

# Decision under risk and uncertainty

(A ridiculously sketchy introduction)

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# Decision: typology

## Decision “under certainty”

- $A$ : set of alternatives (possible decisions)
- $X$ : set of consequences
- $c(a) \in X$ : consequence of implementing  $a \in A$

## Problem

- help someone compare alternatives in  $A$  on the basis of their consequences

## Classic problems

- $|A|$  “large”: combinatorial optimization, mathematical programming
- $x \in X$  such that  $x = (x_1, x_2, \dots, x_m)$ : multicriteria problems

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## Problem

- *la décision ne dispose que pour l'avenir* (cf. art. 2 of the French Civil Code:)
  - $c(a)$  is not known with certainty

## Decision under risk

- $c(a)$  is a probability distribution on  $X$

## Decision under uncertainty

- $c(a)$  is known conditionally upon the occurrence of a number of “scenarios”

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# Plan

- 1 Introduction
- 2 Decision under risk
- 3 Decision under uncertainty
- 4 Extensions



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## 1 Introduction

- Model
- Dominance
- Classic criteria
- Max Min
- Max Max
- Hurwicz
- Savage
- Laplace

## 2 Decision under risk

## 3 Decision under uncertainty

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# Decision under uncertainty

## Context

- impossibility to determine with certainty the consequences of implementing an alternative
- no probability
- **Nature** decide of everything that is not under my control
- the consequences of my decisions depend upon my decisions and Nature's decisions ("states of Nature" or "scenarios")
- Nature **does not care**: dropping a slice of bread on the floor (the "tartine beurrée" experiment)

## Problem

- you must choose an alternative **before** knowing Nature's decision

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# Model

## Model

- $A$ : set of alternatives. An element  $a \in A$  is an alternative that can be implemented
- $E$ : set of states of Nature. An element  $e \in E$  is a decision that Nature can take and that can influence the consequences of at least one alternative in  $A$
- $X$ : set of consequences
- $c$ : mapping from  $A \times E$  to  $X$

# Decision table (finite case: $m$ alternatives, $n$ states)

## Decision table

$c$	$e_1$	$e_2$	$\dots$	$e_i$	$\dots$	$e_n$
$a_1$	$c(a_1, e_1)$	$c(a_1, e_2)$	$\dots$	$c(a_1, e_i)$	$\dots$	$c(a_1, e_n)$
$a_2$	$c(a_2, e_1)$	$c(a_2, e_2)$	$\dots$	$c(a_2, e_i)$	$\dots$	$c(a_2, e_n)$
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$
$a_j$	$c(a_j, e_1)$	$c(a_j, e_2)$	$\dots$	$c(a_j, e_i)$	$\dots$	$c(a_j, e_n)$
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$	$\ddots$	$\vdots$
$a_m$	$c(a_m, e_1)$	$c(a_m, e_2)$	$\dots$	$c(a_m, e_i)$	$\dots$	$c(a_m, e_n)$

## Remark

- obtaining such a “decision table” is a **huge** work in practice



# Exemple: the omelette

## The omelette

$A = \{\text{Bowl, Thrash, Aux. Bowl}\}$

$E = \{\text{Good, Bad}\}$

$c$	Good	Bad
Bowl	O. of 6	No O.
Thrash	O. of 5	O. of 5
Aux. Bowl	O. of 6 + Bowl to wash	O. of 5 + Bowl to wash

## Remarks

- no probabilities
- tastes & beliefs
- possibility to acquire additional information (experimentation)

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# Examples

## Bank

	Default	$\overline{\text{Default}}$
Accept	...	...
Refuse	...	...
Accept with guarantees	...	...

## New product

	Success	$\overline{\text{Success}}$
Launch	...	...
$\overline{\text{Launch}}$	...	...

# Example

## Example

$X = \mathbb{R}$ , preference increases with the numbers (€)

$c$	$e_1$	$e_2$	$e_3$
$a_1$	40	70	-20
$a_2$	-10	40	100
$a_3$	20	40	-5

## Classic criteria

- no information about the likelihood of the states of Nature
- no particular model for tastes

# Example

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## Classic criteria

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# Dominance

## Definition

$a \in A$  (strictly) dominates  $b \in A$  ( $a D b$ ) if:

- $c(a, e) \geq c(b, e), \forall e \in E$ ,
- $\exists e \in E$  such that  $c(a, e) > c(b, e)$

## Remark

$D$  is a transitive and asymmetric binary relation

## Definition

$a \in A$  is **efficient** if it is not dominated by another alternative in  $A$ . When  $A$  and  $E$  are finite, the set of efficient alternatives  $A^* \subseteq A$  defined by:

$$A^* = \{a \in A : \text{Not}[b D a], \forall b \in A\}$$

is always nonempty

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# Dominance

## Remarks

- $a D b \Rightarrow a \succ b$ , whatever the likelihood of the states of Nature
- in real-world problems:  $A^* = A$
- limiting attention to  $A^*$  might not be adequate, e.g., if there are doubts on the feasibility of some alternatives in  $A$ . The set  $A^*$  might not contain “close contenders”
- same problems as in MCDA/MCDM

## Example

$c$	$e_1$	$e_2$	$e_3$	$e_4$	$e_5$	$e_6$	$\dots$	$e_{100}$
$a$	100	100	100	100	100	100	$\dots$	100
$b$	99	99	99	99	99	99	$\dots$	99
$c$	100.5	0	0	0	0	0	$\dots$	0
$d$	0	100.5	0	0	0	0	$\dots$	0

- $A = \{a, b, c, d\}$
- $A^* = \{a, c, d\}$  because  $a D b$
- $b$  is a “close contender”

## Example

$c$	$e_1$	$e_2$	$e_3$	$e_4$	$e_5$	$e_6$	$\dots$	$e_{100}$
$a$	100	100	100	100	100	100	$\dots$	100
$b$	99	99	99	99	99	99	$\dots$	99
$c$	100.5	0	0	0	0	0	$\dots$	0
$d$	0	100.5	0	0	0	0	$\dots$	0

- $A = \{a, b, c, d\}$
- $A^* = \{a, c, d\}$  because  $a D b$
- $b$  is a “close contender”

» go faster

## Remark

Every alternative that is solution of problem  $(P)$

$$\max_{a \in A} \sum_{e \in E} p(e) c(a, e)$$

s.t.

$$\sum_{e \in E} p(e) = 1$$

$$p(e) > 0, e \in E$$

 $(P)$ 

is efficient

Suppose that  $a$  is solution of  $(P)$  and that  $a$  is not efficient. Since  $c(b, e) \geq c(a, e)$ ,  $\forall e \in E$  and  $c(b, e') > c(a, e')$  we have

$$\sum_{e \in E} p(e) c(b, e) > \sum_{e \in E} p(e) c(a, e)$$

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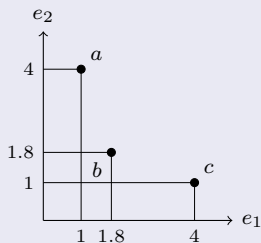
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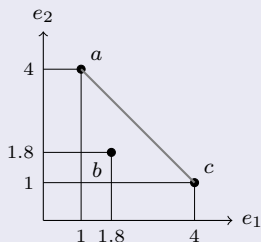
$$\sum_{e \in E} p(e)c(b, e) > \sum_{e \in E} p(e)c(a, e)$$

## Converse



- $A = A^* = \{a, b, c\}$
- $b$  cannot be solution of  $(P)$

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# Wald's criterion (Max Min)

## Idea

- extreme pessimism: base your choice on the **worst** situation (Max Min)
- choose any alternative  $a \in A$  solution of:

$$\max_{a \in A} \min_{e \in E} c(a, e)$$

## Example

choose  $a_3$

(maximum loss = -5)

$a_1$  (maximum loss = -20)

$a_2$  (maximum loss = -10)

$c$	$e_1$	$e_2$	$e_3$	<b>min</b>
$a_1$	40	70	-20	<b>-20</b>
$a_2$	-10	40	100	<b>-10</b>
$a_3$	20	40	-5	<b>-5</b>



