

Warm-up as you log in

Write the pseudo code for breadth first search and depth first search

- Iterative version, not recursive

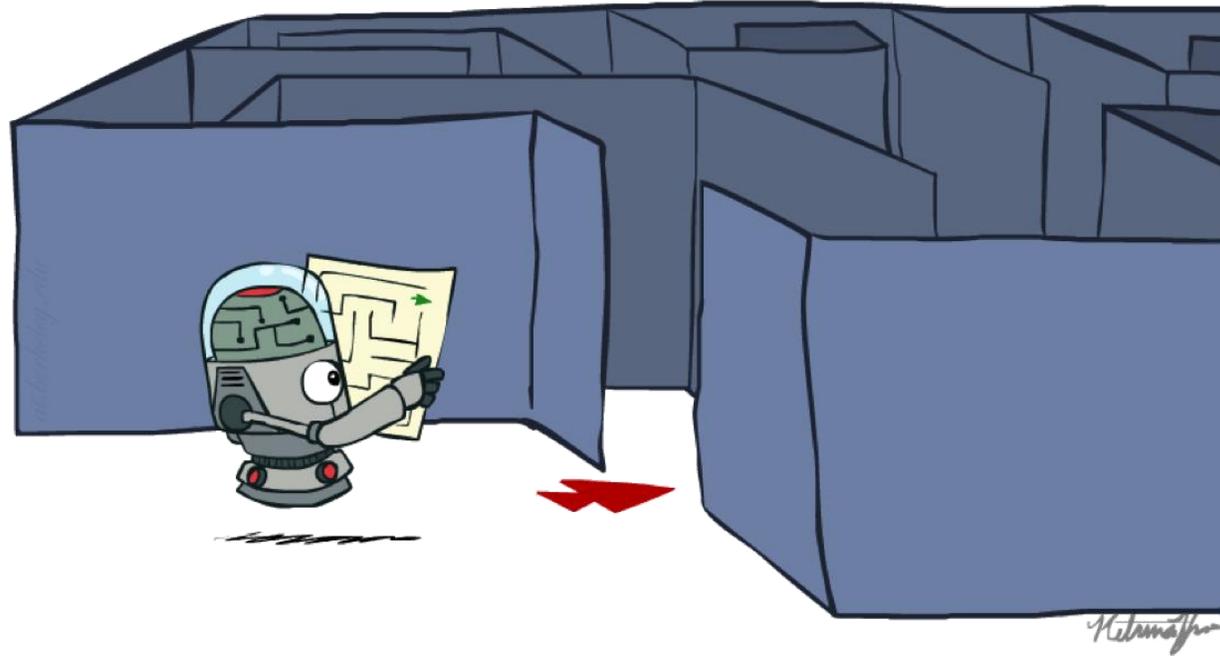
```
class TreeNode
    TreeNode[] children()
    boolean isGoal()
```

```
BFS (TreeNode start)...
```

```
DFS (TreeNode start)...
```

AI: Representation and Problem Solving

Agents and Search



Instructor: Pat Virtue

Slide credits: CMU AI, <http://ai.berkeley.edu>

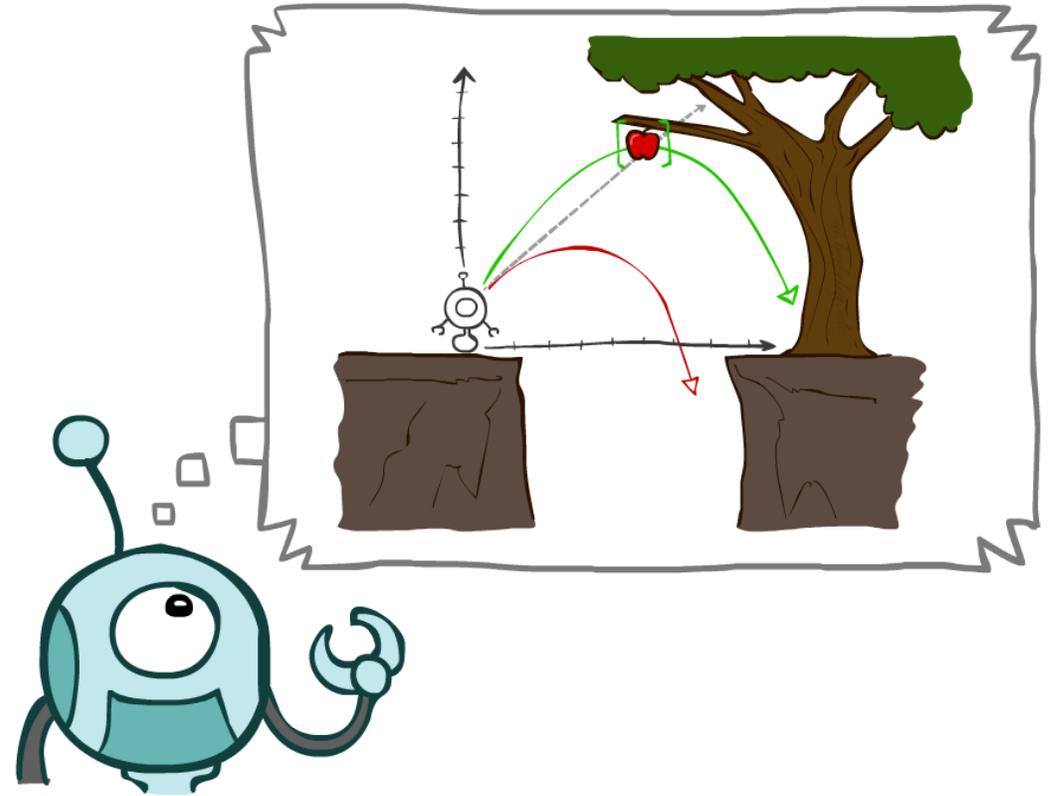
Outline

Agents and Environments

Search Problems

Uninformed Search Methods

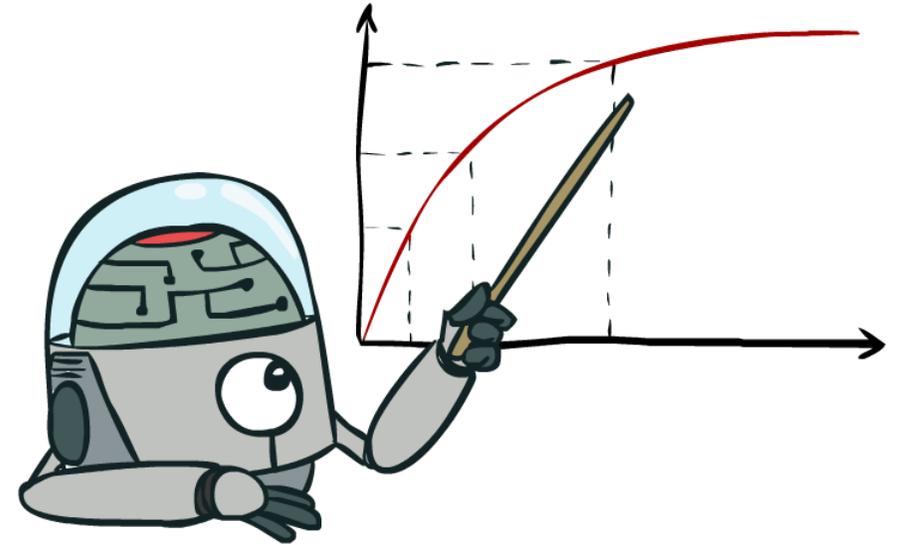
- Depth-First Search
- Breadth-First Search
- Uniform-Cost Search



Rationality, contd.

