

Last update: December 4, 2008

# RATIONAL DECISIONS

## CMSC 421: CHAPTER 16

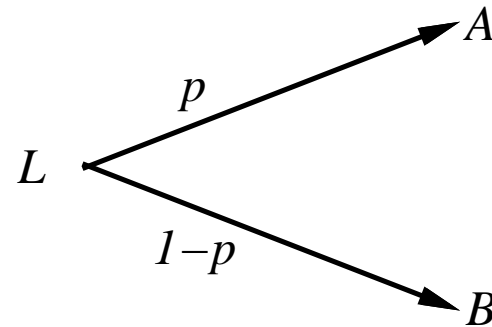
# Outline

- ◇ 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.

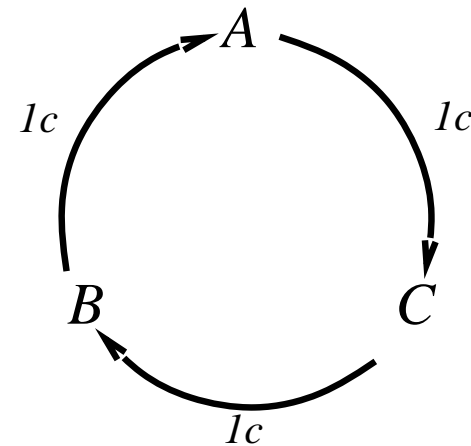
Violating the constraints leads to self-evident irrationality

Example: intransitive preferences

If  $B \succ C$ , then an agent who has  $C$  would trade  $C$  plus some money (e.g., 1 cent) to get  $B$

If  $A \succ B$ , then an agent who has  $B$  would trade  $B$  plus some money (e.g., 1 cent) to get  $A$

If  $C \succ A$ , then an agent who has  $A$  would trade  $A$  plus some money (e.g., 1 cent) to get  $C$



## Rational preferences contd.

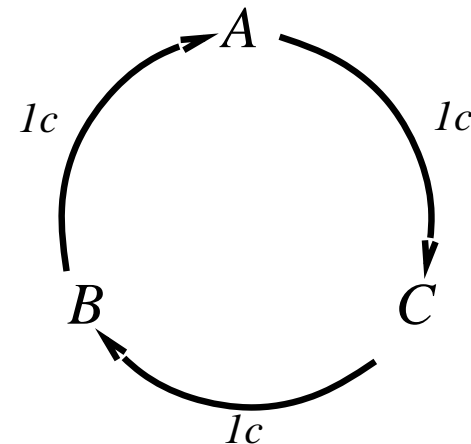
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If  $C \succ A$ , then an agent who has  $A$  would trade  $A$  plus some money (e.g., 1 cent) to get  $C$



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

# Utilities

Utilities map states to real numbers. Which numbers?

Standard approach to assessment of human utilities:

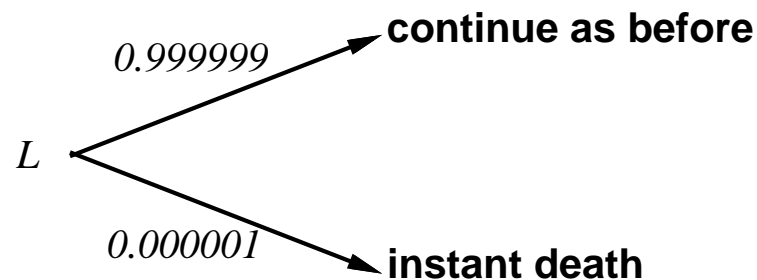
compare a given state  $A$  to a **standard lottery**  $L_p$  that has

“best possible prize”  $u_{\top}$  with probability  $p$

“worst possible catastrophe”  $u_{\perp}$  with probability  $(1 - p)$

adjust lottery probability  $p$  until  $A \sim L_p$

**pay \$30**     $\sim$



# Utility scales

Normalized utilities:  $u_{\top} = 1.0$ ,  $u_{\perp} = 0.0$

Micromorts: one-millionth chance of death

useful for Russian roulette, paying to reduce product risks, etc.

