

On λ -Quantile Dependent Convex Risk Measures

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Outline:

- The class of λ -quantile dependent convex measures of risk:
 - Definition;
 - The λ -quantile Fatou property;
 - The robust representation.
- The λ -quantile law invariant measures of risk:
 - Definition;
 - The robust representation;
 - The λ -quantile uniform preference;
 - Example: the λ -quantile dependent Weighted VaR.

Measures of risk: capital requirements that can be added to a financial position to make it acceptable.

Mathematically, $\rho : \mathcal{X} \rightarrow \mathbb{R} \cup \{\infty\}$. (e.g., $\mathcal{X} = L^p, 1 \leq p \leq \infty$.)

Axioms of convex measures of risk: $\forall X, Y \in \mathcal{X}$,

- Monotonicity: If $X \leq Y$, then $\rho(X) \geq \rho(Y)$.
- Cash invariance: $\rho(X + m) = \rho(X) - m, \forall m \in \mathbb{R}$.
- Convexity: $\rho(\alpha X + (1 - \alpha)Y) \leq \alpha\rho(X) + (1 - \alpha)\rho(Y), \alpha \in [0, 1]$.

ρ is **coherent** if in addition it satisfies

- Positive homogeneity: $\rho(\lambda X) = \lambda\rho(X), \forall \lambda \geq 0, \forall X \in \mathcal{X}$.

Robust Representation of Convex Measures of Risk:

ADEH (1999), Delbaen (2002), Föllmer and Schied (2004), Biagini and Frittelli (2009), Kaina and Rüschendorf (2009):

Notations:

$$L^p := L^p(\Omega, \mathcal{F}, \mathbf{P}), 1 \leq p \leq \infty;$$

$$L^q := L^q(\Omega, \mathcal{F}, \mathbf{P}) \text{ the conjugate space of } L^p;$$

$$\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\} \text{ a proper convex measure of risk;}$$

$$\mathcal{Q}_p := \{\mathbf{Q} \text{ probability measures} : \mathbf{Q} \ll \mathbf{P}, \frac{d\mathbf{Q}}{d\mathbf{P}} \in L^q\}.$$

$$\rho(X) = \sup_{\mathbf{Q} \in \mathcal{Q}_p} (\mathbb{E}_{\mathbf{Q}}[-X] - \rho^*(\mathbf{Q})) \iff \text{lower semicontinuity of } \rho ,$$

$$\rho^*(\mathbf{Q}) = \sup_{X \in L^p} (\mathbb{E}_{\mathbf{Q}}[-X] - \rho(X)), \text{ the Fenchel-Legendre transformation of } \rho.$$

Key: lower semicontinuity of $\rho \Leftrightarrow$ Fatou property.

Literatures on Fatou property:

- *Delbaen (2002)*: For coherent $\rho : L^\infty \rightarrow \mathbb{R}$,

$$\boxed{\text{Fatou(D)}} \quad \forall (X_n) \subset L^\infty \text{ s.t. } |X_n| \leq C,$$

$$X_n \rightarrow X \in L^\infty \mathbf{P}\text{-a.s.} \implies \rho(X) \leq \liminf_{n \rightarrow \infty} \rho(X_n).$$

- *Föllmer and Schied (2004)*: Convex risk measures $\rho : L^\infty \rightarrow \mathbb{R}$.
- *Biagini and Frittelli (2009)*: For proper convex $\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\}$,

$$\boxed{\text{Fatou(BF)}} \quad \forall (X_n) \subset L^p \text{ s.t. } \mathbf{P}\text{-a.s. } |X_n| \leq Y \in L^p,$$

$$X_n \rightarrow X \in L^p \mathbf{P}\text{-a.s.} \implies \rho(X) \leq \liminf_{n \rightarrow \infty} \rho(X_n).$$

Examples of Risk Measures Depending on the Left Tails:

