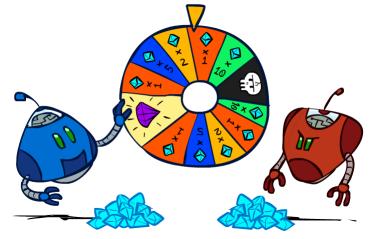
CSEP 573: Artificial Intelligence

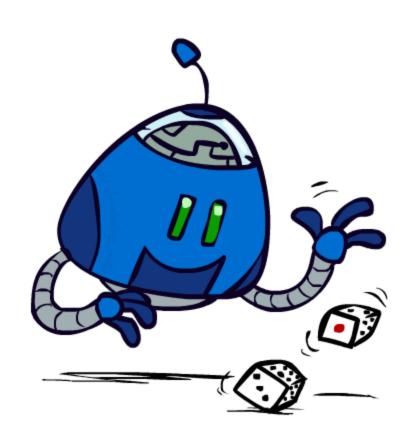
Hanna Hajishirzi

Expectimax – Complex Games

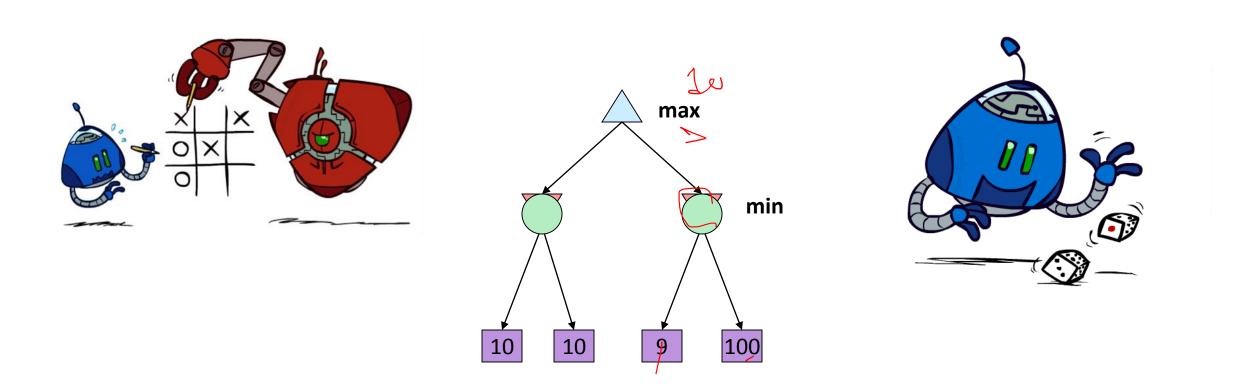
slides adapted from Dan Klein, Pieter Abbeel ai.berkeley.edu And Dan Weld, Luke Zettlemoyer



Uncertain Outcomes



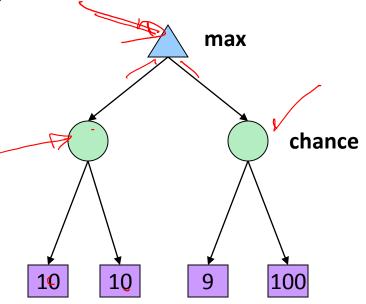
Worst-Case vs. Average Case



Idea: Uncertain outcomes controlled by chance, not an adversary!

