

where $X_F \in \mathcal{X}$ has df F . We call \mathcal{R}_ρ risk functional associated with ρ .

2. Estimation in the context of aggregate risks

Consider a “homogeneous” insurance collective

- ▶ $X_1, \dots, X_n \sim \mu$ i.i.d. individual claims in the next insurance period,
 $\sum_{i=1}^n X_i \sim \mu^{*n}$ total claim in the next insurance period,
individual claim distribution μ be unknown
- ▶ $Y_1, \dots, Y_{u_n} \sim \mu$ i.i.d. indiv. claims in the previous insurance period,
 $u_n/n \rightarrow c \in (0, \infty)$

Goal

- ▶ information on the premium $\rho(\sum_{i=1}^n X_i) = \mathcal{R}_\rho(\mu^{*n})$ for the whole collective

Method

- ▶ choose reasonable estimator $\hat{\mu}_{u_n}^{*n}$ for μ^{*n} based on Y_1, \dots, Y_{u_n}
- ▶ use $\mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})$ as estimator for $\mathcal{R}_\rho(\mu^{*n})$

Questions

1. What is a reasonable estimator $\hat{\mu}_{u_n}^{*n}$ for μ^{*n} ?
2. Simple representation of the (estimated) indiv. premium $\frac{1}{n} \mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})$?
3. What can be said about the relative error $\frac{1}{n} (\mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n}) - \mathcal{R}_\rho(\mu^{*n}))$?

Questions

1. What is a reasonable estimator $\hat{\mu}_{u_n}^{*n}$ for μ^{*n} ?

Answer

The Central Limit Theorem and the Glivenko–Cantelli Theorem suggest

$$\hat{\mu}_{u_n}^{*n} := \mathcal{N}_{n\hat{m}_{u_n}, n\hat{s}_{u_n}^2} \quad \text{and} \quad \hat{\mu}_{u_n}^{*n} := (\hat{\mu}_{u_n})^{*n},$$

respectively, where

\hat{m}_{u_n} = empirical mean of Y_1, \dots, Y_{u_n}

\hat{s}_{u_n} = empirical standard deviation of Y_1, \dots, Y_{u_n}

$\hat{\mu}_{u_n}$ = empirical measure of (uniform distribution on) Y_1, \dots, Y_{u_n} .

Notice that the empirical measure $\hat{\mu}_{u_n}$ can be calculated by the Panjer recursion when $\mu[\{0\}] > 0$.

Questions

2. Simple representation of the (estimated) indiv. premium $\frac{1}{n} \mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})$?

Answer

For every law-invariant coherent risk measure ρ on $\mathcal{X} = L^p(\Omega, \mathcal{F}, \mathbb{P})$ and every $\mu \in \mathcal{M}(\mathcal{X} \cap L^\lambda(\Omega, \mathcal{F}, \mathbb{P}))$ with $\lambda = \lambda(p) > 2$:

$$\begin{aligned}\frac{1}{n} \mathcal{R}_\rho(\mu^{*n}) &= m + \frac{\mathcal{R}_\rho(\mathcal{N}_{0,1})}{\sqrt{n}} s + \mathcal{O}(n^{-1/2-\gamma}) \\ \frac{1}{n} \mathcal{R}_\rho(\mathcal{N}_{n\hat{m}_{u_n}, n\hat{s}_{u_n}^2}) &= \hat{m}_{u_n} + \frac{\mathcal{R}_\rho(\mathcal{N}_{0,1})}{\sqrt{n}} \hat{s}_{u_n} \\ \frac{1}{n} \mathcal{R}_\rho((\hat{\mu}_{u_n})^{*n}) &= \hat{m}_{u_n} + \frac{\mathcal{R}_\rho(\mathcal{N}_{0,1})}{\sqrt{n}} \hat{s}_{u_n} + \mathcal{O}_{\mathbb{P}\text{-a.s.}}(n^{-1/2-\gamma})\end{aligned}$$

Krätschmer/H. Z. (2011)
Lauer/H. Z. (2014)

$$\gamma := \min\{\lambda - 2; 1\}/2$$

Questions

2. Simple representation of the (estimated) indiv. premium $\frac{1}{n} \mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})$?