QALYs: quality-adjusted life years

useful for medical decisions involving substantial risk

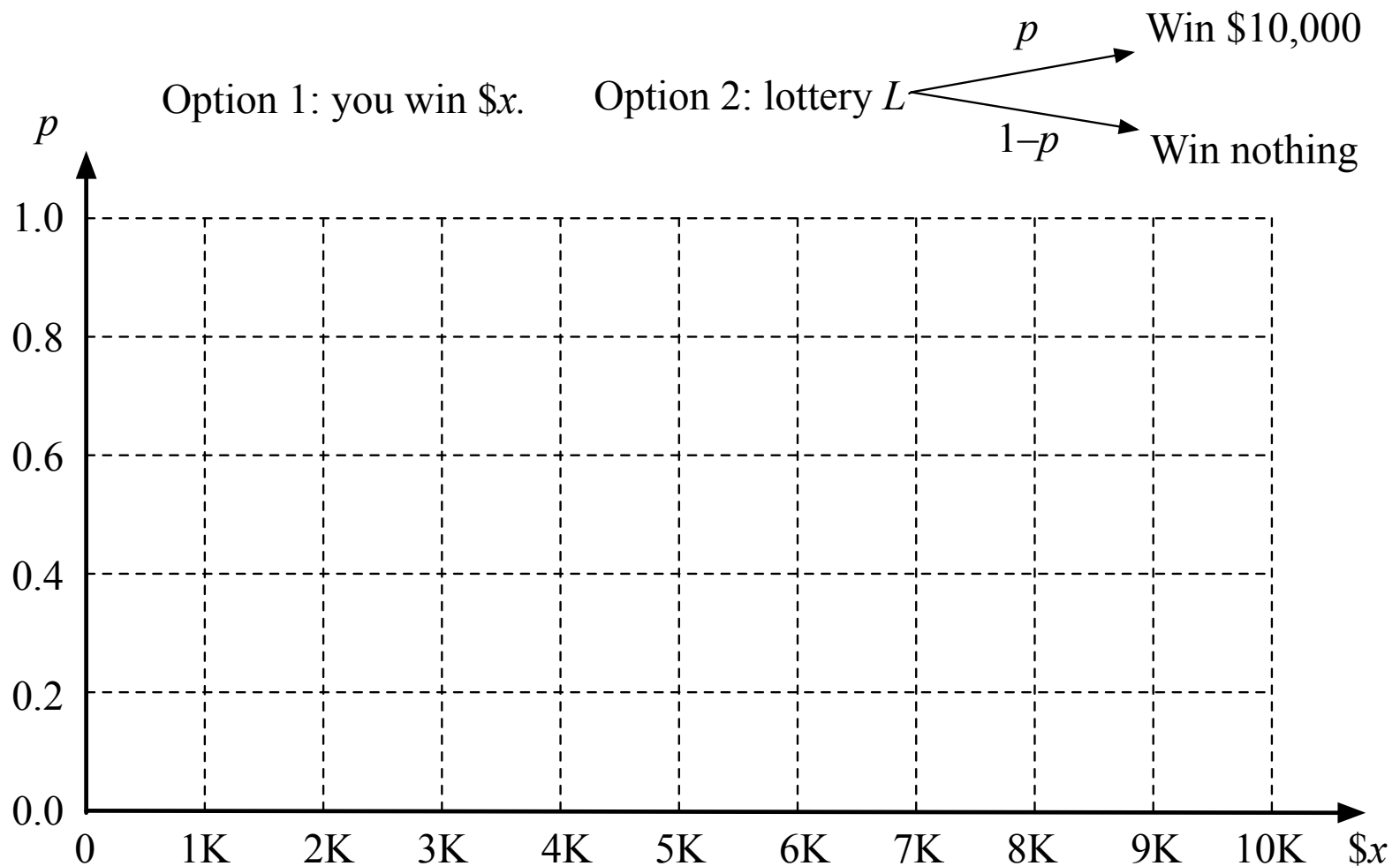
Note: behavior is **invariant** w.r.t. positive linear transformation

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

With deterministic prizes only (no lottery choices), only **ordinal utility** can be determined, i.e., total order on prizes

