Aalto University School of Science

## MS-E2135 Decision Analysis Lecture 2

- Biases in probability assessment
- Expected Utility Theory (EUT)
- Assessment of utility functions


## Last week

$\square$ Decision trees provide a visual and structured way to modelling sequential decision-making problems which involve uncertainties

- Paths of decisions and random events
$\square$ Probabilities are employed to model uncertainties
- Subjective probabilities can be employed even in the absence of data
$\square$ The elicitation of probabilities may involve subjective judgements


## Different kinds of uncertainties

$\square$ We frequently make statements about uncertainty

- "We will have a white Christmas." $\leftrightarrow$ subjective probability
- "The 100 00o $^{\text {th }}$ decimal of $\pi$ is $6 . " \leftrightarrow$ a fact, the uncertainty lies in the available information
- "I win in a lottery with probability $\leftrightarrow$ frequentist or classical 0,00005." probability interpretation
$\square$ Uncertainties are associated with events with unknown outcome
$\square$ Probabilities provide a quantitative measure of this uncertainty


## Classical interpretation

$\square$ Jacob Bernoulli (1685), Pierre-Simon Laplace (1814)

- Probability = The ratio between (i) the number of possible outcomes defining the event and (ii) the total number of possible outcomes which are assumed to be equally likely

$$
P(A)=\frac{\#(A)}{\#(S)}
$$

$\#(A)=$ Number of possible outcomes favourable to $A$
$\#(S)=$ Total number of possible outcomes


Circular definition: Probability defined in terms of "equally likely"
$\square$ Principle of indifference:

- Each event is defined as a collection of outcomes
- Events are "equally likely" if there is no known reason for predicting the occurrence of one event rather than another
- The probability to get " 6 " when tossing a dice is 1/6


## Frequentist interpretation

Leslie Ellis, mid 19 ${ }^{\text {th }}$ century

- Probability = The relative frequency of trials in which the favourable event occurs as the number of trials approaches infinity

$$
P(A)=\lim _{n \rightarrow \infty} \frac{n(A)}{n}
$$

$n(A)=$ Number of times that $A$ occurs
$n=$ Total number of trials

[ You may determine the probability of getting "heads" by tossing a coin (which may not be fair) a very large number of times

- Yet in many cases repeated trials cannot be carried out
- E.g., will there be a recession if the interest rates are raised by $1 \%$ ?


## Subjective (Bayesian) interpretation

ㅁ Bruno De Finetti (1937)

- Probability = An individual's degree of belief in the occurrence of a particular outcome
- The probability may change e.g. when additional information is received
- The event may have already occurred
- Examples
- "I believe there's a 40 \% chance that we will have a white Christmas"
- "I'm $15 \%$ sure that Martin Luther King was 34 years when he died"


## Biases in probability assessment

Subjective judgements by "ordinary people" and "experts" alike are prone to different kinds of biases

- Cognitive bias: Systematic discrepancy between the 'correct' answer and the respondent's actual answer
- E.g., assessments of conditional probabilities differ from the correct value given by the Bayes' rule
- Motivational biases: judgements are influenced by the desirability or undesirability of events, e.g.
- Overoptimism about success probabilities
- Strategic underestimation of failure probabilities

Some biases can be difficult to correct

## Representativeness bias (cognitive)

- If $x$ fits the description of $A$ well, then $\operatorname{Prob}(x \in A)$ is assumed to be large
. The 'base rate' of $A$ in the population (i.e., the probability of $A$ ) is not taken into account
- Example: You see a very tall man in a bar. Is he more likely to be a professional basketball (BB) player or a teacher?

Professional basketball players
$\Delta 5 \begin{aligned} & \text { Aalto University } \\ & \text { School of Science }\end{aligned}$
Teachers