What is rational depends on:

- Performance measure
- Agent's prior knowledge of environment
- Actions available to agent
- Percept sequence to date



Being rational means **maximizing your expected utility**

Rational Agents

Are rational agents **omniscient**?

- No – they are limited by the available percepts

Are rational agents **clairvoyant**?

- No – they may lack knowledge of the environment dynamics

Do rational agents **explore** and **learn**?

- Yes – in unknown environments these are essential

So rational agents are not necessarily successful, but they are **autonomous** (i.e., transcend initial program)

Task Environment - PEAS

Performance measure

- -1 per step; +10 food; +500 win; -500 die; +200 hit scared ghost

Environment

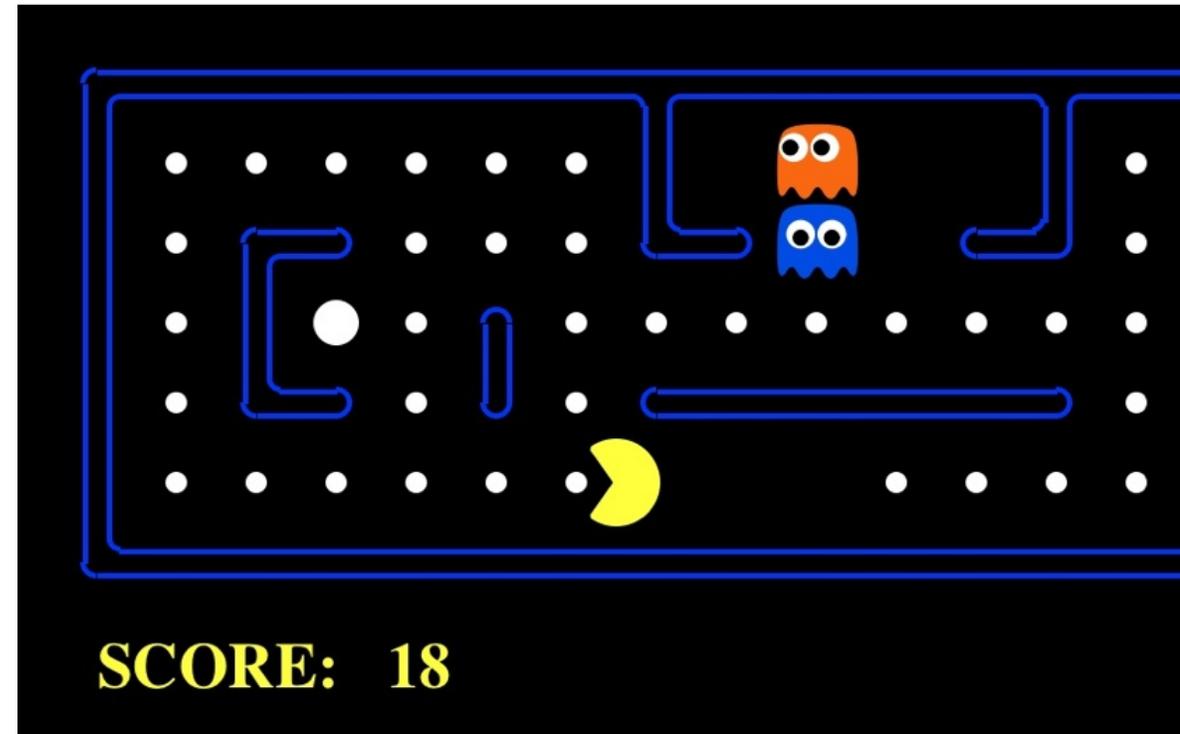
- Pacman dynamics (incl ghost behavior)

Actuators

- North, South, East, West, (Stop)

Sensors

- Entire state is visible



PEAS: Automated Taxi

Performance measure

- Income, happy customer, vehicle costs, fines, insurance premiums

Environment

- US streets, other drivers, customers

Actuators

- Steering, brake, gas, display/speaker

Sensors

- Camera, radar, accelerometer, engine sensors, microphone



Environment Types

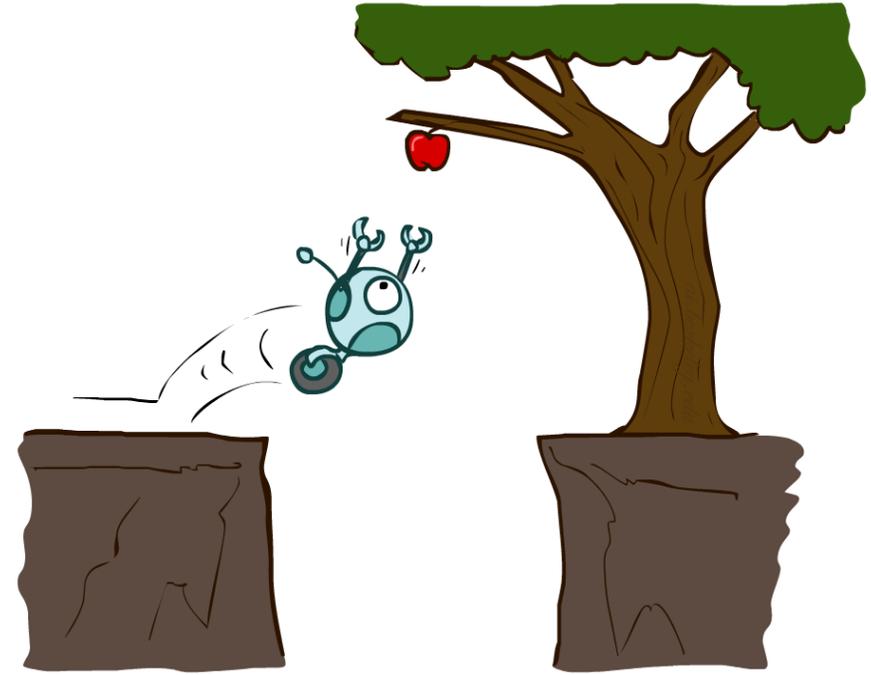
	Pacman	Taxi
Fully or partially observable		
Single agent or multi-agent		
Deterministic or stochastic		
Static or dynamic		
Discrete or continuous		

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 (L2D1)]

[Demo: reflex optimal (L2D2)]

