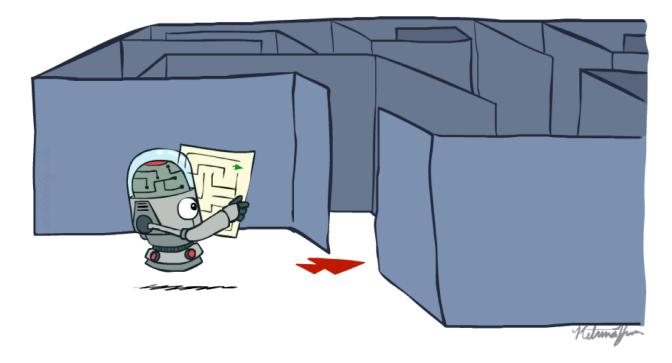
## CS 188: Artificial Intelligence

#### Search



Instructor: Nikita Kitaev

University of California, Berkeley

[These slides adapted from Dan Klein and Pieter Abbeel]

#### Announcements

#### Exam Times

- Midterm 1: July 13, 12pm-2pm (Alternative 12 hours later at 12am-2am)
- Midterm 2: July 29, 12pm-2pm (Alternative 12 hours later at 12am-2am)
- Final: August 13, 12pm-3pm (Alternative 12 hours later at 12am-3am)

#### Sections

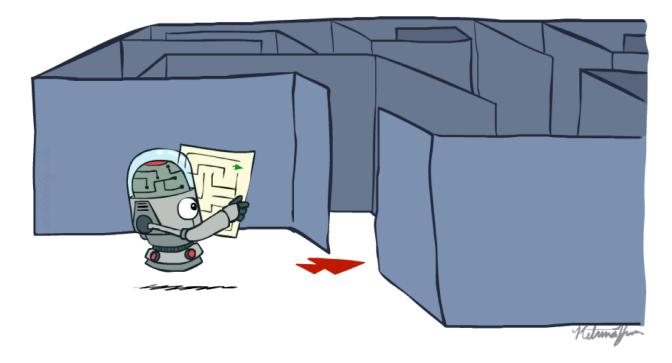
- Start tomorrow, see Piazza for times
- You can go to any section

#### Office Hours

- Start today
- Ryan's OH are today at 3pm-5pm. Cathy's OH today at 11pm-12am. Nathan's OH tomorrow at 7am-9am
- Written Assessment 1 will be released tonight
  - due next Monday before start of lecture
- Lecture 1 slides and recording
  - See Piazza

## CS 188: Artificial Intelligence

#### Search



Instructor: Nikita Kitaev

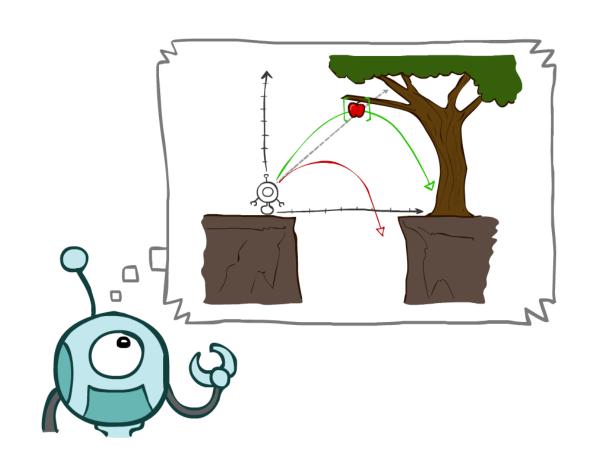
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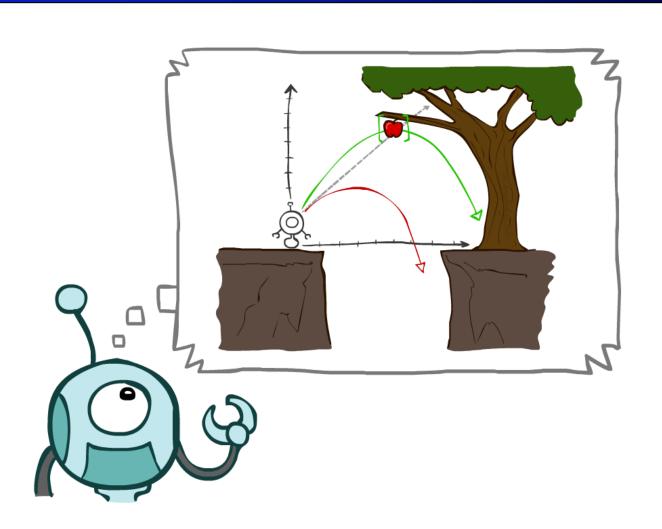
## Today

- Agents that Plan Ahead
- Search Problems

- Uninformed Search Methods
  - Depth-First Search
  - Breadth-First Search
  - Uniform-Cost Search



# Agents that Plan



### Reflex Agents

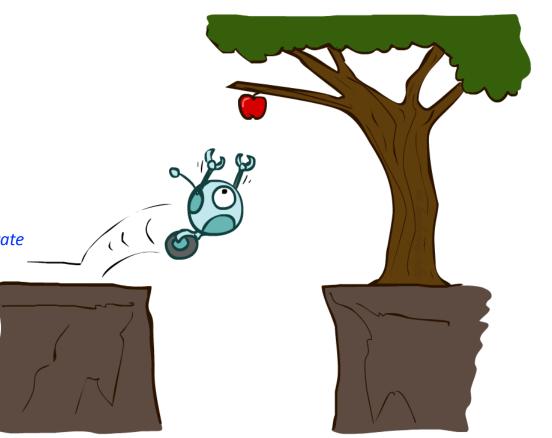
#### Reflex agents:

- 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

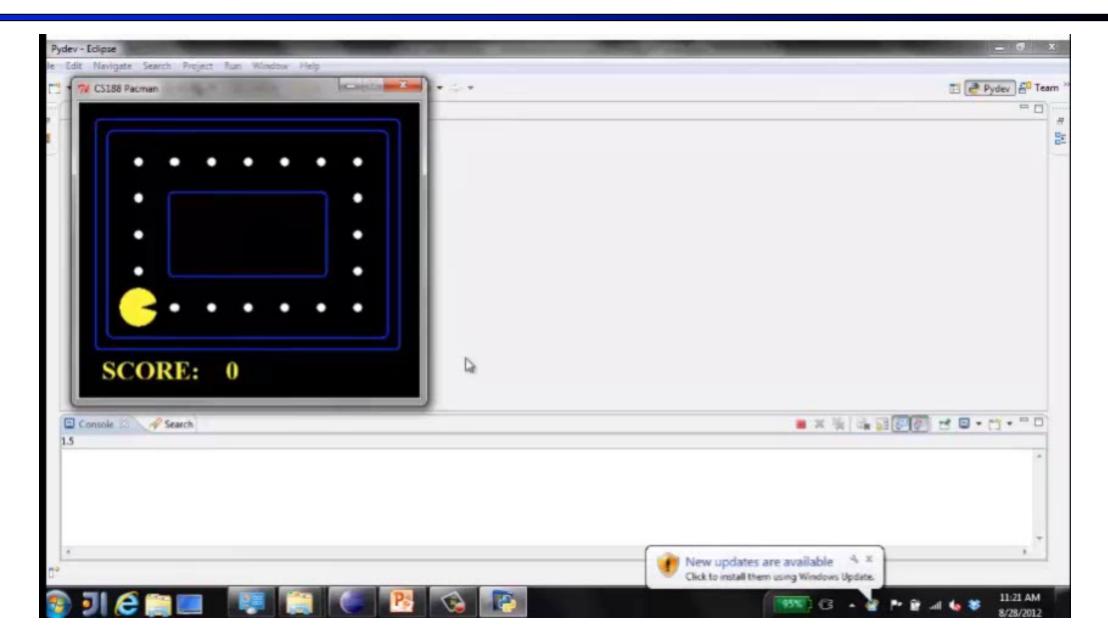
is just how the world is in current state or any given state

Can a reflex agent be rational?

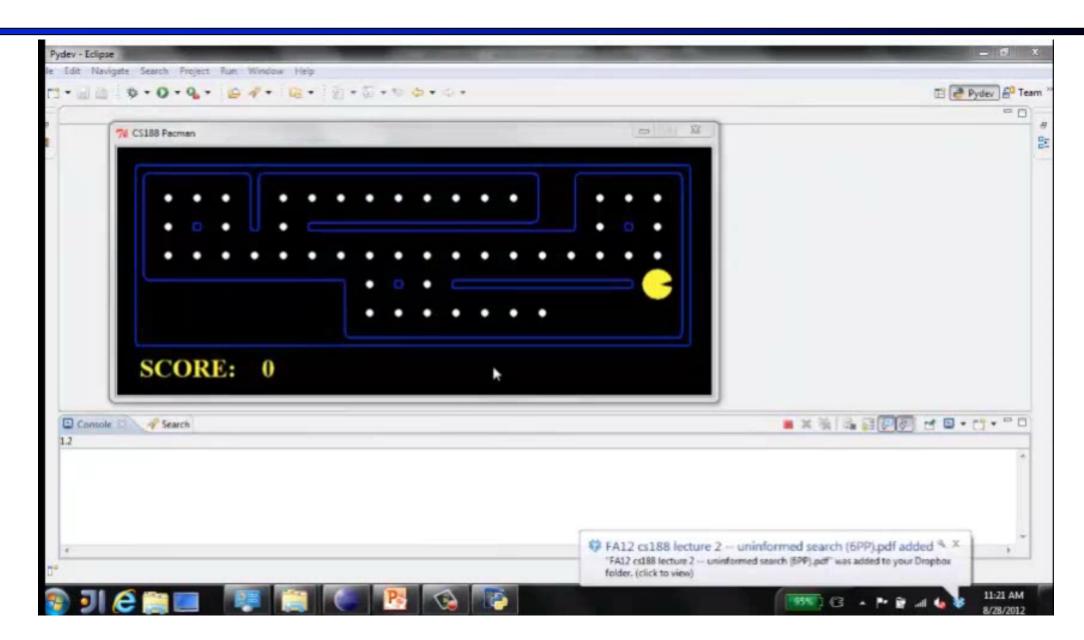
Rationality means do whatever actions to reach whatever goals...