- Value-at-Risk:

$$VaR_\lambda(X) := -q_X^+(\lambda) = q_{-X}^-(1 - \lambda).$$

- Conditional VaR:

$$CVaR_\lambda(X) := \frac{1}{\lambda} \int_0^\lambda VaR_\gamma(X) d\gamma = -\frac{1}{\lambda} \int_0^\lambda q_X^+(t) dt.$$

λ -Quantile Dependent Convex Risk Measures:

Definition 1 For given $\lambda \in (0, 1)$, a convex measure of risk

$$\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\}, \quad 1 \leq p \leq \infty,$$

is λ -quantile dependent, if

$$X \mathbb{I}_{\{X \leq q_X^+(\lambda)\}} = Y \mathbb{I}_{\{Y \leq q_Y^+(\lambda)\}} \quad \mathbf{P}\text{-a.s.} \quad \text{implies} \quad \rho(X) = \rho(Y).$$

Namely, ρ depends on the the random variables up to their λ -quantiles.

Related Fatou property and robust representation?

The λ -quantile Fatou Property:

$$\boxed{\text{Fatou(D)}} \quad \forall (X_n) \subset L^\infty \text{ s.t. } |X_n| \leq C,$$

$$X_n \rightarrow X \in L^\infty \text{ P-a.s.} \implies \rho(X) \leq \liminf_{n \rightarrow \infty} \rho(X_n).$$

$$\boxed{\text{Fatou(BF)}} \quad \forall (X_n) \subset L^p \text{ s.t. P-a.s. } |X_n| \leq Y \in L^p,$$

$$X_n \rightarrow X \in L^p \text{ P-a.s.} \implies \rho(X) \leq \liminf_{n \rightarrow \infty} \rho(X_n).$$

$$\boxed{\lambda\text{-quantile Fatou}} \quad \forall (X_n) \subset L^p \text{ s.t. } q_{X_n}^+(\lambda) \leq c_\lambda, c_\lambda \in \mathbb{R},$$

$$X_n \rightarrow X \in L^p \text{ P-a.s.} \implies \rho(X) \leq \liminf_{n \rightarrow \infty} \rho(X_n).$$

Relations Between the Three Fatou Properties:

- All require the uniform boundedness of (X_n) .
 - **Fatou(D)**: $|X_n| \leq C$ for some $C \in \mathbb{R}$.
 - **Fatou(BF)**: $|X_n| \leq Y$ for some $Y \in L^p$.
 - **λ -quantile Fatou**: $q_{X_n}^+(\lambda) \leq c_\lambda$ for some $c_\lambda \in \mathbb{R}$.
- For $(X_n) \subset L^p$,

$$|X_n| \leq C \implies |X_n| \leq Y \implies q_{X_n}^+(\lambda) \leq c_\lambda.$$

- ρ has λ -quantile Fatou property $\implies \rho$ has Fatou(BF) $\implies \rho$ has Fatou(D).

Is the λ -quantile Fatou property equivalent to the lower semi-continuity as well as to the robust representation of the convex measure of risk?

Yes, if we restrict ourselves to the smaller class of λ -quantile dependent convex risk measures!

Robust Representation of λ -Quantile Dependent Convex Risk

Measures:

Assume $\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\}$ is *proper*, *convex* and *λ -quantile dependent*.

Theorem 2

$$1. \quad \rho(X) = \sup_{\mathbf{Q} \in \mathcal{Q}_p} (\mathbb{E}_{\mathbf{Q}}[-X] - \rho^*(\mathbf{Q})), \quad \rho^* = \sup_{X \in L^p} (\mathbb{E}_{\mathbf{Q}}[-X] - \rho(X)).$$



2. ρ is $\sigma(L^p, L^q)$ lower semicontinuous.



3. ρ is continuous from above.



4. ρ has the λ -quantile Fatou property.

The λ -Quantile Law Invariant Convex Risk Measures:

Assume $\lambda \in (0, 1)$ is given, and

$$\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\}, \quad 1 \leq p \leq \infty,$$

is a convex measure of risk.