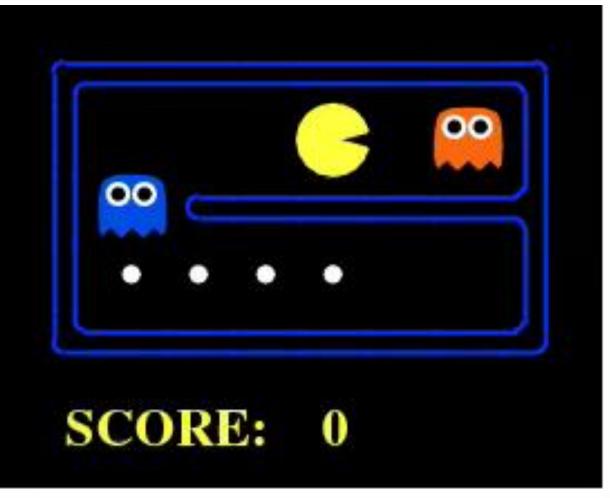
Expectimax Search

- Why wouldn't we know what the result of an action will be?
 - Explicit randomness: rolling dice
 - Unpredictable opponents: the ghosts respond randomly
 - Unpredictable humans: humans are not perfect
 - o Actions can fail: when moving a robot, wheels might slip
- Values should now reflect average-case (expectimax) outcomes, not worst-case (minimax) outcomes
- Expectimax search: compute the average score under optimal play
 - Max nodes as in minimax search
 - o Chance nodes are like min nodes but the outcome is uncertain
 - Calculate their expected utilities
 - I.e. take weighted average (expectation) of children
- Later, we'll learn how to formalize the underlying uncertainresult problems as Markov Decision Processes



Video of Demo Min vs. Exp (Min)



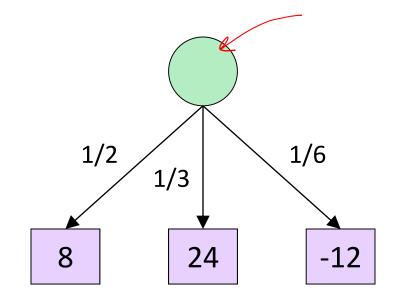


Expectimax Pseudocode

```
def value(state):
                     if the state is a terminal state: return the state's utility
                     if the next agent is MAX: return max-value(state)
                     if the next agent is EXP: return exp-value(state)
def max-value(state):
                                                          def exp-value(state):
   initialize v = -\infty
                                                              initialize v = 0
   for each successor of state:
                                                              for each successor of state:
       v = max(v, value(successor))
                                                                  p = probability(successor)
                                                                 v += p * value(successor)
   return v
                                                              return v
```

Expectimax Pseudocode

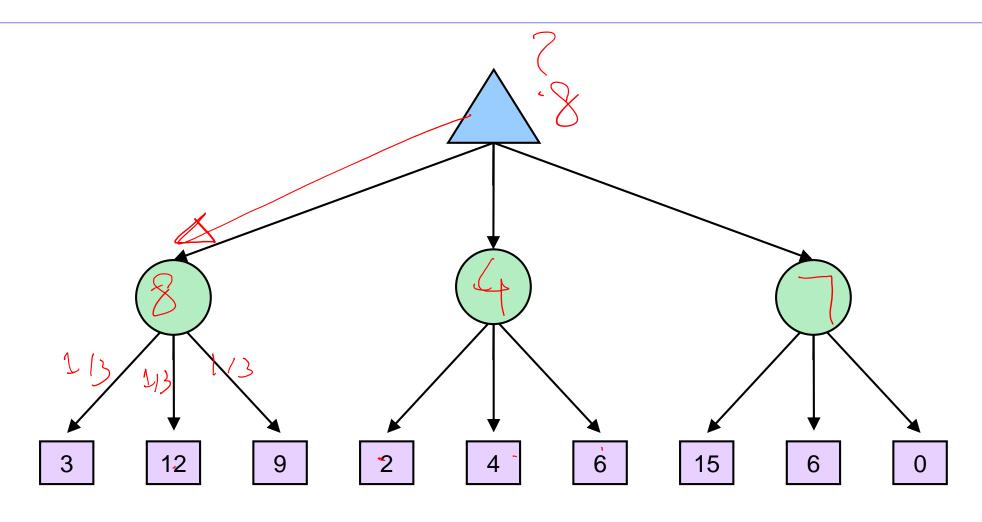
def exp-value(state): initialize v = 0 for each successor of state: p = probability(successor) v += p * value(successor) return v



