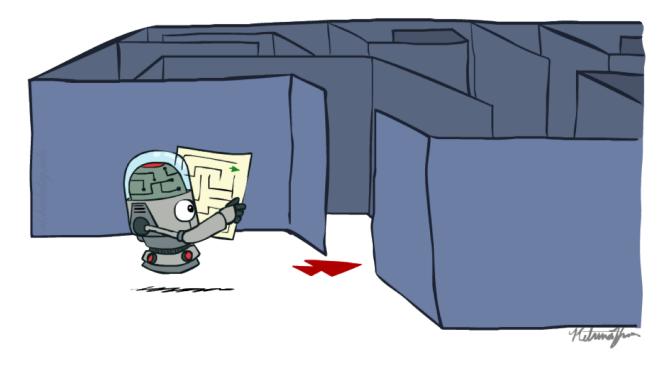
### CS 188: Artificial Intelligence

#### Search



Instructors: Dan Klein and Pieter Abbeel

University of California, Berkeley

## Today

Agents that Plan Ahead

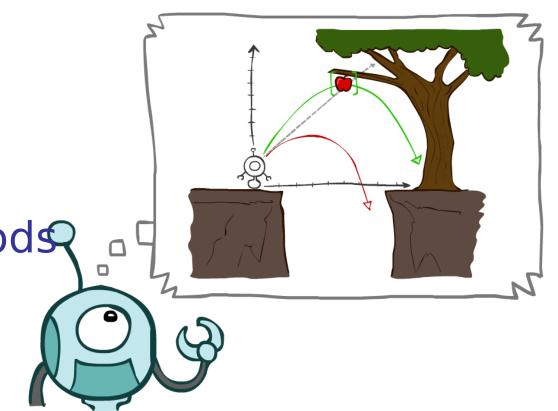
Search Problems

Uninformed Search Methods

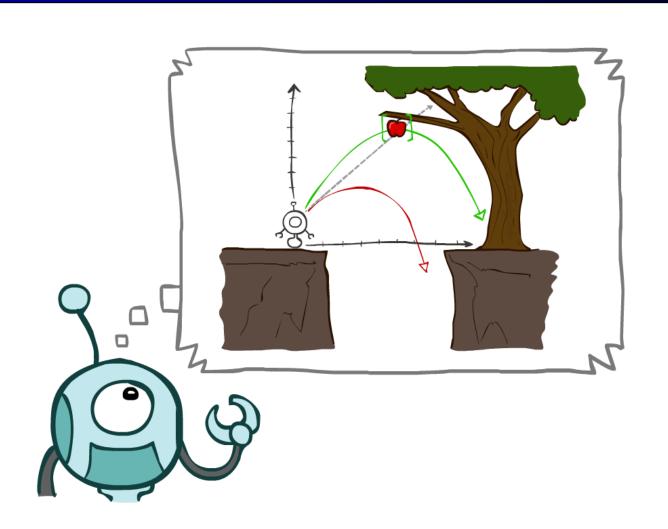
Depth-First Search

Breadth-First Search

Uniform-Cost Search

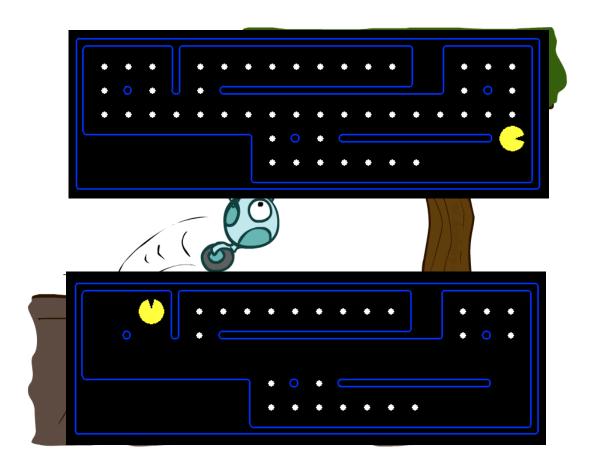


# Agents that Plan



### 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
  - Consider how the world IS
- Can a reflex agent be rational?