ρ is λ -quantile dependent, if

$$X \mathbb{I}_{\{X \leq q_X^+(\lambda)\}} = Y \mathbb{I}_{\{Y \leq q_Y^+(\lambda)\}} \quad \mathbf{P}\text{-a.s.} \quad \text{implies} \quad \rho(X) = \rho(Y).$$

ρ is λ -quantile law invariant, if

$$X \mathbb{I}_{\{X \leq q_X^+(\lambda)\}} \stackrel{\mathcal{D}}{\sim} Y \mathbb{I}_{\{Y \leq q_Y^+(\lambda)\}} \quad \text{implies} \quad \rho(X) = \rho(Y).$$

Robust Representation of The λ -Quantile Law Invariant Convex Risk Measures:

For law invariant convex risk measures: *Kusuoka (2001), Föllmer and Schied (2004), Frittelli and Rosazza Gianin (2005), Jouini, Schachermayer and Touzi (2006).*

Theorem 3 *Suppose $(\Omega, \mathcal{F}, \mathbf{P})$ is atomless, $\rho : L^p \rightarrow \mathbb{R} \cup \{\infty\}$ is proper, convex and λ -quantile law invariant.*

ρ has the λ -quantile Fatou property .

\Updownarrow

$$\rho(X) = \sup_{\mathbf{Q} \in \mathcal{Q}_p} \left(\int_0^\lambda q_X(t) q_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}(t) dt + q_X^+(\lambda) \int_\lambda^1 q_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}(t) dt - \rho^*(\mathbf{Q}) \right),$$

$$\rho^*(\mathbf{Q}) = \sup_{X \in L^p} \left(\int_0^\lambda q_X(t) q_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}(t) dt + q_X^+(\lambda) \int_\lambda^1 q_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}(t) dt - \rho(X) \right).$$

An Example: The λ -quantile Dependent Weighted VaR

Definition 4 (λ -quantile dependent $WVaR$)

$$\rho_{\mu,\lambda} : L^p \rightarrow \mathbb{R} \cup \{\infty\}, \quad 1 \leq p \leq \infty,$$

$$\rho_{\mu,\lambda}(X) := \int_{(0,\lambda]} CVaR_\gamma(X) \mu(d\gamma) = - \int_0^\lambda q_X(t) \phi(t) dt = \int_0^\lambda q_X(t) q_{\nu_\phi}(t) dt.$$

μ : probability measure on $(0, \lambda]$,

$$\phi(t) := \int_{(t,\lambda]} \frac{1}{s} \mu(ds),$$

ν_ϕ : probability distribution measure s.t. $q_{\nu_\phi}(t) := -\phi(t)$, $\forall t \in (0, \lambda]$.

In particular, $\mu(d\gamma) = \mathbb{I}_{\{\lambda\}}(d\gamma) \Rightarrow \phi(t) = \frac{1}{\lambda} \mathbb{I}_{(0,\lambda]}(t)$

$$\Rightarrow \rho_{\mu,\lambda}(X) = CVaR_\lambda(X) = -\frac{1}{\lambda} \int_0^\lambda q_X(t) dt.$$

Robust Representation of λ -quantile Dependent *WVaR*:

$$\rho_{\mu,\lambda}(X) = \max_{\mathbf{Q} \in \mathcal{Q}_\lambda} \mathbb{E}_{\mathbf{Q}}[-X],$$

with $\mathcal{Q}_\lambda := \{\mathbf{Q} \in \mathcal{Q}_p : \nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}} \underset{uni(\lambda)}{\succcurlyeq} \nu_\phi\}$.

Key: $\nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}} \underset{uni(\lambda)}{\succcurlyeq} \nu_\phi$, λ -quantile uniformly preference of $\nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}$ over ν_ϕ .

Note: $\rho_{\mu,\lambda}(X) = \int_0^\lambda q_X(t)q_{\nu_\phi}(t)dt$

- is proper, coherent and λ -quantile law invariant.
- is continuous from above $\Leftrightarrow \rho_{\mu,\lambda}$ has the λ -quantile Fatou property.

