# Lecture 29: Decision Making Under Uncertainty

#### Modeling temporal uncertainty (and failure) with DBNs

$$EU(a|e) = \sum_{s'} P(Result(a) = s'|a, e)U(s')$$

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"A rational agent should choose actions which maximize its expected utility"

 $A \succ B$ Agent prefers A to B $A \sim B$ Agent is indifferent to A and B $A \succeq B$ Agent prefers A to B or is indifferent to them

A and B can be lotteries

# $L = [p_1, C_1; p_2, C_2; \dots; p_n, C_n]$

- Orderability:  $(A \succ B) \lor (B \succ A) \lor (A \sim B)$
- Transitivity:  $(A \succ B) \land (B \succ C) \implies (A \succ C)$
- Continuity:  $A \succ B \succ C \implies \exists p[p, A; 1-p, C] \sim B$
- Substitutability:  $A \sim B \implies [p, A; 1 p, C] \sim [p, B; 1 p, C]$
- Monotonicity:  $A \succ B \implies (p \ge q \iff [p, A; 1-p, B] \succeq [q, A; 1-q, B])$
- Decomposability:

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 $\begin{array}{l} \boldsymbol{A} \succ \boldsymbol{B} \\ \boldsymbol{B} \succ \boldsymbol{C} \\ \boldsymbol{C} \succ \boldsymbol{A} \end{array}$ 

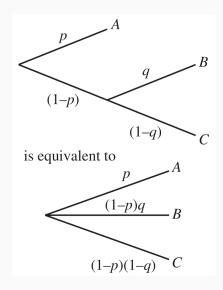
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- Decomposability:

#### Decomposability



- Existence of a Utility Function: There exists a function U such that:  $U(A) > U(B) \iff A \succ B$ ,  $U(A) = U(B) \iff A \sim B$
- Expected Utility of a Lottery: The utility of a lottery is the sum of the probability of each outcome times the utility of that outcome.
  U([p<sub>1</sub>, S<sub>1</sub>, ...; p<sub>n</sub>, S<sub>n</sub>]) = ∑<sub>i</sub> p<sub>i</sub>U(s<sub>i</sub>)

(**Proof?** von Neumann, J. and Morgenstern, O. (1944). Theory of Games and Economic Behavior Princeton University Press)

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# What the Axioms Don't give you 1: Guidance on Arbitrary Preference

"I prefer to have a prime number in my bank account; When i have £10, I will give away £3"

#### Monotonic preference towards money

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What about lotteries?

- A: I give you £1,000,000
- B: Toss a coin: Heads, I give you  $\pounds$ 3,000,000; Tails, you get nothing

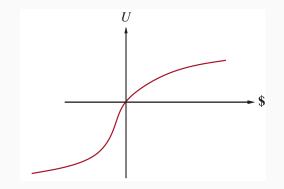
- A: I give you £1,000,000
- B: Toss a coin: Heads, I give you £3,000,000; Tails, you get nothing

Expected monetary value A: 1,000,000

Expected monetary value **B**:  $0.5 \times 0 + 0.5 \times 3,000,000 = 1,500,000$ 

 $S_k$  - state of possessing £k

$$EU(A) = U(S_{k+1M})$$
  
 $EU(B) = \frac{1}{2}U(S_k) + \frac{1}{2}U(S_{k+3M})$ 



#### What the Axioms Don't Give You 2: Uniqueness

$$EU(a_1|e) = \sum_{s'} P(Result(a_1) = s'|a, e)U'(s')$$
$$EU(a_2|e) = \sum_{s'} P(Result(a_2) = s'|a, e)U'(s')$$

$$EU(a_1|e) = \sum_{s'} P(Result(a_1) = s'|a, e)(k_1 + k_2U(s'))$$
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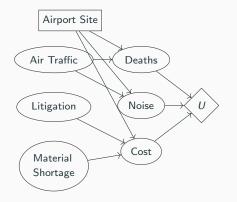
$$EU(a_1|e) = k_1 + k_2 \sum_{s'} P(Result(a_1) = s'|a, e)U(s')$$
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#### Normalized Utility:

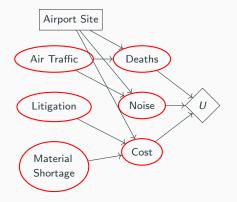
"Best Possible Outcome"  $(u^{\top} = 1)$ "Worst Possible Catastrophe"  $(u^{\perp} = 0)$ 

Figuring out the utility of S:

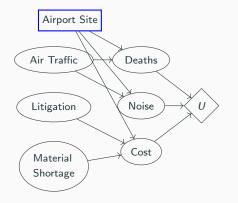
- Offer a lottery:  $[p, u^{\top}; 1 p, u^{\perp}]$
- Adjust p until  $[p, u^{ op}; 1-p, u^{\perp}] \sim S$
- Set U(S) = p



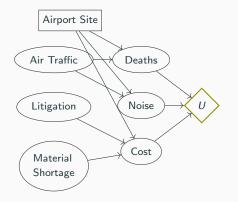
- Chance Nodes (ovals)
- Decision Nodes (rectangles)
- Utility Nodes (diamonds)



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### Evaluating Expected Utility with a Decision Network

#### 1. Set evidence variables for current state

- 2. For each value of decision node
  - 2.1 Set decision node to that value
  - 2.2 Calculate posterior probabilities for parents of utility node
  - 2.3 Calculate resulting expected utility for action
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- Utility Theory, Axioms and Criticisms
- Decision Networks for Expected Utility