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## Remarks

- bad use of information
- no compensation between consequences in different states of Nature
- bias towards *status quo*
- only requires that  $X$  can be ordered

## Example

$c$	$e_1$	$e_2$	$e_3$	$\dots$	$e_{1\,000}$
$a$	-100	10 000	10 000	$\dots$	10 000
$b$	-99	-99	-99	$\dots$	-99

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## Example

$c$	$e_1$	$e_2$	$e_3$	$\dots$	$e_{1\,000}$
$a$	-100	10 000	10 000	$\dots$	10 000
$b$	-99	-99	-99	$\dots$	-99

# Other classic criteria

- Max Max
- Hurwicz
- Min Max Regret
- Laplace

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# Max Max

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- optimism: base your choice on the **best** possible situation (Max Max)
- choose any alternative in  $a \in A$  solution of:

$$\max_{a \in A} \max_{e \in E} c(a, e)$$

## Example

choose  $a_2$

(maximal gain = 100)

$a_1$  (maximal gain = 70)

$a_3$  (maximal gain = 40)

$c$	$e_1$	$e_2$	$e_3$	max
$a_1$	40	70	-20	70
$a_2$	-10	40	100	100
$a_3$	20	40	-5	40

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$c$	$e_1$	$e_2$	$e_3$	<b>max</b>
$a_1$	40	70	-20	<b>70</b>
$a_2$	-10	40	100	<b>100</b>
$a_3$	20	40	-5	<b>40</b>

## Remarks

- bad use of information
- no compensation between consequences in different states of Nature
- only requires that  $X$  can be ordered



# Hurwicz

## Idea

- **compromise** between extreme pessimism (Max Min) and extreme optimism (Max Max)
- let  $\alpha \in [0; 1]$  called “coefficient of pessimism”, choose any alternative  $a \in A$  solution of:

$$\max_{a \in A} \left[ \alpha \min_{e \in E} c(a, e) + (1 - \alpha) \max_{e \in E} c(a, e) \right]$$

$$\alpha = 1/2$$

Choose $a_2$	$c$	$e_1$	$e_2$	$e_3$	min	max	$\alpha = 1/2$
$(90/2 = 45)$	$a_1$	40	70	-20	-20	70	25
$a_1 (50/2)$	$a_2$	-10	40	100	-10	100	45
$a_3 (35/2)$	$a_3$	20	40	-5	-5	40	17,5

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$a_3 \ (35/2)$	$a_3$	20	40	-5	-5	40	17,5

## Remarks

- bad use of information
- compromise between bad solutions
- it must be meaningful to take linear combinations!
- how to assess the coefficient of pessimism  $\alpha$ ?

# Savage (Min Max Regret)

## Idea

- criterion for bureaucrats
- choose  $a_2$  and  $e_1$  obtains
  - best decision:  $a_1$  (40)
  - decision taken:  $a_2$  (-10)
  - regrets:  $40 - (-10) = 50$

$c$	$e_1$	$e_2$	$e_3$
$a_1$	40	70	-20
$a_2$	-10	40	100
$a_3$	20	40	-5

## Definition

Choose any alternative  $a \in A$  solution of:

$$\min_{a \in A} \max_{e \in E} R(a, e)$$

avec

$$R(a, e) = \max_{b \in A} c(b, e) - c(a, e)$$

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## Example

$c$	$e_1$	$e_2$	$e_3$	$R$	$e_1$	$e_2$	$e_3$	max
$a_1$	40	70	-20	$a_1$	0	0	120	120
$a_2$	-10	40	100	$a_2$	50	30	0	50
$a_3$	20	40	-5	$a_3$	20	30	105	105

Choose  $a_2$  (max regret 50)  $a_1$  (120),  $a_3$  (105)

## Remarks

- criterion that is different from Max Min ( $a_3$ )
- it must be meaningful to take differences!
- taking differences is an adequate way to measure regrets
- choice is set dependent. Adding new alternatives can alter choice in an unpredictable way

## Example

$c$	$e_1$	$e_2$	$e_3$	$R$	$e_1$	$e_2$	$e_3$	max
$a_1$	40	70	-20	$a_1$	0	0	120	120
$a_2$	-10	40	100	$a_2$	50	30	0	50
$a_3$	20	40	-5	$a_3$	20	30	105	105

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## Example

$c$	$e_1$	$e_2$	
$a_1$	8	0	
$a_2$	2	4	

$R$	$e_1$	$e_2$	max
$a_1$	0	4	4
$a_2$	6	0	6

- Choice of  $a_1$

## Example (adding $a_3$ )

$c$	$e_1$	$e_2$	
$a_1$	8	0	
$a_2$	2	4	
$a_3$	1	7	

$R$	$e_1$	$e_2$	max
$a_1$	0	7	7
$a_2$	6	3	6
$a_3$	7	0	7

- initial choice:  $a_1$
- choice after addition of  $a_3$ :  $a_2$ !
- risk of “manipulations”

## Example

$c$	$e_1$	$e_2$
$a_1$	8	0
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$a_1$	8	0
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$a_1$	0	7	7
$a_2$	6	3	6
$a_3$	7	0	7

- initial choice:  $a_1$
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- risk of “manipulations”

# Laplace

## Idea

- Principle of “insufficient reason”

## Definition

Choose any alternative in  $A$  solution of:

$$\max_{a \in A} \sum_{e \in E} \frac{1}{|E|} c(a, e)$$

## Example

Choose  $a_2$

(130/3)

$a_1$  (90/3)

$a_3$  (55/3)

$c$	$e_1$	$e_2$	$e_3$	
$a_1$	40	70	-20	90/3
$a_2$	-10	40	100	130/3
$a_3$	20	40	-5	55/3

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$c$	$e_1$	$e_2$	$e_3$	
$a_1$	40	70	-20	90/3
$a_2$	-10	40	100	130/3
$a_3$	20	40	-5	55/3

## Remarks

- it must be meaningful to take linear combinations!
- either you will become the King of the Belgians or not. Are these two events equally likely?
- criterion that depends on the arbitrary model for states of Nature ( $E$  can always be refined: “ $E$  and rain tomorrow” and “ $E$  and no rain tomorrow”)
- Is expected gain a good criterion, even when all states are supposed equally likely?

►► go faster

# Example

## Example

$c$	$e_1$	$e_2$	$e_3$	$e_4$
$a$	2	2	0	1
$b$	1	1	1	1
$c$	0	4	0	0
$d$	1	3	0	0

## Results

- Wald:  $b$
- Max Max:  $c$
- Laplace:  $a$
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# Conclusion

## Classic Criteria

- none really satisfactory!
  - necessity to model likelihood (beliefs)
  - necessity to model desirability of consequences (tastes)

## Central questions

- why is there no probability?
- where probabilities come from?

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# Plan

- 1 Introduction
- 2 Decision under risk
  - Model
  - Classic Criteria
  - Expected Utility Theory
  - Risk aversion
- 3 Decision under uncertainty
- 4 Extensions

# Decision under risk: model

## Model

- $X$ : set of consequences
- $X$  finite =  $\{x_1, x_2, \dots, x_n\}$
- $X \subseteq \mathbb{R}$  (e.g., money)

## Simple lottery on $X$

- discrete r.v. on  $X$
- $\ell = (x_1, p_1; x_2, p_2; \dots; x_n, p_n)$
- $p_\ell(x_i)$ : probability to obtain consequence  $x_i$  with lottery  $\ell$

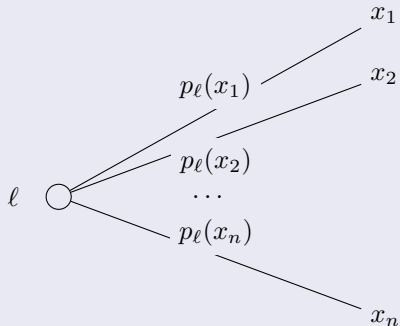
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Simple lottery  $\ell$  on  $X$ 

# Lotteries

## Set of lotteries

- simple lotteries on  $X$
- first order lotteries on  $X$ : lotteries on simple lotteries
- second order lotteries on  $X$ : lotteries on first order lotteries
- etc.
  