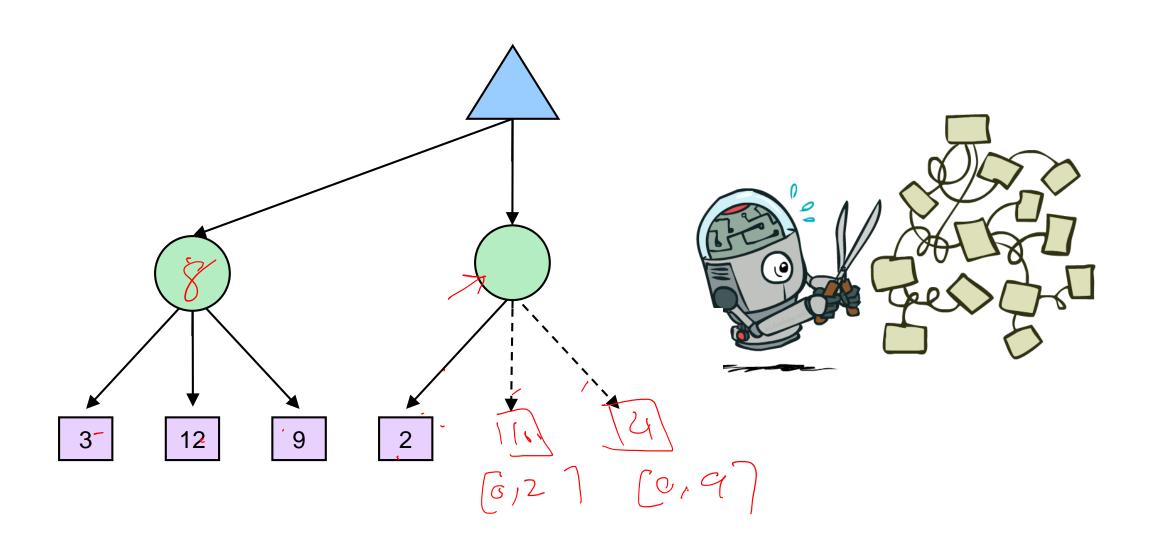
$$v = (1/2)(8) + (1/3)(24) + (1/6)(-12) = 10$$

$$8 \times \frac{1}{2} + 24 \times \frac{1}{3} + (-12) \times \frac{1}{3}$$

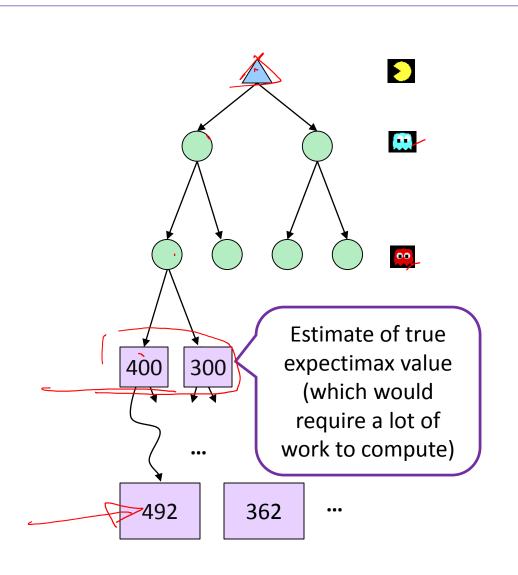
Expectimax Example



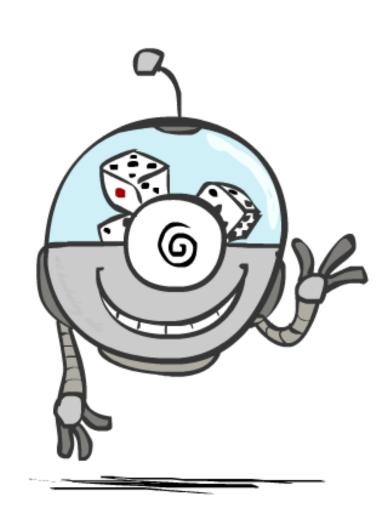
Expectimax Pruning?



Depth-Limited Expectimax



Probabilities

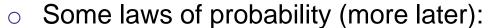


Reminder: Probabilities

- A random variable represents an event whose outcome is unknown.
- A probability distribution is an assignment of weights to outcomes



- Random variable: T ≠ whether there's traffic
- Outcomes: T in {none, light, heavy}
- o Distribution: P(T=none) = 0.25, P(T=light) = 0.50, P(T=heavy) = 0.25



- Probabilities are always non-negative
- Probabilities over all possible outcomes sum to one
- As we get more evidence, probabilities may change:
 - o P(T=heavy) = 0.25, P(T=heavy | Hour=8am) = 0.60
 - We'll talk about methods for reasoning and updating probabilities later



