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RATIONAL DECISIONS

CMSC 421: CHAPTER 16

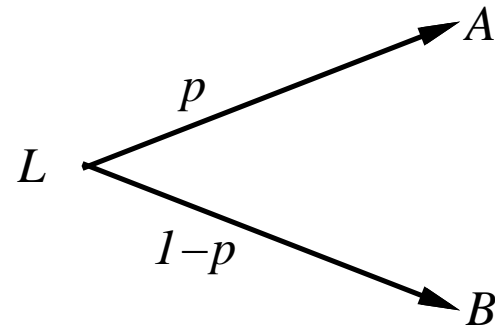
Outline

- ◇ Rational preferences
- ◇ Utilities
- ◇ Money
- ◇ Multiattribute utilities
- ◇ Decision networks
- ◇ Value of information

Preferences

An agent chooses among *prizes* (A , B , etc.) and *lotteries*, i.e., situations with uncertain prizes

Lottery $L = [p, A; (1 - p), B]$



Notation:

- $A \succ B$ A preferred to B
- $A \sim B$ indifference between A and B
- $A \not\succeq B$ B not preferred to A

Rational preferences

Idea: preferences of a rational agent must obey constraints.

Rational preferences \Rightarrow

behavior describable as maximization of expected utility

Constraints:

Orderability:

$$(A \succ B) \vee (B \succ A) \vee (A \sim B)$$

Transitivity:

$$(A \succ B) \wedge (B \succ C) \Rightarrow (A \succ C)$$

Continuity:

$$A \succ B \succ C \Rightarrow \exists p [p, A; 1 - p, C] \sim B$$

Substitutability:

$$A \sim B \Rightarrow [p, A; 1 - p, C] \sim [p, B; 1 - p, C]$$

Monotonicity:

$$A \succ B \Rightarrow (p \geq q \Leftrightarrow [p, A; 1 - p, B] \succsim [q, A; 1 - q, B])$$

Rational preferences contd.

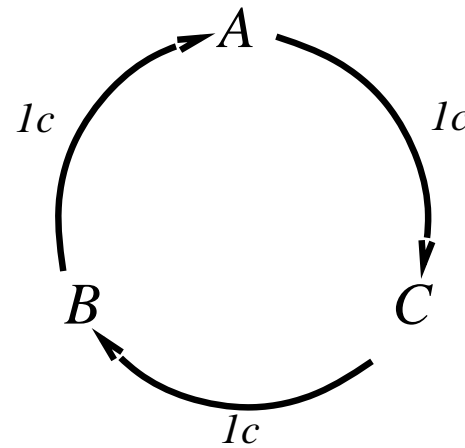
What happens if an agent's preferences violate the constraints?

Example: intransitive preferences

If $B \succ C$, then an agent who has C would trade C plus some money to get B

If $A \succ B$, then an agent who has B would trade B plus some money to get A

If $C \succ A$, then an agent who has A would trade A plus some money to get C



Rational preferences contd.

What happens if an agent's preferences violate the constraints?

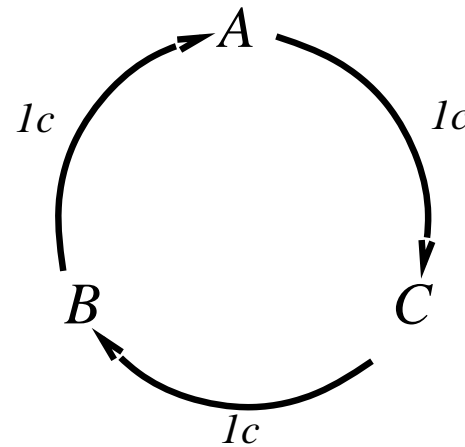
It leads to self-evident irrationality

Example: intransitive preferences

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If $A \succ B$, then an agent who has B would trade B plus some money to get A

If $C \succ A$, then an agent who has A would trade A plus some money to get C



An agent with intransitive preferences can be induced to give away all its money

Maximizing expected utility

Theorem (Ramsey, 1931; von Neumann and Morgenstern, 1944):

Given preferences satisfying the constraints,
there exists a real-valued function U such that

$$U(A) \geq U(B) \Leftrightarrow A \succsim B$$
$$U([p_1, S_1; \dots ; p_n, S_n]) = \sum_i p_i U(S_i)$$

MEU principle:

Choose the action that maximizes the expected utility

Note: an agent can maximize the expected utility without ever representing or manipulating utilities and probabilities

E.g., a lookup table to play tic-tac-toe perfectly

Human utilities

Utilities map states to real numbers. Which numbers?

