# Dual Representation of Quasiconvex Conditional Maps Quasiconvex Dynamic Risk Measures

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Quasiconvex Dynamic Risk Measures

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# On Quasiconvexity (QCO)

• 
$$f: E \to \overline{\mathbb{R}} := \mathbb{R} \cup \{-\infty\} \cup \{\infty\}$$
 is quasiconvex (QCO) if

 $f(\lambda X + (1 - \lambda)Y) \le \max\{f(X), f(Y)\}, \lambda \in [0, 1]$ 

• Equivalently: f is (QCO) if all the lower level sets

 $\{X \in E \mid f(X) \le c\} \quad \forall c \in \mathbb{R}$ 

are convex

- Findings on (QCO) real valued functions go back to De Finetti (1949), Fenchel (1949)...
- On (QCO) real valued functions and their dual representation: J-P Penot 1990 - 2007, Volle 1998, ...

# Dual representation for real valued maps

As a straightforward application of the Hahn-Banach Theorem:

#### Proposition (Volle 98)

Let E be a locally convex topological vector space and E' be its topological dual space. If  $f : E \to \overline{\mathbb{R}} := \mathbb{R} \cup \{-\infty\} \cup \{\infty\}$  is lsc and (QCO) then

 $f(x) = \sup_{x'\in E'} R(x'(x), x'),$ 

where  $R : \mathbb{R} \times E' \to \overline{\mathbb{R}}$  is defined by

 $R(m, x') := \inf \{ f(\xi) \mid \xi \in E \text{ such that } x'(\xi) \ge m \}.$ 

An application of the above result leads to:

# Dual representation of static (QCO) cash-subadditive risk measures

#### Proposition (Cerreia-Maccheroni-Marinacci-Montrucchio, 2009)

A function  $\rho: L^{\infty} \to \overline{\mathbb{R}}$  is quasiconvex cash-subadditive decreasing if and only if

$$\rho(X) = \max_{Q \in ba_{+}(1)} R(E_{Q}[-X], Q),$$
  
 
$$R(m, Q) = \inf \{ \rho(\xi) \mid \xi \in L^{\infty} \text{ and } E_{Q}[-\xi] = m \}$$

where  $R : \mathbb{R} \times ba_+(1) \to \overline{\mathbb{R}}$  and R(m, Q) is the reserve amount required today, under the scenario Q, to cover an expected loss m in the future.

# The conditional setting: let $\mathcal{G}\subseteq \mathcal{F}$

A map

$$\pi: L(\Omega, \mathcal{F}, P) \to L(\Omega, \mathcal{G}, P)$$

is quasiconvex (QCO) if  $\forall X, Y \in L(\Omega, \mathcal{F}, P)$  and for all  $\mathcal{G}$ -measurable r.v.  $\Lambda, 0 \leq \Lambda \leq 1$ ,

 $\pi(\Lambda X + (1 - \Lambda)Y) \leq \pi(X) \vee \pi(Y);$ 

or equivalently if all the lower level sets

$$\mathcal{A}(Y) = \{X \in L(\Omega, \mathcal{F}, P) \mid \pi(X) \leq Y\} \quad \forall Y \in L(\Omega, \mathcal{G}, P)$$

are conditionally convex, i.e. for all  $X_1, X_2 \in \mathcal{A}(Y)$  one has that  $\Lambda X_1 + (1 - \Lambda)X_2 \in \mathcal{A}(Y)$ .

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## The problem

Let  $\mathcal{G} \subseteq \mathcal{F}$  be an arbitrary sub sigma algebra.

Which is the dual representation of a (QCO) conditional map

$$\pi: L(\Omega, \mathcal{F}, P) \to L(\Omega, \mathcal{G}, P)$$
 ?

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 ?

As in the convex case, the dual representation of a (QCO) conditional map turns out to have the same structure of the real valued case,

...but the proof is not a straightforward application of known facts.

# Dynamic (QCO) Risk Measures

- Let  $\Lambda$ ,  $0 \leq \Lambda \leq 1$ , be  $\mathcal{G}$ -measurable random variables
- The convexity of  $\pi: L(\Omega, \mathcal{F}, P) \rightarrow L(\Omega, \mathcal{G}, P)$

$$\pi(\Lambda X + (1 - \Lambda)Y) \leq \Lambda \pi(X) + (1 - \Lambda)\pi(Y)$$

implies:

$$\pi(\Lambda X + (1 - \Lambda)Y) \leq \Lambda \pi(X) + (1 - \Lambda)\pi(Y) \leq \pi(X) \lor \pi(Y).$$

• Quasiconvexity alone:

$$\pi(\Lambda X + (1 - \Lambda)Y) \leq \pi(X) \lor \pi(Y)$$

allows to control the risk of a diversified position.