[Demo: reflex optimal [Demo] reflex optimal

# Video of Demo Reflex Optimal

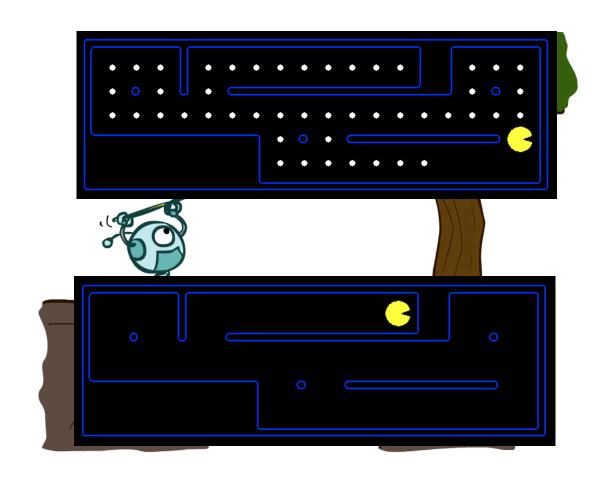


### Video of Demo Reflex Odd



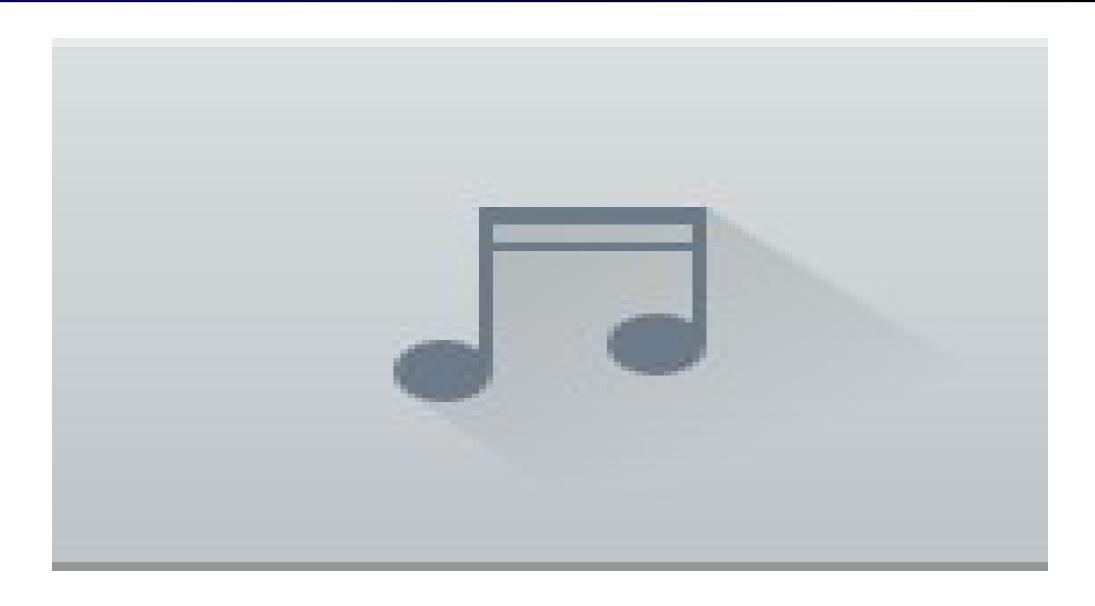
### Planning Agents

- 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
- Optimal vs. complete planning
- Planning vs. replanning



[Demo: replanning (Demo: mastermind

# Video of Demo Replanning



### Video of Demo Mastermind



### Search Problems



### Search Problems

A search problem consists of:

A state space





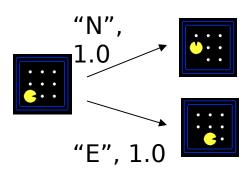








A successor function (with actions, costs)

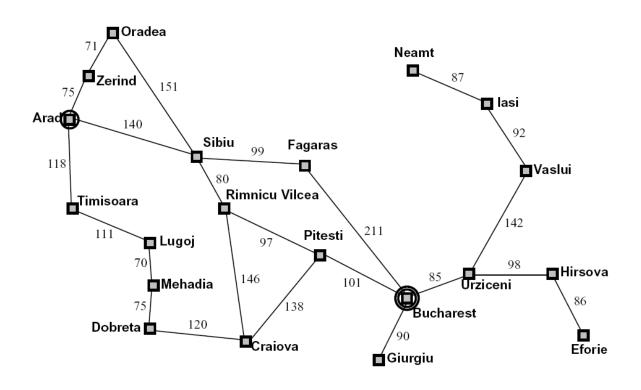


A start state and a goal test

### 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:
  - Is state == Bucharest?
- Solution?

### What's in a State Space?

The world state includes every last detail of the environment



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

- Problem: Pathing
  - States: (x,y) location
  - Actions: NSEW
  - Successor: update location only
  - Goal test: is (x,y)=END

- Problem: Eat-All-Dots
  - States: {(x,y), dot booleans}
  - Actions: NSEW
  - Successor: update location and possibly a dot boolean
  - Goal test: dots all false

## State Space Sizes?

#### World state:

Agent positions: 120

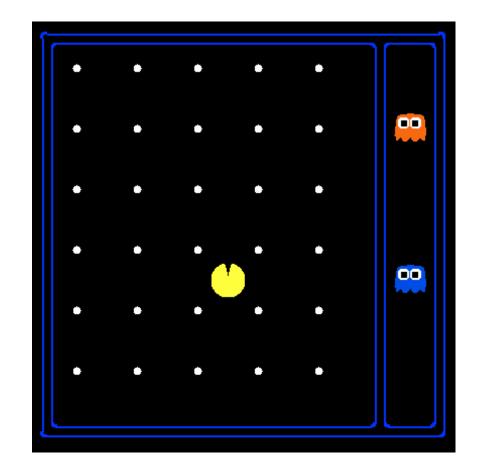
Food count: 30

Ghost positions: 12

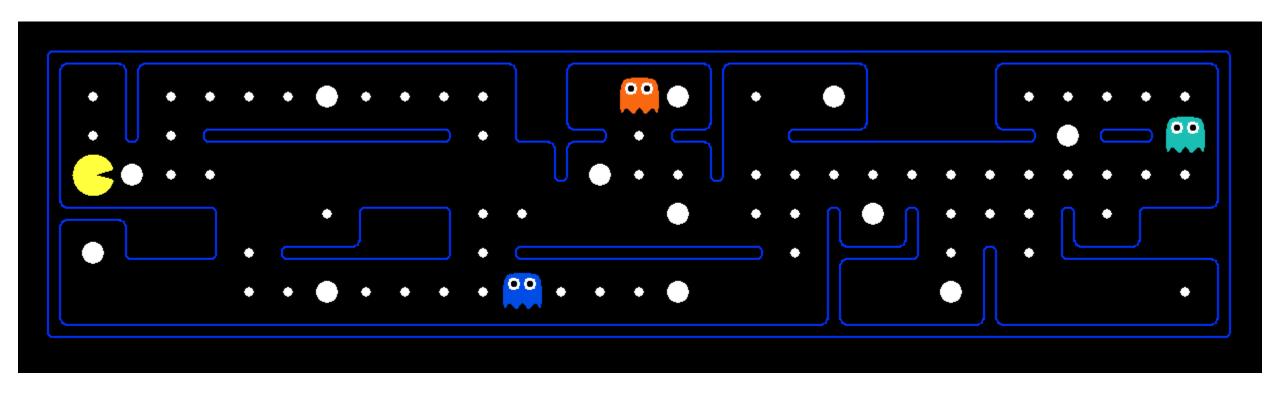
Agent facing: NSEW

#### How many

- World states?120x(2<sup>30</sup>)x(12<sup>2</sup>)x4
- States for pathing?120
- States for eat-all-dots?120x(2<sup>30</sup>)

