

### **Probability Weighting in Recursive Evaluation of Two-Stage Prospects**

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## Probability Weighting in Recursive Evaluation of Two-Stage Prospects

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 $24-25$  September 2014 JJC - Dpt SAE2 - INRA- Nancy  $-1$ 

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# Why investigating the evaluation of Two-stage prospects?

## <span id="page-4-0"></span>Two-stage prospect and reduction of compound lotteries

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Table 1: compound risk and its reduced one-stage lottery

 $\Box$  Rational Decision makers (DMs) reduce compound risks, represented by compound lotteries, into single stage lotteries by using the Reduction of compound prospects axiom (RCP).

⇒ Rational DMs should exhibit a perfect neutrality toward compound risk.

However...

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- 1. Reduction of compound prospects have been descriptively challenged in many empirical investigations:
	- $\Box$  Bar Hillel (1973), Bernasconi & Loomes (1992), Budescu & Fisher (2001), Abdellaoui, Klibanoff & Placido (2014), Nebout & Dubois (2014)...
- 2. Following Becker & Brownson (1964) and Yates and Zukowski (1976), Segal (1987, 1990) represented ambiguous bets as two-stage prospects.
- 3. Prospect Theory (PT) is the most succesful descriptive model of decision making under risk and ambiguity.

 $\Rightarrow$  Would it still be the case when dealing with attitudes toward two-stage prospects?

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# Theoretical and Empirical Background

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- $\Box$   $(x, p; y)$  denotes the one-stage prospect resulting in outcome x with probability p and in outcome y with probability  $1 - p$  with  $x \geq y \geq 0.$ 
	- Probability  $p$  is generated using a known urn containing 100 balls numbered from 1 to 100, i.e. drawing a ball which has a number between 1 and  $p \times 100$ .
- $\Box$   $(x, E_p; y)$  denotes the corresponding ambiguous prospect. The probability  $P(E_p)$  is unknown to the DM.
	- We use an unknown urn containing 100 balls numbered from 1 to 100 in unknown proportions, i.e. drawing a ball which has a number between 1 and  $p \times 100$ . Symmetry arguments imply  $P(E_p) = p$ . (Chew & Sagi, 2006, 2008)



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 $\Box$  Under Expected Utility (EU), prospects are evaluated as follows:  $EU(x, p; y) = pu(x) + (1 - p)u(y)$ 

– Where  $u$  is the utility function (and a risk attitude index). – Violations of EU popularized by Kahneman & Tversky.

 $\Box$  Under Prospect Theory (PT), prospects are evaluated as follows in the gain domain:

$$
PT(x, p; y) = w(p)u(x) + (1 - w(p))u(y)
$$

- $u$  is the utility function.
- w is the probability weighting function. w is strictly increasing and satisfies  $w(0) = 0$  and  $w(1) = 1$ .

 $\Rightarrow$  Many experimental evidence on RDU under risk and ambiguity.

## <span id="page-10-0"></span>Probability weighting under risk and ambiguity



 $\Rightarrow$  Ambiguity increases likelihood insensitivity.

### <span id="page-11-0"></span>How to evaluate two-stage prospects?



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□ Traditional Recursive Expected Utility (TREU):

$$
\Rightarrow \frac{1}{2} \times EU(P) + \frac{1}{2} \times EU(Q)
$$

### How to evaluate two-stage prospects?



$$
\Rightarrow \frac{1}{2} \times \phi \left[ EU(P) \right] + \frac{1}{2} \times \phi \left[ EU(Q) \right]
$$

- Kreps  $&$  Porteus (1978) introduced this transformed EU functionnal to account for delayed resolution of uncertainty.
- Klibanof  $\&$  al. (2005) used the same preference functional to model ambiguity. (Seo, 2009, Ergin & Gul (2008),...)

### How to evaluate two-stage prospects?



$$
\Rightarrow \pi_1 \times \phi \left[ PT(P) \right] + \pi_2 \times \phi \left[ PT(Q) \right]
$$

- Segal (1987) suggested this recursive form of RDU to model ambiguity attitudes through second-order probabilities.
- Abdellaoui & Zank (2014) provide the first axiomatization of this general form of Prospect Theory.

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# Experimental Design

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 $\Rightarrow$  We can compare our results to these benchmark studies.

#### CE1 Id45 (17)





Alternative B

 $\overline{\Box}$ 

 $\overline{\mathbb{C}}$ 

 $50\in$ 

 $\cal G$ 

#### CE1 Id45 (13)

### Alternative A









n° 1 à 100 vous gagnez : 18 €

<span id="page-19-0"></span>

 $\Box$  2 first stage probability levels: 1/3 and 2/3.

 $\Rightarrow$  Elicitation, comparison and test of 4 second stage probability weighting functions.



#### Alternative A

(Deux tirages)

Alternative B

(Un seul tirage)



**Example 100 boules** 

Exemples

 $\mathbb{C}$  .