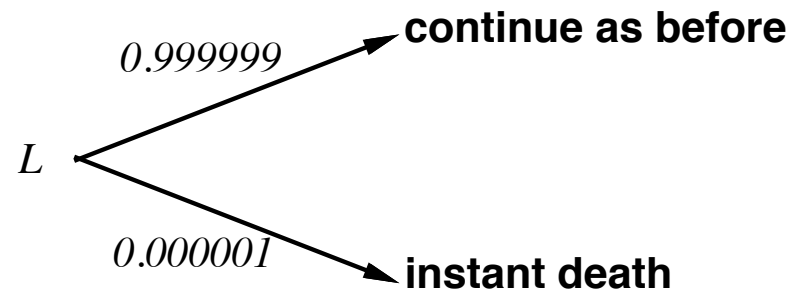
Standard approach to assessing human utilities:

Compare a given state A to a *standard lottery* L_p that has

- “best possible prize” u_{\max} with probability p
- “worst possible catastrophe” u_{\min} with probability $(1 - p)$

Adjust lottery probability p until $A \sim L_p$

**How much
would you pay**
to avoid a
1/1,000,000 chance of death?



Human utilities

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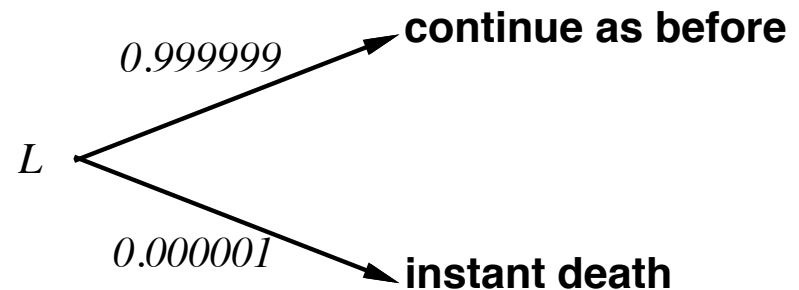
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Judging from people’s **actions**,
they will pay about
\$20 to avoid a
1/1,000,000 chance of death