## Video of Demo Reflex Optimal

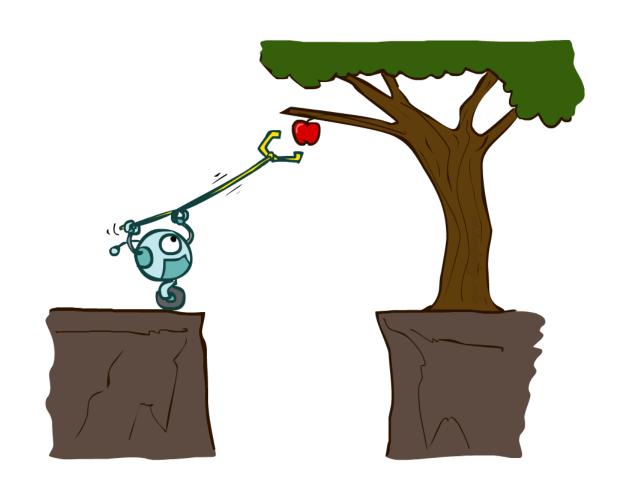


#### Video of Demo Reflex Odd

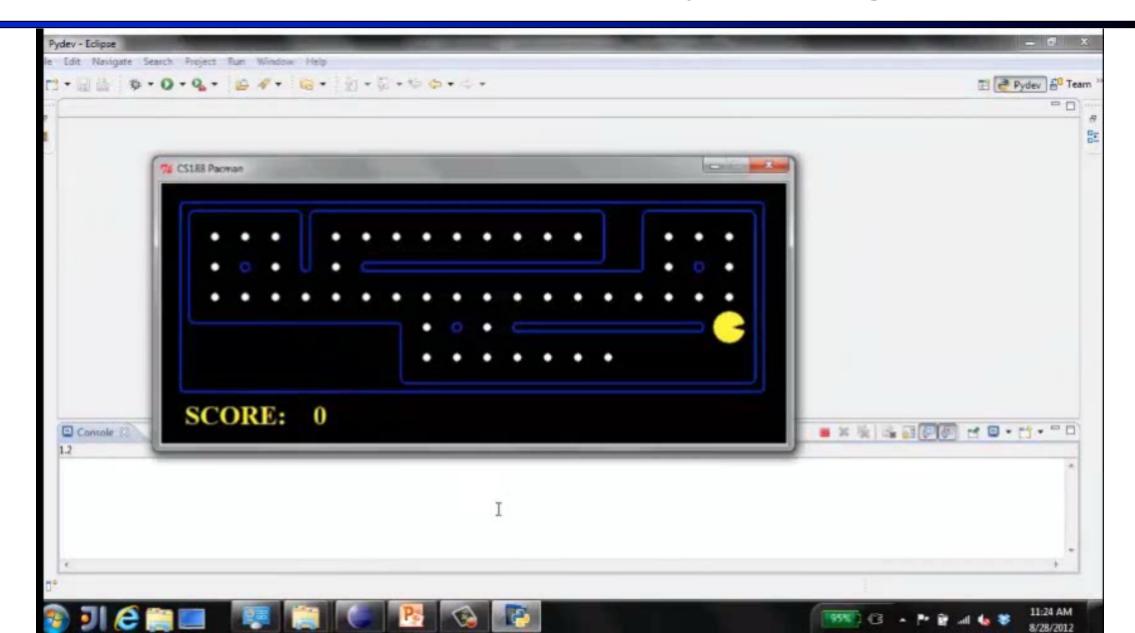


#### Planning Agents

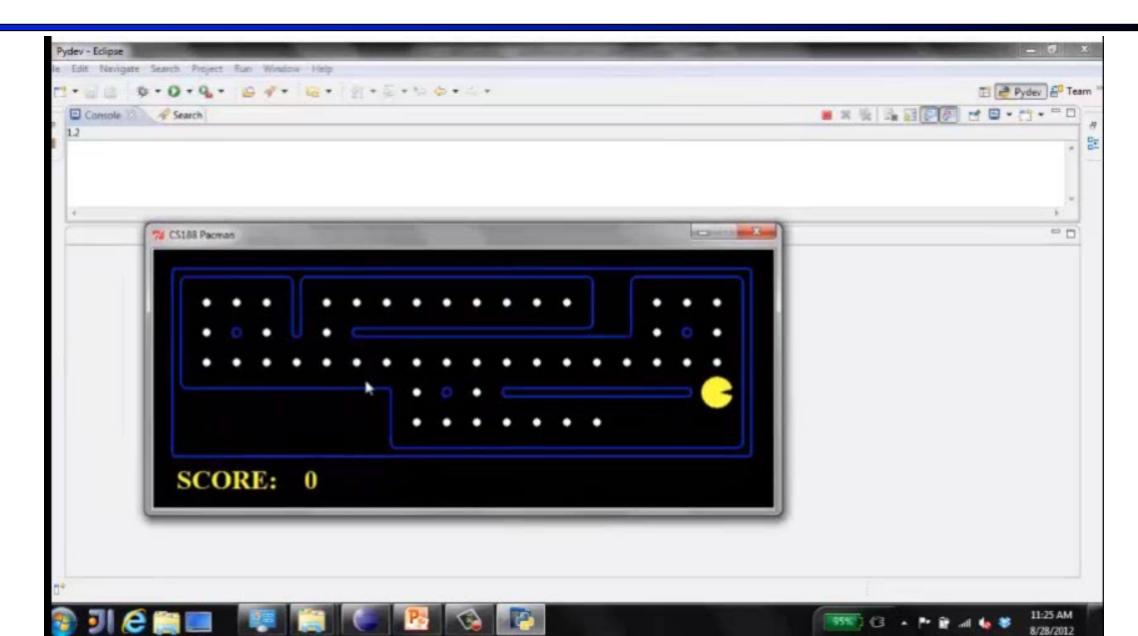
- Planning agents:
  - Ask "what if"
  - Decisions based on (hypothesized) consequences of actions
  - Must have a model of how the world evolves in response to actions
  - Must formulate a goal (test)
  - Consider how the world WOULD BE
- Optimal vs. complete planning
- Planning vs. replanning



## Video of Demo Replanning



#### Video of Demo Mastermind



### Search Problems

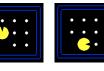


#### Search Problems

- A search problem consists of:
  - A state space





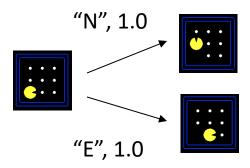








A successor function (with actions, costs)

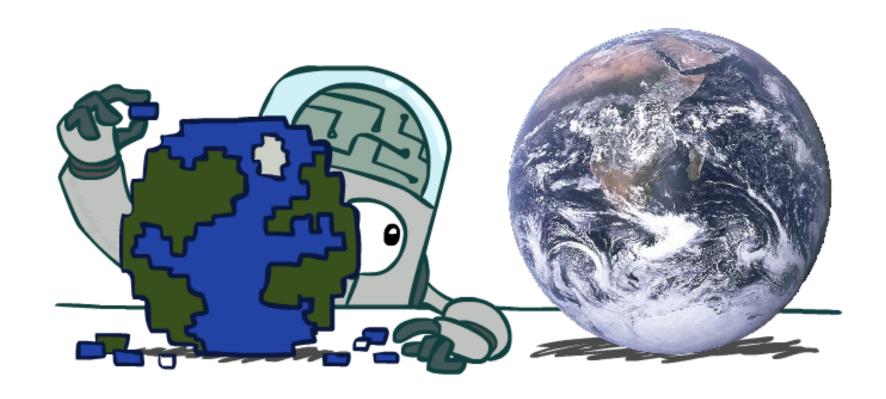


A start state and a goal test

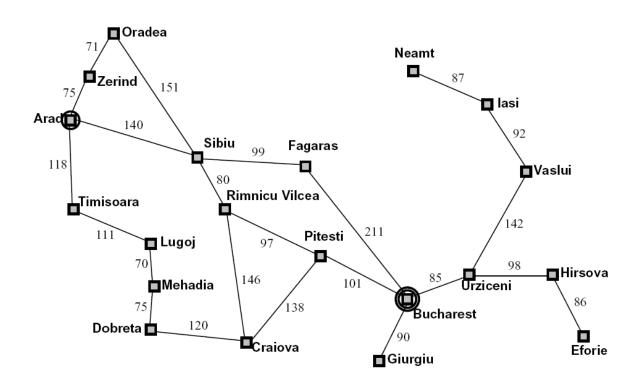
That we have more than one state to accomplish the goal..

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

### Search Problems Are Models



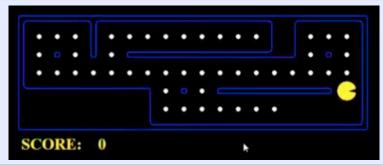
## Example: Traveling in Romania



- State space:
  - Cities
- Successor function:
  - Roads: Go to adjacent city with cost = distance
- Start state:
  - Arad
- Goal test:
  - 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