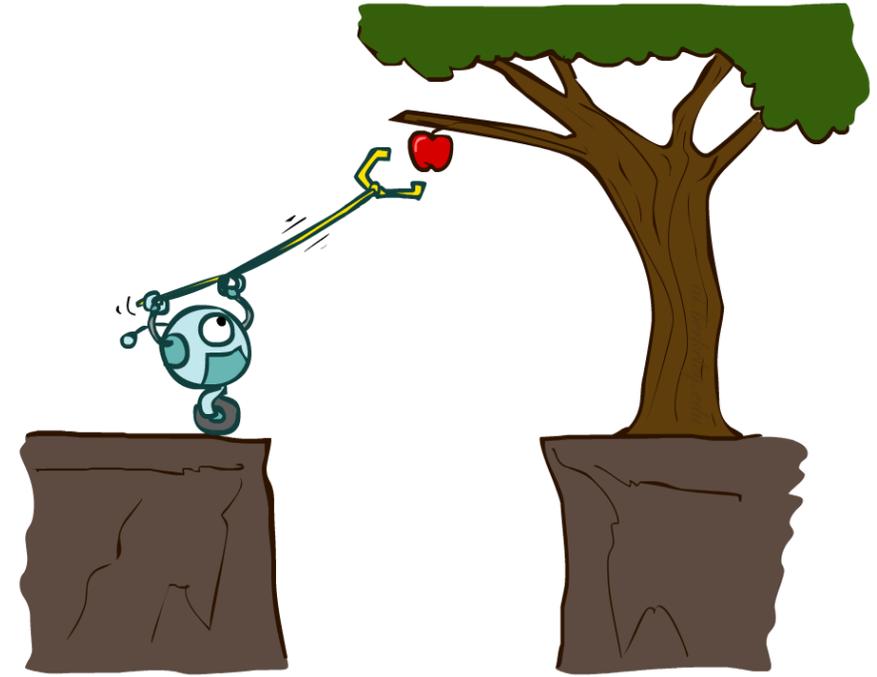
Agents that Plan Ahead

Planning agents:

- Decisions based on *predicted consequences* of actions
- Must have a *transition model*: how the world evolves in response to actions
- Must formulate a goal
- Consider how the world **WOULD BE**

Spectrum of deliberativeness:

- Generate complete, optimal plan offline, then execute
- Generate a simple, greedy plan, start executing, replan when something goes wrong



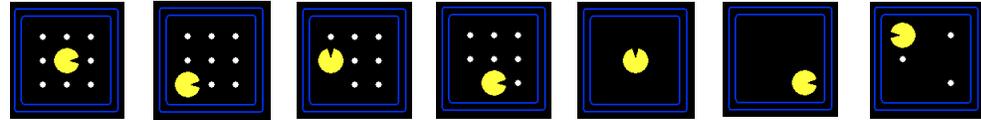
Search Problems



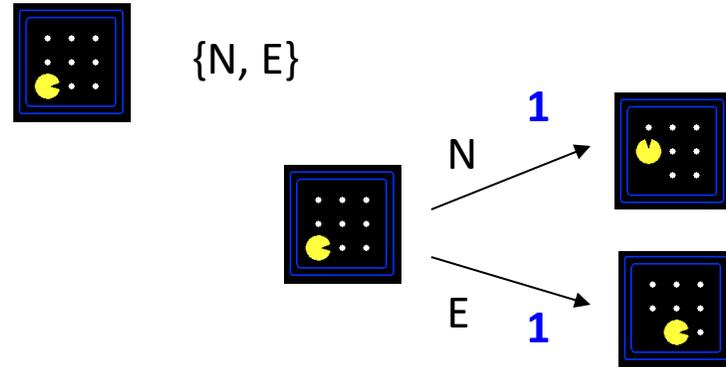
Search Problems

A **search problem** consists of:

- A state space



- For each state, a set
Actions(s) of allowable actions



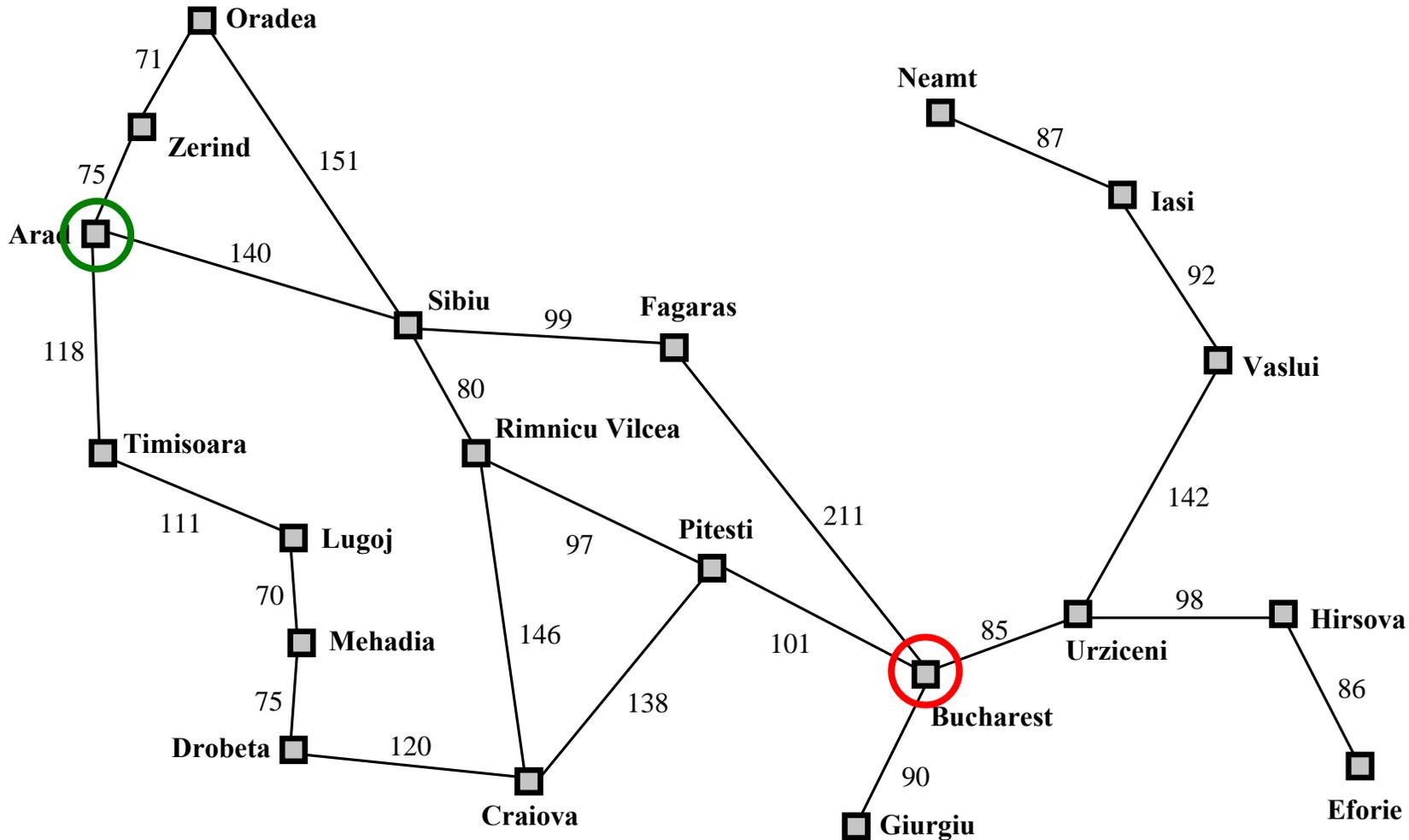
- A transition model $\text{Result}(s,a)$
- A step cost function $c(s,a,s')$
- A start state and a goal test