- $L(X)$ : set of lotteries at all finite orders
  - $L(X)$  is always infinite

# Lotteries

## Remark

- $L(X)$  includes all lotteries that corresponds to the implementation of alternatives in  $A$  and many other “hypothetical” lotteries

## Problem

- help someone compare lotteries in  $L(X)$

## Notation

- $\ell \in L(X)$ : lotteries (simple or not)
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# Classic Criterion

## Expected Value ( $EV$ )

$$\ell \succ \ell' \Leftrightarrow \sum_{x \in X} xp_{\ell}(x) > \sum_{x \in X} xp_{\ell'}(x)$$

$$\ell \sim \ell' \Leftrightarrow \sum_{x \in X} xp_{\ell}(x) = \sum_{x \in X} xp_{\ell'}(x)$$

- $\succ$ : strict preference
- $\sim$ : indifference

# EV

## Advantages

- simple
- good use of information
- can be “decentralized”

## Disadvantages

- limited to numerical consequences
- no clear rationale
- contradict observed behavior of “rational” people (diversification, insurance)

# EV

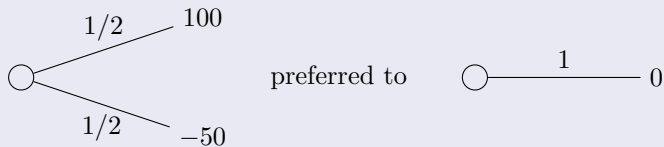
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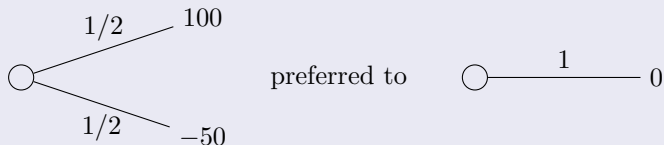


$$E(\ell) = 25 \quad E(\ell') = 0$$

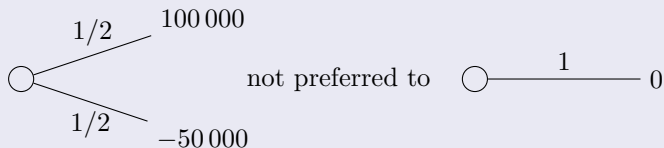


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# Example



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# Saint Petersburg Paradox (D. Bernoulli)

## Game

- a “banker” plays with a “player”. The player must pay a fixed sum to enter the game.
- the banker tosses a coin till “Tails” obtains
- the game stops
- if “tails” obtains at the  $n$ th toss, the banker pays  $2^n \text{ €}$  to the player
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$$EV = 2 \times \frac{1}{2} + 2^2 \times \frac{1}{2^2} + \dots$$

(50% chances of winning only 2 €!)

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# Other classic criterion

## Expected Value + Variance

- add a measure of dispersion to the measure of central tendency

## Problems

- less simple
- how to deal with the two criteria (efficient solutions or synthesis?)
- is variance a good measure of risk? (inter-quartile spreads, semi-variance, etc.)

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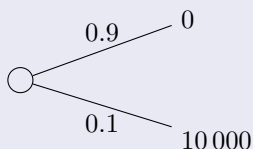
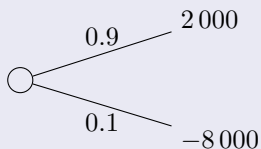
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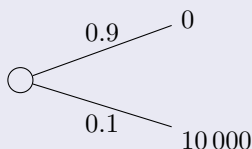
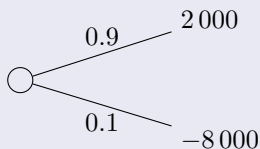


$$\begin{aligned}EV(\ell) &= 0.9 \times 2000 + 0.1 \times -8000 = 1000 \\ &= 0.9 \times 0 + 0.1 \times 10000\end{aligned}$$

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# Limit of classic criteria

## Central problem

- these criteria do not take the **psychology** of the individual towards risk
  - what is her wealth?
  - what is her income?
  - what is her attitude towards risk?
  - etc.



# Pseudo-Solution

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- **directly** ask the individual about her preferences

## Problems

- consistency?
- decentralized decisions?
- cognitive effort!

# Pseudo-Solution

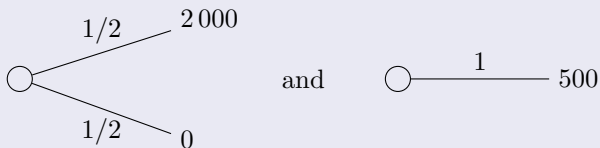
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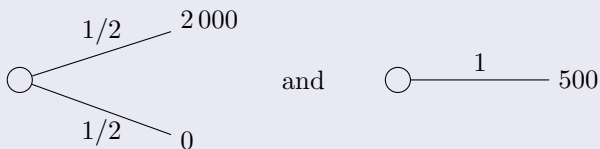
Example: choice between



Example: choice between

- $\mathcal{N}(878.32; 72.45)$  and
- $Bi(1200; 0.75)$

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# Expected Utility Theory

J. von Neumann & O. Morgenstern (1945)

## Idea

- ask the individual about **simple choices**
- model the behavior of the individual using a **mathematical model**
- use the model to process complex choices

## Questions

- what model?
- what rationale?
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# Mathematical model

## Idea

- replace  $EV$  by an “Expected Utility” ( $EU$ )
- the ‘utility’ capture the psychology of the individual towards risk

$$\ell \succ \ell' \Leftrightarrow \sum_{x \in X} \textcolor{red}{u}(x)p_{\ell}(x) > \sum_{x \in X} u(x)p_{\ell'}(x)$$

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## Utility function

- $u : X \rightarrow \mathbb{R}$
- $u(x)$  is the “utility” of consequence  $x \in X$
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## Advantages

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# Theoretical Analysis

## How to justify the model?

- axiomatic analysis

## Interpretation of axioms?

- descriptive
- normative
- prescriptive

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### Axiom (A1: Ranking)

For all  $\ell, \ell' \in L(X)$  at least one the following holds:

- $\ell$  is preferred or indifferent to  $\ell'$  ( $\ell \succsim \ell'$ )
- $\ell'$  is preferred or indifferent to  $\ell$  ( $\ell' \succsim \ell$ )

Moreover,  $\succsim$  is transitive:

$$\ell \succsim \ell' \text{ and } \ell' \succsim \ell'' \Rightarrow \ell \succsim \ell''$$

$$\forall \ell, \ell', \ell'' \in L(X)$$

## Remark

- $\ell \succ \ell' \Leftrightarrow [\ell \succsim \ell' \text{ and } \text{Not}[\ell' \succsim \ell]]$ 
  - strict preference
- $\ell \sim \ell' \Leftrightarrow [\ell \succsim \ell' \text{ and } \ell' \succsim \ell]$ 
  - indifference
- A1 implies that  $\sim$  and  $\succ$  are transitive

# Descriptive Analysis

## Difficulties

- incomplete preference
- nontransitive indifference
- intransitive strict preference

## Complete Preferences?

- the *raison d'être* of the theory is to help structure preferences!