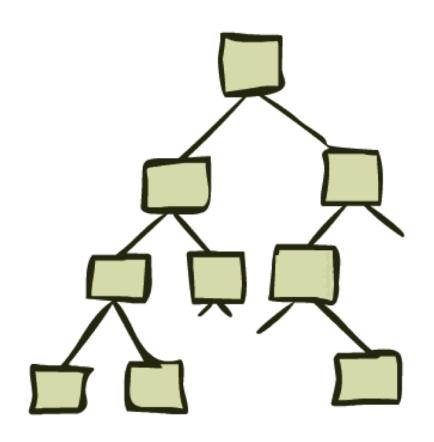


### Quiz: Safe Passage



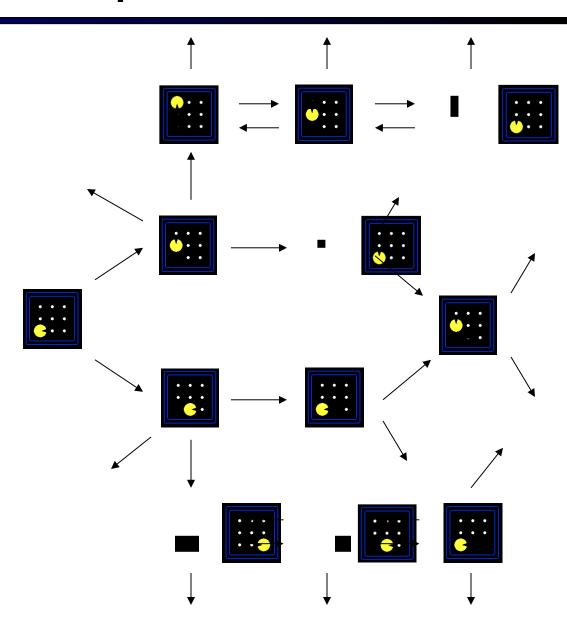
- Problem: eat all dots while keeping the ghosts perma-scared
- What does the state space have to specify?
  - (agent position, dot booleans, power pellet booleans, remaining scared time)

## State Space Graphs and Search Trees



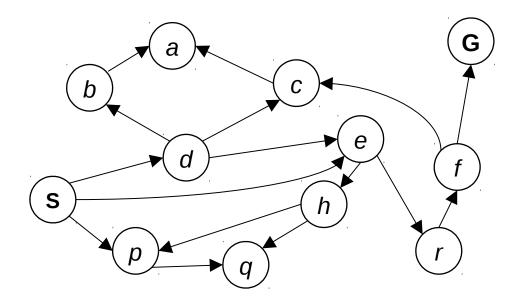
### State Space Graphs

- 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



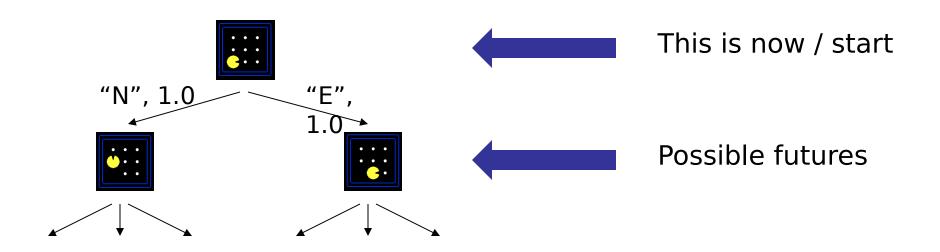
### State Space Graphs

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Tiny search graph for a tiny search problem

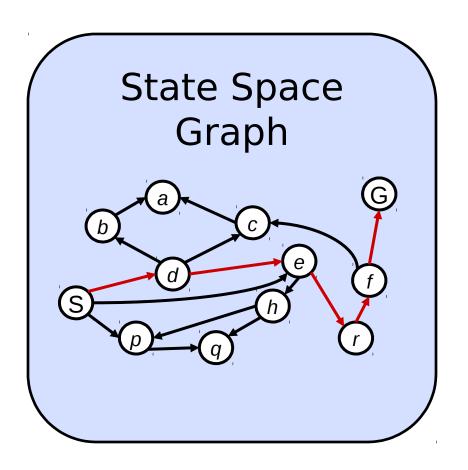
#### Search Trees



#### A search tree:

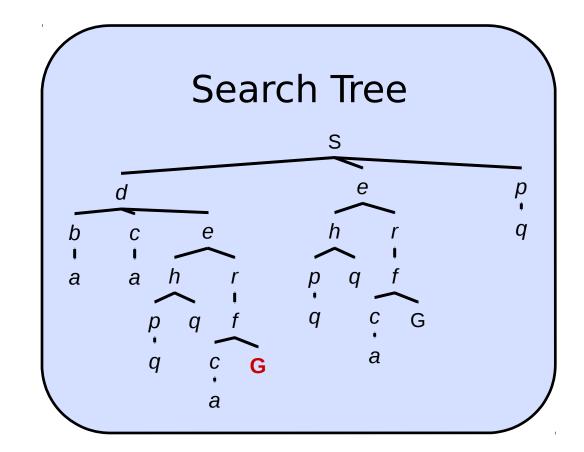
- A "what if" tree of plans and their outcomes
- The start state is the root node
- Children correspond to successors
- Nodes show states, but correspond to PLANS that achieve those states
- For most problems, we can never actually build the whole tree

### State Space Graphs vs. Search Trees



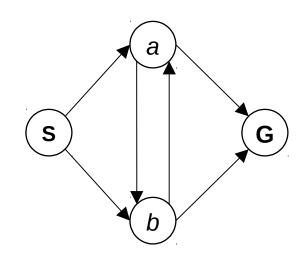
Each NODE in in the search tree is an entire PATH in the state space graph.

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



## Quiz: State Space Graphs vs. Search Trees

Consider this 4-state graph:

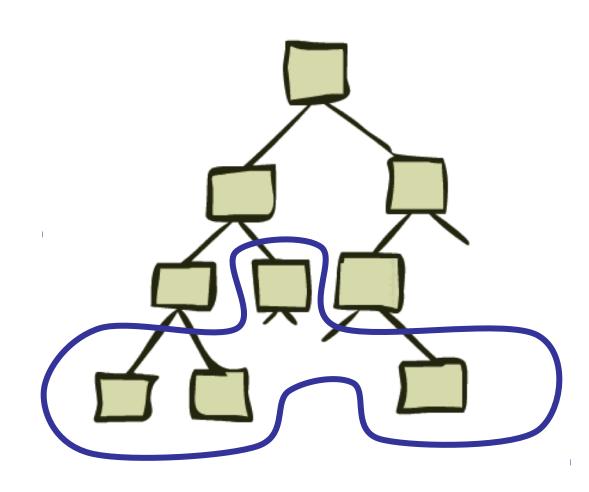


How big is its search tree (from S)?

