

# CS 486 — Lecture 16: Intro to Decision Theory and Decision Networks

## 1 Decision Theory

- Combines probability theory and utility theory.
- How should an agent act in an uncertain world?
- Recall probability theory is what an agent should believe based on evidence, and utility theory is what the agent wants.
- In other words, an agent should choose the action that maximizes the expected utility.
- Unfortunately, it's easier said than done to get the *absolute best utility* — finding if the utility is the best is NP-hard most of the time, and requires checking into the future.
- Hence, we use expected utility.

## 2 Constructing a Decision Network

- Given the situation (robot puts on pads to lower damage, choose between a short or long route, short route may cause it to slip):
  - RV:  $A$  represents whether an accident occurs or not.
  - Decision variables/actions:  $S$  is whether it takes the short route,  $P$  is whether it puts on pads.
- We have 3 types of nodes:
  - Chance nodes — represent random variables (similar to a Bayesian network).
  - Decision nodes — represent actions (decision variables).
  - Utility nodes — represent the agent's utility function on states (the “happiness” in each state).

## 3 Evaluating a Decision Network

- How do we choose an action?
  1. Set evidence variables for the current state.
  2. For each possible value of a decision node:
    - (a) Set decision node to value
    - (b) Calculate posterior probability for parent nodes of the utility node
    - (c) Calculate expected utility for the action
  3. Return the action with the highest expected utility.
- Often, we'll get expected utilities that are functions — we will define *policies* to select which choices to make based on the function's parameters (ie: if  $q > 0.4$  then pick policy A, otherwise pick policy B...).

## 4 Variable Elimination

- In our original example, we can actually combine Short and Pads together.
- For a single-stage decision network:
  - We prune all the nodes that are not ancestors of the utility node.
  - Then, sum out all chance nodes.
  - For the single remaining factor, return the maximum value and the assignment that gives the maximum value.