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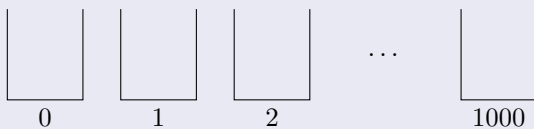
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# Luce (1956)

## Comparison of cups of coffee



$$0 \sim 1, 1 \sim 2, \dots, 999 \sim 1000 \Rightarrow 0 \sim 1000$$

- imperfect senses  $\Rightarrow$  nontransitive indifference

» go faster



# Dominance with threshold

## Example

$$x \succ y \Leftrightarrow \begin{cases} x \text{ at least as good as } y \text{ on all criteria} \\ x \text{ better than } y \text{ on at least one criterion} \end{cases}$$

	$g_1$	$g_2$	$g_3$
$a$	10	10	10
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- threshold = 1.1 (below you do not distinguish)



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► go faster

# Condorcet's Paradox

## Data

- Voter 1:  $a \succ b \succ c$
- Voter 2:  $c \succ a \succ b$
- Voter 3:  $b \succ c \succ a$

majority :  $a \succ b; b \succ c, c \succ a$

# Threshold effects

## Example

1	Car	15 000 €
2	Car + PE1	15 500 €
3	Car + PE1 + PE2	16 000 €
4	Car + PE1 + PE2 + PE3	16 500 €
$\vdots$		
$n$	Car + ...	18 000 €

## Preference of a “naïve” consumer

$$2 \succ 1, 3 \succ 2, 4 \succ 3, n \succ (n-1) \quad \text{but} \quad 1 \succ n$$

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# Analysis

## Prescriptive Approach

- effectiveness
  - it is simple to help someone choose on the basis of complete and transitive preferences

## Normative Approach

- money pump argument
  - exchanges starting with  $c$



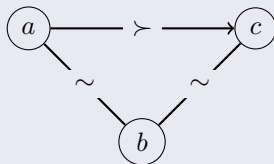
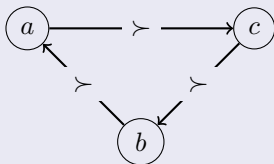
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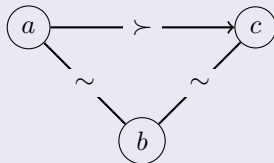
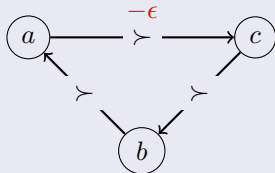
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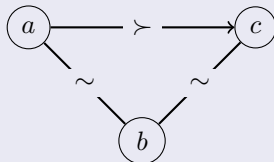
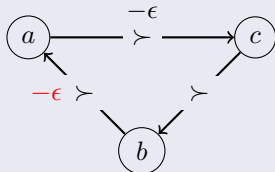
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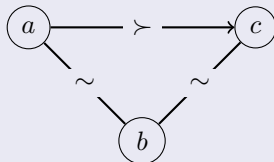
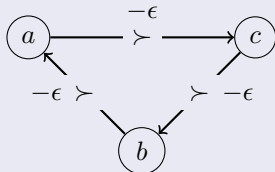
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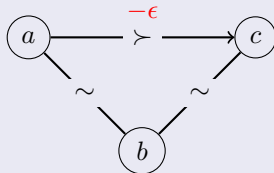
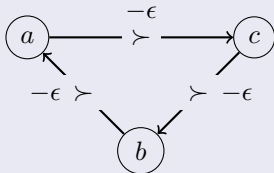
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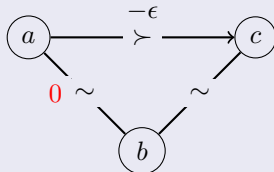
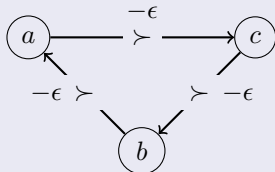
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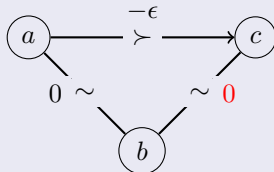
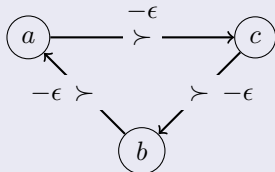
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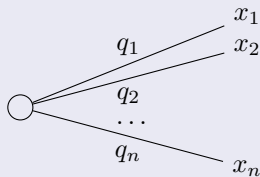
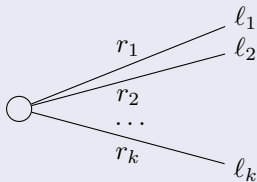




# Axioms

## Axiom (A2 Reduction)

$\ell_j$ : first order lotteries



with  $q_i = \sum_{j=1}^k r_j p_{\ell_j}(x_i)$

$\Rightarrow$  Indifference

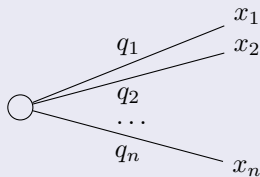
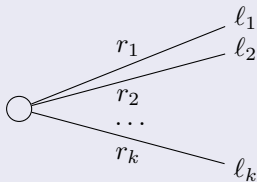
## Interpretation

- “games are played seriously”

# Axioms

## Axiom (A2 Reduction)

$\ell_j$ : first order lotteries



with  $q_i = \sum_{j=1}^k r_j p_{\ell_j}(x_i)$

$\Rightarrow$  Indifference

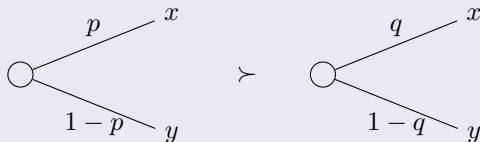
## Interpretation

- “games are played seriously”

# Axioms

## Axiom (A3 Monotonicity)

If  $(x, 1) \succ (y, 1)$  then



iff  $p > q$  ( $\forall x, y \in X$ )

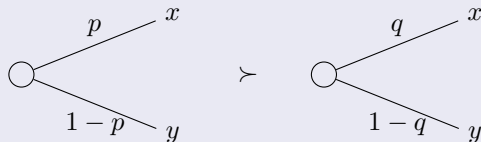
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- greediness
- do not try to outperform randomness

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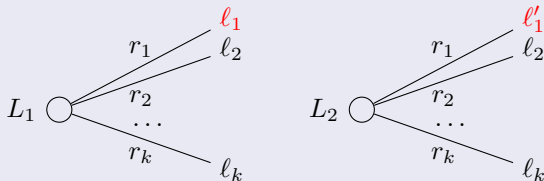
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## Axiom (A4 Independence)



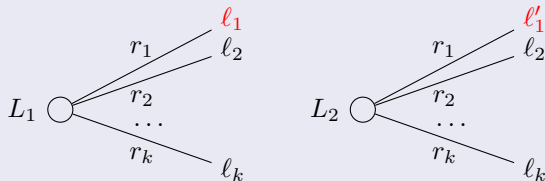
If  $\ell_1 \sim \ell'_1$  then  $L_1 \sim L_2$

## Interpretation

- indifference is indifference

# Axioms

## Axiom (A4 Independence)



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# Axioms

## Axiom (A5 Continuity)

If  $(x, 1) \succ (y, 1) \succ (z, 1)$  then there is a probability  $p \in ]0; 1[$  such that:



## Remark

- A3 implies that this probability is unique

## Interpretation

- you are not naïve with probabilities (continuum between certainty and risk)

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## Example

- $x$ : win 2€
- $y$ : win 1€
- $z$ : be hung tomorrow at dawn

$$(x, 1) \succ (y, 1) \succ (z, 1)$$

## Problem

- is there a probability  $p \in ]0; 1[$  such that:  

$$y \sim (x, p; z, (1 - p))$$
- $p = 1 - 10^{-100}$ ?

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# Consequences of axioms

## Theorem (Representation)

Let  $\succsim$  be a preference relation on  $L(X)$ .

This relation satisfies A1-A5

iff

there is a function  $u : X \rightarrow \mathbb{R}$  such that:

$$\ell \succsim \ell' \Leftrightarrow \sum_{x \in X} u(x)p_{\ell}(x) \geq \sum_{x \in X} u(x)p_{\ell'}(x) \quad (\text{vNM})$$

## Remark

- necessity is obvious
- $u$  is linked to  $\succsim$  and, hence, to the individual

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» skip proof

# Proof

## 5 steps

- constructive!

