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Decisions and Value of Information

Many slides courtesy of Dan Klein, Stuart Russell, or Andrew Moore

CS 5300 / CS 6300
Artificial Intelligence
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Slide 1 CS 5300: Decisions + Information

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Announcements

- Today:
 - Finish inference in "simple" networks
 - How to make decisions based on probabilistic inference
- Coming soon!
 - Reasoning over time
- Due-ness:
 - Homework 8 out today, due next Thursday
 - P4 due the following Tuesday (get going!)
- Grading-ness:
 - Project 2 grades posted
 - Project 3 grades out Θ (this week)

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P3 Winners!

- First place (average score of)
 - x, y and z
- Second place (average score of)
 - x, y and z
- Third place (average score of)
 - x, y and z

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Recap: Inference Example

- Find $P(W|F=bad)$
- Restrict all factors
- No hidden vars to eliminate (this time!)
- Just join and normalize

W	P(W)
sun	0.7
rain	0.3

W	P(F=bad W)
sun	0.2
rain	0.9

F	P(F sun)
good	0.8
bad	0.2

F	P(F rain)
good	0.1
bad	0.9

W	P(W, F=bad)
sun	0.14
rain	0.27

W	P(W F=bad)
sun	0.34
rain	0.66

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Decision Networks

- MEU: choose the action which maximizes the expected utility given the evidence
- Can directly operationalize this with decision diagrams
 - Bayes nets with nodes for utility and actions
 - Lets us calculate the expected utility for each action
- New node types:
 - Chance nodes (just like BNs)
 - Actions (rectangles, must be parents, act as observed evidence)
 - Utilities (depend on action and chance nodes)

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Decision Networks

- Action selection:
 - Instantiate all evidence
 - Calculate posterior over parents of utility node
 - Set action node each possible way
 - Calculate expected utility for each action
 - Choose maximizing action

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Example: Decision Networks

Umbrella = leave

Umbrella = take

Optimal = leave

	A	W	U(A,W)
leave	sun	rain	100
leave	rain	sun	0
take	sun	rain	20
take	rain	sun	70

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Example: Decision Networks

Umbrella = leave

Umbrella = take

Optimal = take

	W	P(W F=bad)
sun	rain	0.34
rain	sun	0.66

	A	W	U(A,W)
leave	sun	rain	100
leave	rain	sun	0
take	sun	rain	20
take	rain	sun	70

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Value of Information

- Idea: compute value of acquiring each piece of evidence
- Can be done directly from decision network
- Example: buying oil drilling rights
 - Two blocks A and B, one has oil, worth k
 - Prior probabilities 0.5 each
 - Current price of each block is k/2
 - MEU = 0 (either action is a maximizer)
- Solution: compute **value of information** = expected gain in MEU from observing new information
- Probe gives accurate survey of A. Fair price?
 - Survey may say "oil in a" or "oil in b," prob 0.5 each
 - If we know O, MEU is k/2 (either way)
 - Gain in MEU?
 - VPI(O) = k/2
 - Fair price: k/2

O	P
a	1/2
b	1/2

D	O	U
a	a	k/2
a	b	-k/2
b	a	-k/2
b	b	k/2

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Value of Information

- Current evidence $E=e$, utility depends on $S=s$

$$MEU(e) = \max_a \sum_s P(s|e) U(s, a)$$
- Potential new evidence E' : suppose we knew $E' = e'$

$$MEU(e, e') = \max_a \sum_s P(s|e, e') U(s, a)$$
- BUT E' is a random variable whose value is currently unknown, so:
 - Must compute expected gain over all possible values

$$VPI_e(E') = \sum_{e'} P(e'|e) (MEU(e, e') - MEU(e))$$

- (VPI = value of perfect information)

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VPI Example

MEU with no evidence

MEU if forecast is bad

MEU if forecast is good

Forecast distribution

F	P(F)
good	0.59
bad	0.41

$$VPI_e(E') = \sum_{e'} P(e'|e) (MEU(e, e') - MEU(e))$$

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VPI Properties

- Nonnegative in expectation

$$\forall E', e : VPI_e(E') \geq 0$$
- Nonadditive ---consider, e.g., obtaining E_j twice

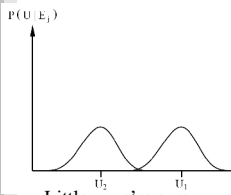
$$VPI_e(E_j, E_k) \neq VPI_e(E_j) + VPI_e(E_k)$$
- Order-independent

$$VPI_e(E_j, E_k) = VPI_e(E_j) + VPI_{e, E_j}(E_k) = VPI_e(E_k) + VPI_{e, E_k}(E_j)$$

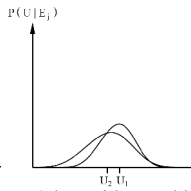
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VPI Scenarios

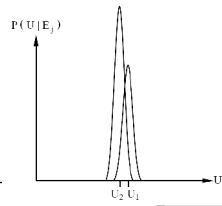
- > Imagine actions 1 and 2, for which $U_1 > U_2$
- > How much will information about E_j be worth?



Little – we're sure action 1 is better.



A lot – either could be much better



Little – info likely to change our action but not our utility