Answer

For instance, the first identity follows from

$$\begin{aligned}\mathcal{R}_\rho(\mu^{*n}) &= \mathcal{R}_\rho(\mathcal{N}_{nm, ns^2}) + (\mathcal{R}_\rho(\mu^{*n}) - \mathcal{R}_\rho(\mathcal{N}_{nm, ns^2})) \\ &= \rho(nm + \sqrt{ns}Z) + (\rho(\sqrt{ns}Z_n + nm) - \rho(\sqrt{ns}Z + nm)) \\ &= nm + \sqrt{ns}\rho(Z) + \sqrt{ns}(\rho(Z_n) - \rho(Z)) \\ &= nm + \sqrt{ns}\mathcal{R}_\rho(\mathcal{N}_{0,1}) + \sqrt{ns}(\mathcal{R}_\rho(\text{law}\{Z_n\}) - \mathcal{R}_\rho(\mathcal{N}_{0,1}))\end{aligned}$$

(with $Z_n := \frac{1}{\sqrt{ns}} \sum_{i=1}^n (X_i - m)$ and $Z \sim \mathcal{N}_{0,1}$) and

$$\begin{aligned}& \sqrt{ns} |\mathcal{R}_\rho(\text{law}\{Z_n\}) - \mathcal{R}_\rho(\mathcal{N}_{0,1})| \\ & \leq \sqrt{ns} \cdot \text{const}_\rho \cdot \sup_{x \in \mathbb{R}} |F_{Z_n}(x) - \Phi_{0,1}(x)| (1 + |x|^\lambda) \\ & \leq \sqrt{ns} \cdot \text{const}_\rho \cdot \text{const}_\lambda \cdot n^{-\gamma}\end{aligned}$$

Questions

3. What can be said about the relative error $\frac{1}{n}(\mathcal{R}_\rho(\widehat{\mu}_{u_n}^{*n}) - \mathcal{R}_\rho(\mu^{*n}))$?

Answer

For every law-invariant coherent risk measure ρ on $\mathcal{X} = L^p(\Omega, \mathcal{F}, \mathbb{P})$ and every $\mu \in \mathcal{M}(\mathcal{X} \cap L^\lambda(\Omega, \mathcal{F}, \mathbb{P}))$ with $\lambda = \lambda(p) > 2$:

$$\begin{aligned}\frac{1}{n}(\mathcal{R}_\rho(\mathcal{N}_{nm, ns^2}) - \mathcal{R}_\rho(\mu^{*n})) &= \mathcal{O}(n^{-1}), \\ \frac{1}{n}(\mathcal{R}_\rho(\mathcal{N}_{n\widehat{m}_{u_n}, n\widehat{s}_{u_n}^2}) - \mathcal{R}_\rho(\mu^{*n})) &= (\widehat{m}_{u_n} - m) + o_{\mathbb{P}\text{-a.s.}}(n^{-1/2}), \\ \frac{1}{n}(\mathcal{R}_\rho((\widehat{\mu}_{u_n})^{*n}) - \mathcal{R}_\rho(\mu^{*n})) &= (\widehat{m}_{u_n} - m) + o_{\mathbb{P}\text{-a.s.}}(n^{-1/2}).\end{aligned}$$

Krätschmer/H. Z. (2011)

Lauer/H. Z. (2014)

$$\gamma := \min\{\lambda - 2; 1\}/2$$

Questions

3. What can be said about the relative error $\frac{1}{n}(\mathcal{R}_\rho(\widehat{\mu}_{u_n}^{*n}) - \mathcal{R}_\rho(\mu^{*n}))$?