Finite case:  $X = \{x_1, x_2, \dots, x_n\}$

Consider a lottery  $\ell \in L(X)$

1)

Using A1, (ranking), A2 (reduction) and A4 (independence), you can always find a **simple lottery** such that:  $\ell \sim (x_1, p_\ell(x_1); x_2, p_\ell(x_2); \dots; x_n, p_\ell(x_1))$

Suppose wlog that:

$$(x_n, 1) \succ (x_{n-1}, 1) \succ \dots \succ (x_1, 1)$$

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2)

A5 (continuity): since

$$(x_n, 1) \succ (x_{n-1}, 1) \succ \cdots \succ (x_1, 1)$$

there is  $u_i \in ]0; 1[$  such that

$$(x_i, 1) \sim [x_n, u_i; x_1; (1 - u_i)]$$

Let:

$$u_n = 1, u_1 = 0$$

3)

Using A4 (independence), A1 (ranking) and A2 (reduction), we know that:

$$\ell \sim (x_1, p_\ell(x_1); x_2, p_\ell(x_2); \dots; x_{n-1}, p_\ell(x_{n-1}); x_n, p_\ell(x_n))$$

$$\ell \sim [x_1, (1 - K_\ell); x_2, 0; \dots; x_{n-1}, 0; x_n, K_\ell]$$

with

$$K_\ell = \sum_{i=1}^n p_\ell(x_i) u_i$$

4)

Use steps 1) to 3) to transform a lottery

$$\ell' \sim (x_1, p_{\ell'}(x_1); x_2, p_{\ell'}(x_2); \dots; x_{n-1}, p_{\ell'}(x_{n-1}); x_n, p_{\ell'}(x_n))$$

We have:

$$\ell' \sim [x_1, (1 - K_{\ell'}); x_2, 0; \dots; x_{n-1}, 0; x_n, K_{\ell'}]$$

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5)

Using A1 (Ranking) and A3 (Monotonicity) we know that:

$$\begin{aligned}
 \ell &\succ \ell' \Leftrightarrow \\
 (x_1, p_\ell(x_1); x_2, p_\ell(x_2); \dots; x_n, p_\ell(x_n)) &\succ \\
 (x_1, p_{\ell'}(x_1); x_2, p_{\ell'}(x_2); \dots; x_n, p_{\ell'}(x_n)) &\Leftrightarrow \\
 [x_1, (1 - K_\ell); x_2, 0; \dots; x_n, K_\ell] &\succ [x_1, (1 - K_{\ell'}); x_2, 0; \dots; x_n, K_{\ell'}] \\
 &\Leftrightarrow \\
 K_\ell &> K_{\ell'} \Leftrightarrow \\
 \sum_{i=1}^n p_\ell(x_i) u_i &> \sum_{i=1}^n p_{\ell'}(x_i) u_i
 \end{aligned}$$

and define  $u$  letting:

$$u(x_i) = u_i$$



# Consequences of axioms

## Theorem (Uniqueness)

If there are two functions  $u$  and  $v$  such that (vNM) holds then there are  $\alpha, \beta \in \mathbb{R}$  with  $\alpha > 0$  such that:

$$v(x) = \alpha u(x) + \beta$$

$\forall x \in X$

## Interpretation

preferences can be measured as temperature

## Proof

obvious: if  $u$  and  $v$  are not linked by a positive affine transformation, you can always find two loteries that will have different expected utilities  $\square$

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# Assessment of a utility function

## Hypotheses

- $X = \mathbb{R}$  (money)
- let  $u(0) = 0$  and  $u(10\,000) = 1$

## Assessment



$$1 \times u(x) = u(x) = 1/2 \times u(10\,000) + 1/2 \times u(0) = 1/2$$

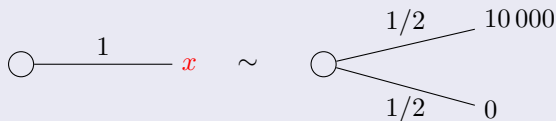


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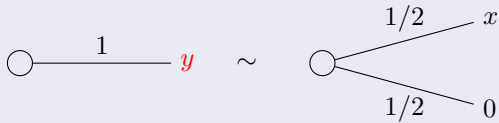
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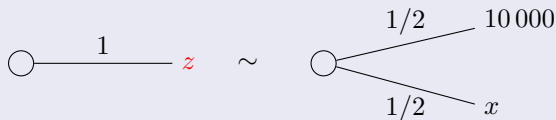
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$$u(y) = 1/2 \times u(x) + 1/2 \times u(0) = 1/4$$

# Assessment of a utility function

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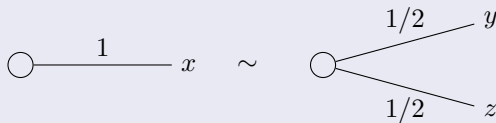


$$u(z) = 1/2 \times u(10000) + 1/2 \times u(x) = 3/4$$

# Assessment of a utility function

## Control question

- we must have:



- if not: go back and check

# Assessment of a utility function

## General case



$$u(x) = pu(z) + (1 - p)u(w)$$

- 4 unknowns
- fix the value of 3 of them and find indifference on the 4th

# Assessment of a utility function

## General case



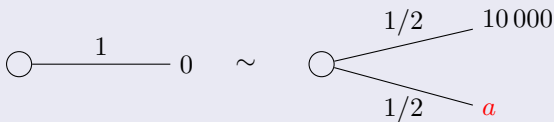
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►► go faster

# Assessment of a utility function

Going down



$$1 \times u(0) = 1/2 \times u(10\,000) + 1/2 \times u(a) \Rightarrow u(a) = -1$$

# Assessment of a utility function

Going up



$$1 \times u(10\,000) = 1/2 \times u(b) + 1/2 \times u(0) \Rightarrow u(b) = 2$$



# Assessment of a utility function

## Remarks

- use of simple probabilities:  $1/2$ ,  $1/3$ ,  $1/4$
- bracketing
- checks are necessary

# Risk aversion

## Context

- let  $X = \mathbb{R}$  (money)
- suppose that the DM satisfies A1-A5
- it is not restrictive to suppose that  $u$  is increasing!
- $P$ : wealth of the DM

## Certainty equivalent of a lottery $\ell$ : $\hat{x}(\ell)$

- amount of money that the DM finds equivalent to owning the lottery  $\ell$   
(minimal selling price of lottery  $\ell$ )

$$E(u(P + \hat{x}(\ell))) = u(P + \hat{x}(\ell)) = E(u(P + \ell))$$

$$\Rightarrow$$

$$\hat{x}(\ell) = u^{-1}[E(u(P + \ell))] - P$$

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# Risk premium $\pi(\ell)$ for lottery $\ell$

## Risk premium

$$\pi(\ell) = E(\ell) - \hat{x}(\ell)$$

## Risk aversion

A DM is

- **risk averse** if  $\hat{x}(\ell) < E(\ell)$
- **risk prone** if  $\hat{x}(\ell) > E(\ell)$
- **risk neutral** if  $\hat{x}(\ell) = E(\ell)$  (EV)

pour toute loterie  $\ell \in L(X)$

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## Theorem (Arrow-Pratt)

A DM is risk averse iff her utility function is concave

### Proof

Risk aversion ( $\hat{x}(\ell) < E(\ell)$ )

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$$u(P + [\alpha x + (1 - \alpha)y]) = u(\alpha(x + P) + (1 - \alpha)(y + P)) > E(u(P + \ell)) = \alpha u(x + P) + (1 - \alpha)u(y + P)$$

$$z = x + P, w = y + P$$

$$u(\alpha z + (1 - \alpha)w) > \alpha u(z) + (1 - \alpha)u(w)$$



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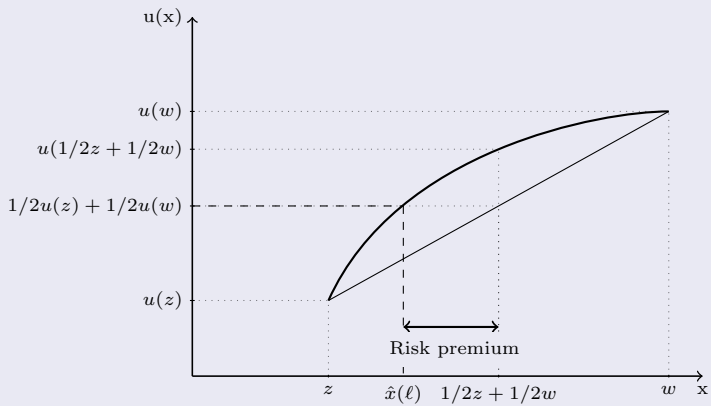
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# Plan

- 1 Introduction
- 2 Decision under risk
- 3 Decision under uncertainty
  - Reminder on Probabilities
  - Classic school
  - Subjectivist school
  - Applications
- 4 Extensions

# Decision under uncertainty

## Subjectivist school

- de Finetti, Savage, Aunscombe & Aumann
- use results in the context of risk in the context of uncertainty

## Idea



## Minimal greediness

- if the first lottery is preferred (not preferred) to the second lottery we have  $P(E) > p$  (resp.  $P(E) < p$ )
- the **subjective probability** of the event is the value  $p$  for which these two lotteries are indifferent

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# Subjectivist school

## Idea

- under uncertainty, use subjective probabilities
- back to the case of decision under risk

## Questions

- are subjective probabilities probabilities?
- are subjective probabilities “true” probabilities?
- rationale?