#### Alternative A

(Deux tirages)

Alternative B (Un seul tirage)

Tirage Preliminaire  $\begin{matrix} \left(\bullet\right)\\ \bullet \end{matrix}$ Rien n'est gagné  $\bullet$  $\boxdot$ <br/> $\boxdot$  $\boxed{\vdots}$  $\bigodot$ Vous tirez une boule dans l'urne  $\overline{?}$  $\overline{\mathbf{r}}$  $\overline{\mathcal{C}}$  $\overline{2}$  $\overline{\phantom{a}}$  $\overline{\mathcal{L}}$  $\overline{\mathcal{L}}$  $\overline{2}$  $\overline{2}$  $\sqrt{2}$  $\overline{?}$  $\overline{\phantom{a}}$  $\overline{?}$  $2222$  $\boldsymbol{\eta}$  $121212$  $\gamma$ ? (? (? (? (? (? ) ?  $? ? ? ?$ 50 € n° 1 à 50 vous gagnez :  $0 \in$ sinon



Exemples

![](_page_21_Picture_95.jpeg)

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# Model Specification

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Using the equivalence revealed by the elicitation of the matching probability r:

 $((\bar{x}, p), q) \sim (\bar{x}, m)$ 

we infer the following equalities:

1. Under TREU, we have:

 $q \times p = m$ 

2. Under REU, we have:

 $q \times \phi(p) = \phi(m)$ 

Where  $\phi$  is a transformation function.

Parametric specification:  $\phi(x) = x^{1/\theta}$ .

<span id="page-24-0"></span>![](_page_24_Picture_313.jpeg)

### e have:

![](_page_24_Picture_314.jpeg)

### $\quad \Box \quad$  Parametric specifications:

$$
u(x) = x^{\alpha}
$$
 and  $w(p) = exp(-(-\ln(p)^{\gamma})^{\delta}$ .

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# Results

## <span id="page-26-0"></span>Results: RCP and TREU

![](_page_26_Picture_214.jpeg)

Table 2: RCP  $(\Delta = m - pq)/pq)$ 

- $\Box$  RCP is globally violated, thus TREU is not descriptively valid for evaluating two-stage prospects.
- $\Box$  Overall, we observe preference for the compound prospect, especially for  $q = 1/3$ .

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<span id="page-27-0"></span>![](_page_27_Picture_159.jpeg)

Table 3: Parameter  $\theta$  empirical distribution characteristics under REU

 $\Box$   $\phi$  is convex for  $q = 1/3$  and linear for  $q = 2/3$ .

 $\Box$  The transformation function  $\varphi$  in REU can not absorb the observed discrepancies from RCP.

<span id="page-28-0"></span>![](_page_28_Picture_232.jpeg)

 $\Box$  Function,  $w^*$ , depends on probability q. While it is close to linearity for  $q = 1/3$ , it is convex for  $q = 2/3$ .

 $\Rightarrow$  Inverse than for REU but same problem (differences both for elevation and curvature between  $w^*_{1/3}$  and  $w^*_{2/3}$ ).

### Results: RPT under risk

![](_page_29_Figure_1.jpeg)

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 Adding ambiguity (first and second stage) does not change our main results i.e.

- Impact of probability q on the shape of function  $\phi$ .
- Stage dependent pwf.
- Dependence of the second stage pwf on the first stage probability.
- $\Box$  Benchmark results are found for the single stage pwf under risk and ambiguity.
- $\Box$  No association between RCP and ambiguity attitudes ( $\neq$  from Halevy, 2008 and Segal).

![](_page_31_Figure_1.jpeg)

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1. Recursive evaluation of two-stage prospects is more complex than allowed by any existing recursive model.

2. In Kreps and Porteus (or KMM) integral, function  $\phi$  is sensitive to the first-stage probability of winning.

3. Second-stage probability weighting is very sensitive to the first-stage winning probability.

 $\Rightarrow$  Abdellaoui & Zank (2014) first axiomatized a RPT model that could account for our experimental findings.

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Thank you for your attention!

![](_page_34_Picture_2.jpeg)

![](_page_34_Picture_3.jpeg)

Alternative B

 $\Box$ 

 $\overline{\Gamma}$ 

 $\mathcal{C}$ 

 $\mathcal{C}$ 

 $\cap$ 

 $\subset$ 

90 boules

100 boules

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Fox & Tversky (1995) introduced the comparative ignorance hypothesis as a condition of observability of ambiguity aversion. For this hypothesis they proposed the following conjecture :

"When evaluating an uncertain event in isolation, people attempt to assess its likelihood – as a good bayesian would – paying relatively little attention to second-order characteristics such as vagueness or weight of evidence. However, when people compare two events about which they have different levels of knowledge, the contrast makes the less familiar bet less attractive or the more familiar bet more attractive". p 588.

 $\Rightarrow$  Many experimental tests of this hypothesis. Non neutrality toward ambiguity is always observed! (Fox & Tversky (1995), Chow & Sarin (2001), Rubaltelli & al. (2010))

 $\Rightarrow$  Complete analysis in Nebout (2011).