Announcements

- Homework 0: Math self-diagnostic **due tonight/Tuesday 9/3 at 11:59pm**
	- Strongly encouraged, optional, but important to check your preparedness for second half
- Project 0: Python Tutorial **due tomorrow/Wednesday 9/4 at 5pm**
	- Strongly encouraged, optional, get to know project/submission system)
- Homework 1: Search
	- Going out this week, **due next week Tuesday 9/10 11:59pm**
	- Two parts: electronic + written
- Project 1: Search
	- Going out this week, **due next week Friday 9/13 5pm**
	- Longer than most, and best way to test your programming preparedness
- Office hours
	- Starting this week, see course calendar on website
- **Sections**
	- Starting next week

CS 188: Artificial Intelligence

Search

Instructors: Pieter Abbeel & Igor Mordatch

University of California, Berkeley

[Many of these slides were originally created by Dan Klein and Pieter Abbeel for CS188 Intro to AI at UC Berkeley (ai.berkeley.edu)]

Today

- Agents that Plan Ahead
- **Search Problems**
- Uninformed Search Methods
	- Depth-First Search
	- Breadth-First Search
	- Uniform-Cost Search

Agents that Plan

Reflex Agents

- **Reflex agents:**
	- Choose action based on current percept (and maybe memory)
	- **May have memory or a model of the world's** current state
	- Do not consider the future consequences of their actions
	- **Example 2 Consider how the world IS**
- Can a reflex agent be rational?

[Demo: reflex optimal (L2D1)] [Demo: reflex optimal (L2D2)]

Demo Reflex: Optimal

Demo Reflex: Suboptimal / Stuck

Planning Agents

Planning agents:

- Ask "what if"
- **Decisions based on (hypothesized)** consequences of actions
- Must have a model of how the world evolves in response to actions
- Must formulate a goal (test)
- Consider how the world WOULD BE
- Planning agents' properties:
	- Completeness (or not)
	- Optimality (or not)
- Planning vs. replanning and the control of the control of

[Demo: mastermind (L2D4)]

Video of Demo Replanning

Video of Demo Mastermind

Search Problems

Search Problems

A search problem consists of:

■ A state space

■ A successor function (with actions, costs)

- A start state and a goal test
- A solution is a sequence of actions (a plan) which transforms the start state to a goal state

Search Problems Are Models

Example: Traveling in Romania

- State space:
	- **Cities**
- **Successor function:**
	- Roads: Go to adjacent city with cost = distance
- Start state:
	- Arad
- Goal test:
	- \blacksquare Is state == Bucharest?
- **Solution?**

What's in a State Space?

A search state keeps only the details needed for planning (abstraction)

- Problem: Pathing
	- States: (x,y) location
	- **Actions: NESW**
	- **Successor: update location** only
	- Goal test: is (x,y) =END
- Problem: Eat-All-Dots
	- States: $\{(x,y), \text{dot boolean}\}$
	- **Actions: NESW**
	- **Successor: update location** and possibly a dot boolean
	- Goal test: dots all false

State Space Sizes?

■ World state:

- **Agent positions: 120**
- **Food presence: 30 booleans**
- Ghost positions: 12
- Agent facing: NESW
- **How many**
	- World states? 120x(2³⁰)x(12²)x4
	- **States for pathing?** 120
	- States for eat-all-dots? 120x(230)

Quiz: Safe Passage

- Problem: eat all dots while keeping the ghosts perma-scared
- What does the state space have to specify?
	- (agent position, dot booleans, power pellet booleans, remaining scared time)

State Space Graphs and Search Trees

State Space Graphs

- **State space graph: A mathematical** representation of a search problem
	- Nodes are (abstracted) world configurations
	- **Arcs represent successors (action results)**
	- The goal test is a set of goal nodes (maybe only one)
- In a state space graph, each state occurs only once!
- We can rarely build this full graph in memory (it's too big), but it's a useful idea

State Space Graphs

- State space graph: A mathematical representation of a search problem
	- Nodes are (abstracted) world configurations
	- **EXPLO Arcs represent successors (action results)**
	- The goal test is a set of goal nodes (maybe only one)
- In a state space graph, each state occurs only once!
- We can rarely build this full graph in memory (it's too big), but it's a useful idea

Tiny state space graph for a tiny search problem

Search Trees

A search tree:

- A "what if" tree of plans and their outcomes
- The start state is the root node
- Children correspond to successors
- Nodes show states, but correspond to PLANS that achieve those states
- For most problems, we can never actually build the whole tree

State Space Graphs vs. Search Trees

Each NODE in in the search tree is an entire PATH in the state space graph.

We construct both on demand – and we construct as little as possible.

Quiz: State Space Graphs vs. Search Trees

Consider this 4-state graph: How big is its search tree (from S)?

Quiz: State Space Graphs vs. Search Trees

Consider this 4-state graph:

How big is its search tree (from S)?

Important: Lots of repeated structure in the search tree!