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# Conditional Certainty Equivalent: CCE [F. Maggis 2010]

Consider a Stochastic Dynamic Utility (SDU)  $u(x, t, \omega)$ 

$$u: \mathbb{R} \times [0,\infty) \times \Omega \to \mathbb{R} \cup \{-\infty\}$$

#### Definition

Let *u* be a SDU and *X* be a  $\mathcal{F}_t$  measurable random variable. For each  $s \in [0, t]$ , the backward Conditional Certainty Equivalent  $C_{s,t}(X)$  of *X* is the  $\mathcal{F}_s$  measurable random variable solution of the equation:

$$u(C_{s,t}(X), s, \omega) = E[u(X, t, \omega)|\mathcal{F}_s].$$

This valuation operator  $C_{s,t}(X) = u^{-1} (E[u(X, t, \omega)|\mathcal{F}_s], s, \omega)$  is the natural generalization to the dynamic and stochastic environment of the classical definition of the certainty equivalent, as given in Pratt 1964. Even if  $u(., t, \omega)$  is concave the CCE is not a concave functional, but it is conditionally quasiconcave.

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- Other applications of real valued quasiconvex maps in finance (static quasiconvex risk measures) can be found in the papers by:
- Cerreia–Voglio, Maccheroni, Marinacci and Montrucchio 2009
- Drapeau and Kupper 2010
  - Dynamic quasiconvex risk measures are studied in:
- F. Maggis 2009 and 2010

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Setting

# Setting for the dual representation

$$\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$$

We now state the assumptions on the spaces of random variables  $L_{\mathcal{F}}$  and  $L_{\mathcal{G}}$  and on the quasiconvex conditional map  $\pi$  in order to obtain the dual representation.

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#### Setting

#### Notations

- $L^p_{\mathcal{F}} := L^p(\Omega, \mathcal{F}, P), \ p \in [0, \infty].$
- L<sub>F</sub> := L(Ω, F, P) ⊆ L<sup>0</sup>(Ω, F, P) is a lattice of F measurable random variables.
- L<sub>G</sub> := L(Ω, G, P) ⊆ L<sup>0</sup>(Ω, G, P) is a lattice of G measurable random variables.
- $L_{\mathcal{F}}^{c} = (L_{\mathcal{F}}, \geq)^{c}$  is the order continuous dual of  $(L_{\mathcal{F}}, \geq)$ , which is also a lattice.

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# Standing assumptions on the spaces

•  $L_{\mathcal{F}}$  (resp.  $L_{\mathcal{G}}$ ) satisfies the property  $1_F$  (resp  $1_G$ ):

$$X \in L_{\mathcal{F}} \text{ and } A \in \mathcal{F} \Longrightarrow (X\mathbf{1}_A) \in L_{\mathcal{F}}.$$
 (1<sub>F</sub>)

#### **2** $(L_{\mathcal{F}}, \sigma(L_{\mathcal{F}}, L_{\mathcal{F}}^{c}))$ is a locally convex TVS.

This condition requires that the order continuous dual  $L_{\mathcal{F}}^c$  is rich enough to separate the points of  $L_{\mathcal{F}}$ .

- $L^{c}_{\mathcal{F}}$  satisfies the property  $1_{F}$

## Examples of spaces satisfying the assumptions

- The  $L^p$  spaces:  $L_{\mathcal{F}} := L^p_{\mathcal{F}}$ , with  $p \in [1, \infty]$ . Then:  $L^c_{\mathcal{F}} = L^q_{\mathcal{F}} \hookrightarrow L^1_{\mathcal{F}}$  (with q = 1 when  $p = \infty$ ).
- The Orlicz spaces  $L_{\mathcal{F}} := L_{\mathcal{F}}^{\Psi}$ , for any Young function  $\Psi$ . Then  $L_{\mathcal{F}}^{c} = L^{\Psi^{*}} \hookrightarrow L_{\mathcal{F}}^{1}$ , where  $\Psi^{*}$  is the conjugate of  $\Psi$ .
- The Morse subspace  $L_{\mathcal{F}} := M^{\Psi}$  for any continuous Young function  $\Psi$ . Then  $L_{\mathcal{F}}^c = L^{\Psi^*} \hookrightarrow L_{\mathcal{F}}^1$ .

