A polarity theory for sets of desirable gambles

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Message of this talk

- Sets of lexicographic probability systems stand to sets of desirable gambles just like sets of probabilities stand to sets of almost desirable gambles
- The key ingredient of the correspondence is polarity (duality) theory for convex sets

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- We assume that the set of outcomes Ω of an experiment (e.g., coin tossing) is finite, say Ω:={1,.., n}.
- A gamble on the experiment is thence a vector g:=(g(1), ..., g(n)) in the real vector space Rⁿ, where g(i) represents the reward the gambler would obtain if i is the actual outcome of the experiment.

Definition: A set $K \subseteq \mathbb{R}^n$ is called a coherent set of almost desirable gambles if it satisfies [linearity] if $f,g \in K$, then $\mu f + \nu g \in K$, for $\mu, \nu > 0$ [accepting partial gain] if $g > 0_n$ then $g \in K$ [avoiding sure loss] $-1_n \notin K$ [closure] if $g+f \in K$, for all $f > 0_n$, then $g \in K$

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Let $\Pi = \{1, ..., m\} \subset \Omega, g \in \mathbb{R}^m$. We define $g \mid^{\Pi c} \in \mathbb{R}^n$ as $g \mid^{\Pi c}(\omega) := \begin{cases} g(\omega), \text{ if } \omega \in \Pi \\ 0, \text{ otherwise.} \end{cases}$

Definition [De Cooman and Quaeghebeur (2012)]: Let $K \subset \mathbb{R}^n$. The conditioned set of K with respect to $\Pi \subset \Omega$ is the set

$$\mathbf{K}|_{\Pi} := \{ g \in \mathbb{R}^m : g | \Pi^{\mathsf{c}} \in \mathbf{K} \}$$

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Definition: Let $M \subset \mathbb{P}_n$ be a credal set. The conditioning of M on $\Pi \subset \Omega$ is the projection on Π of all $q(.|\Pi) \in \mathbb{P}_n$ with $q \in M$, that is

$$\mathbf{M}|_{\Pi} := \{ p \in \mathbf{P}_{\mathrm{m}} : \exists q \in \mathbf{M}, q(.|\Pi) = p|^{\Pi c} \}$$

Central tools:

Classical separation theorem for closed convex sets, and the (positive) polarity operator ()•

If $K \subset \mathbb{R}^n$ is a nonempty <u>closed convex set</u>, then for every $g \notin K$ there exist $v \in \mathbb{R}^n$ (non-null) and $b \in \mathbb{R}$ such that for all $f \in K$

 $v \cdot f \ge b > v \cdot g$

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Given a subset $A \subseteq \mathbb{R}^n$, its *polar* is the closed convex cone

 $A^{\bullet} = \{ g \in \mathbb{R}^{n} : g \cdot f \ge 0, \text{ for every } f \in A \}$

Reformulating the separation theorem:

If $K \subset \mathbb{R}^n$ is a nonempty <u>closed convex cone</u>, then for every $g \notin K$ there exists $v \in \mathbb{R}^n$ (non-null) such that $K \subset \{v\}^\bullet$ but $g \notin \{v\}^\bullet$

Hence: *K* is a *closed convex cone* iff it is the polar of some set $A \subset \mathbb{R}^n$ (i.e. $K = A^{\bullet}$)

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Hence:

M is a *credal set* iff it is the intersection of the set \mathbb{P}_n of all pmf over Ω with the polar of some set.

















Theorem: The map $\mathscr{C}: K \longmapsto K^{\bullet} \cap \mathbb{P}_n$ is an isomorphism between the collection of coherent sets of almost desirable gambles equipped with the conditioning operation and the collection of credal sets equipped with the conditioning operation.