Answer

Note that

$$\begin{aligned} & n^r \left((\widehat{m}_{u_n} - m) + o_{\mathbb{P}\text{-a.s.}}(n^{-1/2}) \right) \\ &= \frac{u_n^r}{n^r} \cdot u_n^r (\widehat{m}_{u_n} - m) + \frac{o_{\mathbb{P}\text{-a.s.}}(n^{-1/2})}{n^{-r}} \\ &\xrightarrow{\text{a.s.}} 0 \quad \text{for all } r < 1/2 \end{aligned}$$

and

$$\begin{aligned} & \sqrt{u_n} \left((\widehat{m}_{u_n} - m) + o_{\mathbb{P}\text{-a.s.}}(n^{-1/2}) \right) \\ &= \sqrt{u_n} (\widehat{m}_{u_n} - m) + \frac{o_{\mathbb{P}\text{-a.s.}}(n^{-1/2})}{u_n^{-1/2}} \\ &\xrightarrow{\text{d}} Z \sim \mathcal{N}_{0,s^2} \end{aligned}$$

Questions

3. What can be said about the relative error $\frac{1}{n}(\mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n}) - \mathcal{R}_\rho(\mu^{*n}))$?

Answer

... in particular,

$$\left[\frac{\mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})}{n} - \frac{\hat{s}_{u_n}}{\sqrt{u_n}} \Phi^{-1}\left(1 - \frac{\alpha}{2}\right), \frac{\mathcal{R}_\rho(\hat{\mu}_{u_n}^{*n})}{n} - \frac{\hat{s}_{u_n}}{\sqrt{u_n}} \Phi^{-1}\left(\frac{\alpha}{2}\right) \right]$$

provides an asymptotic confidence interval for the individual premium $\frac{1}{n}\mathcal{R}_\rho(\mu^{*n})$ at level α for both $\hat{\mu}_{u_n}^{*n} := \mathcal{N}_{n\hat{m}_{u_n}, n\hat{s}_{u_n}^2}$ and $\hat{\mu}_{u_n}^{*n} := (\hat{\mu}_{u_n})^{*n}$.

2. Estimation in the context of time series

Situation

- ▶ X_1, \dots, X_n ($, X_{n+1}, \dots$) strictly stationary observations
- ▶ one-dimensional marginal distribution μ be unknown

Goal

- ▶ information on $\mathcal{R}_\rho(\mu)$

Method

- ▶ $\hat{\mu}_n :=$ empirical measure of X_1, \dots, X_n
- ▶ use $\mathcal{R}_\rho(\hat{\mu}_n)$ as estimator for $\mathcal{R}_\rho(\mu)$

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Questions

1. asymptotic distribution, bootstrap
2. consistency, a.s. rates
3. qualitative robustness

2. Estimation in the context of time series

2.1 Asymptotic distribution, bootstrap

Question

$$\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)) \xrightarrow{d} ?$$

FDM approach

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$
2. determine the “derivative” $\dot{\mathcal{R}}_{\rho,F}(\cdot)$
3. conclude $\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)) \xrightarrow{d} \dot{\mathcal{R}}_{\rho,F}(B^\circ)$

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... in conventional fashion

1. weak convergence in $D_b = D_b(\mathbb{R})$ w.r.t. sup-norm $\|\cdot\|_\infty$

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1. weak convergence in $D_b = D_b(\mathbb{R})$ w.r.t. sup-norm $\|\cdot\|_\infty$
2. Hadamard derivative w.r.t. sup-norm $\|\cdot\|_\infty$

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3. by means of the classical FDM (Gill 1989, etc.)

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2. Hadamard derivative w.r.t. sup-norm $\|\cdot\|_\infty$
3. by means of the classical FDM (Gill 1989, etc.)

... does not cover any law-invariant coherent risk measure ρ

Question

$$\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)) \xrightarrow{d} ?$$

FDM approach ...

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$
2. determine the “derivative” $\dot{\mathcal{R}}_{\rho,F}(\cdot)$
3. conclude $\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)) \xrightarrow{d} \dot{\mathcal{R}}_{\rho,F}(B^\circ)$

... in new fashion ...

1. weak convergence in $D_\phi = D_\phi(\mathbb{R})$ w.r.t. weighted sup-norm $\|\cdot\|_\phi$
2. **quasi-Hadamard** derivative w.r.t. weighted sup-norm $\|\cdot\|_\phi$
3. by means of the **modified FDM** (Beutner/H. Z. 2010)

... covers “every” law-invariant coherent risk measure ρ

Functional Delta-Method (informally)

$$\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F) = \dot{\mathcal{R}}_{\rho,F}(\widehat{F}_n - F) + \text{“rest”}$$

$$\begin{aligned}\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)) &= \dot{\mathcal{R}}_{\rho,F}(\sqrt{n}(\widehat{F}_n - F)) + \sqrt{n} \text{“rest”} \\ &\rightarrow \dot{\mathcal{R}}_{\rho,F}(B^\circ) + 0\end{aligned}$$

$$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}, \quad \mathbb{F}(\mathcal{X}) \subset D_b, \quad \|\cdot\|_\infty \text{ sup-norm on } D_b$$

Functional Delta-Method (informally)

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What kind of "derivative" is expedient?