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

Search Problems Are Models



Example: Travelling in Romania



State space:

- Cities

Actions:

- Go to adjacent city

Transition model

- $\text{Result}(A, \text{Go}(B)) = B$

Step cost

- Distance along road link

Start state:

- Arad

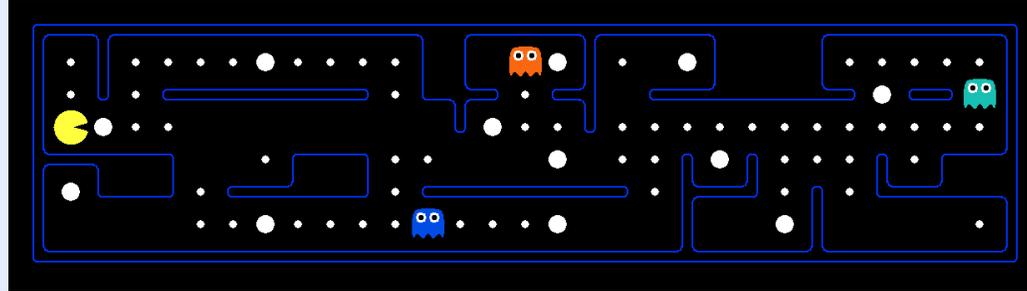
Goal test:

- Is state == Bucharest?

Solution?

What's in a State Space?

The **real world state** includes every last detail of the environment



A **search state** abstracts away details not needed to solve the problem

- Problem: Pathing
 - State representation: (x,y) location
 - Actions: NSEW
 - Transition model: update location
 - Goal test: is $(x,y)=END$
- Problem: Eat-All-Dots
 - State representation: $\{(x,y), \text{dot booleans}\}$
 - Actions: NSEW
 - Transition model: update location and possibly a dot boolean
 - Goal test: dots all false

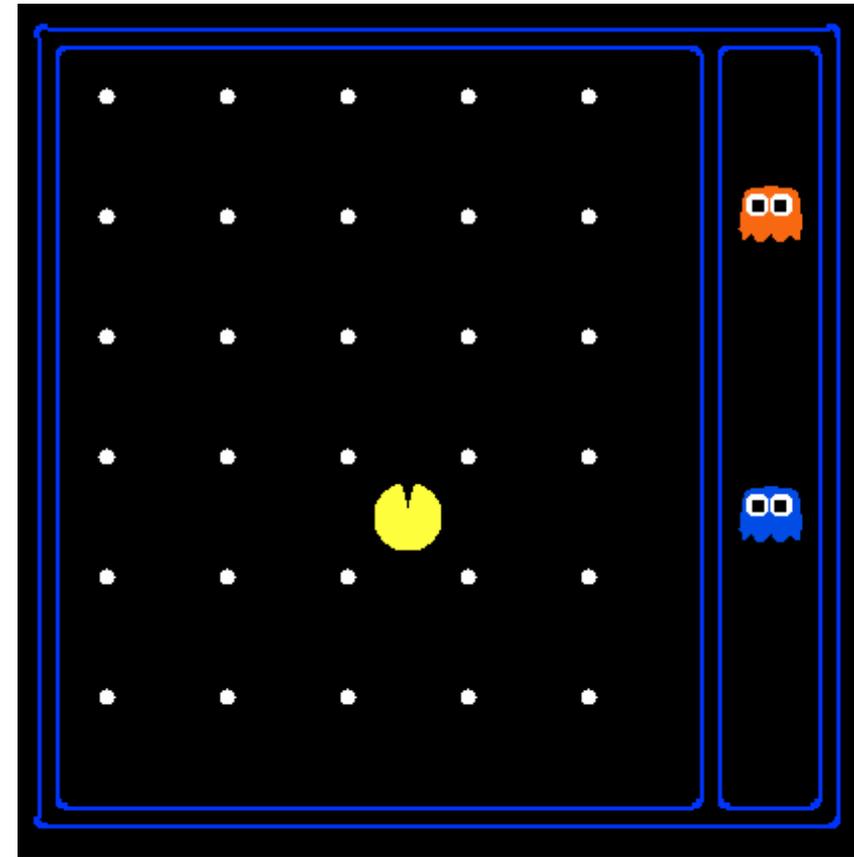
State Space Sizes?

World state:

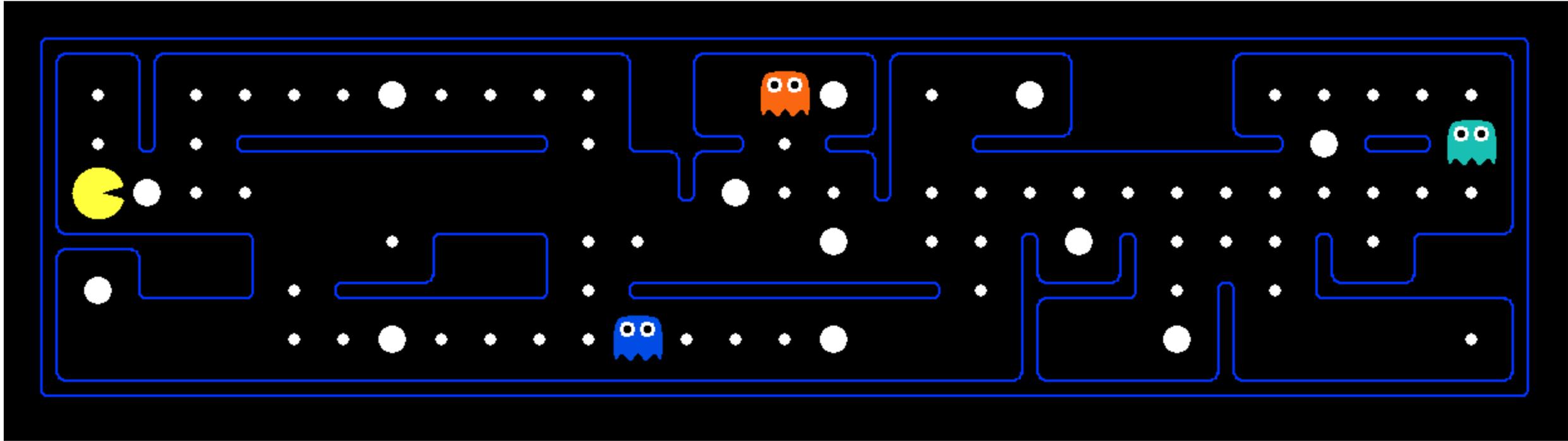
- Agent positions: 120
- Food count: 30
- Ghost positions: 12
- Agent facing: NSEW