## Representativeness bias

- What is meant by 'very tall'?
- 195 cm ?
- Assume all BB players are very tall
- The share of Finnish men taller than 195 cm is about $0.3 \%$
- If BB players go to the bar as often as teachers, it is more probable that the very tall man is a teacher, if the share of very tall men exceeds $0.31 \%$
- Your responses: 87,5\% teacher, 12,5\% basketball player


| Height | Males |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\begin{array}{r} 20-29 \\ \text { years } \\ \hline \end{array}$ | $\begin{array}{r} 30-39 \\ \text { years } \\ \hline \end{array}$ | $\begin{array}{r} 40-49 \\ \text { years } \end{array}$ | $\begin{array}{r} 50-59 \\ \text { years } \\ \hline \end{array}$ | $\begin{array}{r} 60-69 \\ \text { years } \\ \hline \end{array}$ | $\begin{array}{r} \hline 70-79 \\ \text { years } \\ \hline \end{array}$ |
| Percent under- $4^{\prime} 10^{\prime \prime}$ | - | - | - | (B) | - | - |
| 4'11" | - | - | - | (B) | (B) | - |
| $5{ }^{\prime}$. | (B) | - | - | (B) | (B) | - |
| 5'1" | (B) | (B) | (B) | (B) | ${ }^{1} 0.4$ | (B) |
| 5'2" | (B) | (B) | (B) | (B) | (B) | (B) |
| $53^{\prime \prime}$ | (B) | ${ }^{1} 3.1$ | 11.9 | (B) | ${ }^{1} 2.3$ | (B) |
| 5'4" | 3.7 | 14.4 | 3.8 | 14.3 | 4.4 | 5.8 |
| $5{ }^{\prime} 5^{\prime \prime}$ | 7.2 | 6.7 | 5.6 | 7.6 | 7.8 | 12.8 |
| $56^{\prime \prime}$ | 11.6 | 13.1 | 9.8 | 12.2 | 14.7 | 23.0 |
| 5'7" | 20.6 | 19.6 | 19.4 | 18.6 | 23.7 | 35.1 |
| 5'8' | 33.1 | 32.2 | 30.3 | 30.3 | 37.7 | 47.7 |
| 5'9" | 42.2 | 45.4 | 40.4 | 41.2 | 50.2 | 60.3 |
| 5'10" | 58.6 | 58.1 | 54.4 | 54.3 | 65.2 | 75.2 |
| 5'11" | 70.7 | 69.4 | 69.6 | 70.0 | 75.0 | 85.8 |
| $6{ }^{\prime}$. | 79.9 | 78.5 | 79.1 | 81.2 | 84.3 | 91.0 |
|  | 89.0 | 89.0 | 87.4 | 91.6 | 93.6 | 94.9 |
| $6^{\prime} 2$ " | 94.1 | 94.0 | 92.5 | 93.7 | 97.8 | 98.6 |
| $6{ }^{\prime \prime} 3^{\prime \prime}$ | 98.3 | 95.8 | 97.7 | 96.6 | 99.9 | 100.0 |
| $6^{\prime \prime} 4^{\prime \prime}$ | 100.0 | 97.6 | 99.0 | 99.5 | 100.0 | 100.0 |
| $6^{\prime} 5$ " | 100.0 | 99.4 | 99.4 | 99.6 | 100.0 | 100.0 |
| 6'6" . . . . . . . | 100.0 | 99.5 | 99.9 | 100.0 | 100.0 | 100.0 |

## Representativeness bias

Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student, she was deeply concerned with issues of discrimination and social justice, and also participated in antinuclear demonstrations.

Please check the most likely alternative:
a. Linda is a bank teller.
b. Linda is a bank teller and active in the feminist movement.

Many choose $b$, although $b c a$ whereby $\mathrm{P}(\mathrm{b})<\mathrm{P}(\mathrm{a})$

- Your responses: $70 \% \mathrm{a}, 30 \% \mathrm{~b}$.


## Bank tellers

## Bank tellers who are active in the feminist movement

## Conservativism bias (cognitive)

I After obtaining some information about an uncertain event, people typically do not adjust their initial probability estimate about this event as much as they should based on Bayes' theorem.

Example: Consider two bags X and Y . Bag X contains 30 white balls and 10 black balls, whereas bag $Y$ contains 30 black balls and 10 white balls. Suppose that you select one of these bags at random, and randomly draw five balls one-by-one by replacing them in the bag after each draw. Suppose you get four white balls and one black. What is the probability that you selected bag X with mainly white balls?

- Typically people answer something between 70-80\%. Yet, the correct probability is 27/28 $\approx 96 \%$.
$\square$ Your responses: mean response 55\%. Many (20\%) answered 50\%.