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$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}$ Hadamard diff.ble at F if:

There is some $\dot{\mathcal{R}}_{\rho,F} \in L(D_b, \mathbb{R})$ such that for each $v \in D_b$

$$\lim_{n \rightarrow \infty} \left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = 0$$

for all

- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$,
- ▶ $(v_n) \subset D_b$ with $\|v - v_n\|_\infty \rightarrow 0$

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- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$, $h_n := n^{-1/2}$
- ▶ $(v_n) \subset D_b$ with $\|v - v_n\|_\infty \rightarrow 0$ $v_n := \sqrt{n}(\widehat{F}_n - F)$, $v = B^\circ$

What kind of “derivative” is expedient?

$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}$ Hadamard diff.ble at F tangentially to C_b if:

There is some $\dot{\mathcal{R}}_{\rho,F} \in L(C_b, \mathbb{R})$ such that for each $v \in C_b$

$$\lim_{n \rightarrow \infty} \left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = 0$$

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- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$, $h_n := n^{-1/2}$
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for all

- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$, But for coherent ρ there
- ▶ $(v_n) \subset D_b$ with $\|v - v_n\|_\infty \rightarrow 0$ are “too many” (v_n) !!!

What kind of “derivative” is expedient?

$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}$ quasi-Hadam. diff.ble at F tangentially to $C_\phi \langle D_\phi \rangle$ if:

There is some $\dot{\mathcal{R}}_{\rho, F} \in L(C_\phi, \mathbb{R})$ such that for each $v \in C_\phi$

$$\lim_{n \rightarrow \infty} \left| \dot{\mathcal{R}}_{\rho, F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = 0$$

for all

- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$,
- ▶ $(v_n) \subset D_\phi$ with $\|v - v_n\|_\phi \rightarrow 0$ Beutner/H. Z. 2010

What kind of “derivative” is expedient?

$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}$ quasi-Hadam. diff.ble at F tangentially to $C_\phi \langle D_\phi \rangle$ if:

There is some continuous $\dot{\mathcal{R}}_{\rho,F} : C_\phi \longrightarrow \mathbb{R}$ such that for each $v \in C_\phi$

$$\lim_{n \rightarrow \infty} \left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = 0$$

for all

- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$,
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What kind of “derivative” is expedient?

$\mathcal{R}_\rho : \mathbb{F}(\mathcal{X}) \longrightarrow \mathbb{R}$ quasi-Hadam. diff.ble at F tangentially to $C_\phi \langle D_\phi \rangle$ if:

There is some continuous $\dot{\mathcal{R}}_{\rho,F} : C_\phi \longrightarrow \mathbb{R}$ such that for each $v \in C_\phi$

$$\lim_{n \rightarrow \infty} \left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = 0$$

for all

- ▶ $(h_n) \subset \mathbb{R}_{++}$ mit $h_n \rightarrow 0$, $h_n := n^{-1/2}$
- ▶ $(v_n) \subset D_\phi$ with $\|v - v_n\|_\phi \rightarrow 0$ $v_n := \sqrt{n}(\hat{F}_n - F)$, $v = B^\circ$

What kind of “derivative” is expedient?

Example

$$\mathcal{R}_\rho(F) := \int x dF(x)$$

▶ $\dot{\mathcal{R}}_{\rho,F}(v) := - \int v(x) dx, \quad v \in C_b$

▶ $(v_n) \subset D_b$

But $\|v - v_n\|_\infty \rightarrow 0$ does *not* imply

$$\left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = \left| \int (v(x) - v_n(x)) dx \right| \rightarrow 0.$$

What kind of “derivative” is expedient?