How many

- World states?
 $120 \times (2^{30}) \times (12^2) \times 4$
- States for pathing?
120
- States for eat-all-dots?
 $120 \times (2^{30})$



Safe Passage

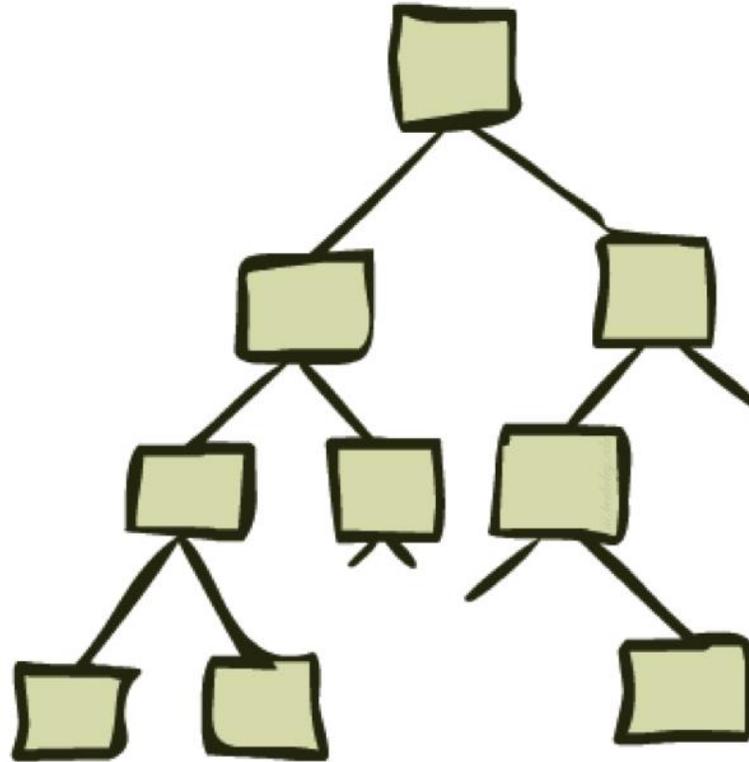


Problem: eat all dots while keeping the ghosts perma-scared

What does the state representation have to specify?

- (agent position, dot booleans, power pellet booleans, remaining scared time)