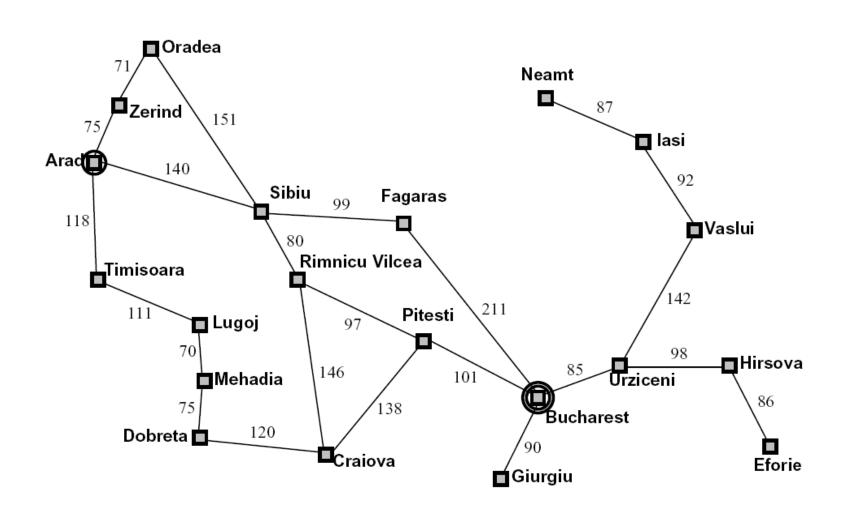


Important: Lots of repeated structure in the search tree!