Example

$$\mathcal{R}_\rho(F) := \int x dF(x)$$

- ▶ $\dot{\mathcal{R}}_{\rho,F}(v) := - \int v(x) dx, \quad v \in C_\phi$
- ▶ $(v_n) \subset D_\phi, \quad \int 1/\phi(x) dx < \infty$

But $\|v - v_n\|_\phi \rightarrow 0$ does ~~not~~ imply

$$\left| \dot{\mathcal{R}}_{\rho,F}(v) - \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} \right| = \left| \int (v(x) - v_n(x)) dx \right| \rightarrow 0.$$

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
2. determine quasi-Hadamard derivative $\dot{\mathcal{R}}_{\rho, F}(\cdot)$ of \mathcal{R}_ρ w.r.t. $\|\cdot\|_\phi$

To do

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Results on 1.

independent	Shorack/Wellner (1986)	
α -mixing	Shao/Yu (1996)	
β -mixing	Arcones/Yu (1994)	ARMA, GARCH, ...
ρ -mixing	Shao/Yu (1996)	
associated	Shao/Yu (1996)	
long-memory	Beutner/Wu/H. Z. (2012) Buchsteiner (2014) <i>(\sqrt{n} has to be replaced by data dependent rate!)</i>	ARFIMA, ...

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
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Results on 1.

The limit in distribution B° is a centered Gaussian process B° with covariance function ...

$$\Gamma(x_0, x_1) = F(x_0 \wedge x_1)(1 - F(x_0 \vee x_1)) \quad \text{for independent data}$$

$$\Gamma(x_0, x_1) = F(x_0 \wedge x_1)(1 - F(x_0 \vee x_1)) \quad \text{for weakly depend. data}$$
$$+ \sum_{i=0}^1 \sum_{k=2}^{\infty} \text{Cov}(\mathbb{1}_{\{X_1 \leq x_i\}}, \mathbb{1}_{\{X_k \leq x_{1-i}\}})$$

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
2. determine quasi-Hadamard derivative $\dot{\mathcal{R}}_{\rho,F}(\cdot)$ of \mathcal{R}_ρ w.r.t. $\|\cdot\|_\phi$

Results on 2.

Quasi-Hadamard differentiability of \mathcal{R}_ρ at F tangentially to $C_\phi \langle D_\phi \rangle$ means that there is some continuous map $\dot{\mathcal{R}}_\rho : C_\phi \rightarrow \mathbb{R}$ such that

$$\lim_{n \rightarrow \infty} \left| \frac{\mathcal{R}_\rho(F + h_n v_n) - \mathcal{R}_\rho(F)}{h_n} - \dot{\mathcal{R}}_{\rho,F}(v) \right| = 0$$

holds for every triplet $(v, (v_n), (h_n)) \in C_\phi \times D_\phi^{\mathbb{N}} \times (0, \infty)^{\mathbb{N}}$ satisfying $\|v_n - v\|_\phi \rightarrow 0$, $(F + h_n v_n) \subset \mathbb{F}(\mathcal{X})$, and $h_n \rightarrow 0$.

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
2. determine quasi-Hadamard derivative $\dot{\mathcal{R}}_{\rho, F}(\cdot)$ of \mathcal{R}_ρ w.r.t. $\|\cdot\|_\phi$

Results on 2.

For every law-invariant coherent risk measure ρ on $\mathcal{X} = L^p(\Omega, \mathcal{F}, \mathbb{P})$ (for some $p \in [1, \infty]$) and every $F \in \mathbb{F}(\mathcal{X})$ satisfying mild conditions, the risk functional \mathcal{R}_ρ is quasi-Hadamard differentiable at F tangentially to $C_\phi \langle D_\phi \rangle$ with

$$\dot{\mathcal{R}}_{\rho, F}(v) = - \sup_{g \in \mathcal{G}_\rho} \int_{F \rightarrow (0)}^{F \leftarrow (1)} g'(F(x)) v(x) dx$$

with \mathcal{G}_ρ from the Kusuoka representation and ϕ depending on ρ and p .

Krätschmer/Schied/H. Z. (2013)

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
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Results on 2.