State Space Graphs and Search Trees



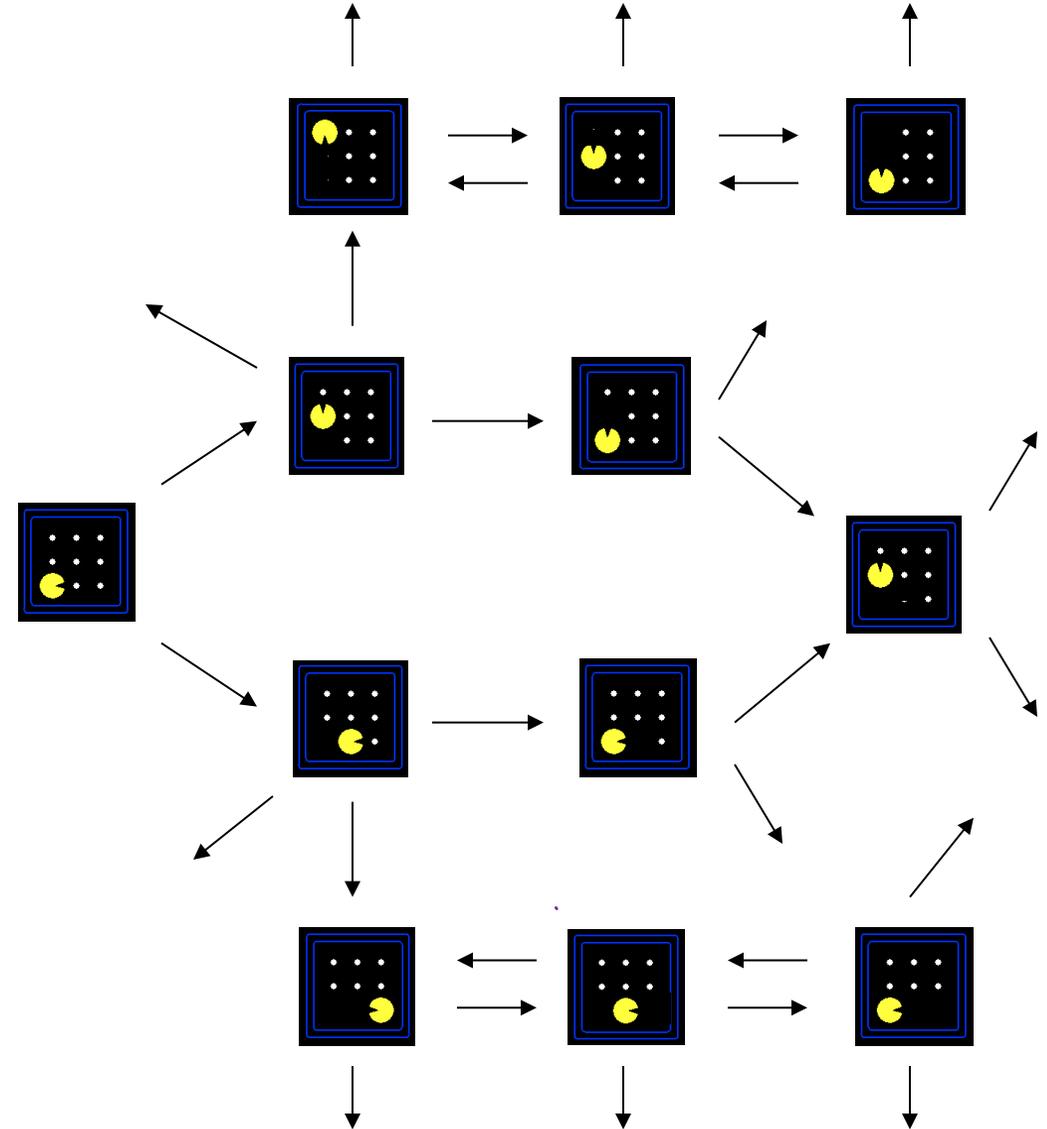
State Space Graphs

State space graph: A mathematical representation of a search problem

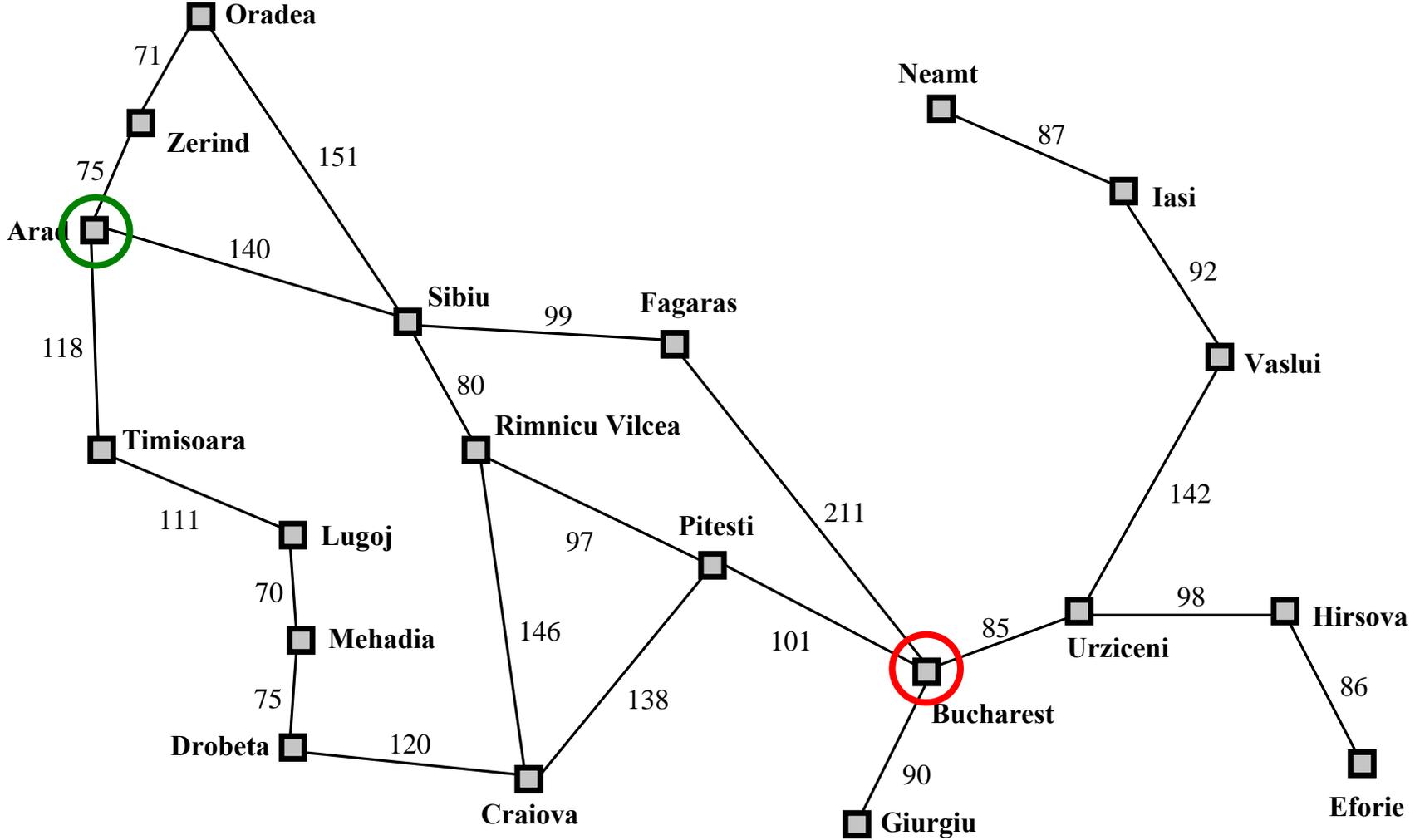
- Nodes are (abstracted) world configurations
- Arcs represent transitions resulting from actions
- The goal test is a set of goal nodes (maybe only one)

In a state space graph, each state occurs only once!

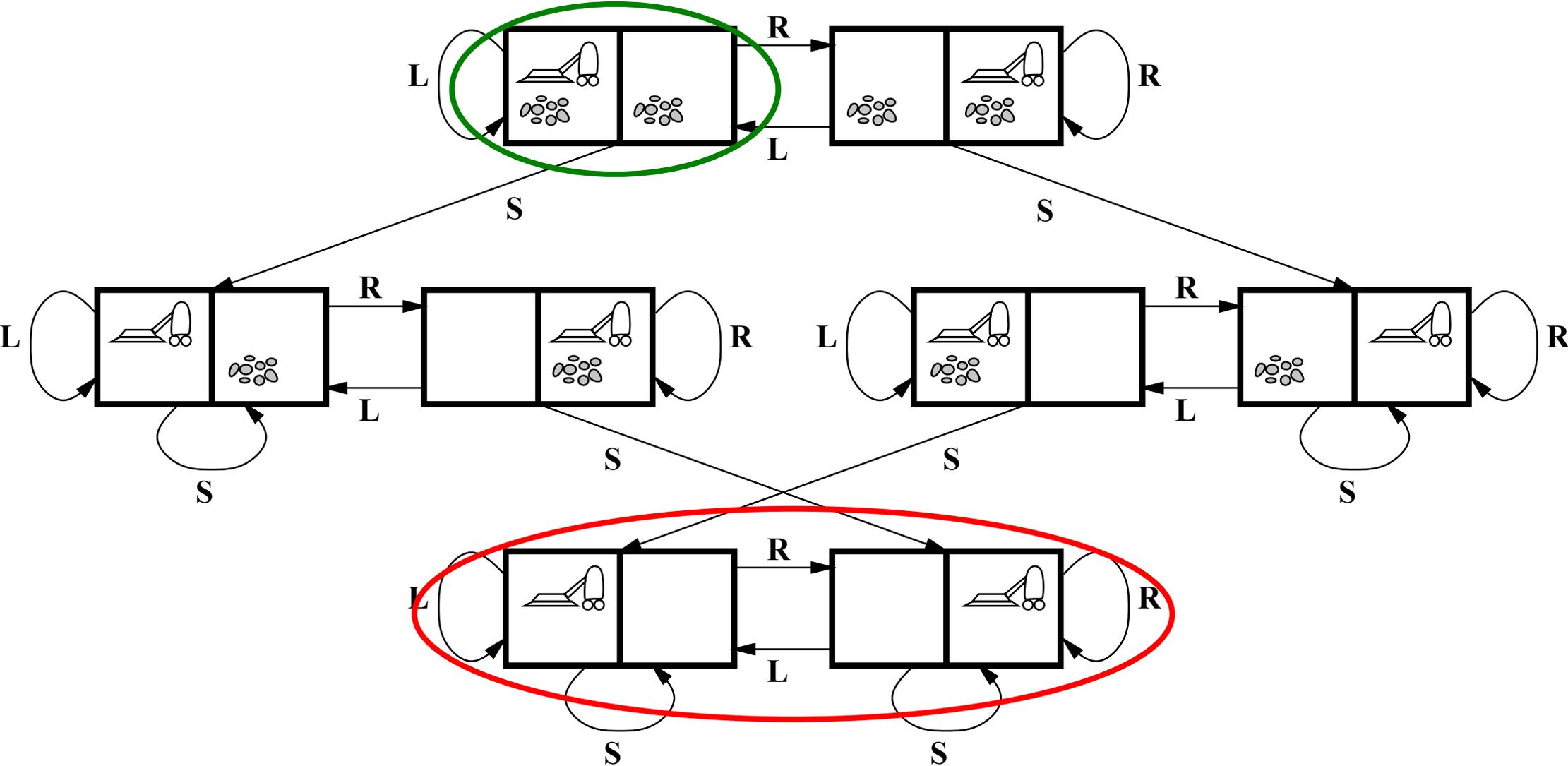
We can rarely build this full graph in memory (it's too big), but it's a useful idea