# Is money a utility function?

For each  $x$ , adjust  $p$  until half the class votes for lottery



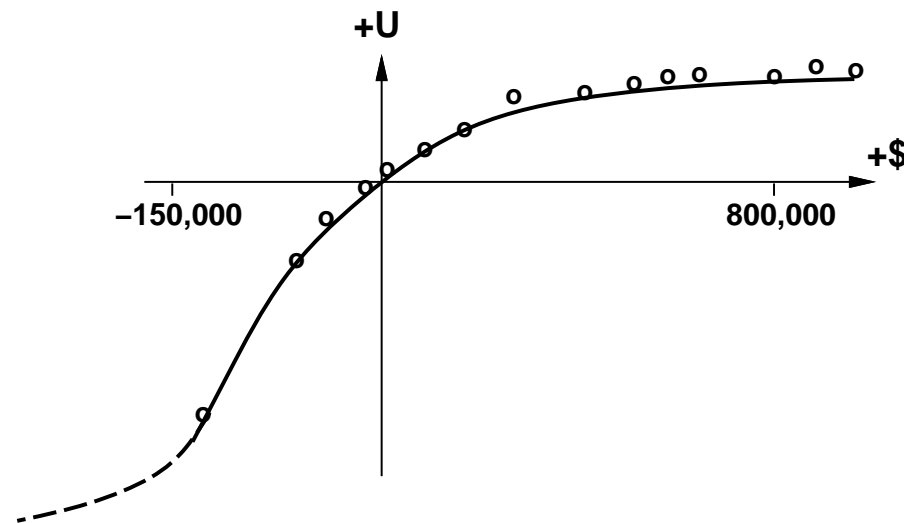
## What the book says

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

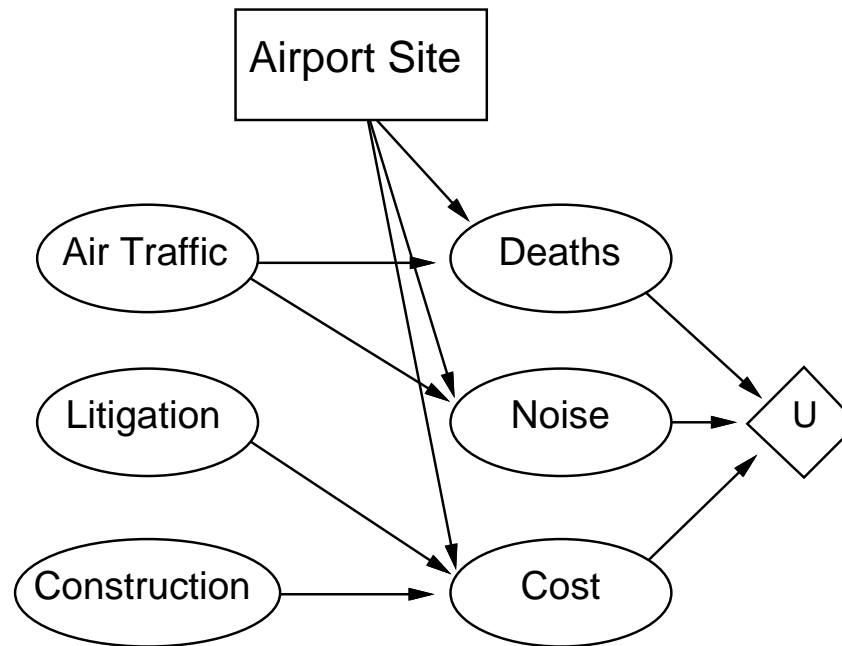
Utility curve: for what probability  $p$  am I indifferent between a prize  $x$  and a lottery  $[p, \$M; (1 - p), \$0]$  for large  $M$ ?

Typical empirical data, extrapolated with risk-prone behavior:



# Decision networks

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



Algorithm:

For each value of action node

    compute expected value of utility node given 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?

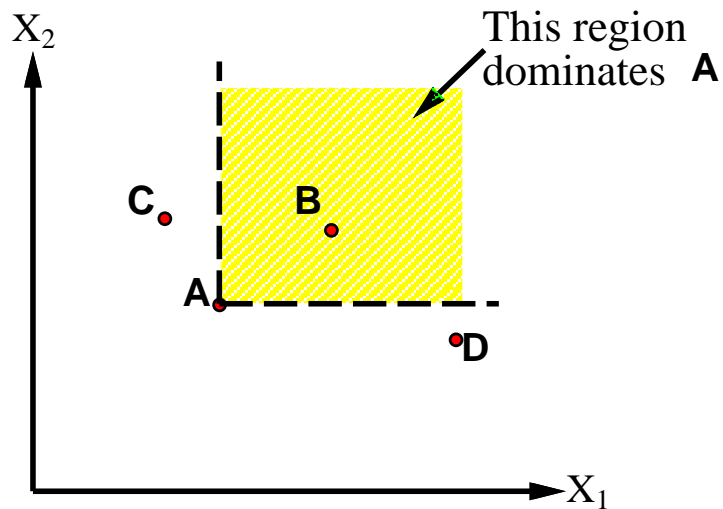
Idea 1: identify conditions under which decisions can be made without complete identification of  $U(x_1, \dots, x_n)$

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

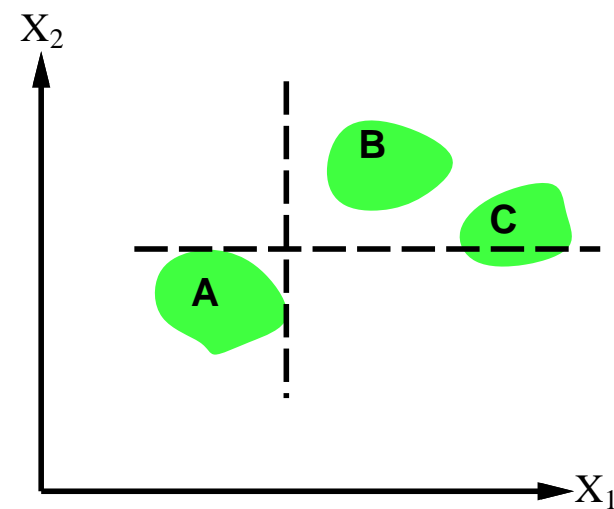
# Strict dominance

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  $U(B) \geq U(A)$ )