## Representativeness and conservativism bias - debiasing

Pay attention to the logic of joint and conditional probabilities and Bayes' rule

- Split the task into an assessment of
- The base rates for the event (i.e., prior probability)
- E.g., what are the relative shares of teachers and pro basketball players?
- The likelihood of the data, given the event (i.e., conditional probabilities)
- E.g., what is the relative share of people active in the feminist movement? Is this share roughly the same among bank tellers as it is among the general population or higher/lower?
- What is the likelihood that a male teacher is taller than 195 cm ? How about a pro basketball player?


## Availability bias (cognitive)

- People assess the probability of an event by the ease with which instances or occurences of this event can be brought to mind.
$\square$ Example: In a typical sample of English text, is it more likely that a word starts with the letter K or that K is the third letter?
- Most (nowadays only many?) people think that words beginning with K are more likely, because it is easier to think of words that begin with " $K$ " than words with " $K$ " as the third letter
- Yet, there are twice as many words with $K$ as the third letter
- Your responses: $35 \%$ first letter, $65 \%$ third letter.
- Other examples:
- Due to media sensationalist reporting in the US, the number of violent crimes such as child murders seems to have increased
- Yet, compared to 2000's, 18 times as many children were killed per capita in 1950's and twice as many in 1990's
- Probabilities offlight accidents after the volcanic eruption in Iceland in 2011


## Availability bias - debiasing

$\square$ Conduct probability training
$\square$ Provide concrete counterexamples
$\square$ Provide statistics
$\square$ Still, based on empirical experimental studies, availability bias is difficult to correct

## Anchoring bias (cognitive)

$\square$ When assessing probabilities, respondents may be guided by reference assessments

- Often, the respondent is anchored to the reference assessment

Example: Is the percentage of African countries in the UN
A. Greater or less than 65 ? What is the exact percentage?

- 'Average' answer: Less, 45\%.
- Your responses: Less, median 39\%, mean 39\%.
B. Greater or less than 10 ? What is the exact percentage?
- 'Average' answer: Greater, mean $25 \%$.


## Anchoring bias - debiasing

$\square$ Avoid providing anchors

- But there are contexts where deliberate attempts to influence answers are made (e.g., marketing)
$\square$ Provide multiple and counteranchors
- If you have to provide an anchor, provide several which differ significantly from each other

Use different experts who use different anchors
$\square$ Based on empirical evidence, anchoring bias is difficult to correct

## Hindsight bias

People falsely believe they could have predicted the outcome of an event

- Once the outcome has been observed, the DM may assume that they are the only ones that could have happened and underestimate the uncertainty
U Undermines possibilities for learning from the past
Alerting people to this bias has little effect


## $\square$ How to mitigate:

- Argue against the inevitability of the reported outcome
- Develop alternative descriptions of how the future might have unfolded differently


## Desirability / undesirability of events (motivational)

. People tend to believe that there is a less than $50 \%$ probability that negative outcomes will occur compared with peers

- "I am less likely to develop long-term symptoms even if I catch COVID-19"
- People tend to believe that there is a greater than $50 \%$ probability that positive outcomes will occur compared with peers
- "I am more likely to become a homeowner / have a starting salary of more than 4,500€"
- Earlier responses on owning a home: 40\% (20\%) more likely, $\mathbf{1 2 \%}$ (12\%) less likely, $\mathbf{4 8 \%}$ (68\%) equally likely
- Earlier responses on salary: 23\% (20 \%) more likely, $\mathbf{1 0 \%}$ ( $10 \%$ ) less likely, $\mathbf{6 7 \%}$ (71\%) equally likely
- People tend to underestimate the probability of negative outcomes and overestimate the probability of positive outcomes
- The estimates are not conservative - the actual risks are higher than estimated


## Desirability / undesirability of events debiasing

$\square$ Use multiple experts with alternative points of view
$\square$ Place hypothetical bets against the desired event
$\square$ "Make the respondent think about monetary consequences"
$\square$ Use decomposition and realistic assessment of partial probabilities

- "Split the events"
$\square$ Yet, empirical evidence suggests that motivational biases are often difficult to correct

Further reading: Montibeller, G., and D. von Winterfeldt, 2015. Cognitive and Motivational Biases in Decision and Risk Analysis, Risk Analysis

## Overconfidence (cognitive)

$\square$ People tend to assign overly narrow confidence intervals to their probability estimates