Definition: A set $K \subseteq \mathbb{R}^n$ is called a coherent set of almost desirable gambles if it satisfies [*linearity*] if $f,g \in K$, then $\mu f + \nu g \in K$, for $\mu, \nu > 0$ [accepting partial gain] if $g > O_n$ then $g \in K$ [avoiding sure loss] $-1_n \notin K$ [*closure*] if $g+f \in K$, for all $f > O_n$, then $g \in K$

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• a coherent set of desirable gambles is a *convex cone omitting its apex and containing all positive vectors.*

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Lexicographic polarity



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Central tools:

Lexicographic separation theorem for convex sets, and the (lexicographic) polarity operator ()

Theorem [Martinez-Legaz (1983)]: If $K \subset \mathbb{R}^n$ is a nonempty convex set, then for every $g \notin K$ there exists $A \in M_n$ (even $A \in O_n$) and $b \in \mathbb{R}^n$ such that for all $f \in K$

 $A(f) >_{\mathrm{L}} b \ge_{\mathrm{L}} A(g).$

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Given $K \subseteq \mathbb{R}^n$, its *L*-polar is the set

$$\mathbf{K}^{\bullet} = \{ \mathbf{A} \in \mathbb{M}_{n} : \mathbf{A}(f) >_{\mathbf{L}} \mathcal{O}, \forall f \in K \}$$

We call a set $M \subseteq M_n$ a *L*-convex cone if it is the L-polar of some (i.e. $M = K^{\bullet}$)

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Given $K \subseteq \mathbb{R}^n$, its *L*-polar is the set

$$K^{\bullet} = \{ A \in \mathbb{M}_n : A(f) >_L o, \forall f \in K \}$$

We call a set $M \subseteq M_n$ a *L-credal set* if it is the intersection of the set \mathbb{T}_n of all stochastic matrices of full rank with some L-convex cone.

Central tools:

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Given $M \subseteq M_n$ we also define the convex cone omitting its apex

$$\mathbf{M}^{\triangle} = \{ g \in \mathbb{R}^{n} : \mathbf{A}(g) >_{\mathbf{L}} \mathbf{O}, \forall \mathbf{A} \in \mathbf{M} \}$$

Proposition: $K \subseteq \mathbb{R}^n$ is a convex cone omitting its apex if and only if $K = (K^{\checkmark})^{\triangle}$

Reformulating the lexicographic separation theorem:

Proposition: Let $K \in \mathbb{D}_n$ and $g \notin K$, then there exists a matrix $A \in \mathbb{O}_n$ with $A >_L O$ such that $K \subset \{A\}^{\triangle}$ but $g \notin \{A\}^{\triangle}$.

If K is maximal, A is unique.

Where $A >_L 0$ means that each column *a* of A is such that $a >_L 0$.

Theorem: The map $K \mapsto \mathscr{G}(K) := K^{\bullet} \cap \mathbb{T}_n$ is a bijection between coherent sets of desirable gambles and Lcredal sets whose inverse is $(...)^{\triangle}$. The definition of conditioning for stochastic matrices is a variation of the definition by Blume et al. (1991) of conditioning for lexicographic systems.

	(ω_1)	(ω_2)	(ω_3)	
P =	0.5	0.25	0.25	we condition with
	0	0.5	0.5	respect to $\Pi = \{\omega_2, \omega_3\}$
	0.25	0.5	0.25	

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1. for each row *p*, we take $p(.|\Pi)$ if defined, else 0

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 $\begin{pmatrix} \omega_1 \end{pmatrix} & (\omega_2) & (\omega_3) \\ P_2 = \begin{pmatrix} 0 & 0.5 & 0.5 \\ 0 & 0.67 & 0.33 \end{pmatrix}$ we condition with respect to $\Pi = \{\omega_2, \omega_3\}$

2. we discard each row which is a linear combination of the rows preceding it

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3. we take the projections over Π

Theorem: The map $K \mapsto \mathscr{G}(K) := K^{\bullet} \cap \mathbb{T}_n$ is an isomorphism between the collection of all coherent sets of desirable gambles equipped with the conditioning operation and the collection of all Lcredal sets equipped with the corresponding conditioning operation.



What next

- Complete the analysis by considering e.g. marginalisation and independence
- Geometrical properties of L-credal sets
- Obtain similar correspondences when the sample space is infinite
- Describe correspondence within Category Theory, hence subsuming both classical and quantum probabilistic cases [joint work w/ F. Zanasi] (see Alessio Benavoli's invited talk for the latter case and the role of polarity/duality for deriving QM)

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