More Examples

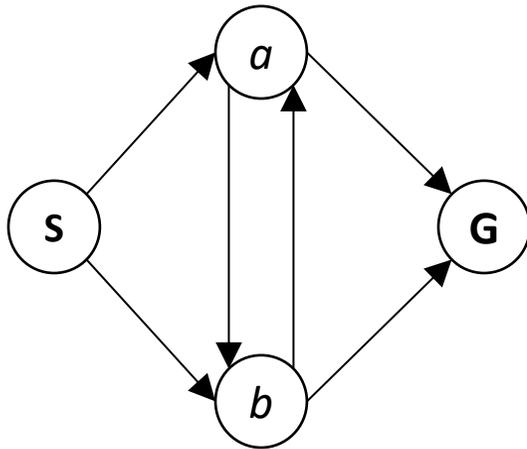


More Examples

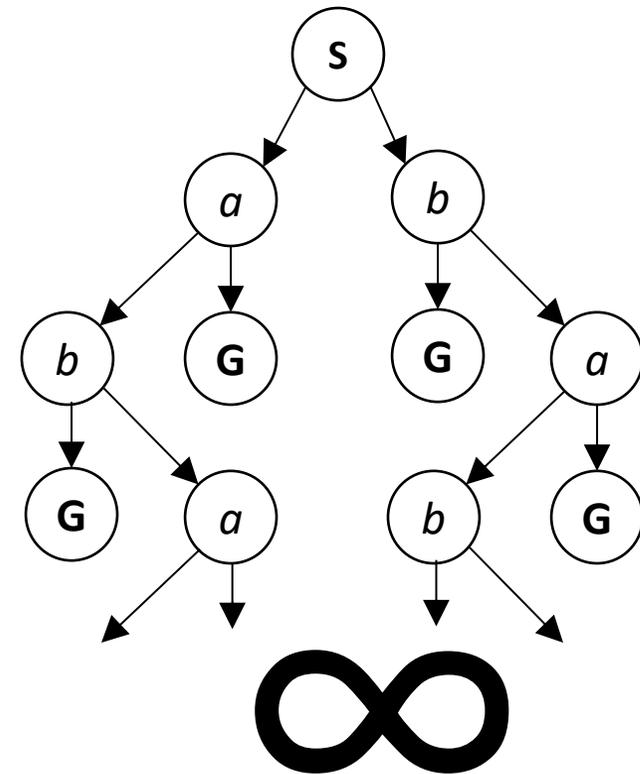


State Space Graphs vs. Search Trees

Consider this 4-state graph:

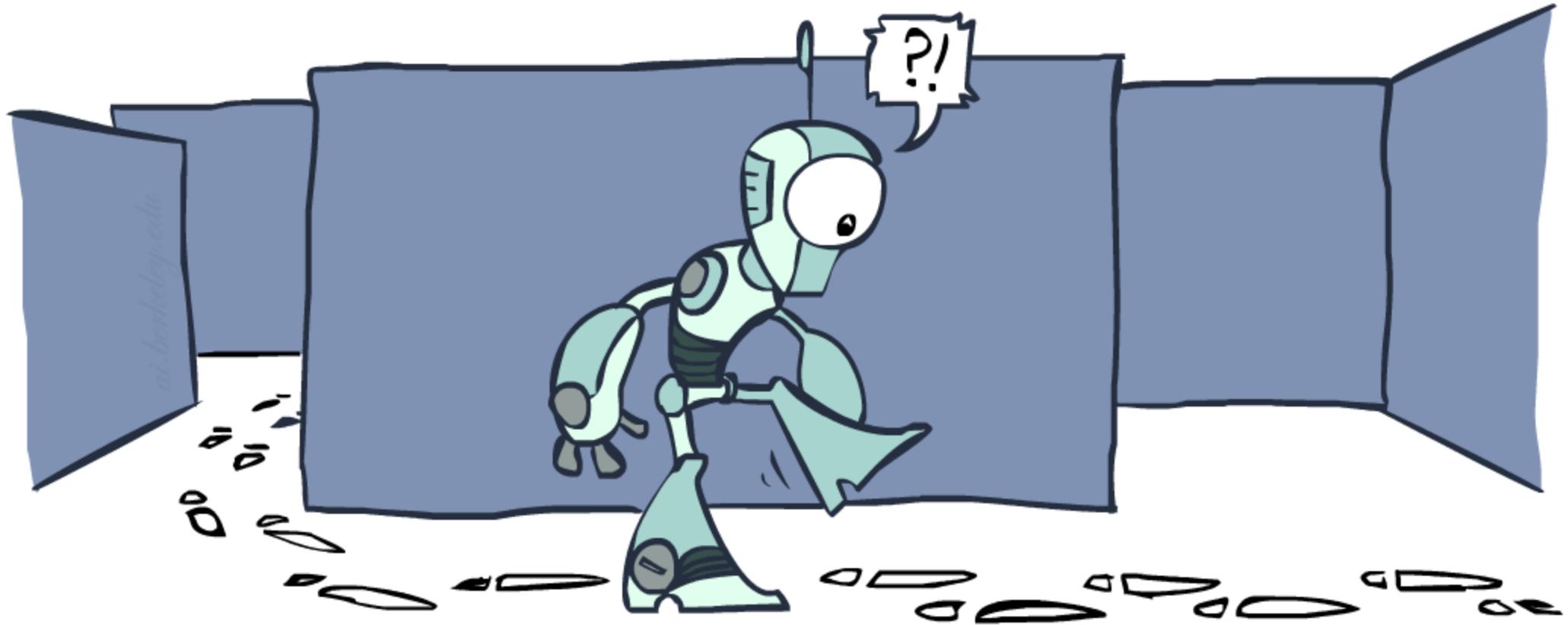


How big is its search tree (from S)?



Important: Lots of repeated structure in the search tree!