1. Martin Luther King's age at death 39 years
2. Length of the Nile River $\mathbf{6 7 3 8} \mathbf{~ k m}$
3. Number of Countries that are members of OPEC 13
4. Number of Books in the Old Testament 39
5. Diameter of the moon $\mathbf{3 4 7 6} \mathbf{~ k m}$
6. Weight of an empty Boeing $747 \mathbf{1 7 6 9 0 0} \mathbf{~ k g}$
7. Year of Wolfgang Amadeus Mozart's birth 1756
8. Gestation period of an Asian elephant 21.5 months
9. Air distance from London to Tokyo $9590 \mathbf{~ k m}$
10. Depth of the deepest known point in the oceans $\mathbf{1 1 0 3 3} \mathbf{~ m}$

Responses by 25 subjects:
Number of questions in which the true value is within the given $90 \%$ confidence interval


## - There are 10 questions with $90 \%$ confidence intervals

- If the intervals are correct, each answer is within the confidence interval with probability 0.9
- The probability that $n$ estimates are within the intervals is $\binom{10}{n} 0.9^{n} 0.1^{10-n}$
- If the intervals are correct, the probability that at least 3 responses lie outside the intervals is $\sum_{n=3}^{10}\binom{10}{n} 0.1^{n} 0.9^{10-n} \approx 7 \% \rightarrow$ The null hypothesis of not being overconfident can be rejected (at the $5 \%$ confidence level)


## Overconfidence - debiasing

$\square$ Provide probability training
I Start with extreme estimates (low and high)
$\square$ Use fixed values instead of fixed probability in elicitations:

- Do not ask: "What is the GDP growth rate $x$ such that the probability of achieving this rate $x$ or less $x$ is $5 \%$ "
- Instead ask : "With what probability will the GDP growth rate be lower than $-3 \%$ ?"
$\square$ Based on empirical evidence, overconfidence is difficult to correct


## Calibration curves

- People tend to assess probabilities best when they have frequent and concrete feedback
- E.g., US weather forecasters
$\square$ Judged probabilities on $x$-axis
$\square$ Observed frequencies on $y$-axis

Can be used for calibration

- Instead of the judged probability, use the corresponding observed frequency

- E.g., in the C case, the actual tail probabilities are more extreme than the judged ones


## Risky or not (so) risky?

- Which one would you choose:
a) Participate in a lottery in which there is a $50 \%$ chance of getting nothing and a $50 \%$ chance of getting $10000 €$
b) Getting $4000 €$ for sure

. Many choose the certain outcome of $4000 €$, although the expected monetary value in alterantive a) is higher


## Option b) involves less risk

## How to compare risky alternatives?

## $\square$ Last week

- We used decision trees to support decision-making under uncertainty assuming that the DM seeks to maximize expected monetary value
- This is valid if the DM is risk neutral, i.e., indifferent between
- obtaining x for sure and
- a gamble with uncertain payoff $Y$ such that $x=E[Y]$
- Many DMs are risk averse $=$ they prefer obtaining $x$ for sure to a gamble with payoff $Y$ such that $x=\mathrm{E}[Y]$


Expectation = 14500

- We accommodate the DM's risk attitude (=preference over alternatives with uncertain outcomes) in decision models


## Expected utility theory (EUT)

] John von Neumann and Oscar Morgenstern, Theory of Games and Economic Behavior, 1944

- Axioms of rationality for preferences over alternatives with uncertain outcomes
- If the DM follows these axioms, she should prefer the alternative with the highest expected utility
- Elements of EUT
- Set of outcomes and "lotteries"
- Preference relation over lotteries which satisfies four axioms
- Representation of preference relation with expected utility


## EUT: Sets of outcomes and lotteries

- Set of possible outcomes $T$ :
- E.g., revenue $T$ euros / demand $T$
- Set of all possible lotteries $L$ :
- A lottery $f \in L$ associates a probability $f(t) \in[0,1]$ with each possible outcome $t \in T$
- Finite number of outcomes with a positive probability $f(t)>0$
- Probabilities add up to one $\sum_{t} f(t)=1$
- Lotteries are discrete probability mass functions (PMFs) / decision trees with a single chance node
- Deterministic outcomes are modeled as degenerate lotteries


## Degenerate lottery

Decision tree PDF
$\bigcirc 10000$

Lottery
Decision tree
Probability mass function (PMF)


## EUT: Compound lotteries

$\square$ Compound lottery:

- Get lottery $f_{X} \in L$ with probability $\lambda$
- Get lottery $f_{Y} \in L$ with probability $1-\lambda$
$\square$ Compound lottery can be modeled as lottery $f_{Z} \in L$ :

$$
f_{Z}(t)=\lambda f_{X}(t)+(1-\lambda) f_{Y}(t) \forall t \in T \simeq f_{Z}=\lambda f_{X}+(1-\lambda) f_{Y}
$$

$\square$ Example:

- You have a 50-50 chance of getting a ticket to lottery $f_{X} \in L$ or to lottery $f_{Y} \in L$



## Preference relation

$\square$ Let $\succcurlyeq$ be preference relation among lotteries in $L$

- Preference $f_{X} \succcurlyeq f_{Y}: f_{X}$ is at least as preferred as $f_{Y}$
- Strict preference $f_{X}>f_{Y}$ defined as $\neg\left(f_{Y} \succcurlyeq f_{X}\right)$
- Indifference $f_{X} \sim f_{Y}$ defined as $f_{X} \succcurlyeq f_{Y} \wedge f_{Y} \succcurlyeq f_{X}$


## EUT axioms A1-A4 for the relation $\succcurlyeq$

- A1: $\geqslant$ is complete
- For any $f_{X}, f_{Y} \in L$, either $f_{X} \succcurlyeq f_{Y}$ or $f_{Y} \geqslant f_{X}$ or both
- A2: $\succcurlyeq$ is transitive
- If $f_{X} \geqslant f_{Y}$ and $f_{Y} \succcurlyeq f_{Z}$, then $f_{X} \geqslant f_{Z}$
$\square$ A3: Archimedean axiom
- If $f_{X} \succ f_{Y} \succ f_{Z}$, then $\exists \lambda, \mu \in(0,1)$ such that

$$
\lambda f_{X}+(1-\lambda) f_{Z}>f_{Y} \text { and } f_{Y} \succ \mu f_{X}+(1-\mu) f_{Z}
$$

$\square$ A4: Independence axiom

- Let $\lambda \in(0,1)$. Then,

$$
f_{X}>f_{Y} \Leftrightarrow \lambda f_{X}+(1-\lambda) f_{Z}>\lambda f_{Y}+(1-\lambda) f_{Z}
$$

## Equivalent formulations of A3 and A4

- A3: Archimedean axiom
- If $f_{X} \succ f_{Y} \succ f_{Z}$, there then exists $p \in(0,1)$ such that $f_{Y} \sim p f_{X}+(1-p) f_{Z}$
$\square$ A4: Independence axiom
- $f_{X} \sim f_{Y} \Leftrightarrow \lambda f_{X}+(1-\lambda) f_{Z} \sim \lambda f_{Y}+(1-\lambda) f_{Z}$
- Any lottery (or outcome = a degenerate lottery) can be replaced by an equally preferred lottery. By A3, such lotteries / outcomes exist

- NOTE: $f_{Z}$ can be any lottery, it can have several possible outcomes


## Main representation theorem for expected utility

$\square \geqslant$ satisfies axioms A1-A4 if and only if there exists a real-valued utility function $u(t)$ over the set of outcomes $T$ such that

$$
f_{X} \succcurlyeq f_{Y} \Leftrightarrow \sum_{t \in T} f_{X}(t) u(t) \geq \sum_{t \in T} f_{Y}(t) u(t)
$$

Implication: a rational DM following axioms A1-A4 selects the alternative with the highest expected utility

$$
E[u(X)]=\sum_{t \in T} f_{X}(t) u(t)
$$

- A similar result can be obtained for continuous distributions:
- $f_{X} \succcurlyeq f_{Y} \Leftrightarrow E[u(X)] \geq E[u(Y)]$, where $E[u(X)]=\int f_{X}(t) u(t) d t$


## Computing expected utility

Example: Joe's utility function for the number of apples is $u(1)=2, u(2)=5, u(3)=7$.

$$
E[u(X)]=u(2)=5
$$ Which alternative would he prefer?

$$
\begin{aligned}
& E[u(Y)]=0.5 u(1)+0.5 u(3) \\
& \quad=0.5 \cdot 2+0.5 \cdot 7=4.5
\end{aligned}
$$

- X: Two apples for certain
- Y: A 50-50 gamble between 1 and 3 apples
- Example: Jane's utility function for money is $u(t)=t^{2}$. Which alternative would she prefer?

$$
\begin{gathered}
E[u(X)]=0.5 u(3)+0.5 u(5) \\
\quad=0.5 \cdot 9+0.5 \cdot 25=17
\end{gathered}
$$