0.25



0.50



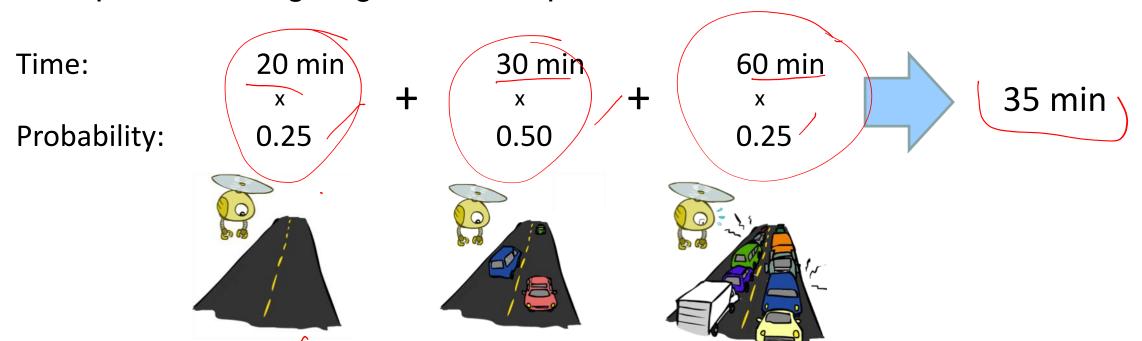
0.25

Reminder: Expectations

45

 The expected value of a function of a random variable the average, weighted by the probability distribution ov outcomes

Example: How long to get to the airport?



What Probabilities to Use?

In expectimax search, we have a probabilismodel of how the opponent (or environment behave in any state

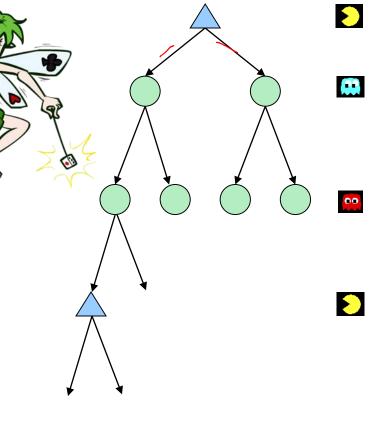
o Model could be a simple uniform distribution (roll a die

 Model could be sophisticated and require a great deak of computation

We have a chance node for any outcome out of our control: opponent or environment

The model might say that adversarial actions are likely!

 For now, assume each chance node magically comes along with probabilities that specify the distribution over its outcomes

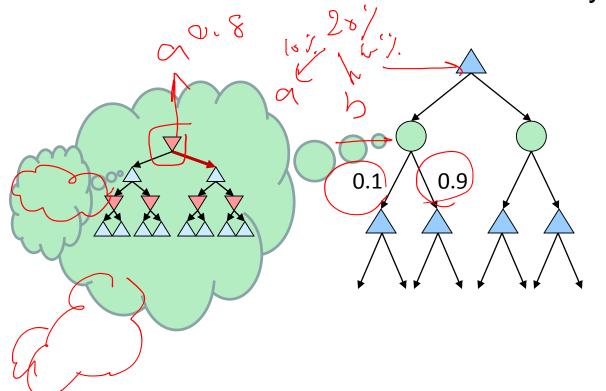


Having a probabilistic belief about another agent's action does not mean that the agent is flipping any coins!

Quiz: Informed Probabilities

Let's say you know that your opponent is actually running a depth 2 minimax, using the result 80% of the time, and moving randomly otherwise

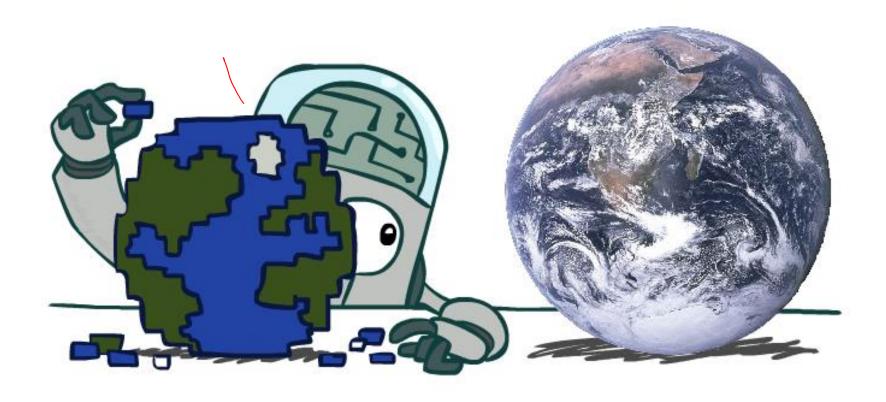
Question: What tree search should you use?