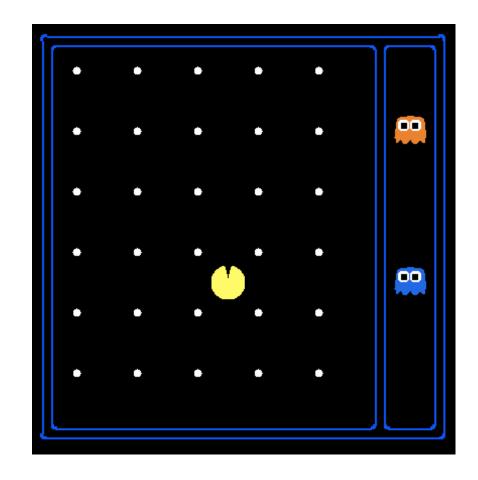
■ Food count: 30

Ghost positions: 12

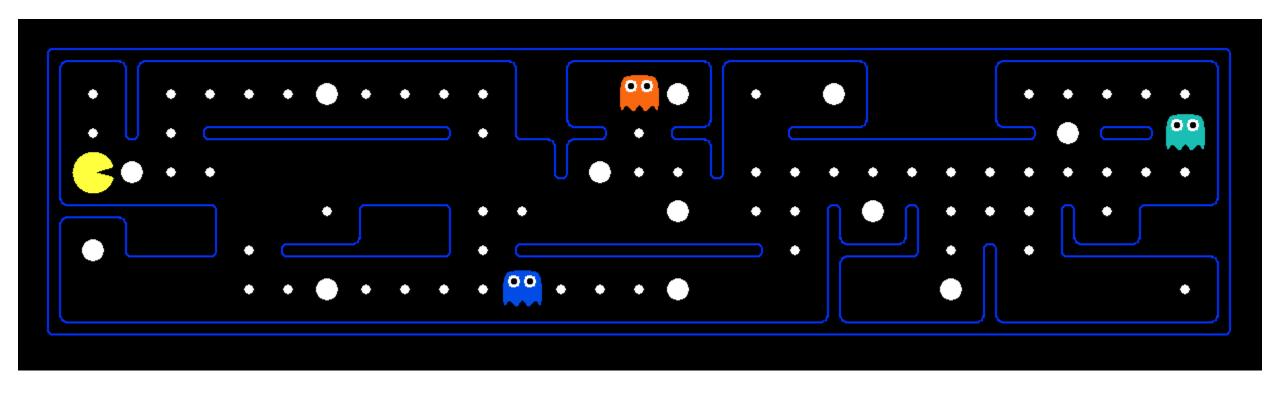
Agent facing: NSEW

#### 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>)

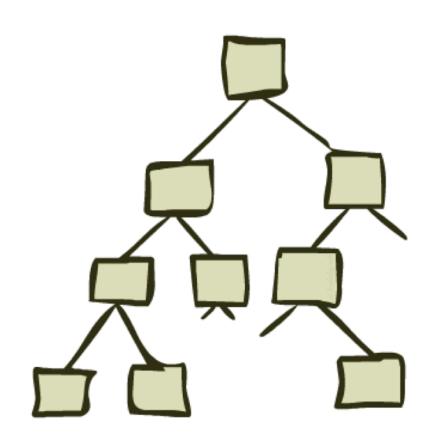


#### Quiz: Safe Passage



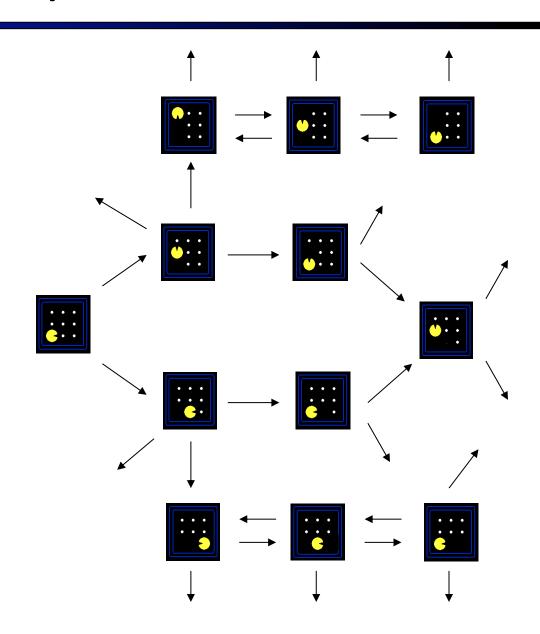
- Problem: eat all dots while keeping the ghosts perma-scared
- What does the state space have to specify?
  - (agent position, dot booleans, power pellet booleans, remaining scared time)

# State Space Graphs and Search Trees



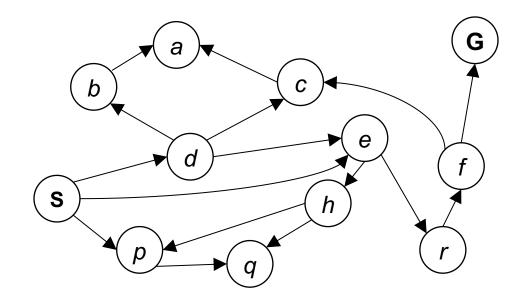
#### State Space Graphs

- 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



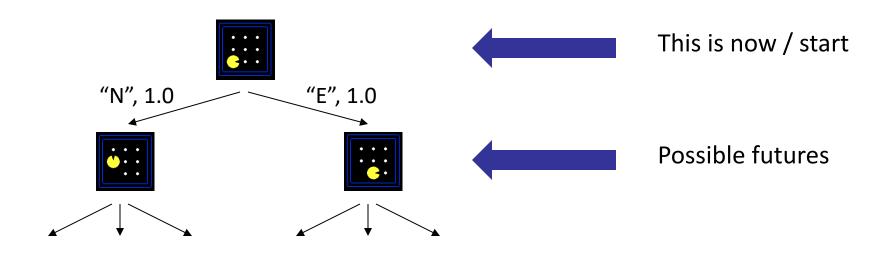
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Tiny state space graph for a tiny search problem

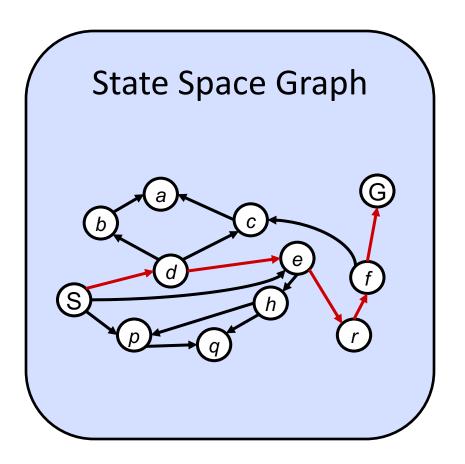
#### Search Trees



#### A search tree:

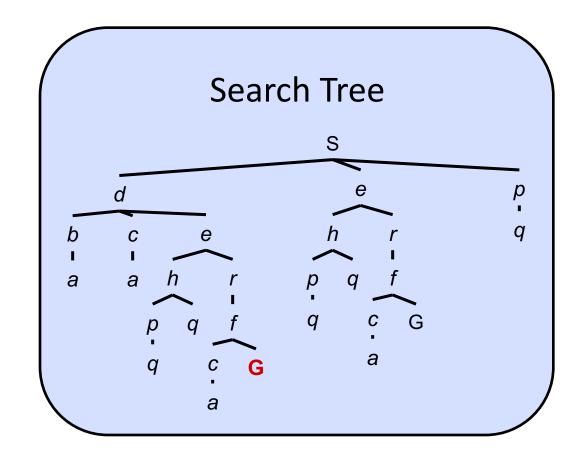
- A "what if" tree of plans and their outcomes
- The start state is the root node
- Children correspond to successors
- Nodes show states, but correspond to PLANS that achieve those states
- For most problems, we can never actually build the whole tree