**λ -Quantile Uniform Preference of Two Measures μ and ν :
(Second Order Stochastic Dominance)**

Definition 5 $\mu \underset{uni(\lambda)}{\succcurlyeq} \nu$ if

for any “ λ -quantile utility function” u defined by

$$u(x) = u_0(x)\mathbb{I}_{\{x \leq q_\nu(\lambda)\}} + u_0(q_\nu(\lambda))\mathbb{I}_{\{x > q_\nu(\lambda)\}},$$

with u_0 a utility function, we have

$$\int_0^\lambda u d\mu \geq \int_0^\lambda u d\nu.$$

Proposition 6

$$\begin{aligned} \mu \underset{uni(\lambda)}{\succcurlyeq} \nu &\iff \int_0^t q_\mu(s) ds \geq \int_0^t q_\nu(s) ds, \forall t \in [0, \lambda] \\ &\iff \int_0^\lambda h(t) q_\mu(t) dt \geq \int_0^\lambda h(t) q_\nu(t) dt, \forall \text{ decreasing } h : [0, \lambda] \rightarrow \mathbb{R}^+. \end{aligned}$$

Theorem 7 (*Robust Representation of the λ -quantile dependent WVaR : $\rho_{\mu,\lambda}$*)

$$\rho_{\mu,\lambda}(X) = \max_{\mathbf{Q} \in \mathcal{Q}_\lambda} \mathbb{E}_{\mathbf{Q}}[-X],$$

with $\mathcal{Q}_\lambda = \{\mathbf{Q} \in \mathcal{Q}_p, \nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}} \succcurlyeq_{uni(\lambda)} \nu_\phi\}$,

ν_ϕ : probability distribution measure s.t. $q_{\nu_\phi}(t) := -\phi(t), \forall t \in (0, \lambda]$,

$$\phi(t) := \int_{(t,\lambda]} \frac{1}{s} \mu(s).$$

The maximum is obtained by choosing $\mathbf{Q}_X \in \mathcal{Q}_\lambda$ such that $\frac{d\mathbf{Q}_X}{d\mathbf{P}} = f(X)$, where the decreasing function f is given by:

$$f(x) = \begin{cases} \phi(F_X(x)) & \text{if } x \text{ is a continuity point of } F_X, \\ \frac{1}{F_X(x) - F_X(x-)} \int_{F_X(x-)}^{F_X(x)} \phi(t) dt & \text{if } x \text{ is a discrete point of } F_X, \end{cases}$$

for $F_X(x) \leq \lambda$ and $f(x) = 0$, otherwise.

Special Case $CVaR_\lambda$:

$$CVaR_\lambda(X) = \rho_{\mu, \lambda}(X), \quad \mu(d\gamma) = \mathbb{I}_{\{\lambda\}}(d\gamma), \quad q_{\nu_\phi}(t) = -\frac{1}{\lambda} \mathbb{I}_{(0, \lambda]}(t).$$

From Theorem 7,

$$CVaR_\lambda(X) = \max_{\mathbf{Q} \in \mathcal{Q}_\lambda} \mathbb{E}_{\mathbf{Q}}[-X],$$

$$\mathcal{Q}_\lambda = \left\{ \mathbf{Q} \in \mathcal{Q}_p, \quad \nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}} \underset{uni(\lambda)}{\succcurlyeq} \nu_\phi \right\},$$

$$\nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}} \underset{uni(\lambda)}{\succcurlyeq} \nu_\phi \iff \int_0^t q_{\nu_{-\frac{d\mathbf{Q}}{d\mathbf{P}}}}(s) ds \geq \int_0^t q_{\nu_\phi}(s) ds = -\frac{t}{\lambda}$$

The set \mathcal{Q}_λ coincides with the set

$$\left\{ \mathbf{Q} \text{ probability measure} : \mathbf{Q} \ll \mathbf{P}, \quad \frac{d\mathbf{Q}}{d\mathbf{P}} \leq \frac{1}{\lambda} \quad \mathbf{P} - a.s. \right\}.$$

THANK YOU!