For every coherent *distortion risk measure* ρ_g on $\mathcal{X} = L^p(\Omega, \mathcal{F}, \mathbb{P})$ (for some $p \in [1, \infty]$) and every $F \in \mathbb{F}(\mathcal{X})$ satisfying mild conditions, the risk functional \mathcal{R}_ρ is quasi-Hadamard differentiable at F tangentially to $C_\phi \langle D_\phi \rangle$ with

$$\dot{\mathcal{R}}_{\rho, F}(v) = - \int_{F \rightarrow (0)}^{F \leftarrow (1)} g'(F(x)) v(x) dx$$

with ϕ depending on ρ_g and p .

Beutner/H. Z. (2010), Krättschmer/Schied/H. Z. (2013)

To do

1. show $\sqrt{n}(\widehat{F}_n - F) \xrightarrow{d} B^\circ$ in D_ϕ w.r.t. $\|\cdot\|_\phi$
2. determine quasi-Hadamard derivative $\dot{\mathcal{R}}_{\rho,F}(\cdot)$ of \mathcal{R}_ρ w.r.t. $\|\cdot\|_\phi$

Then

The limit in distribution of $\sqrt{n}(\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F))$ is

$$\dot{\mathcal{R}}_{\rho,F}(B^\circ) = - \int_{F \rightarrow (0)}^{F \leftarrow (1)} g'(F(x)) B^\circ(x) dx.$$

If B° is a mean zero Gaussian process with covariance function Γ , then

$$\dot{\mathcal{R}}_{\rho,F}(B^\circ) \sim \mathcal{N}_{0,s^2}$$

with

$$s^2 = s^2(F) := \iint g'(F(x_0)) \Gamma(x_0, x_1) g'(F(x_1)) dx_0 dx_1.$$

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Problem: The estimation of the asymptotic variance $s^2(F)$ often works only moderately well or is “impossible”.

Way out: Bootstrapping the asymptotic distribution.

Beutner/H. Z. (2014) extended the *modified* FDM to the bootstrap, implying that for coherent distortion risk measures ρ :

$$\sqrt{n}(\mathcal{R}_\rho(\hat{F}_n) - \mathcal{R}_\rho(F)) \stackrel{d}{\approx} \sqrt{n}(\mathcal{R}_\rho(\hat{F}_n^*) - \mathcal{R}_\rho(\hat{F}_n)) \mid \hat{F}_n, \quad n \rightarrow \infty.$$

Here \hat{F}_n^* is a bootstrapped version of \hat{F}_n according to, for instance, ...

- ... Efron's bootstrap, the Bayesian bootstrap, or the wild bootstrap for independent data.
- ... the blockwise bootstrap for β -mixing data.

Efron's bootstrap

$$\widehat{F}_n^*(x; \omega, \tilde{\omega}) := \frac{1}{n} \sum_{i=1}^n W_{n,i}(\tilde{\omega}) \mathbb{1}_{[X_i(\omega), \infty)}(x),$$

where $(W_{n,1}, \dots, W_{n,n})$ is multinomially distributed according to the parameters n and $p_1 = \dots = p_n = 1/n$.

Informally, \widehat{F}_n^* is the empirical distribution function of $X_{n,1}^*, \dots, X_{n,n}^*$, where $X_{n,1}^*, \dots, X_{n,n}^*$ are drawn (with replacement) from the "urn" $\{X_1, \dots, X_n\}$.

2. Estimation in the context of time series

2.2 Consistency, a.s. rates

“Differentiability” implies “continuity” and even “Lipschitz continuity”. This motivates the fact that it suffices to show

$$1. \|\widehat{F}_n - F\|_\phi \xrightarrow{\text{a.s.}} 0$$

$$2. n^r \|\widehat{F}_n - F\|_\phi \xrightarrow{\text{a.s.}} 0$$

in order to obtain respectively

$$1. |\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)| \xrightarrow{\text{a.s.}} 0$$

$$2. n^r |\mathcal{R}_\rho(\widehat{F}_n) - \mathcal{R}_\rho(F)| \xrightarrow{\text{a.s.}} 0$$

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Respective results:

1. For independent data: Andersen/Giné/Zinn (1988), ...

For α -mixing data: H. Z. (2013)

2. For independent data: Andersen/Giné/Zinn (1988), ...

For α -mixing data, $\phi \equiv 1$: H. Z. (2013), ...