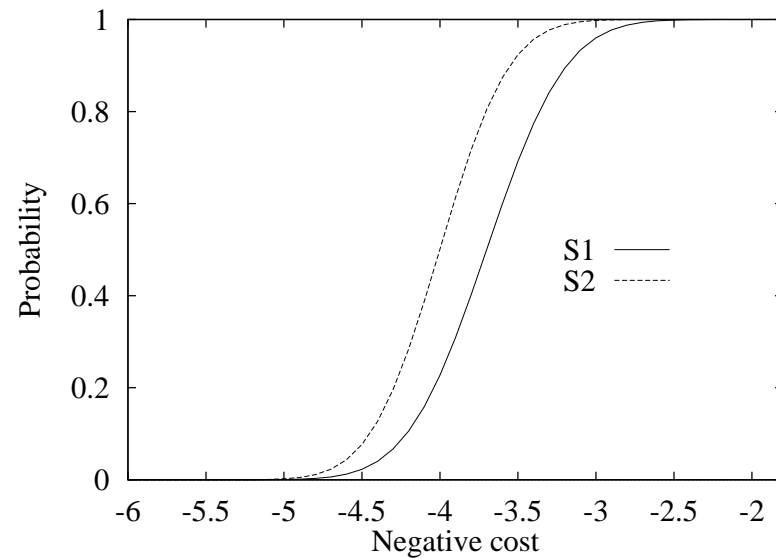
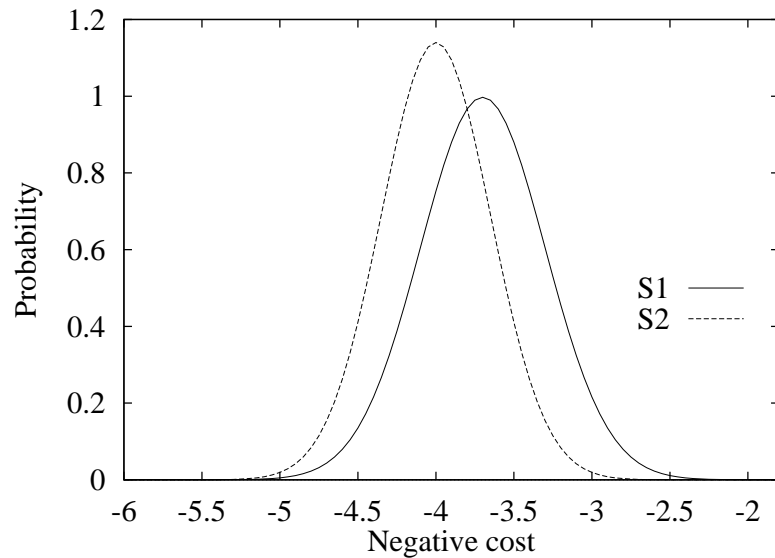
Deterministic attributes



Uncertain attributes

Strict dominance seldom holds in practice

# Stochastic dominance



Choices  $A_1$  and  $A_2$  with continuous distributions  $p_1$  and  $p_2$

$A_1$  stochastically dominates  $A_2$  iff  $\forall t \ P(A_1 \leq t) \leq P(A_2 \leq t)$ ,  
 i.e.,  $\forall t \ \int_{-\infty}^t p_1(x)dx \leq \int_{-\infty}^t p_2(t)dt$

If  $A_1$  stochastically dominates  $A_2$  and  $U$  is monotonic in  $x$ , then

$$EU(A_1) = \int_{-\infty}^{\infty} p_1(x)U(x)dx \geq \int_{-\infty}^{\infty} p_2(x)U(x)dx = EU(A_2)$$