### State Space Graphs vs. Search Trees



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

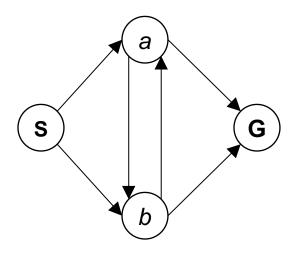
We construct both on demand – and we construct as little as possible.



### Quiz: State Space Graphs vs. Search Trees

Consider this 4-state graph:

How big is its search tree (from S)?

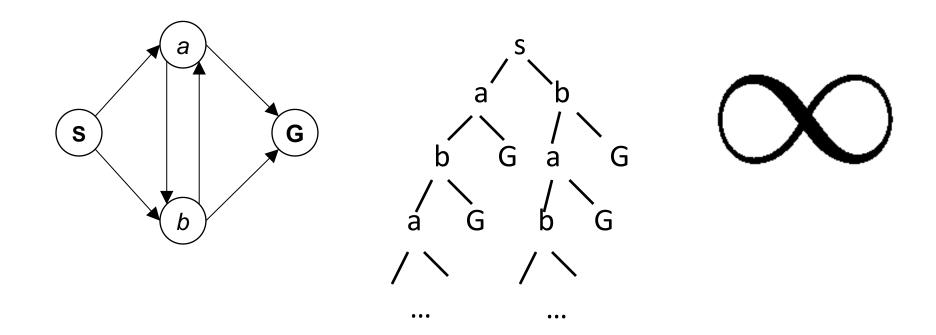




## Quiz: State Space Graphs vs. Search Trees

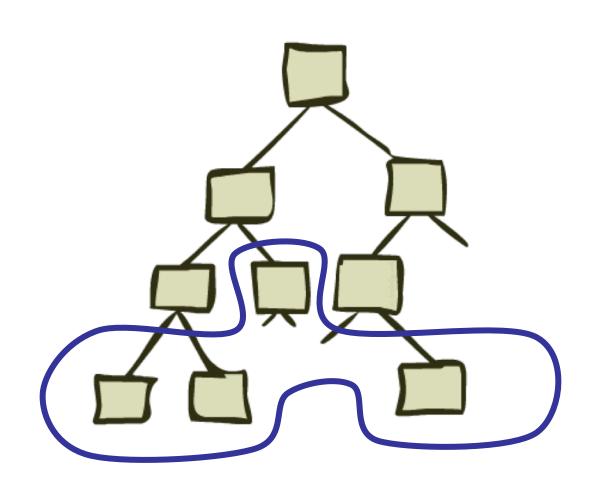
Consider this 4-state graph:

How big is its search tree (from S)?

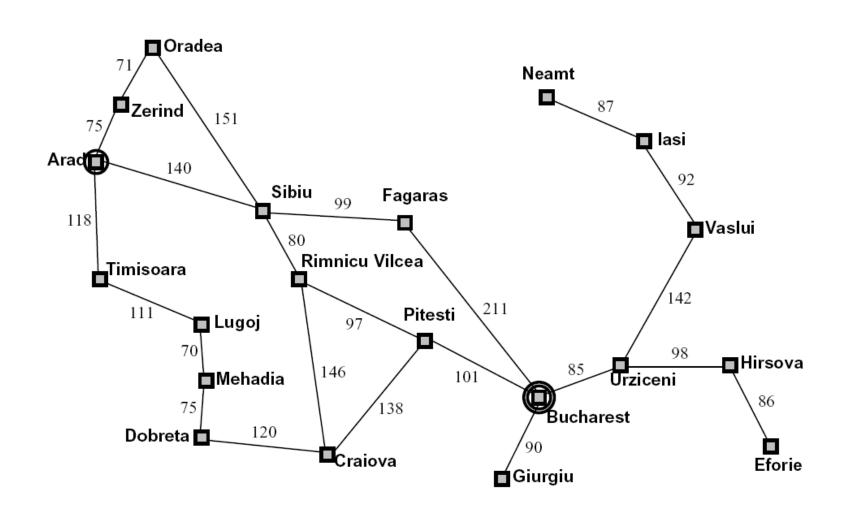


Important: Lots of repeated structure in the search tree!

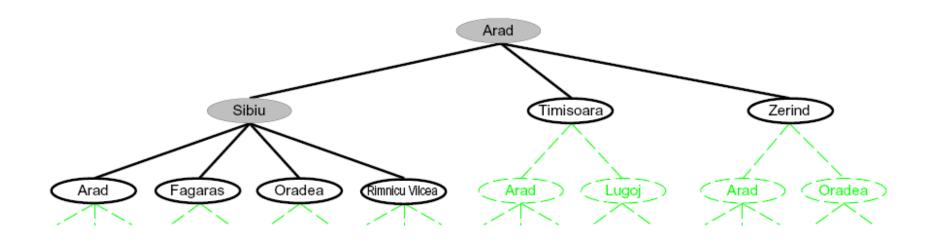
### Tree Search



## 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**

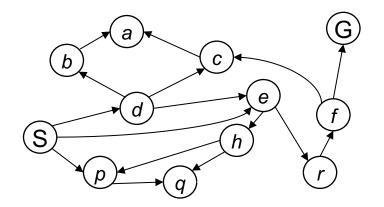
```
function TREE-SEARCH( problem, strategy) returns a solution, or failure initialize the search tree using the initial state of problem loop do

if there are no candidates for expansion then return failure choose a leaf node for expansion according to strategy

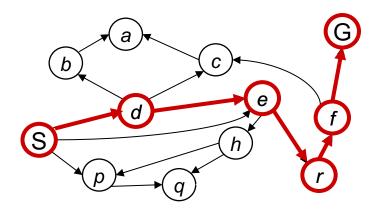
if the node contains a goal state then return the corresponding solution else expand the node and add the resulting nodes to the search tree end
```

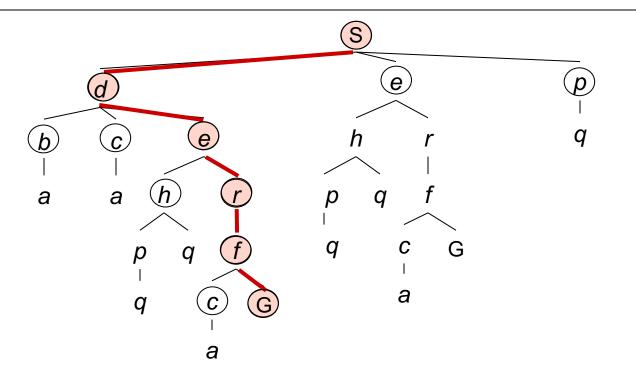
- Important ideas:
  - Fringe
  - Expansion
  - Exploration strategy
- Main question: which fringe nodes to explore?