↖ One *micromort*

$\approx P(\text{accidental death in 370 km of car travel})$

$\approx P(\text{accidental death in 9700 km of train travel})$

Utility scales

Note: behavior is **invariant** w.r.t. positive linear transformation

Let

$$U'(x) = k_1 U(x) + k_2 \quad \text{where } k_1 > 0$$

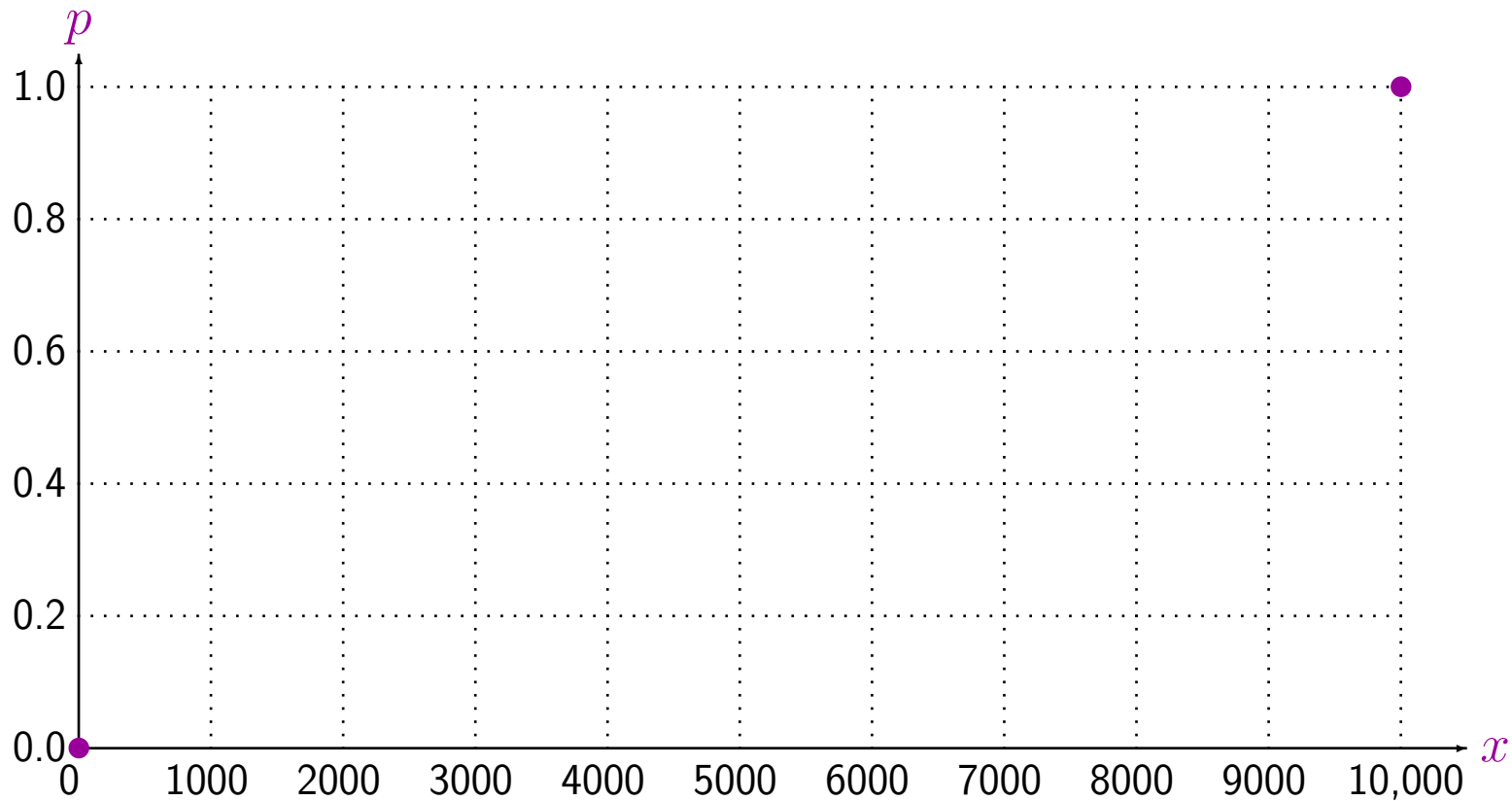
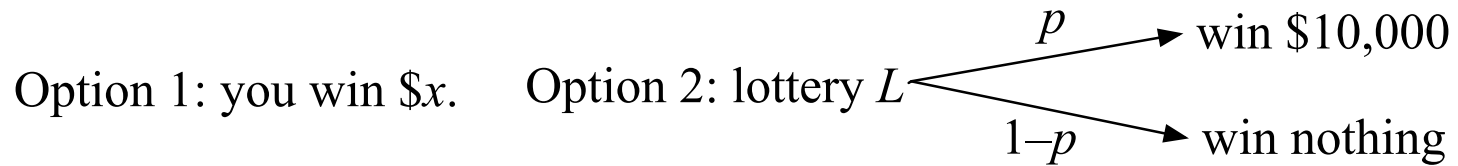
Then U' models the same preferences that U does.