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

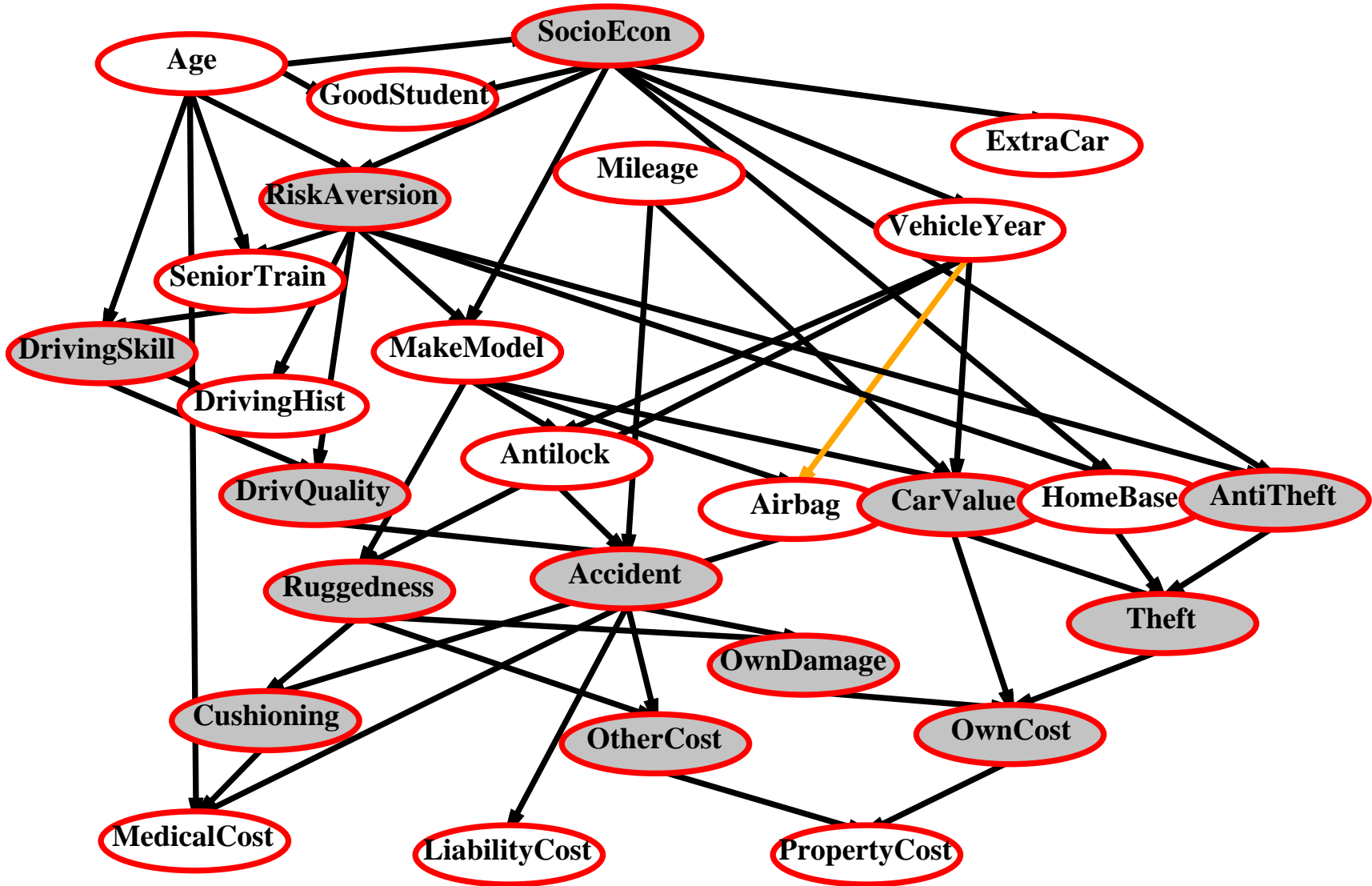
Can annotate belief networks with stochastic dominance information:

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

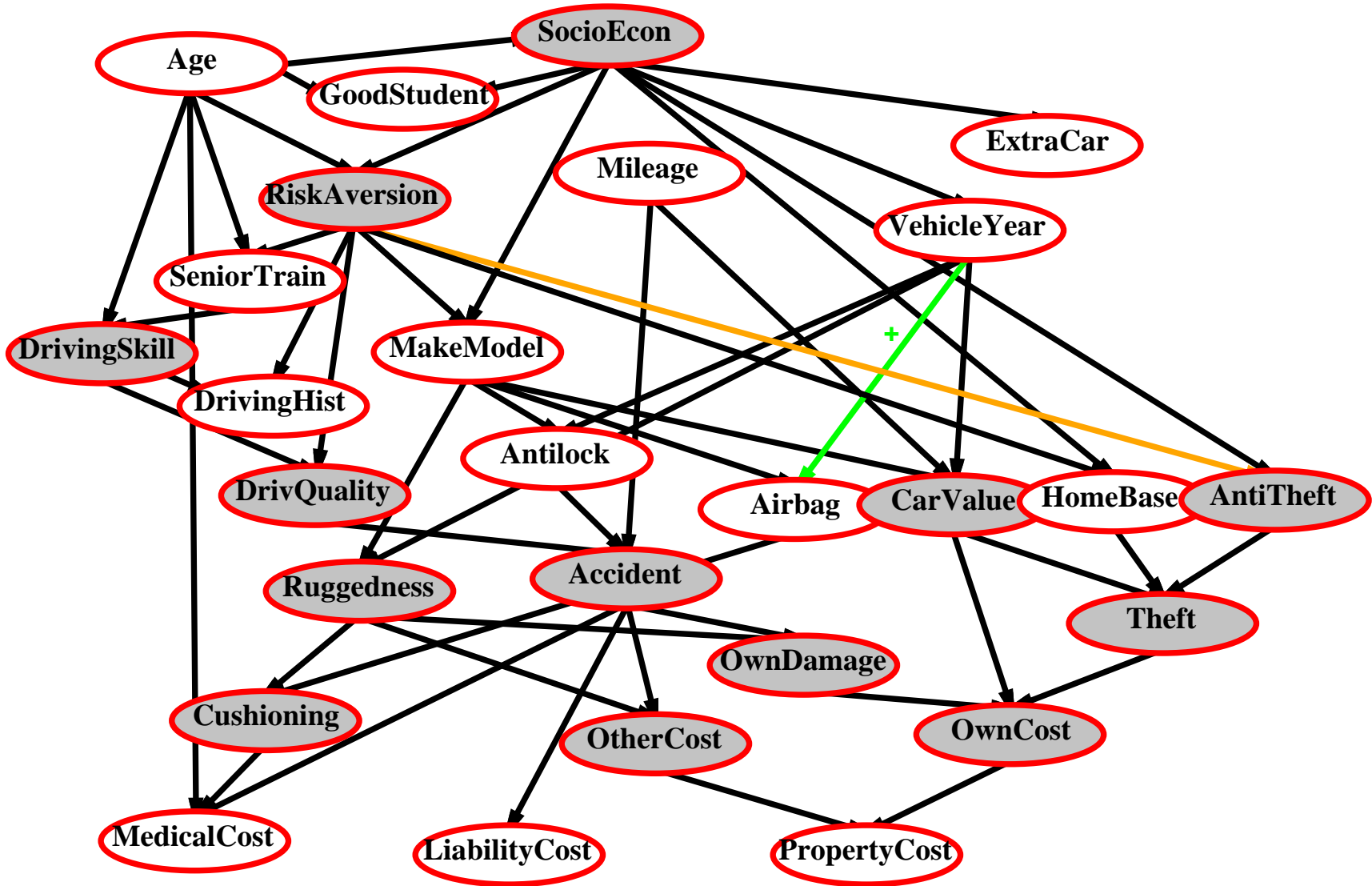
For every value  $\mathbf{z}$  of  $Y$ 's other parents  $\mathbf{Z}$