- X: 50-50 gamble between $3 \mathrm{M} €$ and $5 \mathrm{M} €$
- Y: A random amount of money from the uniform distribution over the interval [3,5]
- What if her utility function was $u(t)=\frac{t^{2}-9}{25-9}$ ?

$$
\begin{gathered}
E[u(Y)]=\int_{3}^{5} f_{Y}(t) u(t) d t=\int_{3}^{5} \frac{1}{2} t^{2} d t \\
=\frac{1}{6} 5^{3}-\frac{1}{6} 3^{3}=16.33333
\end{gathered}
$$

## Uniqueness up to positive affine transformations

$\square$ Let $f_{X} \succcurlyeq f_{Y} \Leftrightarrow E[u(X)] \geq E[u(Y)]$. Then $E[\alpha u(X)+\beta]=\alpha E[u(X)]+\beta \geq$ $\alpha E[u(Y)]+\beta=E[\alpha u(Y)+\beta]$ for any $\alpha>0$ and arbitrary $\beta$
$\square$ Two utility functions $u_{1}(t)$ and $u_{2}(t)=\alpha u_{1}(t)+\beta,(\alpha>0)$ establish the same preference order over lotteries

$$
E\left[u_{2}(X)\right]=E\left[\alpha u_{1}(X)+\beta\right]=\alpha E\left[u_{1}(X)\right]+\beta .
$$

$\square$ Implications

- Any linear utility function $u_{L}(t)=\alpha t+\beta,(\alpha>0)$ that is a positive affine transformation of the identity function $u_{1}(t)=t \Rightarrow u_{L}(t)$ establishes the same preference order as the expected value
- Utilities for two outcomes can be chosen freely:
- E.g., if utilities are represented by $u_{1}$, the normalized utility such that $u_{2}\left(t^{*}\right)=1$ and $u_{2}\left(t^{0}\right)=0$ can be derived through

$$
u_{2}(t)=\frac{u_{1}(t)-u_{1}\left(t^{0}\right)}{u_{1}\left(t^{*}\right)-u_{1}\left(t^{0}\right)}=\underbrace{\frac{1}{u_{1}\left(t^{*}\right)-u_{1}\left(t^{0}\right)}}_{=\alpha>0} u_{1}(t)-\frac{u_{1}\left(t^{0}\right)}{u_{1}\left(t^{*}\right)-u_{1}\left(t^{0}\right)}
$$

## Let's oractice! https://presemo.aalto.fi/drcuckoo

The utility function of Dr. Cuckoo is $u(t)=\sqrt{ } t$. Would he
a) Participate in a lottery A with 50-50 chance of getting either 0 or $400 €$ ?
b) Participate in a lottery $B$ in which the probability of getting $\mathbf{9 0 0} \boldsymbol{€}$ is $\mathbf{3 0 \%}$ and getting $\mathbf{0} \boldsymbol{€}$ is $\mathbf{7 0 \%}$ ?
$u(0)=0, u(400)=20, u(900)=30$
a) $E[u(A)]=0.5 \cdot 0+0.5 \cdot 20=10$
b) $E[u(B)]=0.7 \cdot 0+0.3 \cdot 30=9$

NOTE! The expectation of lottery A = $200 €$ is smaller than that of B = 270€

## Reference lottery revisited

$\square$ Assume that an expected utility maximizer with utility function $u$ uses a reference lottery to assess the probability of event $A$
$\square$ She thus adjusts $p$ such that she is indifferent between lottery X and the reference lottery Y

$$
E[u(X)]=E[u(Y)]
$$



$$
\begin{gathered}
\Leftrightarrow P(A) u\left(t^{+}\right)+(1-P(A)) u\left(t^{-}\right)=p u\left(t^{+}\right)+(1-p) u\left(t^{-}\right) \\
\Leftrightarrow P(A)\left(u\left(t^{+}\right)-u\left(t^{-}\right)\right)=p\left(u\left(t^{+}\right)-u\left(t^{-}\right)\right) \\
\Leftrightarrow P(A)=p
\end{gathered}
$$