# Example: Tree Search



### Example: Tree Search





```
s \rightarrow d

s \rightarrow e

s \rightarrow p

s \rightarrow d \rightarrow b

s \rightarrow d \rightarrow c

s \rightarrow d \rightarrow e

s \rightarrow d \rightarrow e \rightarrow h

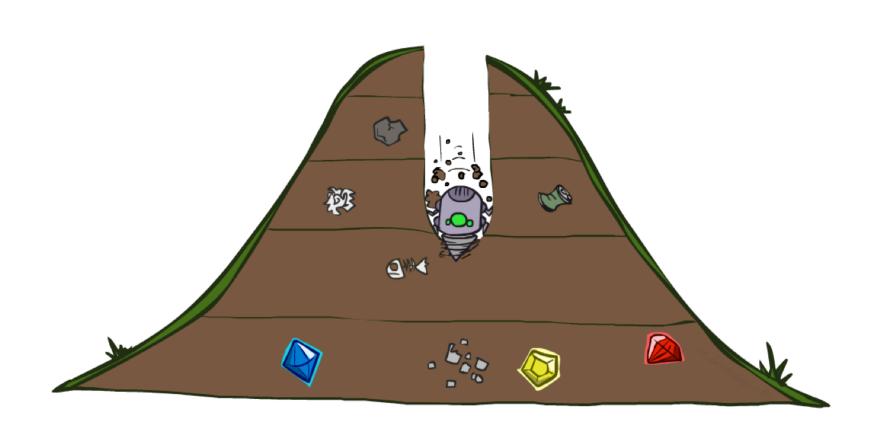
s \rightarrow d \rightarrow e \rightarrow r

s \rightarrow d \rightarrow e \rightarrow r \rightarrow f

s \rightarrow d \rightarrow e \rightarrow r \rightarrow f \rightarrow c

s \rightarrow d \rightarrow e \rightarrow r \rightarrow f \rightarrow c
```