Normalized utilities:

define U' such that $0 \leq U'(x) \leq 1$ for all x

The utility of money

For each amount x , adjust p until half the class votes for each option:



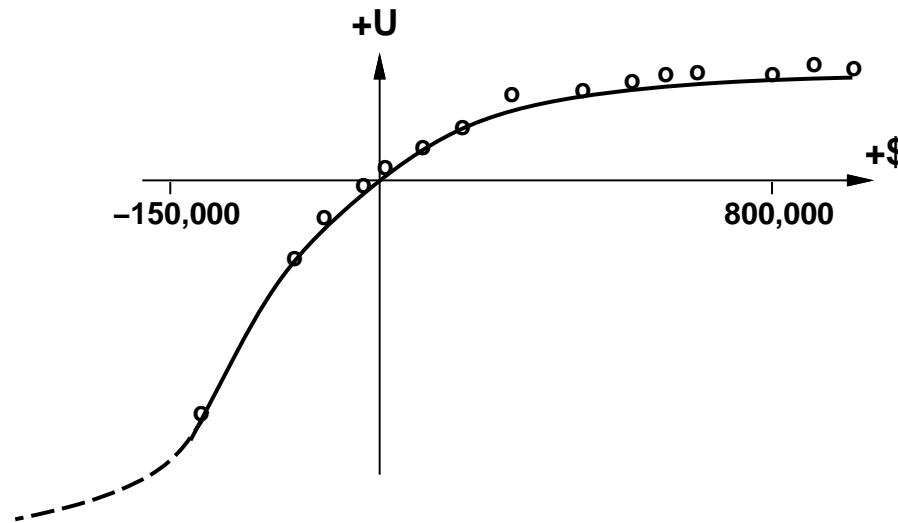
What the book says

Money does **not** behave as a utility function

Given a lottery L with expected monetary value $EMV(L)$, usually $U(L) < U(EMV(L))$, i.e., people are *risk-averse*

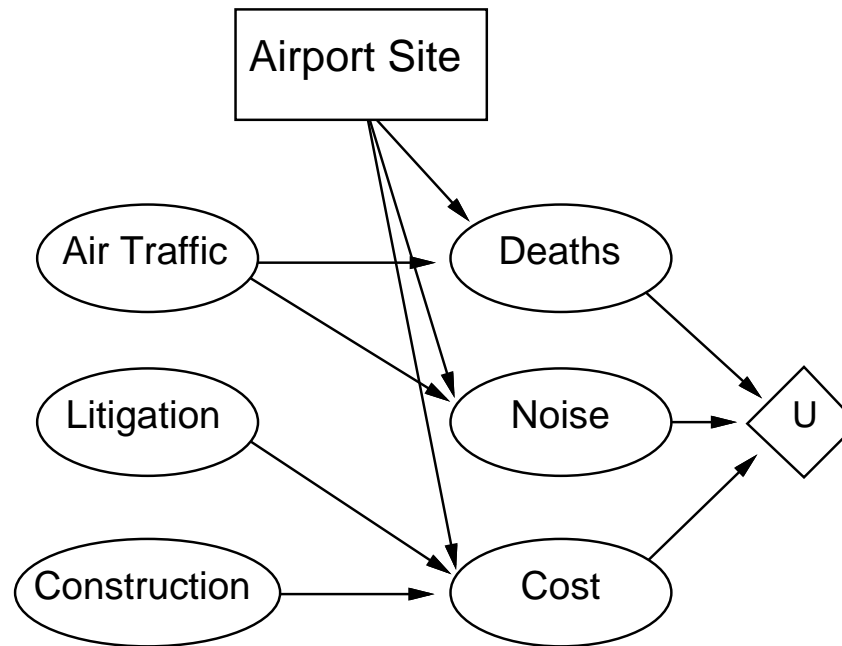
Utility curve: for what probability p am I indifferent between a prize x and a lottery $[p, \$M; (1 - p), \$0]$ for large M ?

Typical empirical data, extrapolated with *risk-prone* behavior:



Decision networks

Add *action nodes* and *utility nodes* to causal networks to enable rational decision making



Algorithm:

For every possible value of the action node
compute $E(\text{utility node} \mid \text{action, evidence})$
Return MEU action

Multiattribute utility

How can we handle utility functions of many variables $X_1 \dots X_n$?
E.g., what is $U(\text{Deaths}, \text{Noise}, \text{Cost})$?