Answer: Expectimax!

- To figure out EACH chance node's probabilities, you have to run a simulation of your opponent
- This kind of thing gets very slow very quickly
- Even worse if you have to simulate your opponent simulating you...
- ... except for minimax and maximax, which have the nice property that it all collapses into one game tree

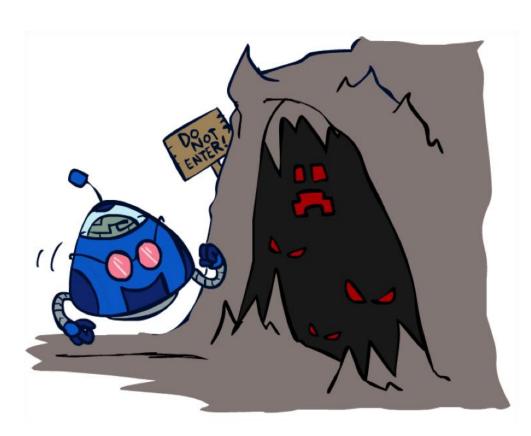
Modeling Assumptions



The Dangers of Optimism and Pessimism

Dangerous Optimism

Assuming chance when the world is adversarial

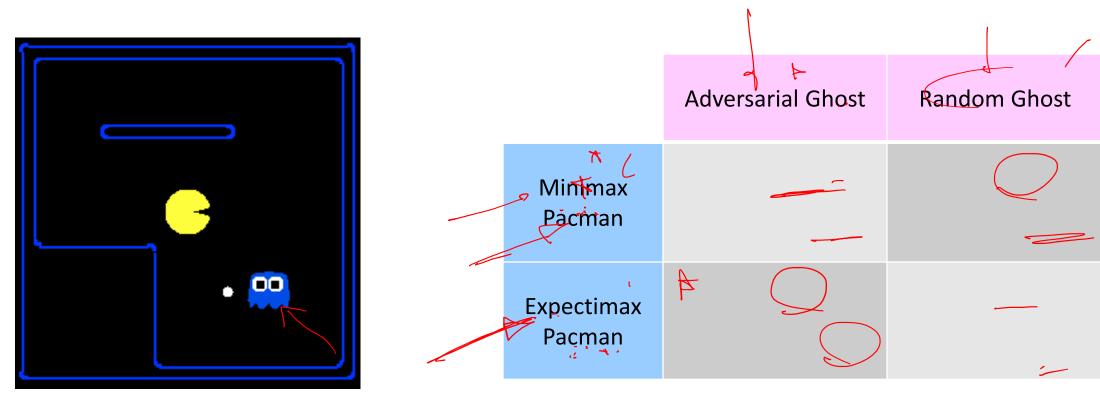


Dangerous Pessimism

Assuming the worst case when it's not likely



Assumptions vs. Reality



Results from playing 5 games

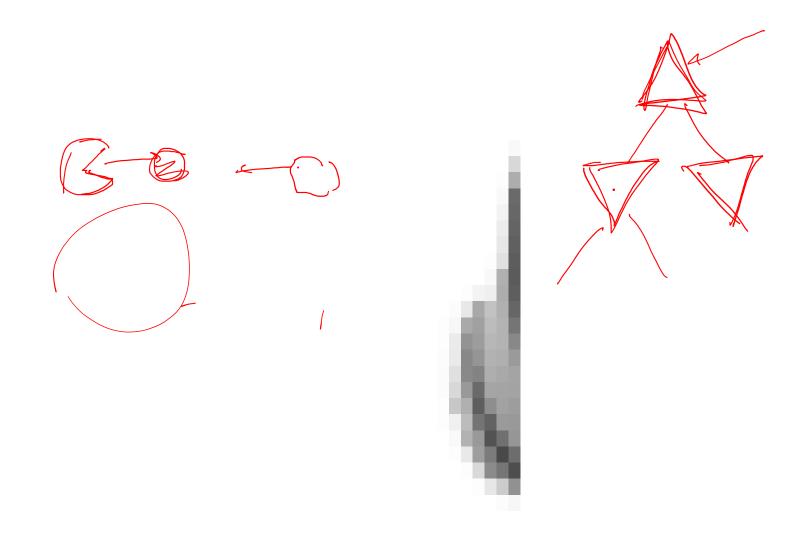
Pacman used depth 4 search with an eval function that avoids trouble Ghost used depth 2 search with an eval function that seeks Pacman