Tree Search

Search Example: Romania

Searching with a Search Tree

Search:

- Expand out potential plans (tree nodes)
- Maintain a fringe of partial plans under consideration
- Try to expand as few tree nodes as possible

General Tree Search

function TREE-SEARCH (problem, strategy) returns a solution, or failure initialize the search tree using the initial state of $problem$ loop do

if there are no candidates for expansion then return failure choose a leaf node for expansion according to *strategy* if the node contains a goal state then return the corresponding solution else expand the node and add the resulting nodes to the search tree end

Important ideas:

- **Fringe**
- **Expansion**
- **Exploration strategy**
- Main question: which fringe nodes to explore?

Example: Tree Search

Example: Tree Search

Depth-First Search

Depth-First Search

deepest node first

Implementation: Fringe is a LIFO stack

Search Algorithm Properties

Search Algorithm Properties

- Complete: Guaranteed to find a solution if one exists?
- Optimal: Guaranteed to find the least cost path?
- Time complexity?
- **Space complexity?**
- Cartoon of search tree:
	- b is the branching factor
	- m is the maximum depth
	- solutions at various depths
- Number of nodes in entire tree?
	- \blacksquare 1 + b + b² + b^m = O(b^m)

Depth-First Search (DFS) Properties

- What nodes does DFS expand?
	- Some left prefix of the tree.
	- Could process the whole tree!
	- If m is finite, takes time $O(b^m)$
- How much space does the fringe take?
	- Only has siblings on path to root, so O(bm)
- Is it complete?
	- m could be infinite, so only if we prevent cycles (more later)
- Is it optimal?
	- No, it finds the "leftmost" solution, regardless of depth or cost

Breadth-First Search

Breadth-First Search

Strategy: expand a shallowest node first

Implementation: Fringe is a FIFO queue

Breadth-First Search (BFS) Properties

- What nodes does BFS expand?
	- Processes all nodes above shallowest solution
	- Let depth of shallowest solution be s
	- Search takes time $O(b^s)$
- How much space does the fringe take?
	- Has roughly the last tier, so $O(b^s)$
- Is it complete?
	- s must be finite if a solution exists, so yes!
- Is it optimal?
	- Only if costs are all 1 (more on costs later)

Quiz: DFS vs BFS

Quiz: DFS vs BFS

When will BFS outperform DFS?

When will DFS outperform BFS?

[Demo: dfs/bfs maze water (L2D6)]

Video of Demo Maze Water DFS/BFS (part 1)

Video of Demo Maze Water DFS/BFS (part 2)

Iterative Deepening

- Idea: get DFS's space advantage with BFS's time / shallow-solution advantages
	- Run a DFS with depth limit 1. If no solution...
	- Run a DFS with depth limit 2. If no solution...
	- Run a DFS with depth limit 3. …..
- Isn't that wastefully redundant?
	- Generally most work happens in the lowest level searched, so not so bad!

Cost-Sensitive Search

BFS finds the shortest path in terms of number of actions. It does not find the least-cost path. We will now cover a similar algorithm which does find the least-cost path.

Uniform Cost Search

Uniform Cost Search

Strategy: expand a cheapest node first:

Fringe is a priority queue comple is a priority queue
(priority: cumulative cost) (S)

Uniform Cost Search (UCS) Properties

- What nodes does UCS expand?
	- Processes all nodes with cost less than cheapest solution!
	- If that solution costs C^* and arcs cost at least ε , then the "effective depth" is roughly C^{\ast}/ε
	- Takes time O(b^{C*/ε}) (exponential in effective depth)
- How much space does the fringe take?
	- Has roughly the last tier, so O(b^{C*/ε})
- Is it complete?
	- Assuming best solution has a finite cost and minimum arc cost is positive, yes!
- Is it optimal?
	- Yes! (Proof next lecture via A*)

Uniform Cost Issues

■ Remember: UCS explores increasing cost contours

■ The good: UCS is complete and optimal!

- **The bad:**
	- **Explores options in every "direction"**
	- No information about goal location

We'll fix that soon!

[Demo: empty grid UCS (L2D5)] [Demo: maze with deep/shallow water DFS/BFS/UCS (L2D7)]

Video of Demo Empty UCS

Video of Demo Maze with Deep/Shallow Water --- DFS, BFS, or UCS? (part 1)

Video of Demo Maze with Deep/Shallow Water --- DFS, BFS, or UCS? (part 2)

Video of Demo Maze with Deep/Shallow Water --- DFS, BFS, or UCS? (part 3)

The One Queue

- All these search algorithms are the same except for fringe strategies
	- Conceptually, all fringes are priority queues (i.e. collections of nodes with attached priorities)
	- Practically, for DFS and BFS, you can avoid the log(n) overhead from an actual priority queue, by using stacks and queues
	- Can even code one implementation that takes a variable queuing object

Search Gone Wrong?

Estimated Total Time: 47 hours, 31 minutes

Search and Models

- **Search operates over** models of the world
	- The agent doesn't actually try all the plans out in the real world!
	- **Planning is all "in"** simulation"
	- Your search is only as good as your models…