How can complex utility functions be assessed from preference behavior?

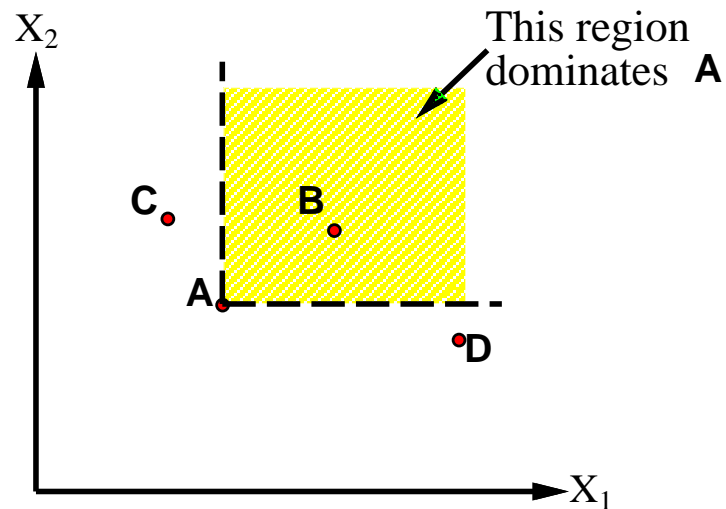
Idea 1: identify conditions (e.g., **dominance**) under which decisions can be made without complete identification of $U(x_1, \dots, x_n)$

Idea 2: identify various types of **independence** in preferences and derive consequent canonical forms for $U(x_1, \dots, x_n)$

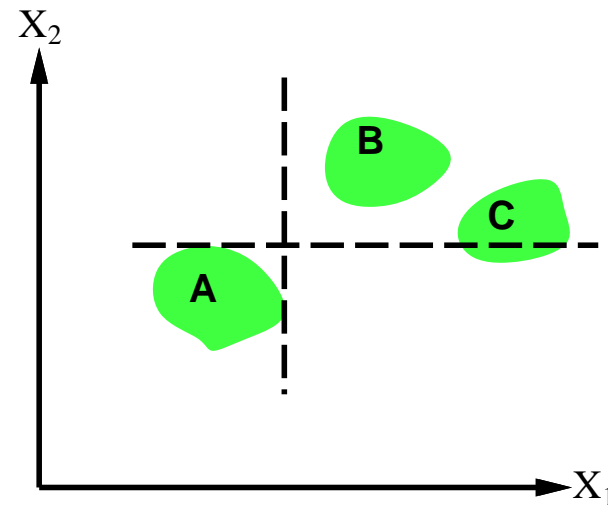
Strict dominance

Typically define attributes such that U is *monotonic* in each attribute

Strict dominance: choice B strictly dominates choice A iff
 $\forall i X_i(B) \geq X_i(A)$ (and hence $U(B) \geq U(A)$)



