

# Some Remarks on Sets of Lexicographic Probabilities and Sets of Desirable Gambles

Fabio G. Cozman  
Universidade de São Paulo – Brazil

## Goals:

- to examine the equivalence between sets of lexicographic probabilities and sets of desirable gambles;
- to study their relationship with full conditional probabilities, with non-uniqueness, with convexity, with independence.

## Two equivalent languages

- Preference  $\succ$ : strict partial order, with admissibility, “independence” axiom: set of desirable gambles  $\mathcal{D}$ .
- $\succ / \mathcal{D}$  is equivalent to set of lexicographic probabilities (Seidenfeld et al. 1989, with some additional work):

$$f \succ g \Leftrightarrow \forall \begin{bmatrix} \mathbb{P}_1 \\ \dots \\ \mathbb{P}_K \end{bmatrix} : \begin{bmatrix} \mathbb{E}_{\mathbb{P}_1}[f] \\ \dots \\ \mathbb{E}_{\mathbb{P}_K}[f] \end{bmatrix} \succ_L \begin{bmatrix} \mathbb{E}_{\mathbb{P}_1}[g] \\ \dots \\ \mathbb{E}_{\mathbb{P}_K}[g] \end{bmatrix}.$$

### Examples:

	H	T	S	B		H	T
layer 0	$\alpha$	$(1-\alpha)$	0	0	layer 0	$\alpha$	$(1-\alpha)$
layer 1	0	0	$\beta$	$(1-\beta)$	layer 1	$\gamma$	$(1-\gamma)$

- Marginalization: do it by layers; do it by cylindrical extension.
- Conditioning: do it by layers; define  $\mathcal{D}|A = \{f : Af \in \mathcal{D}\}$ .

## Full conditional probabilities

- Full conditional probabilities also allow “zero” conditioning events. These probabilities can be represented through layers.

- $\mathbb{P}(\cdot|A)$  is a probability measure, and  $\mathbb{P}(A|A) = 1$ , for each nonempty  $A$ ;
- $\mathbb{P}(A \cap B|C) = \mathbb{P}(A|B \cap C)\mathbb{P}(B|C)$  whenever  $B \cap C \neq \emptyset$ .

### But consider *admissibility*:

$$f(\omega) \geq g(\omega), \text{ and some } f(\omega) > g(\omega), \text{ then } f \succ g.$$

- Lexicographic probabilities satisfy admissibility, but full conditional probabilities fail admissibility (first example above).

### Another example:

	Y = 0	Y = 1	Y = 2
X = 0	$(1/5)_0$	$(1/5)_0$	$(1/5)_0$
X = 1	$(1/5)_0$	$(1/4)_1$	$(1/4)_1$
X = 2	$(1/5)_0$	$(1/4)_1$	$(1/4)_1$

## Convexity?

- Partial preferences (set of desirable gambles) can be represented by a (unique maximal convex) set of lexicographic probabilities.

### Consider:

$\mathbb{P}_1(\omega_i)$	$\omega_1$	$\omega_2$	$\omega_3$
	$(\alpha)_0$	$(1-\alpha)_0$	$1_1$
	$(\gamma)_2$	$(1-\gamma)_2$	

$\mathbb{P}_2(\omega_i)$	$\omega_1$	$\omega_2$	$\omega_3$
	$(1)_0$	$(\beta)_1$	$(1-\beta)_1$

Their half-half convex combination is:

$\mathbb{P}_{1/2}(\omega_i)$	$\omega_1$	$\omega_2$	$\omega_3$
	$(1+\alpha/2)_0$	$((1-\alpha)/2)_0$	$(1-\beta/2)_1$
	$(\gamma/2)_2$	$((1-\gamma)/2)_2$	

## Conclusion

- Sets of lexicographic probabilities and sets of desirable gambles represent the same objects (not really full conditional probabilities!).
- Convexity and independence require some thought; non-uniqueness is everywhere (perhaps good, but is it?).

## What to do about convexity?

- Use representation as set of *total* orders (cumbersome!).
- Normalize after convex combination (why?).
- Forget normalization; work with linear utilities all the way (?).  
Question: is this a problem for sets of desirable gambles?

## Non-uniqueness, deep down

### Marginal and conditional:

		Y = 0	Y = 1	Y = 2
X = 0	$(1/2)_0$	$(1/2)_0$	$(1/2)_1$	$(1/2)_1$
X = 1	$(1/2)_0$	$(1/2)_1$	$(1/2)_1$	$(1/2)_0$
X = 2	$(1/2)_0$	$1_1$		$(1/2)_0$

### How to combine them? Many possibilities!

Two examples:

	Y = 0	Y = 1	Y = 2
X = 0	$(1/4)_0$	$(1/4)_0$	$(1/4)_1$
X = 1	$(1/4)_1$	$(1/4)_0$	$(1/4)_0$
X = 2	$(1/4)_3$	$(1/2)_3$	$(1/4)_2$

	Y = 0	Y = 1	Y = 2
X = 0	$(1/4)_{0,1}$	$(1/4)_{0,3}$	$(1/4)_{2,3}$
X = 1	$(1/4)_1$	$(1/4)_{0,7}$	$(1/4)_{0,2}$
X = 2	$(1/4)_4$	$(1/2)_5$	$(1/4)_{4,7}$
	$(1/4)_6$	$(1/2)_7$	$(1/4)_6$

- Message: once we move to lexicographic probabilities, we should move to *sets* of them, from the outset!

### ... but do we really want all this flexibility?

- Desirable gambles: it does not really matter, so YES.
- Lexicographic probabilities: ?? Note: marginalization may erase layers, so how to recover the “depth”?

## Independence

- No “factorization” here. Possible definitions:

- $[f_1(X) \succ_{\{Y=y_1\}} f_2(X)] \Leftrightarrow [f_1(X) \succ_{\{Y=y_2\}} f_2(X)]$ , and vice-versa (Blume et al. 1991).

- $[f_1(X) \succ_{B(Y)} f_2(X)] \Leftrightarrow [f_1(X) \succ f_2(X)]$ , and vice-versa (h-independence).

- The former fails Weak Union, the latter fails Contraction; also, uniqueness is lost completely.

### Also, consider:

	Y = 0	Y = 1
X = 0	$(1)_0$	$(1)_2$
X = 1	$(1)_1$	$(1)_4$

Should X and Y be independent?

- How to produce this? Does it concern desirable gambles at all?

	W = 0, Y = 0	W = 1, Y = 0	W = 0, Y = 1	W = 1, Y = 1
X = 0	$(1/2)_0$	$(1/2)_0$	$(1)_2$	$(1)_3$
X = 1	$(1/2)_1$	$(1/2)_1$	$(1/2)_4$	$(1/2)_4$

Marginal (?):

	Y = 0	Y = 1
X = 0	$(1)_0$	$(1)_2$
X = 1	$(1)_1$	$(1)_4$