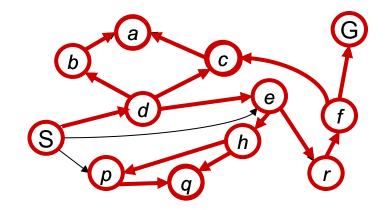
# Depth-First Search

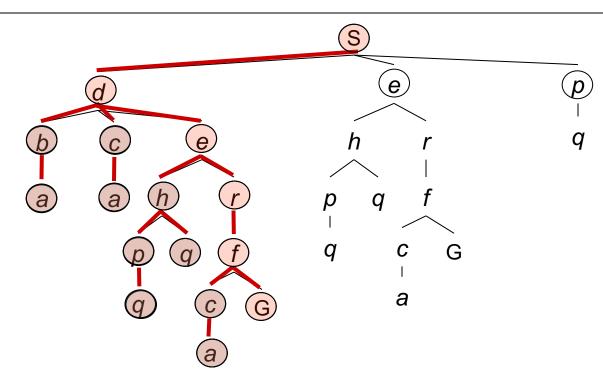


## Depth-First Search

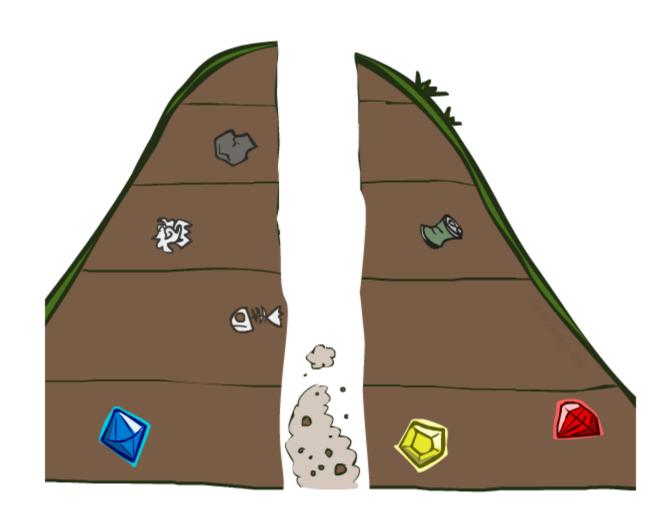
Strategy: expand a deepest node first

Implementation:
Fringe is a <u>LIFO stack</u>



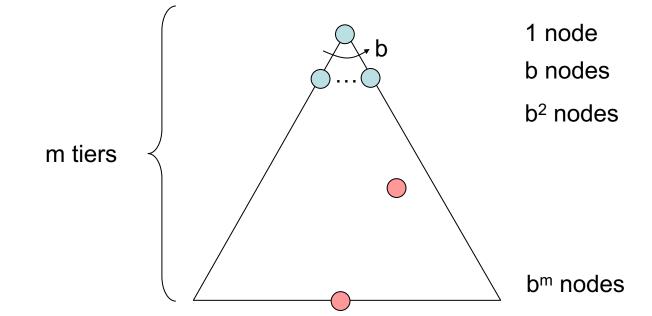


## Search Algorithm Properties



#### 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

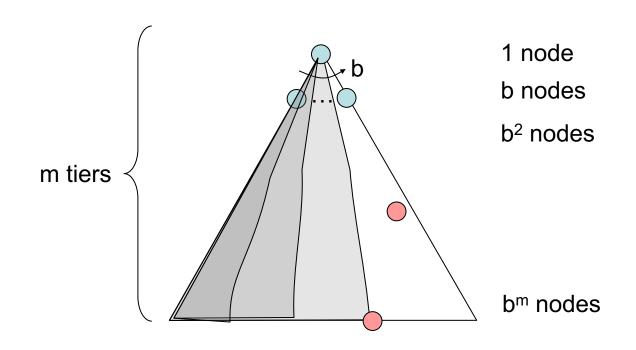


- Number of nodes in entire tree?
  - $1 + b + b^2 + .... b^m = O(b^m)$

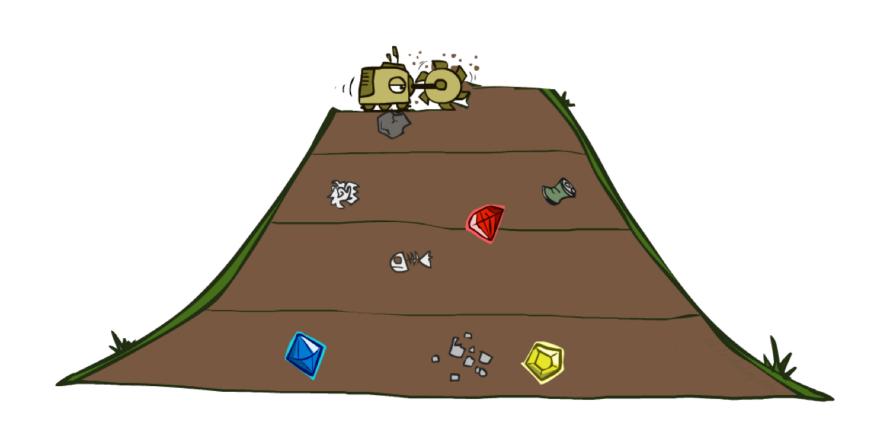
### Depth-First Search (DFS) Properties

#### 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?
  - No, it finds the "leftmost" solution, regardless of depth or cost



## **Breadth-First Search**

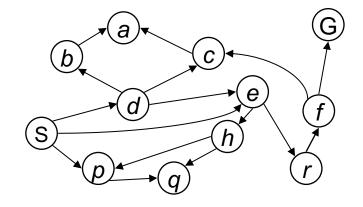


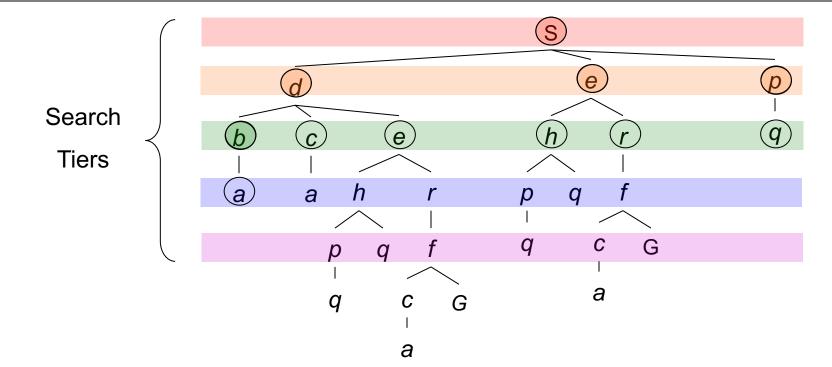
#### **Breadth-First Search**

Strategy: expand a shallowest node first

*Implementation: Fringe* 

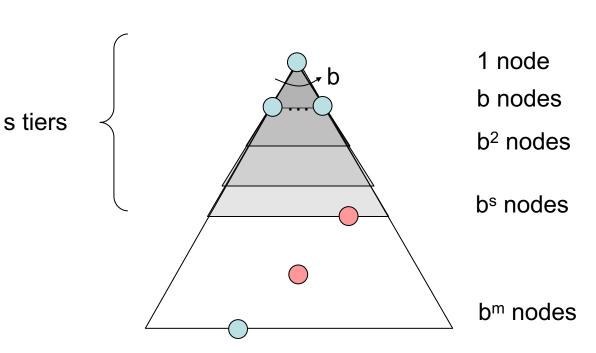
is a FIFO queue



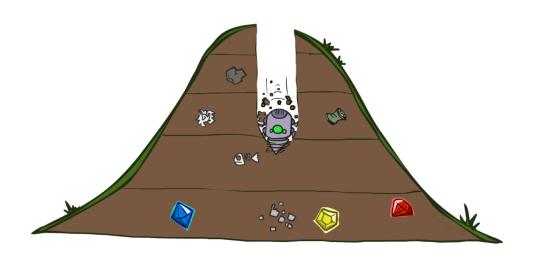


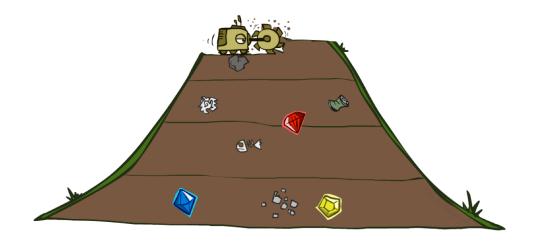
## Breadth-First Search (BFS) Properties

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





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



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



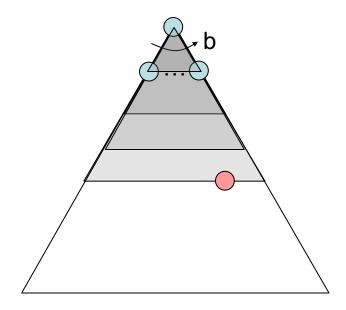
#### Quiz: DFS vs BFS

When will BFS outperform DFS?

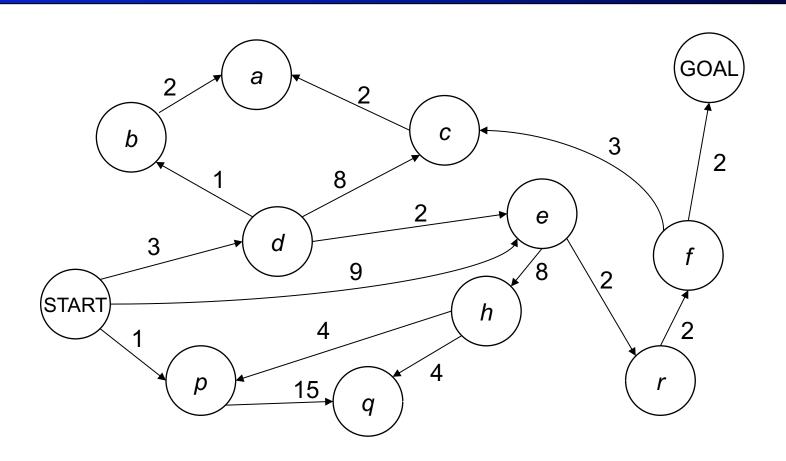
When will DFS outperform BFS?