Video of Demo World Assumptions Random Ghost – Expectimax Pacman

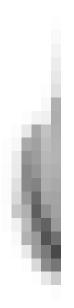
Video of Demo World Assumptions Adversarial Ghost – Minimax Pacman

Video of Demo World Assumptions Random Ghost – Minimax Pacman

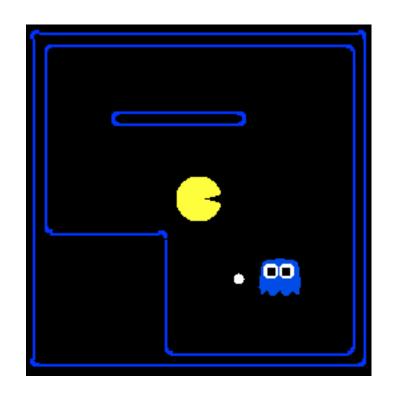




Video of Demo World Assumptions Adversarial Ghost – Expectimax Pacman



Assumptions vs. Reality



	Adversarial Ghost	Random Ghost
Minimax Pacman	Won 5/5 Avg. Score: 483	Won 5/5 Avg. Score: 493
Expectimax Pacman	Won 1/5 Avg. Score: -303	Won 5/5 Avg. Score: 503

Results from playing 5 games



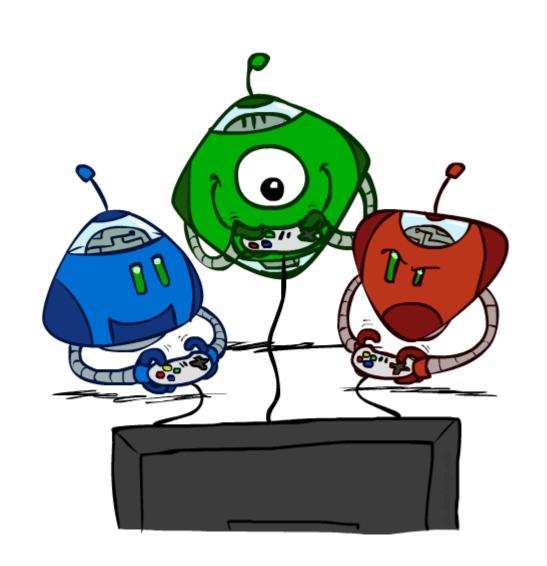
Pacman used depth 4 search with an eval function that avoids trouble Ghost used depth 2 search with an eval function that seeks Pacman

Why not minimax?

- Worst case reasoning is too conservative
- Need average case reasoning

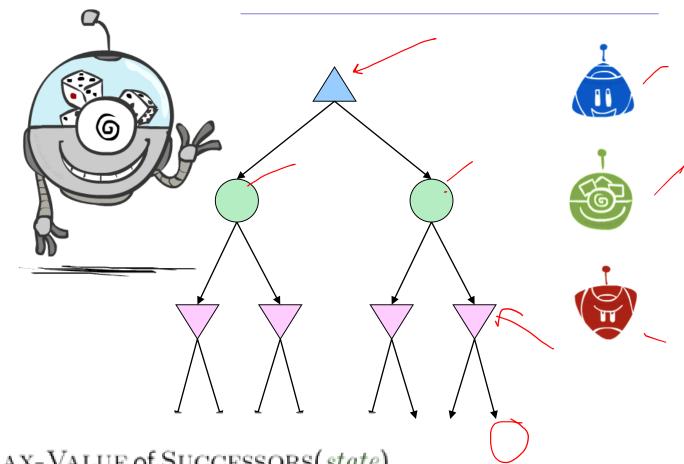


Other Game Types



Mixed Layer Types

- E.g. Backgammon
- Expecti-minimax
 - Environment is an extra "random agent" player that moves after each min/max agent
 - Each node computes the appropriate combination of its children