## Conditions on $\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$

Let  $X_1, X_2 \in L_F$ 

#### (MON) $X_1 \leq X_2 \Longrightarrow \pi(X_1) \leq \pi(X_2)$

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## Conditions on $\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$

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 $(\tau$ -LSC) the lower level set

$$\mathcal{A}_Y = \{X \in L_{\mathcal{F}} \mid \pi(X) \leq Y\}$$

is  $\tau$  closed for each  $\mathcal{G}$ -measurable Y

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Conditions on 
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is  $\tau$  closed for each  $\mathcal{G}$ -measurable Y

(REG)  $\forall A \in \mathcal{G}, \ \pi(X_1\mathbf{1}_A + X_2\mathbf{1}_A^C) = \pi(X_1)\mathbf{1}_A + \pi(X_2)\mathbf{1}_A^C$ 

# On continuity from below (CFB)

(CFB)  $\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$  is continuous from below if

 $X_n \uparrow X \quad P \text{ a.s.} \quad \Rightarrow \quad \pi(X_n) \uparrow \pi(X) \quad P \text{ a.s.}$ 

Under a very weak assumption on  $\sigma(L_{\mathcal{F}}, L_{\mathcal{F}}^{c})$ , that is satisfied in all cases of interest, we have:

#### Proposition

If  $\pi : L_{\mathcal{F}} \to L_{\mathcal{G}}$  satisfies (MON) and (QCO), then are equivalent: (i)  $\pi$  is  $\sigma(L_{\mathcal{F}}, L_{\mathcal{F}}^{c})$ -(LSC) (ii)  $\pi$  is (CFB) (iii)  $\pi$  is order-(LSC) (i.e. the Fatou property)

Conclusion: in the following results, we may replace the condition  $\sigma(L_F, L_F^c)$ -(LSC) with (CFB).

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# The dual representation of conditional quasiconvex maps

#### Theorem

If  $\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$  is (MON), (QCO), (REG) and  $\sigma(L_{\mathcal{F}}, L_{\mathcal{F}}^c)$ -LSC then

$$\pi(X) = ess \sup_{Q \in L^c_{\mathcal{F}} \cap \mathcal{P}} K(X, Q)$$

#### where

$$\begin{split} \mathcal{K}(X,Q) &:= ess \inf_{\xi \in L_{\mathcal{F}}} \left\{ \pi(\xi) \mid E_Q[\xi|\mathcal{G}] \ge_Q E_Q[X|\mathcal{G}] \right\} \\ \mathcal{P} &=: \left\{ Q << P \text{ and } Q \text{ probability} \right\} \end{split}$$

Exactly the same representation of the real valued case, but with conditional expectations!

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# Q = P on $\mathcal{G}$

#### Corollary

Suppose that the assumptions of the Theorem hold true. If for  $X \in L_F$  there exists  $\eta \in L_F$  and  $\varepsilon > 0$  such that  $\pi(\eta) + \varepsilon < \pi(X)$ , then

$$\pi(X) = ess \sup_{Q \in L^c_{\mathcal{F}} \cap \mathcal{P}_{\mathcal{G}}} K(X, Q),$$

where

$$\mathcal{P}_{\mathcal{G}} =: \left\{ Q \in \mathcal{P} \text{ and } Q = P \text{ on } \mathcal{G} 
ight\}.$$

NOTE: The (weak) additional assumption allows us to replace  $\mathcal{P} =: \{Q \le P \text{ and } Q \text{ probability}\}$  with the same set  $\mathcal{P}_{\mathcal{G}}$  that is used in the convex conditional case.

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## Cash additivity

# • A map $\pi: L_{\mathcal{F}} \to L_{\mathcal{G}}$ is said to be (CAS) cash additive if for all $X \in L_{\mathcal{F}}$ and $\Lambda \in L_{\mathcal{G}}$

$$\pi(X + \Lambda) = \pi(X) + \Lambda.$$

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• Note: (CAS) and (QCO) implies Convexity.

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$$\pi(X + \Lambda) = \pi(X) + \Lambda.$$

- Note: (CAS) and (QCO) implies Convexity.
- Next, we show that we recover the result of Detlefsen Scandolo 05 for convex conditional maps.