$\square$ The utility function $u$ does not affect the result

## Expected utility in decision trees

- Carry out everything as before, except:
- Chance node: compute the expected utility
- Decision node: select the alternative corresponding to maximum expected utility
- Cf. the umbrella example, in which the 'magic numbers' represented preferences


$$
u(t)=2-e^{\frac{-t}{1000}}
$$

## Expected utility in Monte Carlo

$\times \vee f x=2-\operatorname{EXP}(-\mathrm{F} 12 / 1000)$
$\square$ Generate a sample $x_{1}, \ldots, x_{n}$ of realizations from the probability density function
$\square$ Comput corresponding utilities for $u\left(x_{i}\right)$ for each $x_{i}$
$\square$ Mean of the sample utilities $u\left(x_{1}\right), \ldots, u\left(x_{n}\right)$ provides an estimate for $E[u(X)]$

|  | C | D | E | F | G | H |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  | , |  |
|  |  |  | Col.mean | Col.mear | Col.mean |  |
|  |  |  | 0.502964 | 990.3014 | 1.580972 |  |
|  |  |  |  |  |  |  |
|  |  | Sample | u | x | Utility |  |
|  |  | 1 | 0.464077 | 954.9167 | 1.615156 |  |
| 1 |  | 2 | 0.704234 | 1268.308 | 1.718693 |  |
|  |  | 3 | 0.777865 | 1382.501 | 1.74905 |  |
|  |  | 4 | 0.534927 | 1043.831 | 1.647897 |  |
|  |  | 5 | 0.4426 | 927.8094 | 1.604581 |  |
|  |  | 6 | 0.916252 | 1690.147 | 1.815508 |  |
|  |  | 7 | 0.649453 | 1191.922 | 1.696363 |  |
|  |  | 8 | 0.65278 | 1196.418 | 1.697725 |  |
|  |  | 9 | 0.110887 | 389.0874 | 1.322325 |  |
|  |  | 10 | 0.189275 | 559.714 | 1.428628 |  |
|  |  | 11 | 0.902882 | 1649.073 | 1.807772 |  |

## Summary

$\square$ Probability elicitation is prone to cognitive and motivational biases

- Some cognitive biases can be easy to correct, but...
- Some other cognitive biases and all motivational biases can be difficult to overcome
- The DM's preferences over alternatives with uncertain outcomes can be described by a utility function

A rational DM (according to the four axioms of rationality) should choose the alternative with the highest expected utility
$\square$ This is NOT necessarily the alternative for which the utility associated with the expected monetary consequences is highest

## EUT for normative decision support

$\square$ EUT is a normative theory: if the DM is rational (as defined by the axioms), she should select the alternative with the highest expected utility

- Not descriptive or predictive: EUT does not describe or predict how people actually do select among alternatives with uncertain outcomes
$\square$ The four axioms characterize properties that can be associated with rational decision makers
- E.g., if the transivity axiom A2 is violated so that $f_{X}>f_{Y}, f_{Y}>f_{Z}, f_{Z}>f_{X}$, one would be willing to pay in order exchange $f_{X}$ for $f_{Z}$, then $f_{Z}$ for $f_{Y}$ and finally $f_{Y}$ for $f_{X}$, thus becoming a "money pump"
- If these rationality axioms are accepted, then the DM should abide by them

Question 1

Which of the following alternatives would you choose?

1. A sure gain of $1 \mathrm{M} €$
2. A gamble in which there is a

- $1 \%$ probability of getting nothing,
- $89 \%$ probability of getting $1 M €$, and
- $10 \%$ probability of getting $5 \mathrm{M} €$
$\square$ A rare disease breaks out in a community, killing as many as 600 people. Which one of the following two programs for addressing the threat would you choose:
- Program A: 200 people will be saved for sure.
- Program B: There is a $33 \%$ probability that all 600 will be saved and a $67 \%$ probability that no one will be saved.
Which program will you choose?

1. Program A
2. Program B
$\square$ Which of the below alternatives would you choose?
3. A lottery in which there is a

- $89 \%$ probability of getting nothing
- $11 \%$ probability of getting $1 M €$

2. A lottery gamble in which there is a

- $90 \%$ probability of getting nothing
- $10 \%$ probability of getting $5 \mathrm{M} €$
$\square$ Imagine that a rare disease is breaking out in some community and is expected to kill 600 people. Two different programs are available to deal with the threat.
- Program C: 400 of the 600 people will die.
- Program D: There is a $33 \%$ probability that nobody will die and a $67 \%$ probability that 600 people will die.
Which program will you choose?

1. Program C
2. Program D