if state is a MAX node then

return the highest ExpectiMinimax-Value of Successors(state)

if state is a MIN node then

return the lowest ExpectiMinimax-Value of Successors(state)

if state is a chance node then

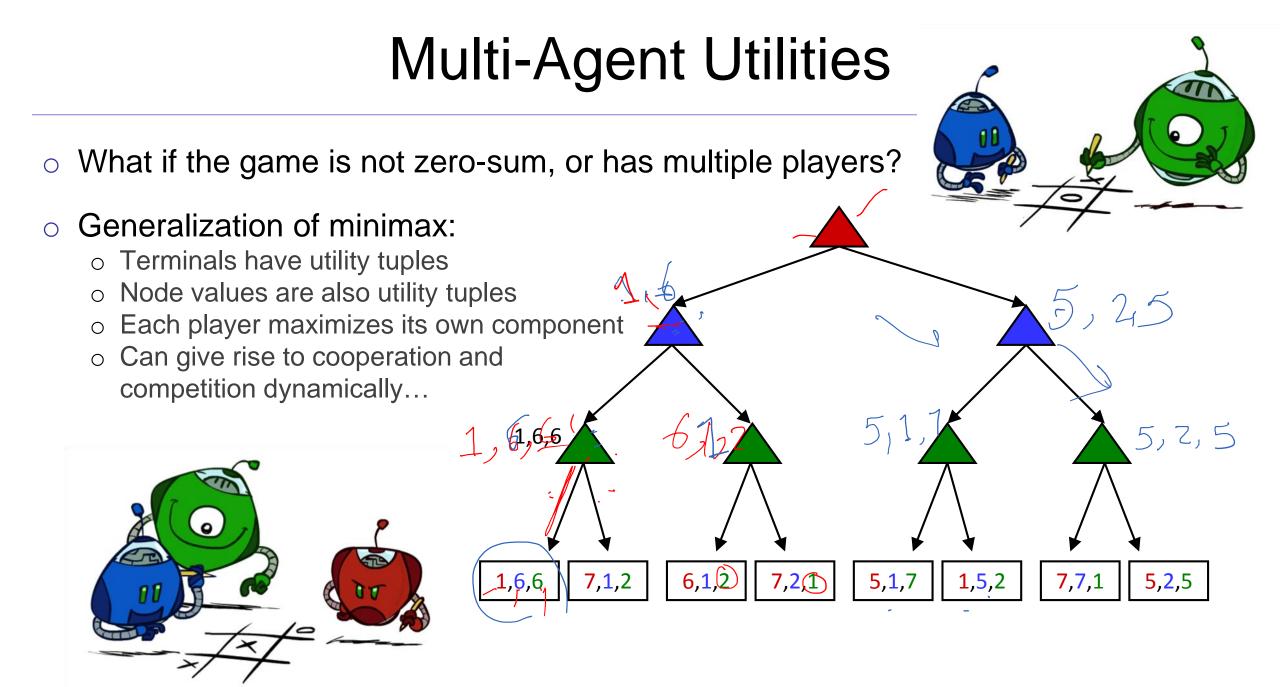
return average of ExpectiMinimax-Value of Successors(state)

Example: Backgammon

- Dice rolls increase b: 21 possible rolls with 2 dice
 - Backgammon ≈ 20 legal moves
 - o Depth 2 = $20 \times (21 \times 20)^3 = 1.2 \times 10^9$
- As depth increases, probability of reaching a given search node shrinks
 - So usefulness of search is diminished
 - So limiting depth is less damaging
 - But pruning is trickier...
- Historic AI: TDGammon uses depth-2 search + very good evaluation function + reinforcement learning: world-champion level play

