# Artificial Intelligence csc 665

# SearchII

8.29.2023

- Search: make decisions by looking ahead
- Logic: deduce new facts from existing facts
- Constraints: find a way to satisfy a given specification
- Probability: reason quantitatively about uncertainty
- Learning: make future predictions from past observations

# Recap

### Homework O

- Due yesterday at midnight.
- Reminder that the late policy allows you to submit up to 5 days late with a 10% penalty per day.
- Homework o is free points!

### Modeling (last time)

Start state:  $s_0 \in S$ 

Possible actions:  $Actions(s) \subseteq A$ 

Action cost:  $Cost(s, a) \in \mathbb{R}_{\geq 0}$ 

Transition model:  $Succ(s, a) \in S$ 

Goal test:  $IsEnd(s) \in \{True, False\}$ 

state space S, action set A, non-negative real numbers  $\mathbb{R}_{>0}$ 

### Backtracking search (last time)

Global state: minimum cost path, set of explored nodes

function search(s, path):

- if IsEnd(s):
  - update the minimum cost path
- for each action  $a \in Actions(s)$ :
  - if Succ(s, a) hasn't been explored yet:
    - add it to the explored set
    - extend path with Succ(s, a) and Cost(s, a)
    - recurse: search(Succ(s, a), path)

[fix goat.py]

# More inference algorithms

### Breadth-first and depth-first search

- Last time: backtracking search implemented recursively
- Today: BFS and DFS implemented iteratively
- Every iterative program can be implemented recursively, and vice-versa

### General approach

- Start with a frontier that contains  $s_0$ , and an empty explored set
- While the frontier is nonempty:
  - Pop a node s from the frontier
  - If IsEnd(s): return solution
  - Add s to the explored set
  - Expand s, adding Succ(s, a) to the frontier **for** each  $a \in Actions(s)$ , as long as it's neither in the frontier nor already explored

### BFS vs. DFS

- Breadth-first search (BFS)
  - Expands the shallowest node in the frontier
  - Explores nodes in order of increasing depth
  - Frontier is a queue (FIFO)
- Depth-first search (DFS)
  - Expands the deepest node in the frontier
  - Equivalent to a backtracking search that stops after the first solution
  - Frontier is a stack (LIFO)

### [maze examples]

### Two ways to analyze algorithms

#### Correctness

- Exact or approximate?
- If approximately correct, how far off from exactness?
- If exactly correct, under what conditions?

#### Efficiency

- Asymptotic analysis (big-oh)
- Time
- Space

### Correctness of search algorithms

- Backtracking search: returns shortest path for any cost function
- BFS: returns shortest path for (non-negative) constant cost function
- DFS: returns shortest path for zero cost function

### Efficiency of search algorithms

- Backtracking search: O(D) space,  $O(b^D)$  time
- BFS:  $O(b^d)$  space,  $O(b^d)$  time
- **DFS**: O(D) space,  $O(b^D)$

b actions per state, solution depth d, maximum depth D

### Summary

algorithm	cost function	space	time
backtracking	any	linear	exponential
BFS	constant	exponential	exponential
DFS	zero	linear	exponential

### Layered search

- BFS works because it explores in layers of equal depth
- But only if the cost function is constant
- Can we make the idea of a layered search work with non-constant action costs?

Yes, thanks to Dijkstra!

### Uniform Cost Search (UCS, Dijkstra's Algorithm)

- Start with a frontier that contains  $s_0$ , and an empty explored set
- While the frontier is nonempty:
  - Pop the node s with smallest priority p from the frontier
  - If IsEnd(s): return solution
  - Add s to the explored set
  - For each  $a \in Actions(s)$ ,
    - Get  $s' = \operatorname{Succ}(s, a)$
    - If s' is already explored: continue
    - Add s' to frontier with priority p + Cost(s, a)

### Correctness of UCS

**Theorem:** Assume action costs are non-negative. If a node s is popped from the frontier with priority p, then p is the cost of the min-cost path from  $s_0$  to s.

Proof: Take CSC 510 (or come to office hours).

Corollary: UCS computes the min-cost path to the goal node.

# Informed search

### Using domain knowledge

- So far: uninformed search
  - Algorithms that don't use problem-specific information
  - **Pro:** completely generic same algorithm works for all search problems
  - Con: can't useful domain knowledge
- Next: informed search
  - Use a heuristic function  $h: S \to \mathbb{R}$  to estimate progress toward goal