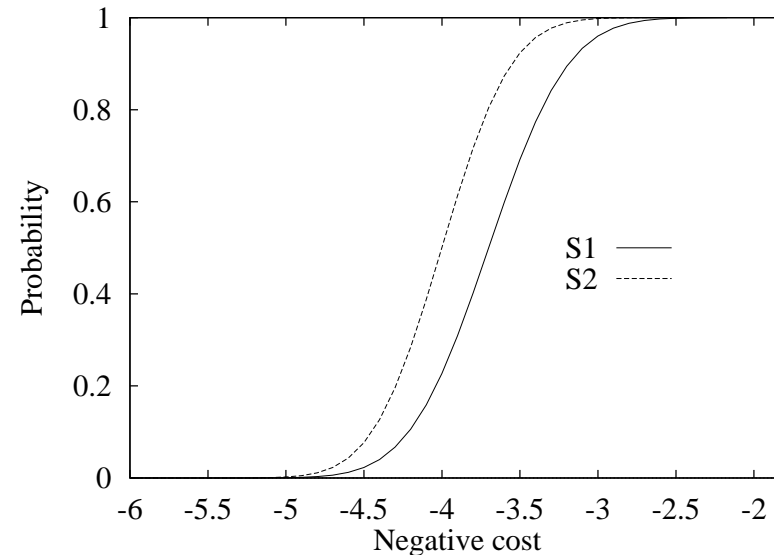
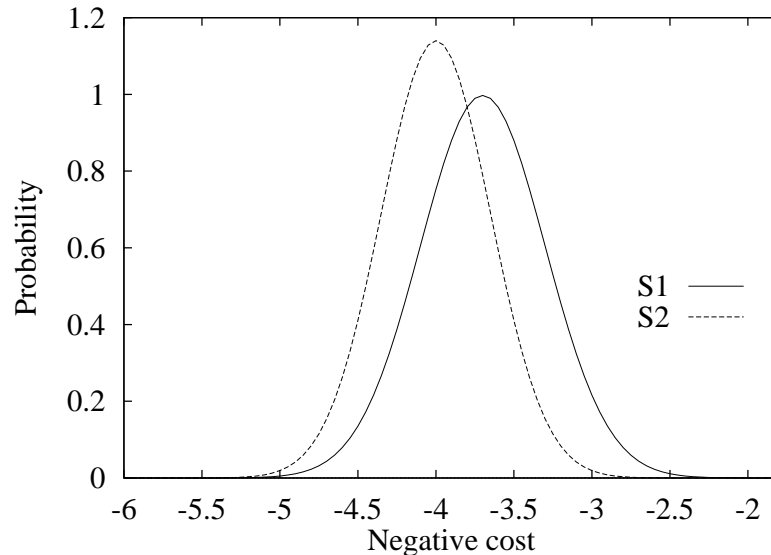
Deterministic attributes



Uncertain attributes

Strict dominance seldom holds in practice

Stochastic dominance



Choices S_1 and S_2 with continuous distributions p_1 and p_2

S_1 stochastically dominates S_2 iff $\forall t \ P(S_1 \leq t) \leq P(S_2 \leq t)$,
 i.e., $\forall t \ \int_{-\infty}^t p_1(x)dx \leq \int_{-\infty}^t p_2(t)dt$

If S_1 stochastically dominates S_2 and U is monotonic in x , then

$$EU(S_1) = \int_{-\infty}^{\infty} p_1(x)U(x)dx \geq \int_{-\infty}^{\infty} p_2(x)U(x)dx = EU(S_2)$$

If p_1, p_2 are discrete, use sums instead of integrals

Multiattribute case: stochastic dominance on all attributes \Rightarrow optimal

Stochastic dominance contd.

Stochastic dominance can often be determined without exact distributions using **qualitative** reasoning

E.g., construction cost increases with distance from city

S_1 is closer to the city than S_2

$\Rightarrow S_1$ stochastically dominates S_2 on cost

E.g., injury increases with collision speed

Can annotate belief networks with stochastic dominance information:

$X \xrightarrow{+} Y$ (X positively influences Y) means that

For every value \mathbf{z} of Y 's other parents \mathbf{Z}

$\forall x_1, x_2 \quad x_1 \geq x_2 \Rightarrow \mathbf{P}(Y|x_1, \mathbf{z})$ stochastically dominates $\mathbf{P}(Y|x_2, \mathbf{z})$

Preference structure: Deterministic

X_1 and X_2 *preferentially independent* of X_3 iff
preference between $\langle x_1, x_2, x_3 \rangle$ and $\langle x'_1, x'_2, x_3 \rangle$
does not depend on x_3

E.g., $\langle \text{Noise, Cost, Safety} \rangle$:

$\langle 20,000 \text{ suffer, } \$4.6 \text{ billion, } 0.06 \text{ deaths/mpm} \rangle$ vs.
 $\langle 70,000 \text{ suffer, } \$4.2 \text{ billion, } 0.06 \text{ deaths/mpm} \rangle$