## **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!

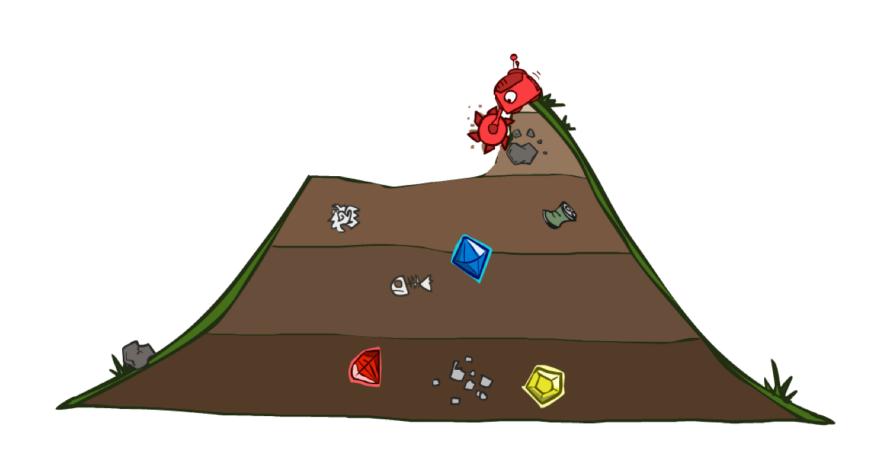


#### **Cost-Sensitive Search**



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.

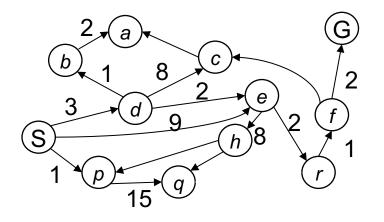
#### **Uniform Cost Search**

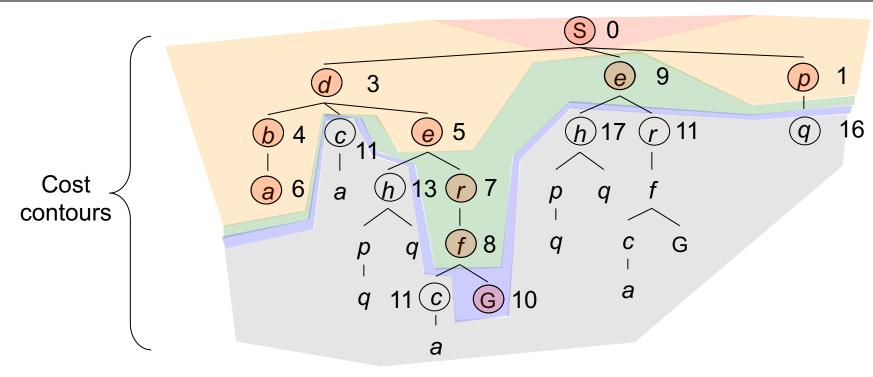


#### **Uniform Cost Search**

Strategy: expand a cheapest node first:

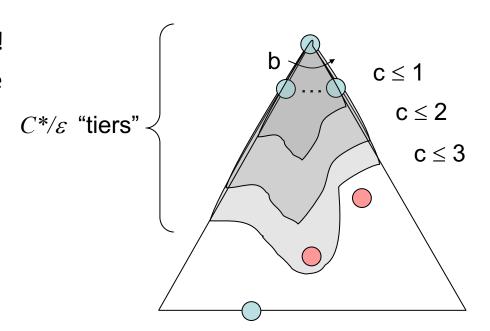
Fringe is a priority queue (priority: cumulative cost)





## 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  $\varepsilon$ , then the "effective depth" is roughly  $C^*/\varepsilon$
  - Takes time  $O(b^{C*/\varepsilon})$  (exponential in effective depth)
- How much space does the fringe take?
  - Has roughly the last tier, so  $O(b^{C*/\varepsilon})$
- 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\*)

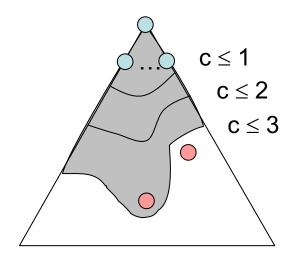


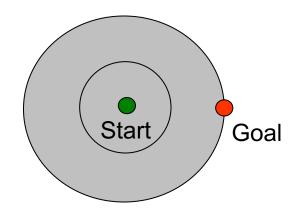
#### **Uniform Cost Issues**

 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!

# Video of Demo Empty UCS



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



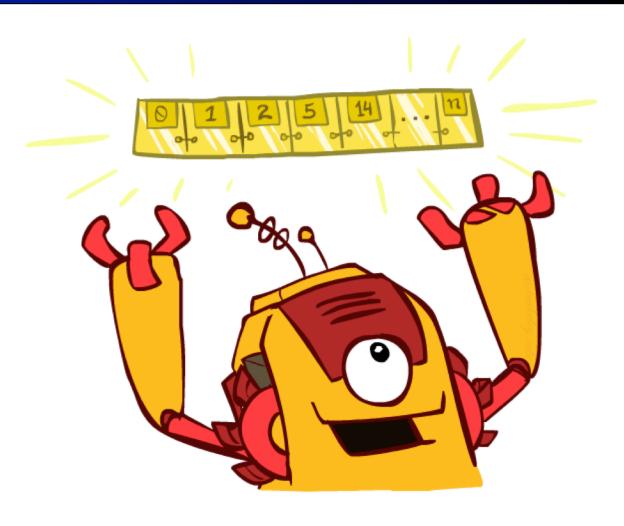


Video of Demo Maze with Deep/Shallow Water --- DFS, BFS, or UCS? (part 3)



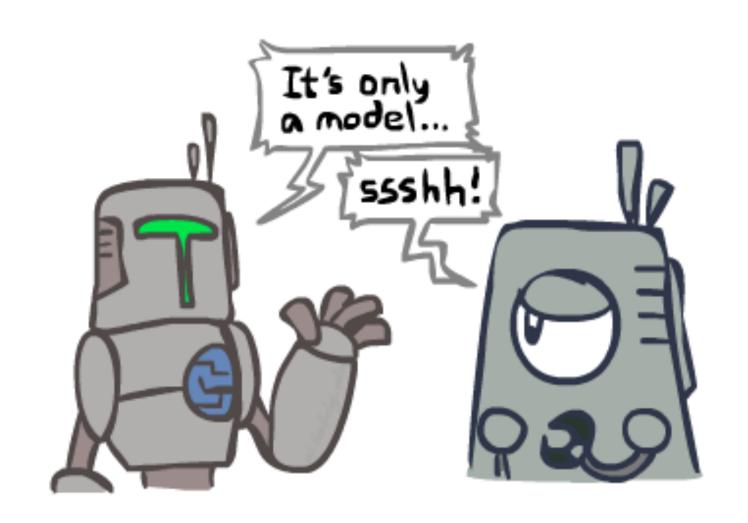
#### The One Queue

- 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 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...



# Search Gone Wrong?