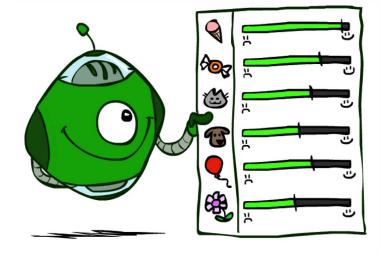


Image: Wikipedia



Utilities

Utilities: values that we assign to every state



- Why should we average utilities? Why not minimax?
- Principle of maximum expected utility:
 - A rational agent should choose the action that maximizes its expected utility, given its knowledge





Utilities

- Utilities are functions from outcomes (states of the world) to real numbers that describe an agent's preferences
- O Where do utilities come from?
 - In a game, may be simple (+1/-1)
 - Utilities summarize the agent's goals

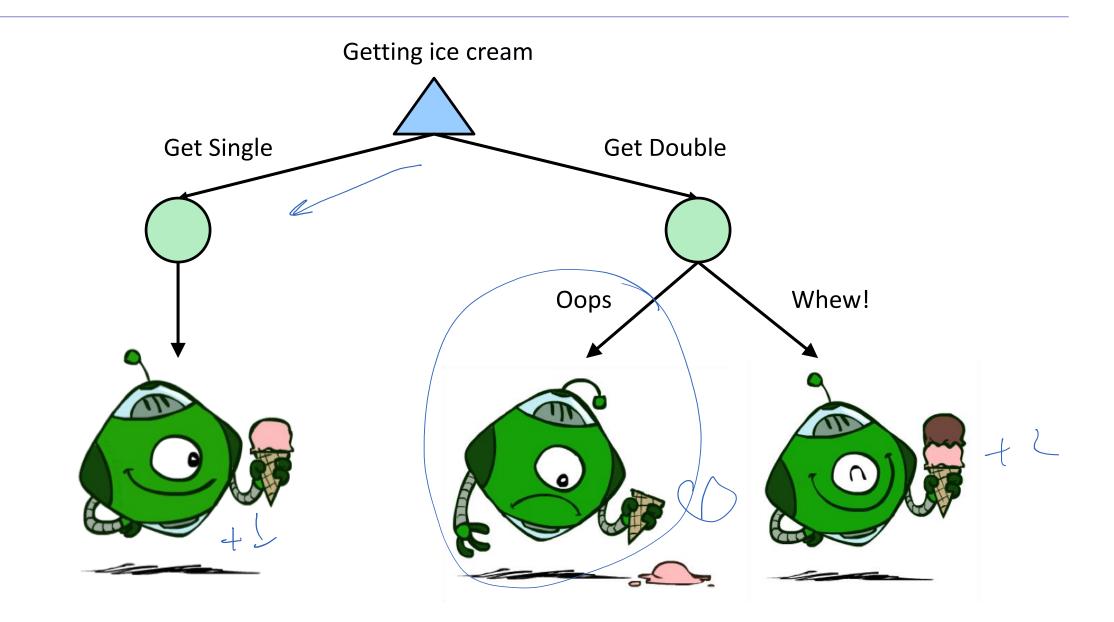




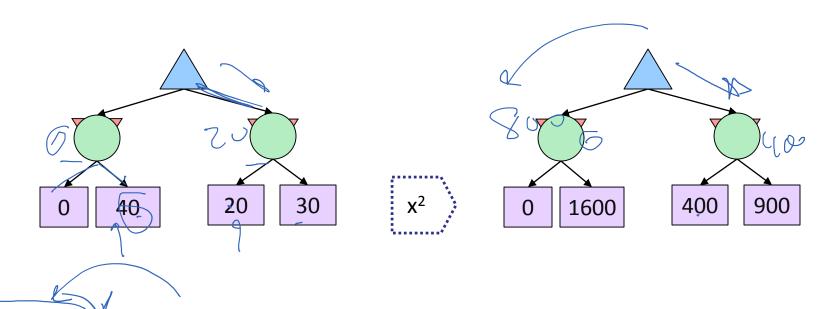




Utilities: Uncertain Outcomes



What Utilities to Use?



- For worst-case minimax reasoning, terminal function scale doesn't matter
 - We just want better states to have higher evaluations (get the ordering right)
 - We call this insensitivity to monotonic transformations
- For average-case expectimax reasoning, we need magnitudes to be meaningful

Next Topic: MDPs!