Theorem (Leontief, 1947): if every pair of attributes is P.I. of its complement, then every set of attributes is P.I. of its complement: *mutual P.I.*

Theorem (Debreu, 1960): mutual P.I. $\Rightarrow \exists$ *additive* value function:
If the attributes of S are X_1, X_2, \dots, X_n , then

$$V(S) = \sum_i V_i(X_i(S))$$

Hence assess n single-attribute functions; often a good approximation

Preference structure: Stochastic

Need to consider preferences over lotteries:

\mathbf{X} is *utility-independent* of \mathbf{Y} iff

preferences over lotteries in \mathbf{X} do not depend on \mathbf{y}

The lotteries in $\mathbf{X} = \{X_1, \dots, X_k\}$ are mutually U.I. if every subset of \mathbf{X} is U.I. of its complement

$\Rightarrow \exists$ *multiplicative* utility function:

$$\begin{aligned} U &= k_1U_1 + k_2U_2 + k_3U_3 \\ &+ k_1k_2U_1U_2 + k_2k_3U_2U_3 + k_3k_1U_3U_1 \\ &+ k_1k_2k_3U_1U_2U_3 \end{aligned}$$

Routine procedures and software packages for generating preference tests to identify various canonical families of utility functions

Value of information

Idea: compute value of acquiring each possible piece of evidence
Can be done **directly from decision network**

Example: buying oil drilling rights

Two blocks A and B , exactly one has oil, worth k

Prior probabilities 0.5 each, mutually exclusive

Current price of each block is $k/2$

How much to pay a consultant for an accurate survey of A ?

Solution: compute expected value of information

= expected value of best action given the information

minus expected value of best action without information

Survey may say “oil in A ” or “no oil in A ”, **prob. 0.5 each** (from above)

$$\begin{aligned} &= [0.5 \times \text{value of “buy } A \text{” given “oil in } A \text{”} \\ &\quad + 0.5 \times \text{value of “buy } B \text{” given “no oil in } A \text{”}] \\ &\quad - 0 \end{aligned}$$

$$= (0.5 \times k/2) + (0.5 \times k/2) - 0 = k/2$$

General formula

Current evidence E , current best action α

Possible action outcomes $\{S_1, S_2, \dots\}$

$$EU(\alpha|E) = \max_a EU(a|E) = \max_a \sum_i U(S_i) P(S_i|E, a)$$

Potential new evidence E_j

If we knew $E_j = e_j$, then we would choose α_{e_j} s.t.

$$EU(\alpha_{e_j}|E, E_j = e_j) = \max_a \sum_i U(S_i) P(S_i|E, a, E_j = e_j)$$

E_j is a random variable whose value is currently unknown

\Rightarrow must compute expected gain over all possible values:

$$VPI_E(E_j) = \left(\sum_k P(E_j = e_j|E) EU(\alpha_{e_j}|E, E_j = e_j) \right) - EU(\alpha|E)$$

(VPI = value of perfect information)

Properties of VPI

Nonnegative—in **expectation**, not **post hoc**

$$\forall j, E \quad VPI_E(E_j) \geq 0$$

Nonadditive—consider, e.g., obtaining E_j twice

$$VPI_E(E_j, E_k) \neq VPI_E(E_j) + VPI_E(E_k)$$

Order-independent

$$VPI_E(E_j, E_k) = VPI_E(E_j) + VPI_{E,E_j}(E_k) = VPI_E(E_k) + VPI_{E,E_k}(E_j)$$

Note: when more than one piece of evidence can be gathered, greedy selection (next one to gather = the one of maximum VPI) isn't always optimal

Can have situations where

$$VPI(E_1|E) > VPI(E'_1|E) \text{ and } VPI(E_2|E, E_1) > VPI(E'_2|E, E_1)$$

but $VPI(E_1, E_2|E) < VPI(E'_1, E'_2|E)$

⇒ evidence-gathering becomes a **sequential** decision problem

Qualitative behaviors

- a) Choice is obvious, information worth little
- b) Choice is nonobvious, information worth a lot
- c) Choice is nonobvious, information worth little