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

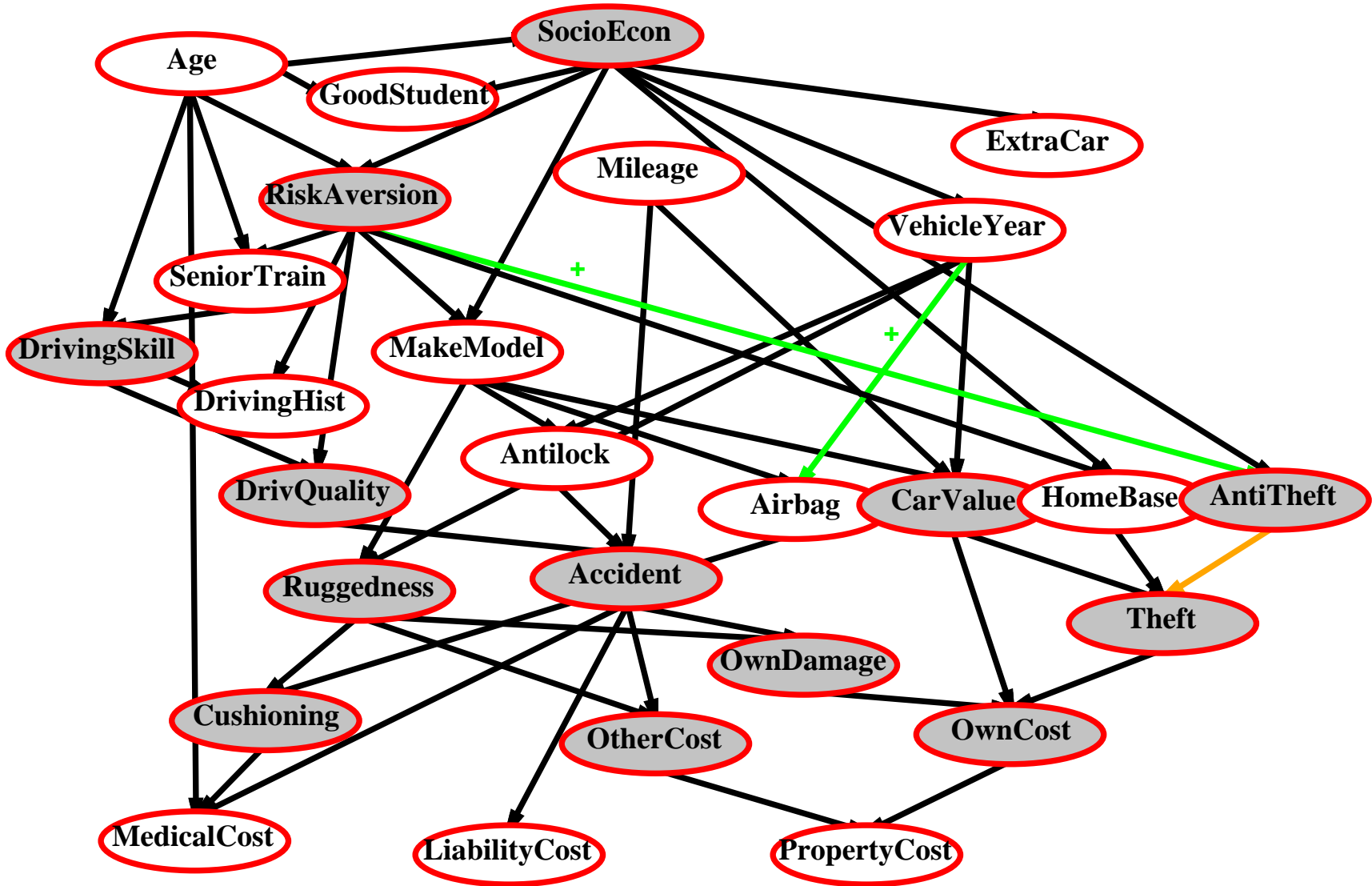
Label the arcs + or -



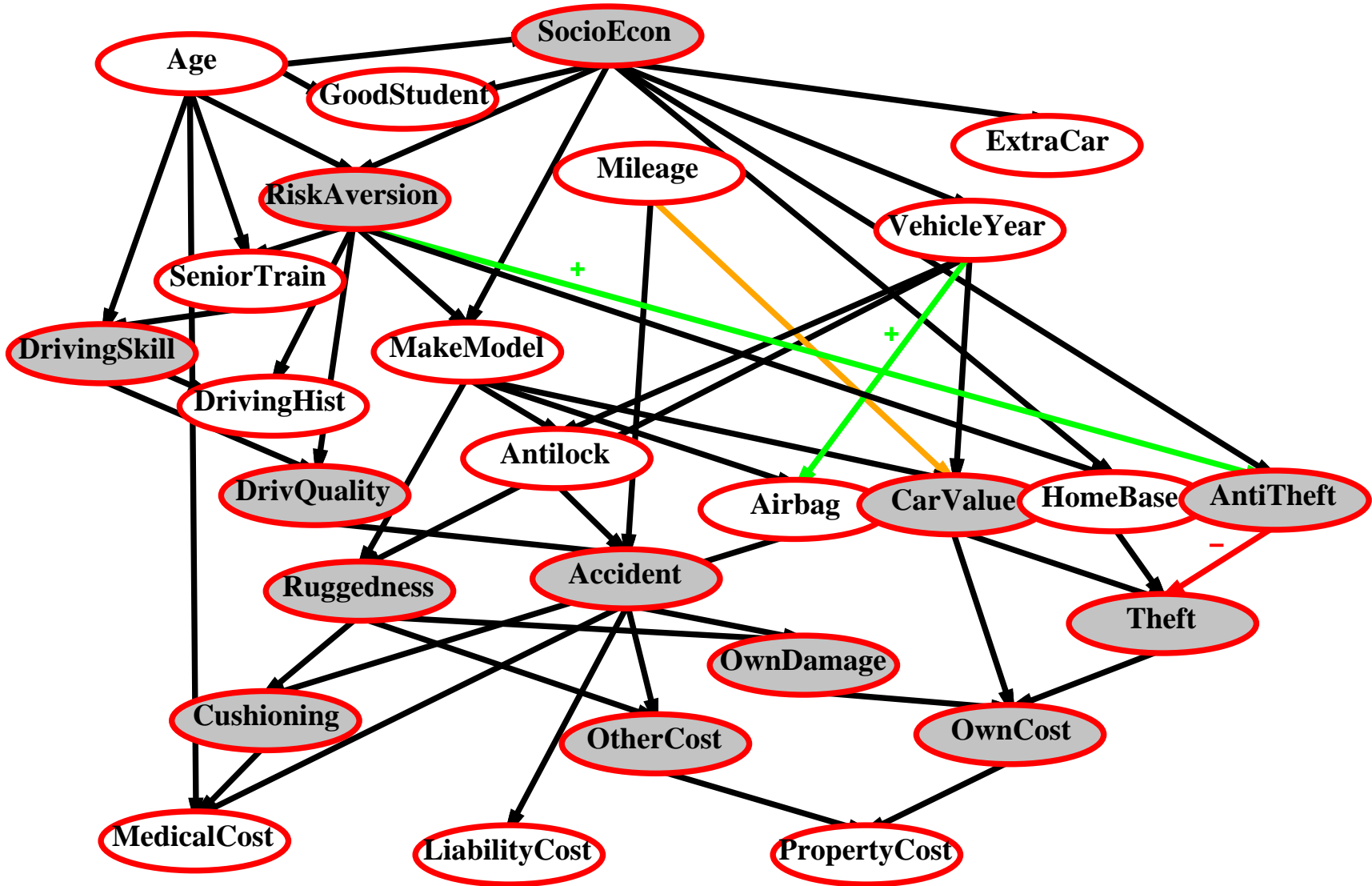
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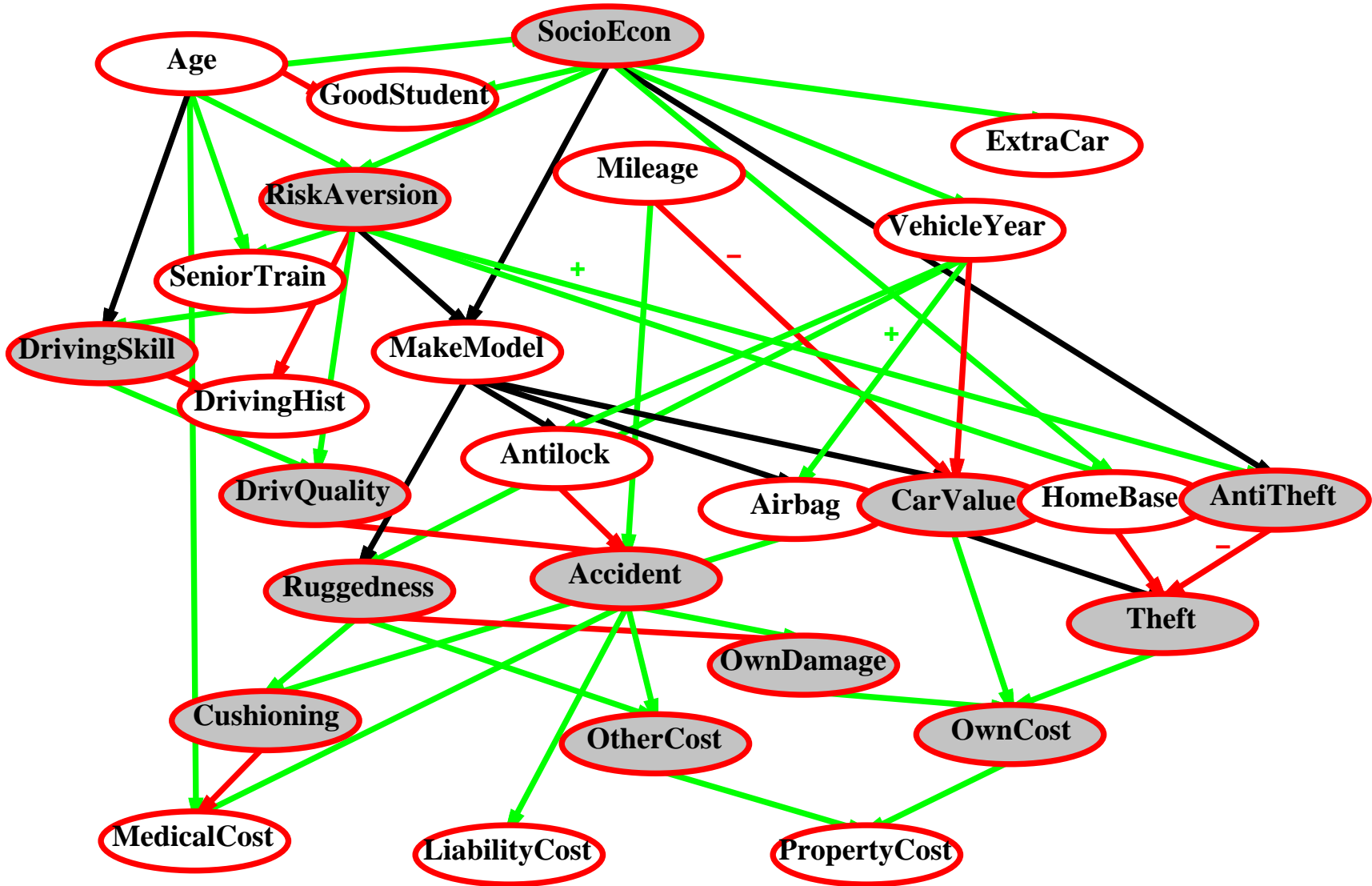


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## Preference structure: Deterministic

$X_1$  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.,  $\langle \text{Noise}, \text{Cost}, \text{Safety} \rangle$ :

$\langle 20,000 \text{ suffer}, \$4.6 \text{ billion}, 0.06 \text{ deaths/mpm} \rangle$  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:

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

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  $k/2$

“Consultant” offers accurate survey of  $A$ . Fair price?

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** (given!)

=  $[0.5 \times$  value of “buy  $A$ ” given “oil in  $A$ ”  
+  $0.5 \times$  value of “buy  $B$ ” given “no oil in  $A$ ”]

– 0

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

## General formula

Current evidence  $E$ , current best action  $\alpha$

Possible action outcomes  $S_i$

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

Potential new evidence  $E_j$

If we knew  $E_j = e_{jk}$ , then we would choose  $\alpha_{e_{jk}}$  s.t.

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

$E_j$  is a random variable whose value is *currently* unknown

$\Rightarrow$  must compute expected gain over all possible values:

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

(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, maximizing VPI for each to select one is not always optimal

⇒ 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