2. Estimation in the context of time series

2.3 Qualitative robustness

Let ρ be a law-invariant coherent risk measure on $\mathcal{X} = L^p = L^p(\Omega, \mathcal{F}, \mathbb{P})$ for some $p \in [1, \infty]$, and equip $\mathcal{M}(\mathcal{X})$ with any metric d which generates the weak topology. For instance, $d = d_{\text{Lévy}}$ or $d = d_{\text{Proh}}$.

- ▶ $\mathbb{P}^\mu := \mu^{\otimes \mathbb{N}}$ for $\mu \in \mathcal{M}(\mathcal{X})$.
- ▶ X_1, X_2, \dots coordinate projections on $\mathbb{R}^{\mathbb{N}}$.
- ▶ $\widehat{\mathcal{R}}_n := \mathcal{R}_\rho(\widehat{\mu}_n)$ with $\widehat{\mu}_n$ the empirical measure of X_1, \dots, X_n .

Definition

The sequence $(\widehat{\mathcal{R}}_n)$ is said to be **robust at** $\mathcal{M}_0 \subset \mathcal{M}(\mathcal{X})$ when for each $\mu_1 \in \mathcal{M}(\mathcal{X})$ and $\varepsilon > 0$ there exists some $\delta > 0$ such that

$$\mu_2 \in \mathcal{M}_0, \quad d(\mu_1, \mu_2) \leq \delta \quad \implies \quad d_{\text{Proh}}(\mathbb{P}_{\widehat{\mathcal{R}}_n}^{\mu_1}, \mathbb{P}_{\widehat{\mathcal{R}}_n}^{\mu_2}) \leq \varepsilon \quad \text{for all } n \in \mathbb{N}.$$

Let ρ be a law-invariant coherent risk measure on $\mathcal{X} = L^p = L^p(\Omega, \mathcal{F}, \mathbb{P})$ for some $p \in [1, \infty]$, and equip $\mathcal{M}(\mathcal{X})$ with any metric d which generates the weak topology. For instance, $d = d_{\text{Lévy}}$ or $d = d_{\text{Proh}}$.

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Hampel's theorem

$$(\widehat{\mathcal{R}}_n) \text{ robust at } \mathcal{M}_0 \quad \iff \quad \mathcal{R}_\rho|_{\mathcal{M}_0} \text{ continuous for weak topology}$$

Recall: p -weak topology

Let $p \in [0, \infty]$. On

$$\mathcal{M}_1^p(\mathbb{R}) := \left\{ \mu \in \mathcal{M}_1(\mathbb{R}) : \int |x|^p \mu(dx) < \infty \right\}$$

we may impose the p -weak topology, that is, the coarsest topology for which all mappings $\mu \mapsto \int f d\mu$, $f \in C_p(\mathbb{R})$, are continuous, where

$$C_p(\mathbb{R}) := \left\{ f \in C(\mathbb{R}) : |f(x)| \leq c(1 + |x|^p) \text{ for some } c \in (0, \infty) \right\}.$$

Notice that

$\mu_n \rightarrow \mu$ p -weakly

$$\iff \int f d\mu_n \rightarrow \int f d\mu \text{ for all } f \in C_p(\mathbb{R})$$

$$\iff \mu_n \rightarrow \mu \text{ weakly and } \int |x|^p \mu_n(dx) \rightarrow \int |x|^p \mu(dx)$$

$$\iff d_{\text{Proh}}(\mu_n, \mu) \rightarrow 0 \text{ and } \int |x|^p \mu_n(dx) \rightarrow \int |x|^p \mu(dx)$$

In original work on monetary risk measures the domain of $\rho : \mathcal{X} \rightarrow \mathbb{R}$ was assumed to be $\mathcal{X} = L^\infty$. In view of

$$L^\infty \subset L^p \subset L^1 \subset L^0 \quad (1 < p < \infty),$$

this is, of course, too restrictive for practical aspects. But:

Filipovic/Svindland (2012):

For every law-invariant coherent risk measure $\rho : L^\infty \rightarrow \mathbb{R}$ there exists a unique ($\|\cdot\|_1$ -lower-semicontinuous) law-invariant coherent risk measure

$$\bar{\rho} : L^1 \rightarrow \mathbb{R} \cup \{+\infty\}$$

such that $\bar{\rho}$ is an extension of ρ .