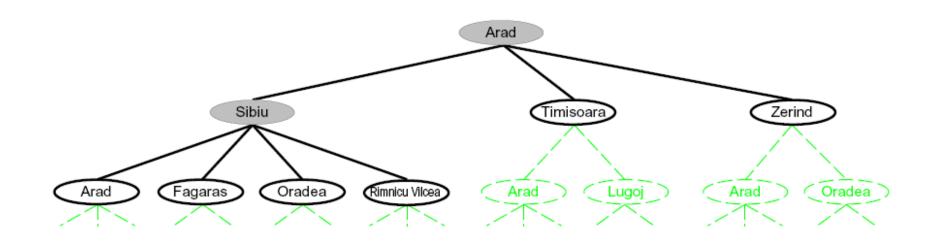
### Tree Search



# 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

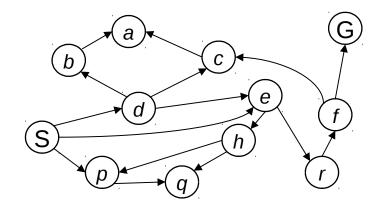
#### General Tree Search

```
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

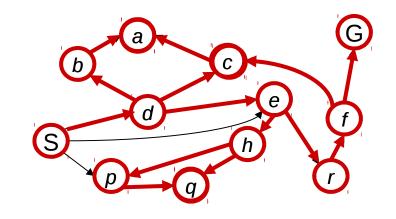


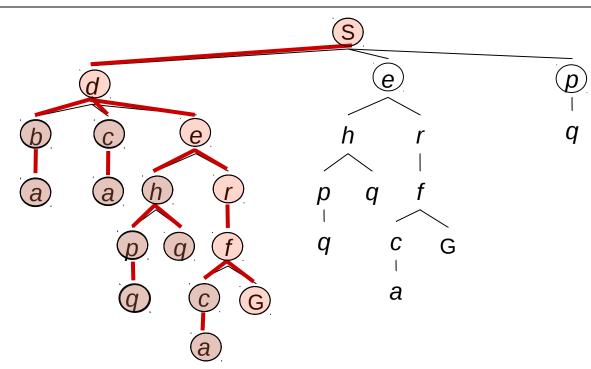
# Depth-First Search



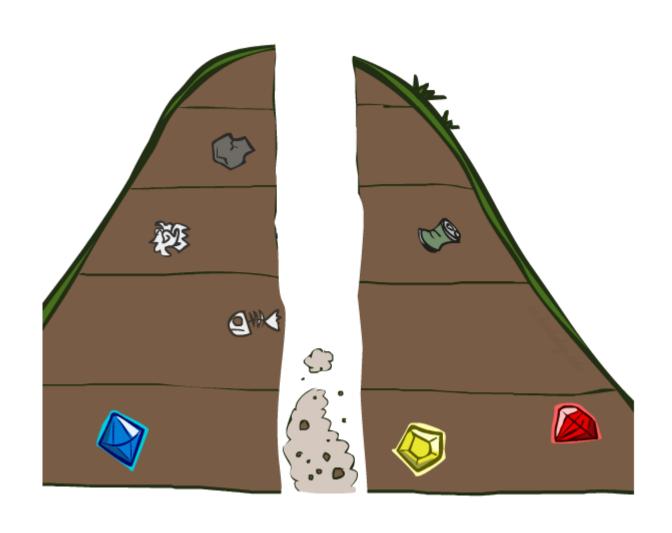
# Depth-First Search

Strategy: expand a deepest node first Implementation: Fringe is a LIFO stack



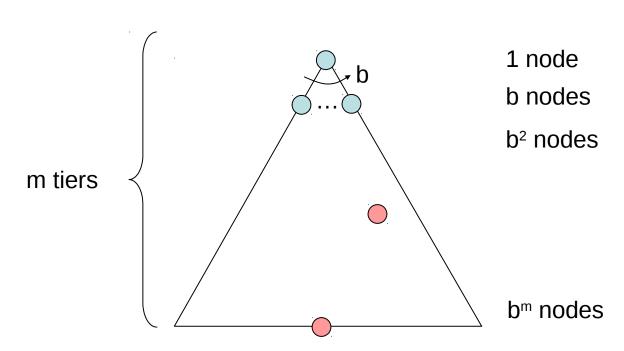


### Search Algorithm Properties



### Search Algorithm Properties

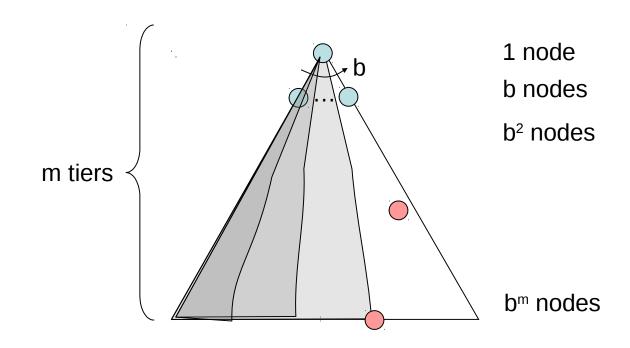
- Complete: Guaranteed to find a solution if one exists?
- Optimal: Guaranteed to find the least cost path?
- Time complexity?
- Space complexity?
- Cartoon of search tree:
  - b is the branching factor
  - m is the maximum depth
  - solutions at various depths



- Number of nodes in entire tree?
  - $1 + b + b^2 + \dots b^m = O(b^m)$

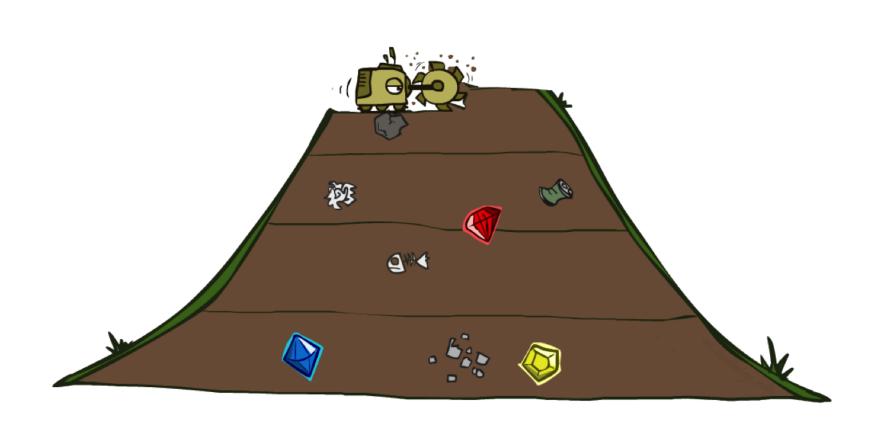
### Depth-First Search (DFS) Properties

- What nodes DFS expand?
  - Some left prefix of the tree.