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# Reminder

## Random experiment

- experiment for which it is impossible to predict the result before you run it

## Examples

- toss of a coin
- price of \$/€ in one year
- next month sales



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## Definitions

- $S$ : set of elementary events = sets of all possible results of the experiment
  - toss of a coin:  $S = \{H, T\}$
  - rolling a dice:  $S = \{1, 2, 3, 4, 5, 6\}$
  - price of \$/€ in one year:  $S = \mathbb{R}_+$
- $\mathcal{S}$ : sets of events based on  $S$  (allowing to speak of the union, intersection, of elementary events)

## Properties

- $S \in \mathcal{S}$
- $A \in \mathcal{S} \Rightarrow S \setminus A \in \mathcal{S}$
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## Definition

A probability  $P$  is a mapping from  $\mathcal{S}$  to  $\mathbb{R}$ , associating to each event  $A$  its probability  $P(A)$  and such that:

- ❶  $P(A) \geq 0$
- ❷  $P(S) = 1$
- ❸  $A \cap B = \emptyset \Rightarrow P(A \cup B) = P(A) + P(B)$

(Kolmogorov's axioms)

» go faster

# Reminder

## Definition

The probability of “ $A$  given  $B$ ”, denoted by  $P(A/B)$ , is defined by:

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## Total probabilities

If  $A_1, A_2, \dots, A_n$  partition  $S$  then:

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## Bayes' formula

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# Probabilities?

## Source of probabilities?

- two main schools
  - classic school
  - subjectivist school

# Classic school

## Sources of probabilities

- logical source
- frequentist source

## Logical source

### Principle of “insufficient reason”

- $P(\text{Tails}) = P(\text{Heads}) = 1/2$
- $P(E) = \text{Number of favorable cases} / \text{Number of cases}$

## Problems

- restricted to “perfect objects”
- problem of proof: what is a “perfect object”?
- limited practical impact (ex.: price of \$/€ in one year?)

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## Frequentist source

### Law of large numbers

- $n$  identical and independent repetitions of event  $E$  of probability  $p$ :

$$\lim_{n \rightarrow \infty} P \left( \left| \frac{n(E)}{n} - p \right| > \varepsilon \right) = 0$$

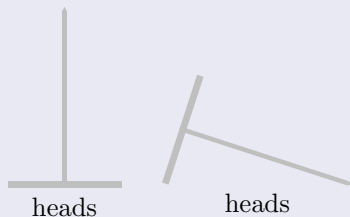
- “if you do not  $p$ , experiment”

## Problems

- events that cannot be “repeated”
- proof problem (past = future)
- different results if “identical”?
- rôle of information unclear

# Exemple: thumbtack

- 10 000 € if Tails
- -5 000 € if Heads

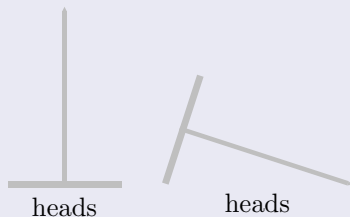


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would you accept to perform 1 000 tosses of the thumbtack before taking a decision?

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## Facts

- 2 different and equally rational persons can have different probability judgements on the same event
- experiments transform probability judgements: thumbtack
- language has a great variety of expression to speak of likelihood

## Definition: subjective probability

- a probability is a **subjective measure** of the likelihood of an event

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# Subjective Probabilities

## Problems

- why probabilities? (Kolmogorov's axioms)
- how to assess probabilities?

## Idea

- comparing “uncertain” lotteries with “risky lotteries” (with probabilities coming from consensual random experiments; coins, cards, dices, urns, etc.)

# Subjective Probabilities

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## Comparison



## Minimal greediness

- if the first lottery is preferred to the second:  $P(E) > p$
- if the second lottery is preferred to the first:  $P(E) < p$
- subjective probability of  $E$  is the value for which they are indifferent

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## Problems

- are these numbers probabilities?
- are these numbers **true** probabilities?
- rationale?
- use?

# Results

## Idea

- add to  $L(X)$  “uncertain lotteries”
- impose (modified) axioms A1-A5 to this new set of lotteries

## Theorem (Savage, Aunscombe & Aumann)

There are a function  $u : X \rightarrow \mathbb{R}$  and a Probability measure  $P$  on  $\mathcal{S}$  such that the comparison of two lotteries (risky or uncertain) is done comparing their Expected Utilities ( $p$  for risky lotteries,  $P$  for risky lotteries)

The probability measure is **unique**. The utility function is unique up to a positive affine transformation.

- tastes:  $u$  (subjective)
- beliefs:  $P$  (subjective)



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## Consequences

- same analysis as in the risky case replacing (whenever needed) “probabilities” by “subjective probabilities”
- exchangeable events and subjective vs objective probabilities (de Finetti)

# Idea of proof

Example ( $A \cap B = \emptyset$ )

$L_1$	$F$	$P$	$L_2$	$F$	$P$
$A$	$G$	$L$	$A$	$G$	$L$
$B$	$L$	$G$	$B$	$G$	$L$
$A \text{ or } B$	$L$	$L$	$A \text{ or } B$	$L$	$L$

A2  $\Rightarrow [G \text{ if Tails}; L \text{ if Heads}] \sim [L \text{ if Tails}; G \text{ if Heads}]$

A4  $\Rightarrow L_1 \sim L_2$

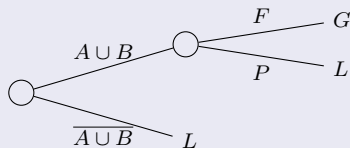
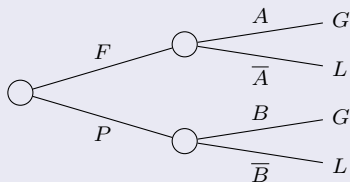
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$B$	$L$	$G$	$B$	$G$	$L$
$A \text{ or } B$	$L$	$L$	$A \text{ or } B$	$L$	$L$

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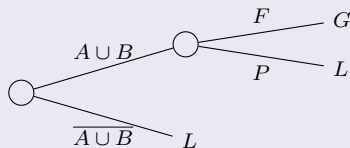
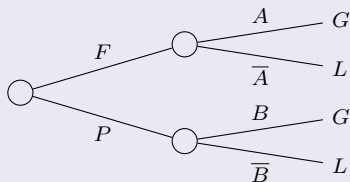
A4  $\Rightarrow L_1 \sim L_2$



## Consequences

- $A4 \Rightarrow L_1 \sim L_2$
- Let  $u(G) = 1$  and  $u(P) = 0$
- We have:

$$\begin{aligned}
 E[u(L_1)] &= 0.5[P(A)u(G) + (1 - P(A))u(L)] + \\
 &\quad 0.5[P(B)u(G) + (1 - P(B))u(L)] = \\
 &\quad 0.5[P(A) + P(B)] \quad \Rightarrow \quad P(A) + P(B) = P(A \cup B) \\
 E[u(L_2)] &= 0.5P(A \cup B)
 \end{aligned}$$



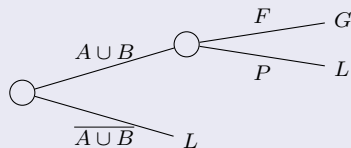
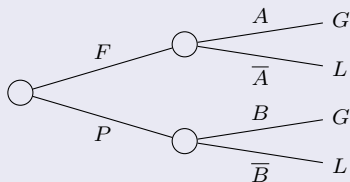
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# Bayesian Decision Theory

creativity:  $A$   
objectives:  $X$   
beliefs:  $P$   
tastes:  $u$   
evaluation:  $(S)EU$