- ▶ For the **Average Value at Risk** at level $\alpha \in (0, 1)$

$$\text{AV@R}_\alpha(X) := \frac{1}{\alpha} \int_0^\alpha \text{V@R}_s(X) ds$$

the Filipovic–Svindland extension $\overline{\text{AV@R}_\alpha}$ is finite on L^1 .

- ▶ For the **distortion risk measure** associated with continuous convex distortion function g

$$\rho_g(X) := - \int_{-\infty}^0 g(F_X(x)) dx + \int_0^\infty (1 - g(F_X(x))) dx$$

the Filipovic–Svindland extension $\overline{\rho_g}$ is finite on L^p iff

$$\int_0^1 g'_+(t)^{\frac{p}{p-1}} dt < \infty; \quad p > 1,$$
$$g'_+ \text{ is bounded; } \quad p = 1.$$

- ▶ For the **risk measure based on one-sided p th moments** ($a \in [0, 1]$)

$$\rho_{p,a}(X) := \mathbb{E}[X] + a \mathbb{E}[(X - \mathbb{E}[X])^+]^p$$

the Filipovic–Svindland extension $\overline{\rho_{p,a}}$ is finite on L^q iff $q \geq p$.

Theorem (Krätschmer/Schied/H. Z. (2014))

For every law-invariant coherent risk measure $\rho : L^\infty \rightarrow \mathbb{R}$ and $1 \leq p < \infty$ the following four statements are equivalent:

- (i) $\bar{\rho}$ is finite on L^p .
- (ii) $\mathcal{R}_{\bar{\rho}}$ is (finite and) p -weakly continuous on $\mathcal{M}(L^p)$.
- (iii) \mathcal{R}_ρ is p -weakly continuous on $\mathcal{M}(L^\infty)$.
- (iv) ρ is continuous w.r.t. $\|\cdot\|_p$ at $X \equiv 0$.

Consequence

For a law-invariant coherent risk measure ρ with “natural” domain L^p ($p \in (1, \infty]$) the associated risk functional \mathcal{R}_ρ is *not* weakly continuous but only p -weakly continuous.

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In particular

For a law-invariant coherent risk measure ρ with “natural” domain L^p the sequence $(\widehat{\mathcal{R}}_n)$ is robust on a set $\mathcal{M}_0 \subset \mathcal{M}(L^p)$ if and only if the weak topology and the p -weak topology on \mathcal{M}_0 coincide.

Consequence

For a law-invariant coherent risk measure ρ with “natural” domain L^p ($p \in (1, \infty]$) the associated risk functional \mathcal{R}_ρ is *not* weakly continuous but only p -weakly continuous.

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Note

The following statements are equivalent:

- (i) The weak and the p -weak topology on \mathcal{M}_0 coincide.
- (ii) \mathcal{M}_0 is locally uniformly $|\cdot|^p$ -integrating:

For each $\mu_1 \in \mathcal{M}_0$ and $\varepsilon > 0$ there are $\delta > 0$ and $a_0 > 0$ such that

$$\mu_2 \in \mathcal{M}_0, d(\mu_1, \mu_2) \leq \delta \implies \int |x|^p \mathbb{1}_{\{|x|^p \geq a\}} \mu_2(dx) \leq \varepsilon \text{ for } a \geq a_0.$$

Consequence

For a law-invariant coherent risk measure ρ with “natural” domain L^p ($p \in (1, \infty]$) the associated risk functional \mathcal{R}_ρ is *not* weakly continuous but only p -weakly continuous.

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For a law-invariant coherent risk measure ρ with “natural” domain L^p the sequence $(\widehat{\mathcal{R}}_n)$ is robust on a set $\mathcal{M}_0 \subset \mathcal{M}(L^p)$ if and only if the weak topology and the p -weak topology on \mathcal{M}_0 coincide.

That is

For a law-invariant coherent risk measure ρ with “natural” domain L^p :

The smaller p is, the larger are the sets on which the sequence $(\widehat{\mathcal{R}}_n)$ is robust!

Thank you!

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