  - Could process the whole tree!
  - If m is finite, takes time O(b<sup>m</sup>)
- 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)



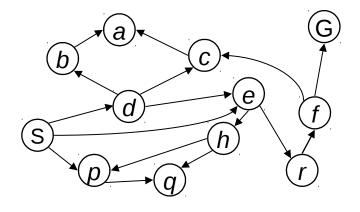
Is it optimal?

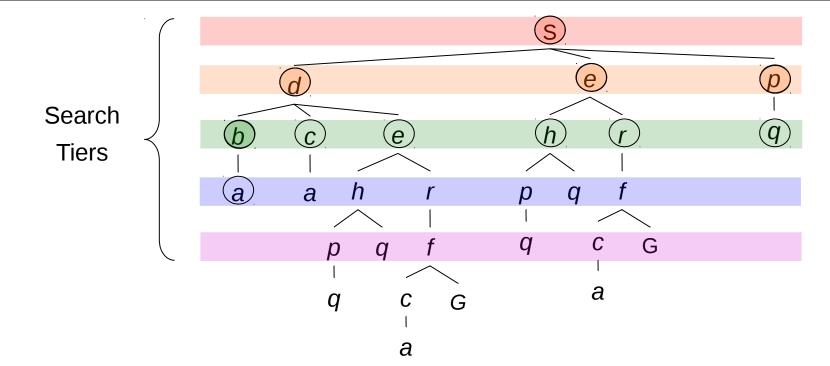
### Breadth-First Search



### Breadth-First Search

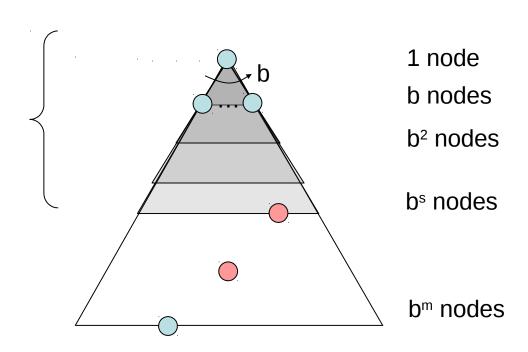
Strategy: expand a shallowest node first Implementation: Fringe is a FIFO queue





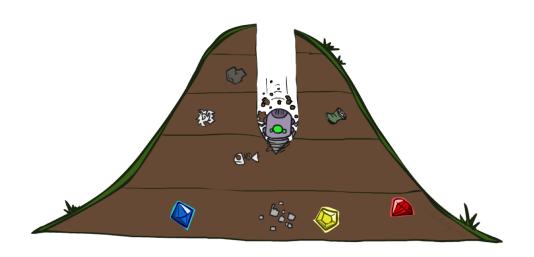
# Breadth-First Search (BFS) Properties

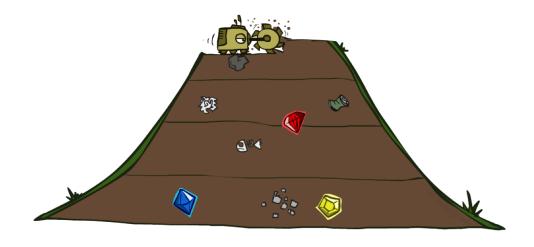
- What nodes does BFS expand?
  - Processes all nodes above shallowest solution
  - Let depth of shallowest solution be s
  - Search takes time O(bs)
- How much space does the fringe take?
  - Has roughly the last tier, so O(b<sup>s</sup>)
- Is it complete?
  - so yes!
- s must be finite if a solution exists,



Is it optimal?

# Quiz: DFS vs BFS



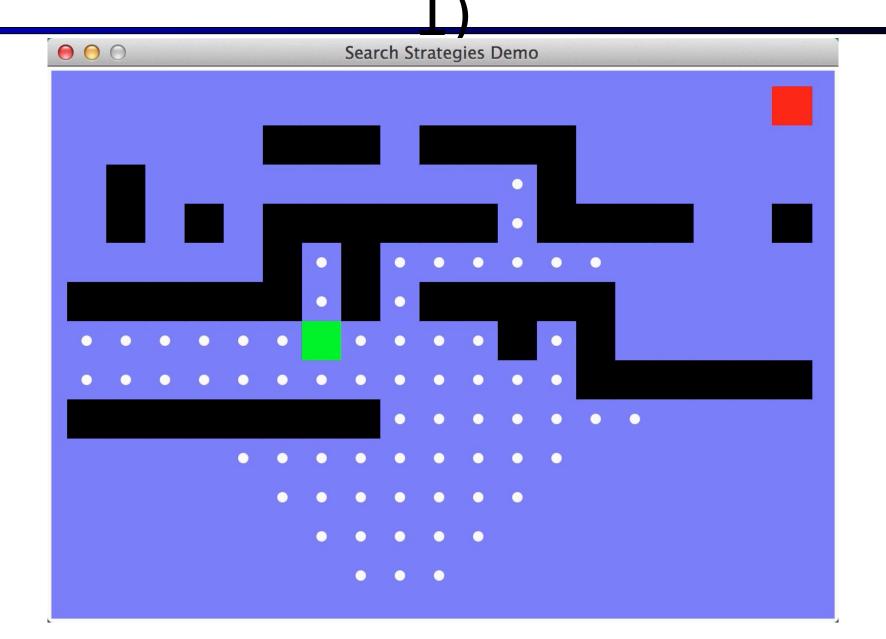


### Quiz: DFS vs BFS

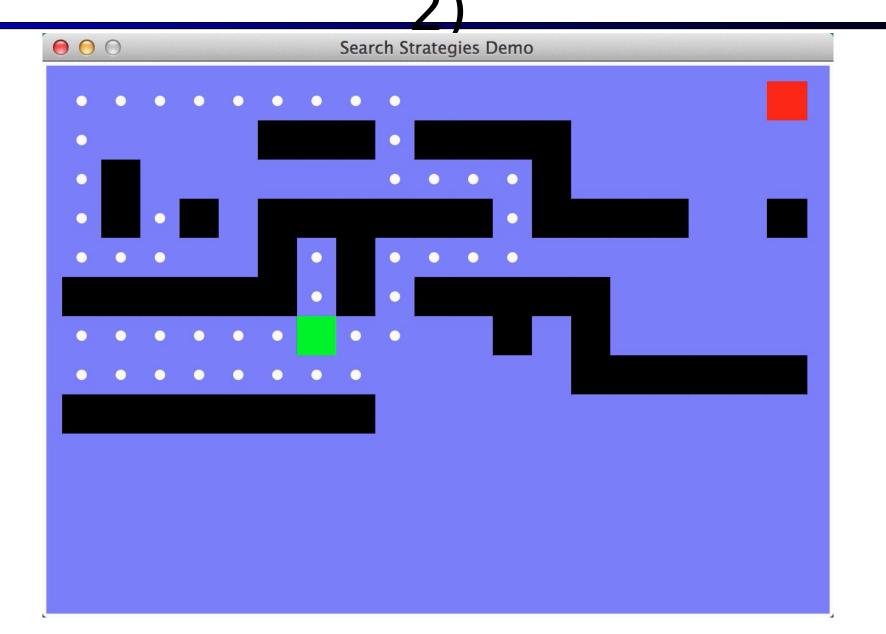
When will BFS outperform DFS?

When will DFS outperform BFS?