# Applications

- risk aversion
- stochastic dominance
- economics of finance and insurance
- value of information
- Bayesian statistics

• all this must be skipped today      :-)

# Applications

- risk aversion
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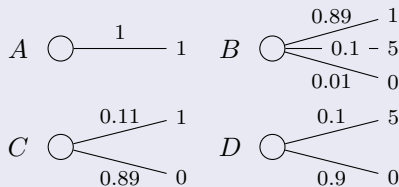
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# Plan

- 1 Introduction
- 2 Decision under risk
- 3 Decision under uncertainty
- 4 Extensions
  - Decision under risk
  - Decision under uncertainty

# Allais' paradox ( $10^6$ €)

## Experiment



## Modal Result

- $A \succ B$  and  $D \succ C$

## Interpretation

- $A \succ B \Rightarrow$   

$$u(1) > 0.89u(1) + 0.1u(5) + 0.01u(0)$$

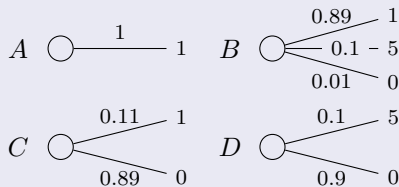
$$\Rightarrow u(1) > 10/11$$
- $D \succ C \Rightarrow$   

$$0.1u(5) + 0.9u(0) > 0.11u(1) + 0.89u(0)$$

$$\Rightarrow u(1) < 10/11$$

# Allais' paradox ( $10^6$ €)

## Experiment



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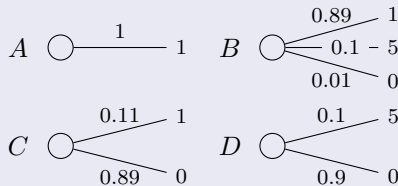
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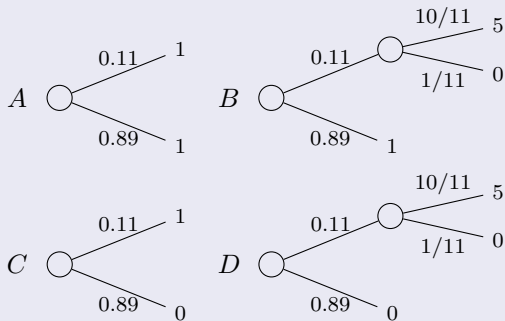
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# Allais' paradox

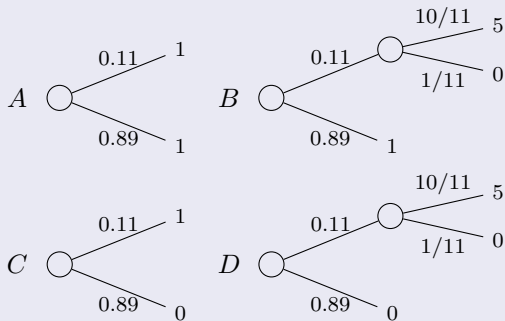
## Reformulation





# Allais' paradox

## Reformulation



► skip examples

# Allais's paradox

## Reformulation

- urn with 89 balls  $R$  and 11 balls  $N$

	$R$	$N$
$X$	$Q$	1
$Y$	$Q$	5 with 10/11 0 with 1/11

Should the choice between  $X$  and  $Y$  depend upon  $Q$ ?

- $Q = 0$ :  $X = C$  and  $Y = D$
- $Q = 1$ :  $X = A$  and  $Y = B$

# Violation of independence

## Normative analysis

- $(x; 1) \succ (y; 1)$  and  $(y, p; z, (1 - p)) \succ (x, p; z, (1 - p))$
- $(x; 1) \succ (y; 1) \Rightarrow (x - \varepsilon; 1) \succ (y; 1)$
- $(y, p; z, (1 - p)) \succ (x, p; z, (1 - p)) \Rightarrow$   
 $(y, p; z - \varepsilon, (1 - p)) \succ (x, p; z, (1 - p))$

## Exchanges starting with $(x, p; z, (1 - p))$

- $(x, p; z, (1 - p))$  exchanged for  $(y, p; z - \varepsilon, (1 - p))$ 
  - if  $E$  does not obtain: gain =  $(z - \varepsilon)$
  - if  $E$  obtains: gain =  $y$  and  $y$  is exchanged for  $(x - \varepsilon)$
- you had  $(x, p; z, (1 - p))$  you have  $(x - \varepsilon, p; z - \varepsilon, (1 - p))!$

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# Rational violation of independence

## Mom's decision

- two kids:  $A$  and  $B$
- only one gift to be given to either  $A$  or  $B$

## Mom's preferences

- $(A, 1) \sim (B, 1)$
- $(B, 1/2; A, 1/2) \succ (A, 1/2; A, 1/2)$

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# EURDP

## Definition

- distortion of probabilities according to their **rank**
- $\phi$  increasing bijection on  $[0, 1]$

$$(x_1, p_1; x_2, p_2; \dots; x_n, p_n)$$

$$x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$$

where  $(\cdot)$  is a permutation on  $\{1, 2, \dots, n\}$

$$(x_{(1)}, p_{(1)}; x_{(2)}, p_{(2)}; \dots; x_{(n)}, p_{(n)})$$

# EURDP

## Computation

- $\phi$  increasing bijection on  $[0, 1]$

$$\begin{aligned} EURDP = & u(x_{(1)})\phi(p_{(1)} + p_{(2)} + \dots p_{(n)}) + \\ & (u(x_{(2)}) - u(x_{(1)}))\phi(p_{(2)} + \dots p_{(n)}) + \\ & (u(x_{(3)}) - u(x_{(2)}))\phi(p_{(3)} + \dots p_{(n)}) + \\ & (u(x_{(4)}) - u(x_{(3)}))\phi(p_{(4)} + \dots p_{(n)}) + \\ & \dots \\ & (u(x_{(n)}) - u(x_{(n-1)}))\phi(p_{(n)}) \end{aligned}$$

## Motivation

- distorting cumulative probabilities is inevitable if inconsistencies are to be avoided



# EURDP

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# Allais' paradox and EURDP

## Allais

- $A : (1, 1)$  preferred to  $B : (0, 0.01; 1, 0.89; 5, 0.1)$
- $D : (0, 0.9; 5, 0.1)$  preferred to  $C : (0, 0.89; 1, 0.11)$

$$EURDP(A) = u(1)$$

$$EURDP(B) = u(0) + (u(1) - u(0))\phi(0.99) + (u(5) - u(1))\phi(0.1)$$

$$EURDP(C) = u(0) + (u(1) - u(0))\phi(0.11)$$

$$EURDP(D) = u(0) + (u(5) - u(0))\phi(0.1)$$

# Allais' paradox and EURDP

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$$EURDP(C) = u(0) + (u(1) - u(0))\phi(0.11)$$

$$EURDP(D) = u(0) + (u(5) - u(0))\phi(0.1)$$

- $u(5) = 1, u(0) = 0, \phi(x) = x^2$

$$EURDP(A) = u(1) \quad EURDP(B) = 0.01 - 0.9701u(1)$$

$$EURDP(C) = 0.11^2 u(1) \quad EURDP(D) = 0.1^2$$

$$u(1) > 0.01 - 0.9701u(1) \Rightarrow u(1) > 0.33$$

$$0.1^2 > 0.11^2 u(1) \Rightarrow u(1) < 0.826$$

# Interpretation

## EU & EURDP

- $\phi(x) = x^2$
- $(x_1, p; x_2, (1 - p))$  avec  $x_1 < x_2$
- $EU = pu(x_1) + (1 - p)u(x_2)$

$$\begin{aligned}EURDP &= u(x_1) + (u(x_2) - u(x_1))\phi(1 - p) \\ &= (1 - \phi(1 - p))u(x_1) + \phi(1 - p)u(x_2)\end{aligned}$$

- $\phi(x) = x^2$  convex  $\Rightarrow$
- $\phi(1 - p) < (1 - p)$  and  $(1 - \phi(1 - p)) > p$
- bad consequences are overweighed
- good consequences are underweighed