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# The conditional convex case

#### Corollary

Suppose that the assumptions of the Theorem hold true. Suppose that for every  $Q \in L^{c}_{\mathcal{F}} \cap \mathcal{P}_{\mathcal{G}}$  and  $\xi \in L_{\mathcal{F}}$  we have  $E_{Q}[\xi|\mathcal{G}] \in L_{\mathcal{F}}$ . If  $\pi : L_{\mathcal{F}} \to L_{\mathcal{G}}$  satisfies in addition (CAS) then

$$K(X,Q) = E_Q[X|\mathcal{G}] - \pi^*(Q)$$

and

$$\pi(X) = ess \sup_{Q \in L^c_{\mathcal{F}} \cap \mathcal{P}_{\mathcal{G}}} \left\{ E_Q[X|\mathcal{G}] - \pi^*(Q) \right\}$$

where

$$\pi^*(Q) = ess \sup_{\xi \in L_F} \left\{ E_Q[\xi|\mathcal{G}] - \pi(\xi) \right\}.$$

# Why the proofs of the real valued case and convex case do not work

 We cannot directly apply Hahn-Banach to π : L<sub>F</sub> → L<sub>G</sub>, as it happened in the real case, since

$$\{\xi \in L_{\mathcal{F}} \mid \pi(\xi) \leq \pi(X) - \varepsilon\}^{C}$$

is not any more convex!

• Scalarization does not work! Convexity is preserved by the map:

$$\pi_0: L_{\mathcal{F}} \to \mathbb{R} \quad \pi_0(X) := E[\pi(X)]$$

but not quasiconvexity!

# Approximation argument

The idea is to approximate  $\pi$  with combinations of quasiconvex real valued functions  $\pi_{A}$ 

$$\pi_{\mathcal{A}}(X) := \mathop{ess\,\, ext{sup}}_{\omega \in \mathcal{A}} \pi(X), \,\, \mathcal{A} \in \mathcal{G}.$$

We consider finite partitions  $\Gamma = \left\{ A^{\Gamma} \right\}$  of  ${\cal G}$  measurable sets  $A^{\Gamma}$  and

$$\pi^{\Gamma}(X) := \sum_{A^{\Gamma} \in \Gamma} \pi_{A^{\Gamma}}(X) \mathbf{1}_{A^{\Gamma}},$$

$$H^{\Gamma}(X) := \sup_{Q \in L^{c}_{\mathcal{F}} \cap \mathcal{P}} \inf_{\xi \in L_{\mathcal{F}}} \left\{ \pi^{\Gamma}(\xi) \mid E_{Q}[\xi|\mathcal{G}] \geq E_{Q}[X|\mathcal{G}] \right\}$$

## Steps of the proof

First we show 
$$H^{\Gamma}(X) = \pi^{\Gamma}(X)$$
.

II Then it is a simple matter to deduce

$$\pi(X) = \inf_{\Gamma} \pi^{\Gamma}(X) = \inf_{\Gamma} H^{\Gamma}(X)$$

III Finally we prove that

$$\inf_{\Gamma} H^{\Gamma}(X) = \inf_{\Gamma} \sup_{Q \in L_{t}^{c} \cap \mathcal{P}} \inf_{\xi \in L_{t}} \left\{ \pi^{\Gamma}(\xi) | E_{Q}[\xi|\mathcal{F}_{s}] \ge E_{Q}[X|\mathcal{F}_{s}] \right\}$$
$$= \sup_{Q \in L_{t}^{c} \cap \mathcal{P}} \inf_{\xi \in L_{t}} \left\{ \pi(\xi) | E_{Q}[\xi|\mathcal{F}_{s}] \ge E_{Q}[X|\mathcal{F}_{s}] \right\}$$

that is based on a uniform approximation result.

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Following [Filipovic, Kupper, Vogelpoth 2009-2010] we may consider maps

$$ho: L^p_{\mathcal{G}}(\mathcal{F}) \to \overline{L}^0(\mathcal{G})$$

where

$$L^p_{\mathcal{G}}(\mathcal{F}) = L^0(\mathcal{G})L^p(\mathcal{F}) = \{YX \mid Y \in L^0(\mathcal{G}), X \in L^p(\mathcal{F})\}$$

is an  $L^0(\mathcal{G})$  normed module.

- We showed that the dual representation of a quasiconvex dynamic risk measure defined on  $L^p_{\mathcal{C}}(\mathcal{F})$  also works in this setting.
- The proof is easier: it is similar to the real valued case, since it uses the conditional Hahn Banach Th., as developed in [FKV09]
- Quasiconvex dynamic risk measures defined on vector spaces  $L^p_{\mathcal{F}}$  or on  $L^0(\mathcal{G})$  normed module  $L^p_{\mathcal{G}}(\mathcal{F})$  are different objects (satisfy different properties) and therefore the results in the two cases are different.

# Thank you for your attention

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