## Video of Demo Maze Water DFS/BFS (part



# Video of Demo Maze Water DFS/BFS (part

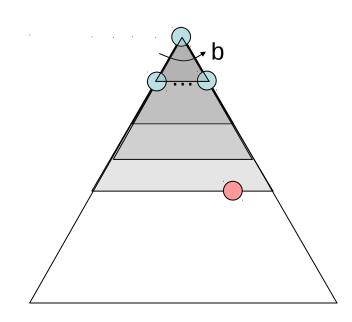


## Iterative Deepening

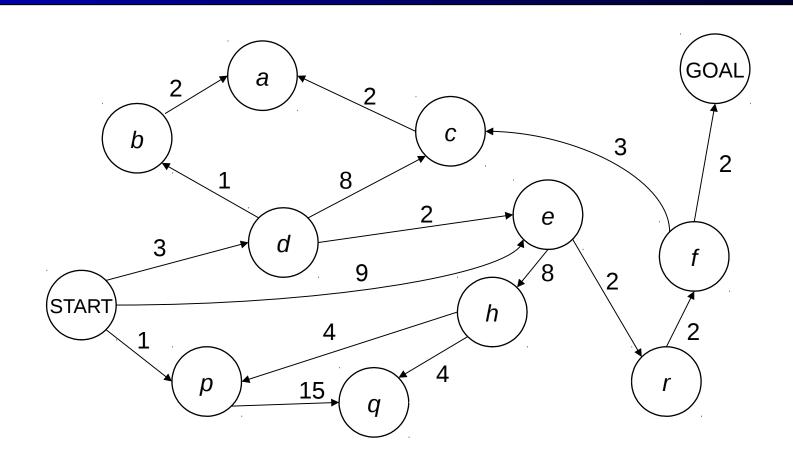
- Idea: get DFS's space advantage with BFS's time / shallow-solution advantages
  - Run a DFS with depth limit 1. If no solution...
  - Run a DFS with depth limit 2. If no solution...
  - Run a DFS with depth limit 3. .....



Generally most work happens in the lowest level searched, so not so bad!

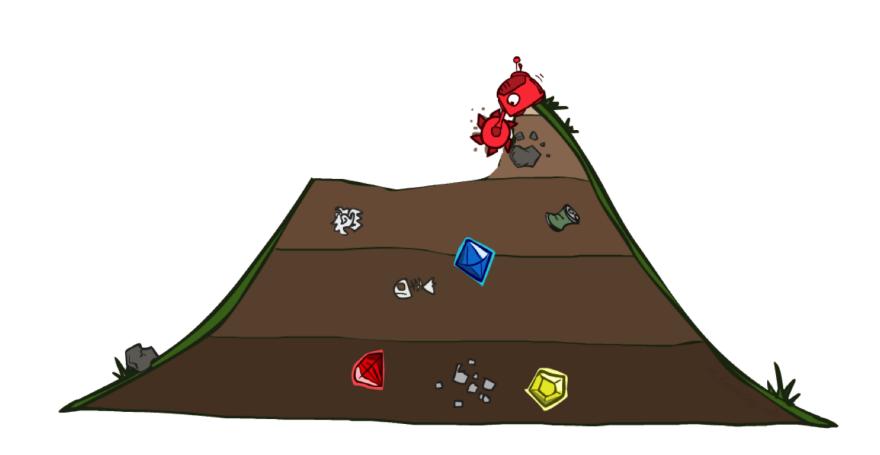


#### Cost-Sensitive Search



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.

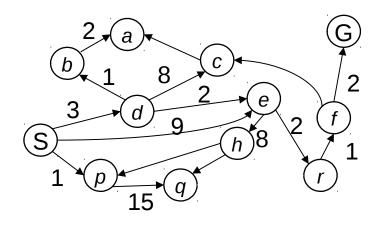
### Uniform Cost Search

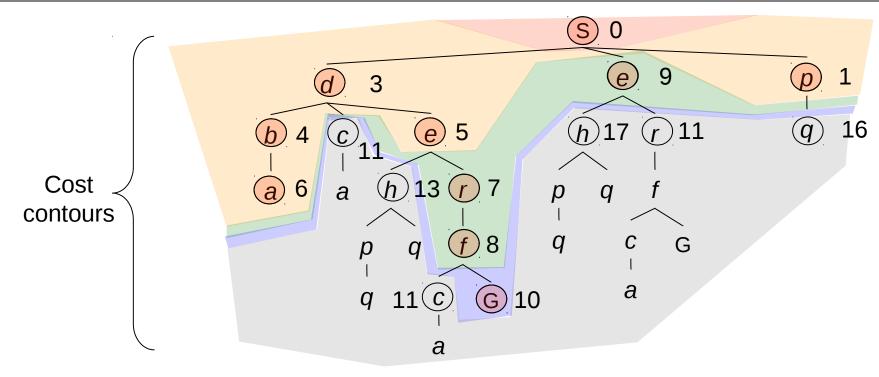


#### Uniform Cost Search

Strategy: expand a cheapest node first:

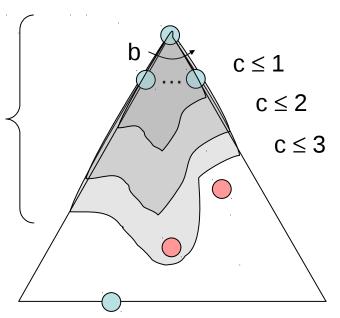
Fringe is a priority queue (priority: cumulative cost)





## 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  $\varepsilon$  ,  $C^*/\varepsilon$  "tiers" then the "effective depth" is roughly  $C^*/\varepsilon$
  - Takes time  $O(b^{C*/\epsilon})$  (exponential in effective depth)
- How much space does the fringe take?