Pessimism

# Interpretation

## Interpretation: EURDP

- $u$  captures attitude towards consequences
- $\phi$  captures attitude towards risk
- with EU  $u$  plays both rôles

## EURDP

- huge literature on the axiomatic foundations and experimental validity
- **distorting** probabilities
- envisage other models of likelihood

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# Belief functions

## Example

- you know that a proportion  $\alpha_k$  of vehicles are of type  $k$
- you know that a vehicle of type  $k$  has a length between  $\ell_k^-$  and  $\ell_k^+$
- an available slot in which a vehicle of length  $L$  can park

## Probability that a car can park?

- between

$$\sum_{k: \ell_k^+ \leq L} \alpha_k$$

and

$$\sum_{k: \ell_k^- \leq L} \alpha_k$$

# Belief functions

$$\varphi([\ell_k^-, \ell_k^+]) = \alpha_k$$

- inferior probability

$$p^-([0, L]) = \sum_{[\ell_k^-, \ell_k^+] \subseteq [0, L]} \varphi([\ell_k^-, \ell_k^+])$$

- superior probability

$$p^+([0, L]) = \sum_{[\ell_k^-, \ell_k^+] \cap [0, L] \neq \emptyset} \varphi([\ell_k^-, \ell_k^+])$$

- inferior probability and Möbius masses

$$p^-(A) = \sum_{B \subseteq A} \varphi(B) \Leftrightarrow \varphi(A) = \sum_{B \subseteq A} (-1)^{|A \setminus B|} p^-(B)$$



# Belief functions

$$p^-(A) = \sum_{B \subseteq A} \varphi(B) \quad p^+(A) = \sum_{B \cap A \neq \emptyset} \varphi(B)$$

- knowing  $p^-$  is equivalent to knowing  $p^+$

$$p^+(A) = 1 - p^-(\bar{A})$$

- knowing  $p^-$  is equivalent to knowing  $\varphi$

$$\varphi(A) = \sum_{B \subseteq A} (-1)^{|A \setminus B|} p^-(B)$$

# Belief functions

## Definition

- $p^-$ : Belief function (Bel)
- $p^+$ : Plausibility function (Pl)
- $\varphi$ : Möbius masses

- $p^-(\emptyset) = 0, p^-(S) = 1$
- $B \subseteq A \Rightarrow p^-(B) \leq p^-(A)$
- $p^-(A \cup B) \geq p^-(A) + p^-(B) - p^-(A \cap B)$
- $p^-(A \cup B \cup C) \geq p^-(A) + p^-(B) + p^-(C) - p^-(A \cap B) - p^-(A \cap C) - p^-(B \cap C) + p^-(A \cap B \cap C)$
- $p^-$  monotone at all orders

- $\varphi(\emptyset) = 0, \varphi(B) \geq 0$
- $\sum_{B \subseteq S} \varphi(B) = 1$

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# Belief functions

## Probabilities

- $\varphi(A) > 0 \Rightarrow |A| = 1$ : masses only on singletons
- $p^- = p^+$ : probabilities

## Possibilities

- $\varphi(A) > 0$  and  $\varphi(B) > 0 \Rightarrow A \subseteq B$  or  $B \subseteq A$ : embedded masses
- $p^-$ : necessity measure
- $p^+$ : possibility measure

$$Nec(A \cup B) = \min(Nec(A), Nec(B))$$

$$Pos(A \cup B) = \max(Pos(A), Pos(B))$$

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# Ellsberg's Paradox

## Experiment

- 90 balls in an urn: 30 are  $R$  and 60 are  $B$  or  $Y$
- $a1$ : 100 if  $R$  or  $a2$ : 100 if  $B$
- $a3$ : 100 if  $R$  or  $Y$  or  $a4$ : 100 if  $B$  or  $Y$
- model answer:  $a1 \succ a2$  and  $a4 \succ a3$
- incompatible with SEU!

$$E[u(a1)] = P(R)u(100) \quad E[u(a2)] = P(B)u(100)$$

$$a1 \succ a2 \Rightarrow P(R) > P(B)$$

$$E[u(a3)] = P(R)u(100) + P(Y)u(100) \quad E[u(a4)] = P(B)u(100) + P(Y)u(100)$$

$$a4 \succ a3 \Rightarrow P(B) > P(R)$$

Aversion to ambiguity

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$$a4 \succ a3 \Rightarrow P(B) > P(R)$$

Aversion to ambiguity



# Choquet Expected Utility (CEU)

## Capacity

$$C : 2^S \rightarrow [0, 1]$$

$$A \subseteq B \Rightarrow C(A) \leq C(B)$$

$$C(\emptyset) = 0, C(S) = 1$$

Capacity: belief function that is not necessarily monotone

## CEU

$$\begin{aligned} & (x_{(1)}, E_{(1)}; x_{(2)}, E_{(2)}; \dots; x_{(n)}, E_{(n)}) \text{ avec } x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)} \\ CEU = & u(x_{(1)})C(E_{(1)} \cup E_{(2)} \cup \dots \cup E_{(n)}) + \\ & (u(x_{(2)}) - u(x_{(1)}))C(E_{(2)} \cup E_{(3)} \cup \dots \cup E_{(n)}) + \\ & (u(x_{(3)}) - u(x_{(2)}))C(E_{(3)} \cup E_{(4)} \cup \dots \cup E_{(n)}) + \\ & \vdots + \\ & (u(x_{(n)}) - u(x_{(n-1)}))C(E_{(n)}) \end{aligned}$$

# CEU and Ellsberg

- 90 balls in an urn: 30 are  $R$  and 60 are  $B$  or  $Y$
- $a1$ : 100 if  $R$  or  $a2$ : 100 if  $B$
- $a3$ : 100 if  $R$  or  $Y$  or  $a4$ : 100 if  $B$  or  $Y$
- $a1 \succ a2$  and  $a4 \succ a3$

$$a1 \succ a2 \Rightarrow C(R) > C(B)$$

$$a4 \succ a3 \Rightarrow C(B \cup Y) > C(R \cup B)$$

## Particular cases

- $A \cap B = \emptyset \Rightarrow C(A) + C(B) = C(A \cup B)$ : Probability
- $C(A \cap B) = \max(C(A), C(B))$ : Possibility
- $C(A \cap B) = f(C(A), C(B))$ : Decomposable Capacity
- $C$  monotone: Belief functions

# Capacity

## Möbius masses

$$\varphi(A) = \sum_{B \subseteq A} -1^{|A \setminus B|} C(B)$$

## Ellsberg

	$\emptyset$	$R$	$B$	$Y$	$R \cup B$	$R \cup Y$	$B \cup Y$	$S$
$C$	0	1/3	0	0	1/3	1/3	2/3	1
$\varphi$	0	1/3	0	0	0	0	2/3	0

# Extensions

## Cumulative prospect theory

- rank dependence
- sign dependence: reference effects

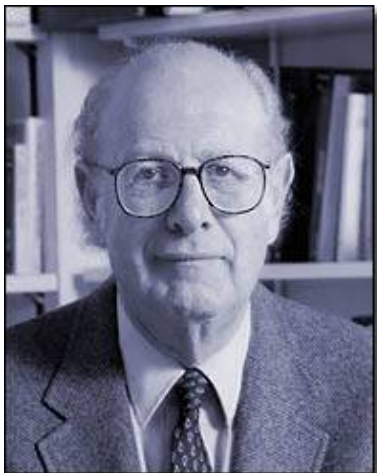
Bruno de Finetti



Leonard J. Savage



Howard Raiffa










Daniel Kahneman



# References

-  Gilboa, I. (2009)  
Theory of Decision under Uncertainty  
Cambridge University Press.
-  Kahneman, D. & Tversky, A. (Eds) (2000)  
Choices, Values and Frames  
Cambridge University Press.
-  Raiffa, H. (1968)  
Decision analysis: Introductory lectures on choices under uncertainty  
Addison-Wesley.
-  Wakker, P. P. (1989)  
Additive representations of preferences  
Kluwer, Dordrecht.
-  Wakker, P. P. (2010)  
Prospect Theory for risk and ambiguity  
Cambridge University Press.

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