CS 581 – ADVANCED ARTIFICIAL INTELLIGENCE

TOPIC: DECISION MAKING UNDER UNCERTAINTY





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UNCERTAINTY

- The agent needs reason in an uncertain world
- Uncertainty can be due to
 - Noisy sensors (e.g., temperature, GPS, camera, etc.)
 - Imperfect data (e.g., low resolution image)
 - Missing data (e.g., lab tests)
 - Imperfect knowledge (e.g., medical diagnosis)
 - Exceptions (e.g., all birds fly except ostriches, penguins, birds with injured wings, dead birds, ...)
 - Changing data (e.g., flu seasons, traffic conditions during rush hour, etc.)
 - •
- The agent still must act (e.g., step on the breaks, diagnose a patient, order a lab test, ...)

A FEW EXAMPLES

- Spam filtering
- Medical diagnosis
- Loan approval
- Automated driving

O ...

RATIONAL AGENT

- Given
 - World states, a utility function, actions, transitions, evidence, and probabilities
- A rational agent chooses the action that maximizes expected utility

$$action = \underset{a}{\operatorname{argmax}} EU(a|e)$$

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UTILITY THEORY

- Lottery: *n* possible outcomes with probabilities
 - $[p_1, S_1; p_2, S_2; \dots p_n, S_n]$
 - Each S_i can be an atomic state or another lottery
- Expected utility of a lottery
 - $EU([p_1, S_1; p_2, S_2; ... p_n, S_n]) = \sum_{i=1}^n p_i S_i$

UTILITY ≠ MONEY

- Most agents prefer more money to less money,
 - But this does not mean money behaves as a utility function
- For example, which lottery would you prefer
 - L₁: [1, \$1 Million]
 - L₂: [0.5, \$0; 0.5, \$2.5 Million]
- If money served as a utility function, then you'd prefer L_2 no matter what, but the answer *often* depends on how much money you currently have
 - The utility of money depends on what you <u>prefer</u>
 - If you are short on cash, a little more certain money can help
 - If you are already billionaire, you might take the risk
 - o Or if you are swimming in debt, you might like to gamble

UTILITY ≠ MONEY

- \bullet Let's say you currently have k and let S_k represent the state of having k
- $\bullet \ \mathrm{EU}(\mathrm{L}_1) = \mathrm{U}(\mathrm{S}_{\mathrm{k+1M}})$
- \bullet EU(L₂) = 0.5*U(S_k) + 0.5*U(S_{k+2.5M})
- \bullet The rational choice depends on your preferences for $S_k,$ $S_{k+1M},$ and $S_{k+2.5M}$
 - i.e, it depends on the values of $U(S_k)$, $U(S_{k+1M})$, and $U(S_{k+2.5M})$
- U(S_i) does not have to be a linear function of i, and for people it often is not

WE'LL COVER

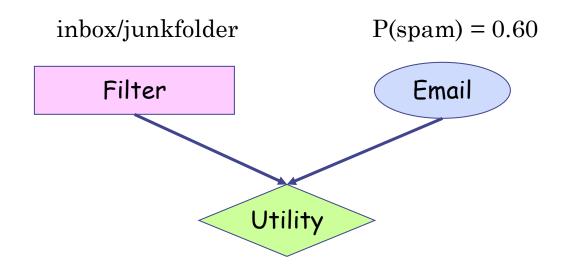
- Single decision
 - Influence diagrams
- Sequence of actions
 - Markov decision processes

INFLUENCE DIAGRAMS

Influence Diagrams / Decision Networks

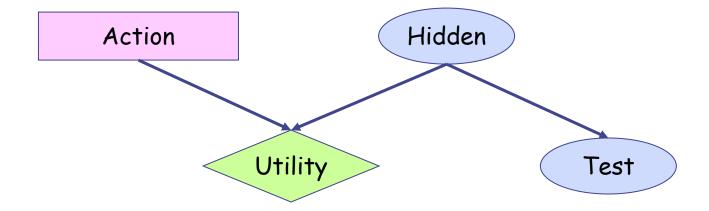
- Builds on Bayesian networks
- In addition to the chance nodes (ovals), decision networks have
 - Decision nodes square
 - Represents actions
 - Utility nodes diamond
 - Represents utilities for possible states and actions

SPAM FILTERING



U(inbox, spam) = -300 U(inbox, ~spam) = 100 U(junkfolder, spam) = 200 U(junkfolder, ~spam) = -400

A COMMON PATTERN



DECISION NETWORKS - APPLICATIONS

Used for

- What action to take
- What information to gather
- How much to pay for a piece of information

• For example:

- Medical diagnosis: which test to perform, which treatment to prescribe, ...
- Marketing: which project to invest in, how much to spend on marketing, how much to spend on user surveys, ...

EVALUATING DECISION NETWORKS

- Set evidence nodes **E** to their values **e**
- For each choice **a** of action **A**
 - Set **A**=**a**
 - Compute the posterior probability of the <u>parent chance</u> nodes of the <u>utility node</u>; i.e., compute P(Pa(Utility) | e,
 a)
 - Compute expected utility using the utility node and the probability distribution P(Pa(Utility) | **e**, **a**)
- Choose action **a** with the maximum expected utility

EXAMPLES

• See OneNote

VALUE OF INFORMATION

- If I am allowed to observe the value of a chance node, how much valuable is that information to me?
- Value of information
 - Expected utility after the information is acquired
 Minus
 - Expected utility before the information is acquired
- There is one catch: we do not know the content of the information before we acquire it
 - Solution: take an expectation over the possible outcomes