  - Has roughly the last tier, so  $O(b^{C*/\epsilon})$
- Is it complete?
  - Assuming best solution has a finite cost and minimum arc cost is positive, yes!
- Is it optimal?

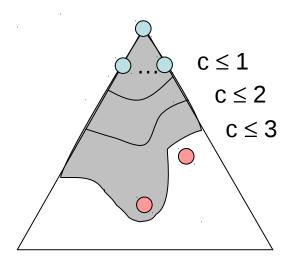


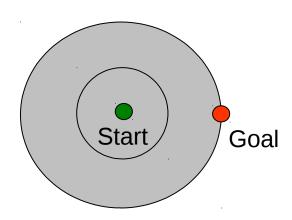
#### Uniform Cost Issues

Remember: UCS explores increasing cost contours

The good: UCS is complete and optimal!

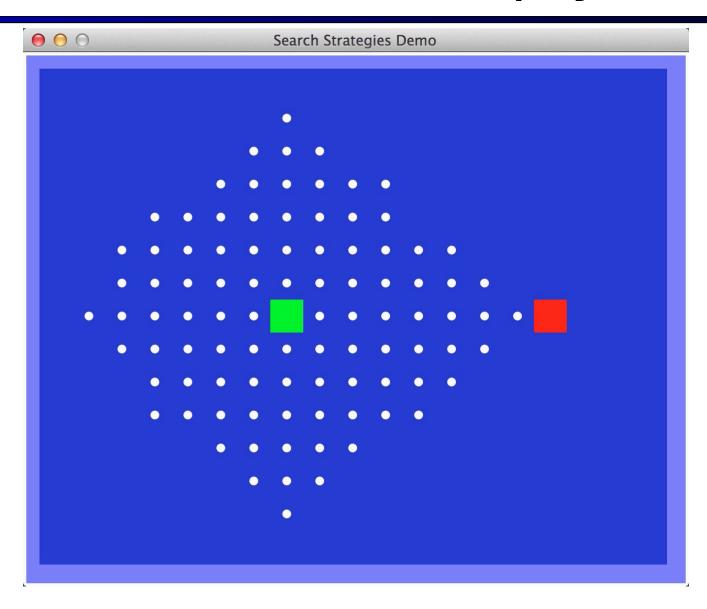
- The bad:
  - Explores options in every "direction"
  - No information about goal location



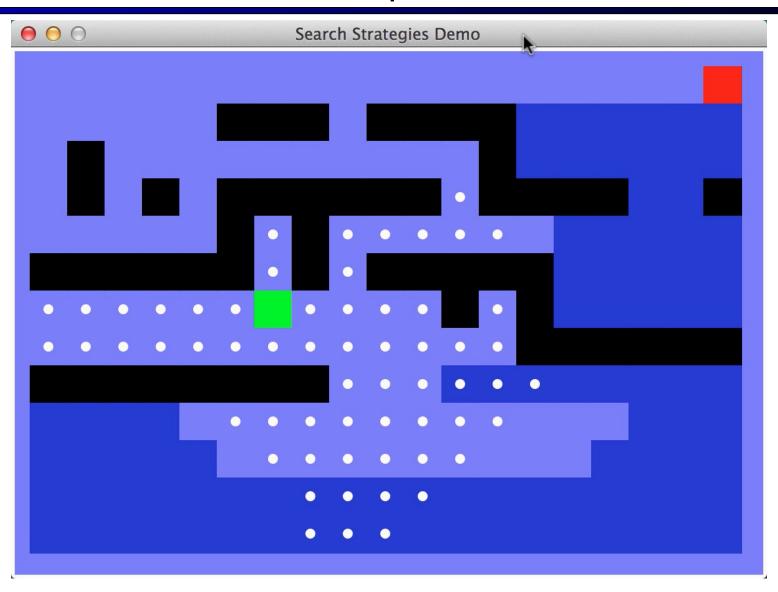


[Demo: empty grid UCS (L2D5)] [Demo: maze with deep/shallow water DFS/BFS/UCS (L2D7)]

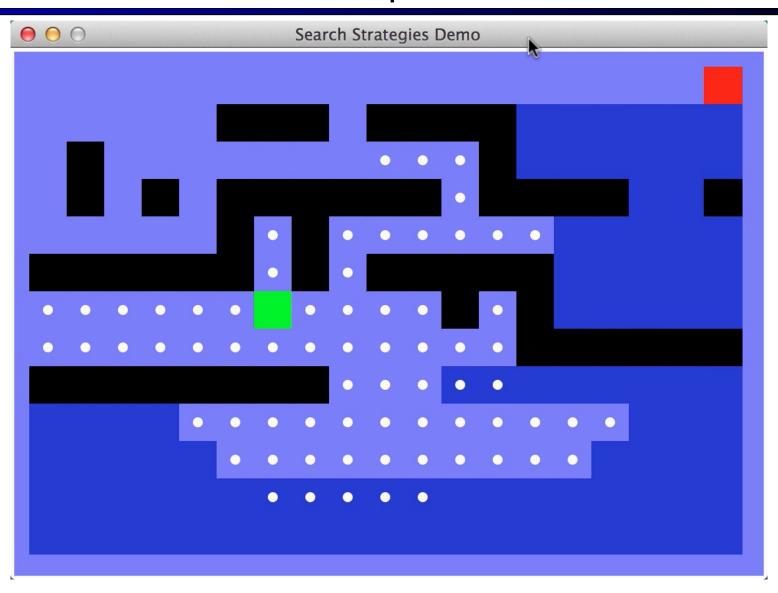
## Video of Demo Empty UCS



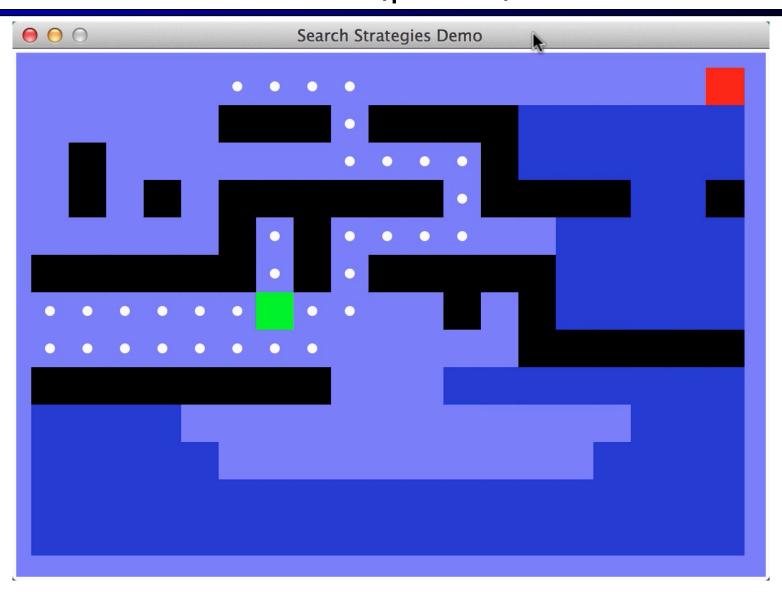
# Video of Demo Maze with Deep/Shallow Water --- DFS, BFS, or UCS? (part 1)



# 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)



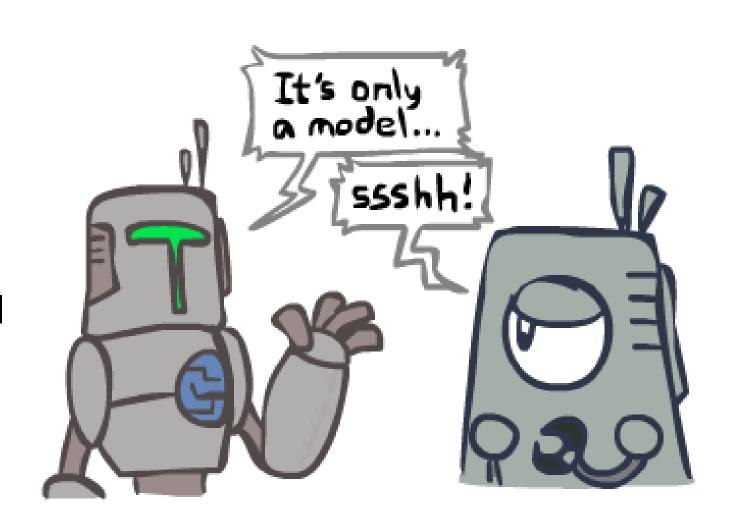
### The One Queue

- 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



#### 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



## Search Gone Wrong?

