

Decision Theory

CPS 270
Ronald Parr

Decision Theory

What does it mean to make an optimal decision?

- Asked by economists to study consumer behavior
- Asked by MBAs to maximize profit
- Asked by leaders to allocate resources
- Asked in OR to maximize efficiency of operations
- Asked in AI to model intelligence

- Asked (sort of) by any intelligent person every day

Utility Functions

- A *utility function* is a mapping from world states to real numbers
- Sometimes called a *value function*
- Rational or optimal behavior is typically viewed as maximizing expected utility:

$$\max_a \sum_s P(s | a)U(s)$$

a = actions, s = states

Are Utility Functions Natural?

- Some have argued that people don't really have utility functions
 - What is the utility of the current state?
 - What was your utility at 8:00pm last night?
 - *Utility elicitation* is difficult problem
- It's easy to communicate *preferences*
- Theorem (sorta): Given a plausible set of assumptions about your preferences, there must exist a consistent utility function

Axioms of Utility Theory

- **Orderability:** $(A \succ B) \vee (A \prec B) \vee (A \sim B)$
- **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:** $A \succ B \Rightarrow (p \geq q \Leftrightarrow [p, A; 1-p, B] \geq [q, A; 1-q, B])$
- **Decomposability:**
 $[p, A; (1-p), [q, B; (1-q), C]] \sim [p, A; (1-p)q, B; (1-p)(1-q), C]$

Consequences of Preference Axioms

- **Utility Principle**
 - There exists a real-valued function U:
$$U(A) > U(B) \Leftrightarrow A \succ B$$
$$U(A) = U(B) \Leftrightarrow A \sim B$$
- **Expected Utility Principle**
 - The utility of a lottery can be calculated as:

$$U([p_1, S_1; \dots; p_n, S_n]) = \sum_i p_i U(S_i)$$

More Consequences

- Scale invariance
- Shift invariance

Maximizing Utility

- Suppose you want to be famous
- You can be either (N,M,C)
 - Nobody
 - Modestly Famous
 - Celebrity
- Your utility function:
 - $U(N) = 20$
 - $U(M) = 50$
 - $U(C) = 100$
- You have to decide between going to grad school and becoming a professor (G) or going to Hollywood and becoming an actor (A)

Outcome Probabilities

- $P(N|G)=0.5, P(M|G)=0.4, P(C|G)=0.1$
- $P(N|H)=0.6, P(M|H)=0.2, P(C|H)=0.2$
- Maximize expected utility:
 - $U(N) = 20, U(M) = 50, U(C) = 100$

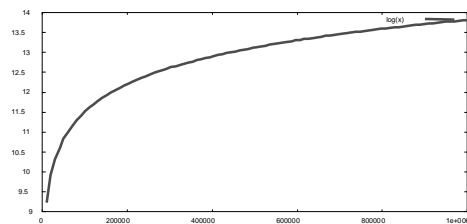
$$EU_G = 0.5(20) + 0.4(50) + 0.1(100) = 40$$

$$EU_H = 0.6(20) + 0.2(50) + 0.2(100) = 42$$

Hollywood wins!

Utility of Money

- How much happier are you with an extra \$1M?
- How much happier is Bill Gates with an extra \$1M?
- Some have proposed:

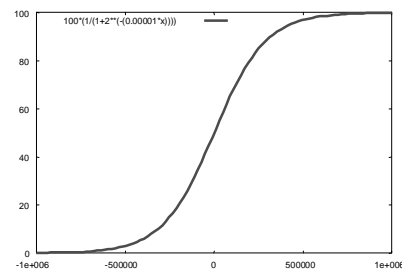


Utility of Money

- $U(\text{money})$ should drop slowly in negative region too
- If you're solvent, losing \$1M is pretty bad
- If already \$10M in debt, losing another \$1M isn't that bad
- Utility of money is probably sigmoidal

A Sigmoidal Utility Function

$$U(\$X) = 100 \frac{1}{1 + 2^{0.00001X}}$$



Utility & Gambling

- Suppose $U(\$X)=X$, would you spend \$1 for a 1 in a million chance of winning \$1M?
- Suppose you start with c dollars:
 - $EU(\text{gamble})=1/1000000(1000000-1+c)+(1-1/1000000)(c-1)=c$
 - $EU(\text{do_nothing})=c$
- Starting amount doesn't matter
- You have no expected benefit from gambling

Sigmoidal Utility & Gambling

- Suppose: $U(\$X) = 100 \frac{1}{1 + 2^{-0.00001X}}$
- Suppose you start with \$1M
 - $EU(\text{gamble}) - EU(\text{do_nothing}) = -5.7 * 10^{-7}$
 - Winning is worthless
- Suppose you start with -\$1M
 - $EU(\text{gamble}) - EU(\text{do_nothing}) = +4.9 * 10^{-5}$
 - Gambling is rational because losing doesn't hurt

Additive Independence

- Suppose it makes me happy to have my car clean
- Suppose it makes me happy to have coffee
- $U = U(\text{coffee}) + U(\text{clean})$
- It seems that these don't interact
- However, suppose there's a tea variable
- $U = U(\text{coffee}) + U(\text{tea}) + U(\text{clean})$???
- Probably not. I'd need $U(\text{coffee,tea}) + U(\text{clean})$
- Often implicit!

Value of Information

- Expected utility of action a with evidence E :

$$EU_E(A | E) = \max_a \sum_i P(S_i | E, a) U(S_i)$$

- Expected utility given new evidence E'

$$EU_{E,E'}(A | E, E') = \max_a \sum_i P(S_i | E, E', a) U(S_i)$$

- Value of knowing E' (value of perfect information)

$$VPI_{E'}(E') = \left(\sum_{E'} P(E' | E) EU_{E,E'}(a | E, E') \right) - EU(a | E)$$

Expected utility given
New information
(weighted)
Previous
Expected
utility

Properties of VOI

- VOI is non-negative!
- VOI is order independent
- VOI is not additive
- VOI is easy to compute and is often used to determine how much you should pay for one extra piece of information. Why is this myopic?

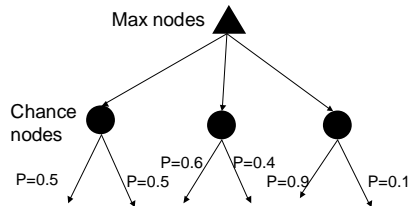
For example, knowing X AND Y together may be useful, while knowing just one alone may be useless.

More Properties of VOI

- Acquiring information optimally is very difficult
- Need to construct a conditional plan for every possible outcome before you ask for even the first piece of information
 - Suppose you're a doctor planning to treat a patient
 - Picking the optimal test to do first requires that you consider all subsequent tests and all possible treatments as a result of these tests
- General versions of this problem are intractable!

Decision Theory as Search

- Can view DT probs as search probs
- States = atomic events



DT as Search

- Attach costs to arcs, leaves
- Path(s) w/lowest expected cost = optimal
- Minimizing expect cost = maximizing expected utility
- Expectimax:

$$V(n_{\max}) = \max_{s \in \text{successors}(n)} V(s)$$

$$V(n_{\text{chance}}) = \sum_{s \in \text{successors}(n)} V(s)p(s)$$

The Form of DT Solutions

- The solution to a DT problem with many steps isn't linear in the number of steps. (Why?)
- What does this say about computational costs?
- What does this say about the hope for exploiting heuristics?

Conclusions

- Decision theory provides a framework for optimal decision making
- Principle: Maximize Expected Utility
- Easy to describe in principle
- Application to complex problems can require advanced planning and probabilistic reasoning techniques