Tree Search vs Graph Search



function TREE_SEARCH(problem) returns a solution, or failure

initialize the frontier as a specific work list (stack, queue, priority queue)

add initial state of problem to frontier

loop do

if the frontier is empty then

return failure

choose a node and remove it from the frontier

if the node contains a goal state then

return the corresponding solution

for each resulting child from node

add child to the frontier

function GRAPH_SEARCH(**problem**) **returns** a solution, or failure

initialize the explored set to be empty

initialize the **frontier** as a specific work list (stack, queue, priority queue)

add initial state of **problem** to **frontier**

loop do

if the **frontier** is empty **then**

return failure

choose a **node** and remove it from the **frontier**

if the **node** contains a goal state **then**

return the corresponding solution

add the node state to the explored set

for each resulting **child** from node

if the **child** state is not already in the **frontier** or **explored set** **then**

add **child** to the **frontier**

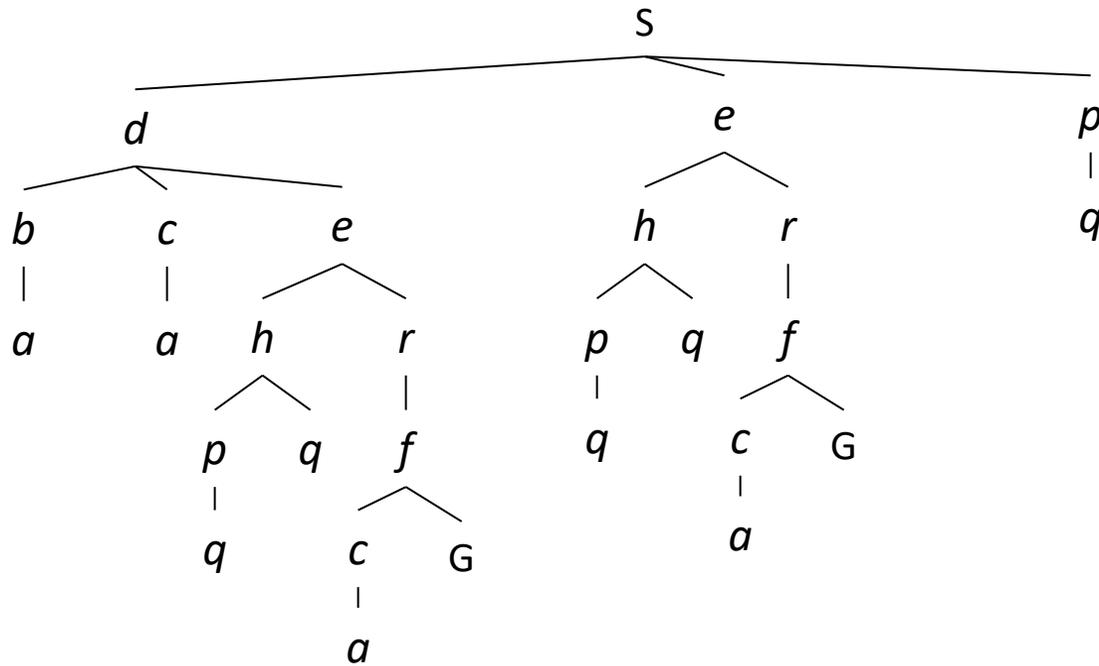
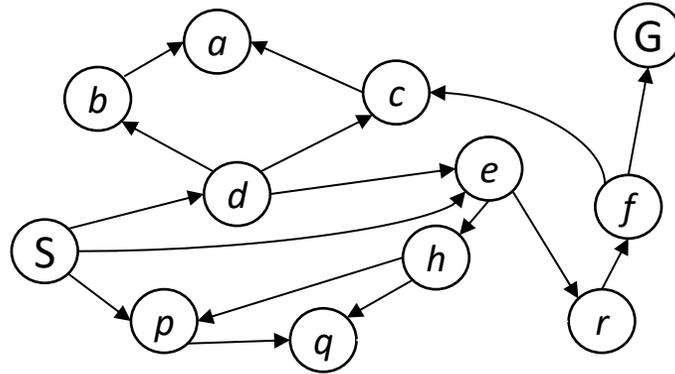
Depth-First (**Graph**) Search

Strategy: expand a deepest node first

Implementation:

Frontier is a LIFO stack

Explored set prevents loops and repeated work



Poll 1

function GRAPH-SEARCH(**problem**) **returns** a solution, or failure

initialize the explored set to be empty

initialize the **frontier** as a specific work list (stack, queue, priority queue)

add initial state of **problem** to **frontier**

loop do

if the **frontier** is empty **then**

return failure

choose a **node** and remove it from the **frontier**

if the **node** contains a goal state **then**

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add the node state to the explored set

for each resulting **child** from node

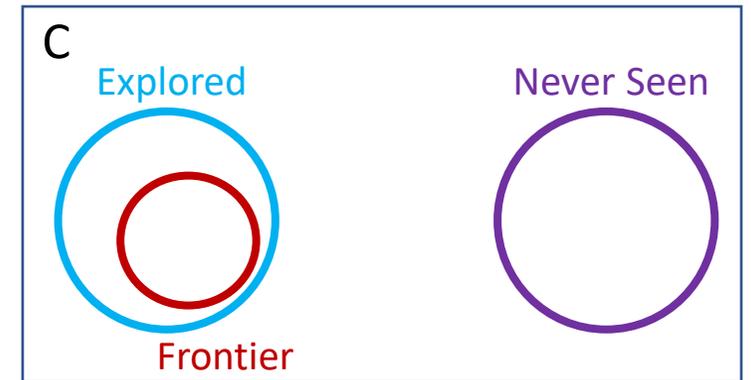
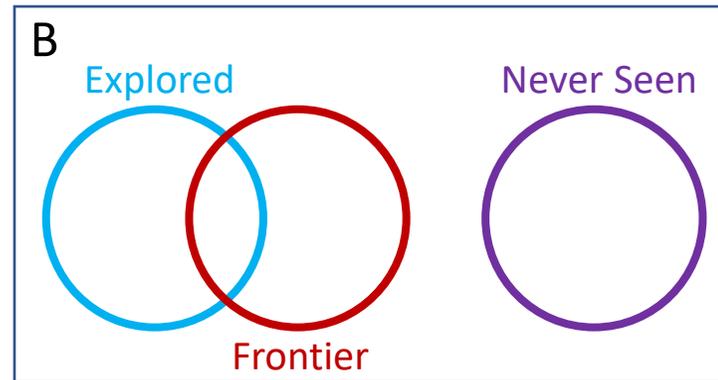
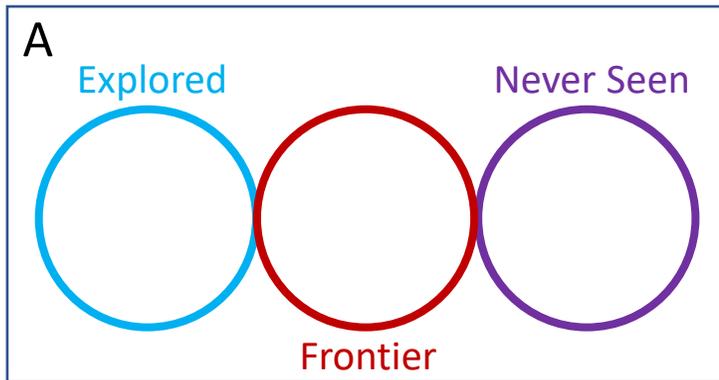
if the **child** state is not already in the **frontier** or **explored set** **then**

add **child** to the **frontier**

Poll 1

What is the relationship between these sets of states after each loop iteration in **GRAPH_SEARCH**?

(Loop invariants!!!)



A Note on Implementation

Nodes have

`state`, `parent`, `action`, `path-cost`

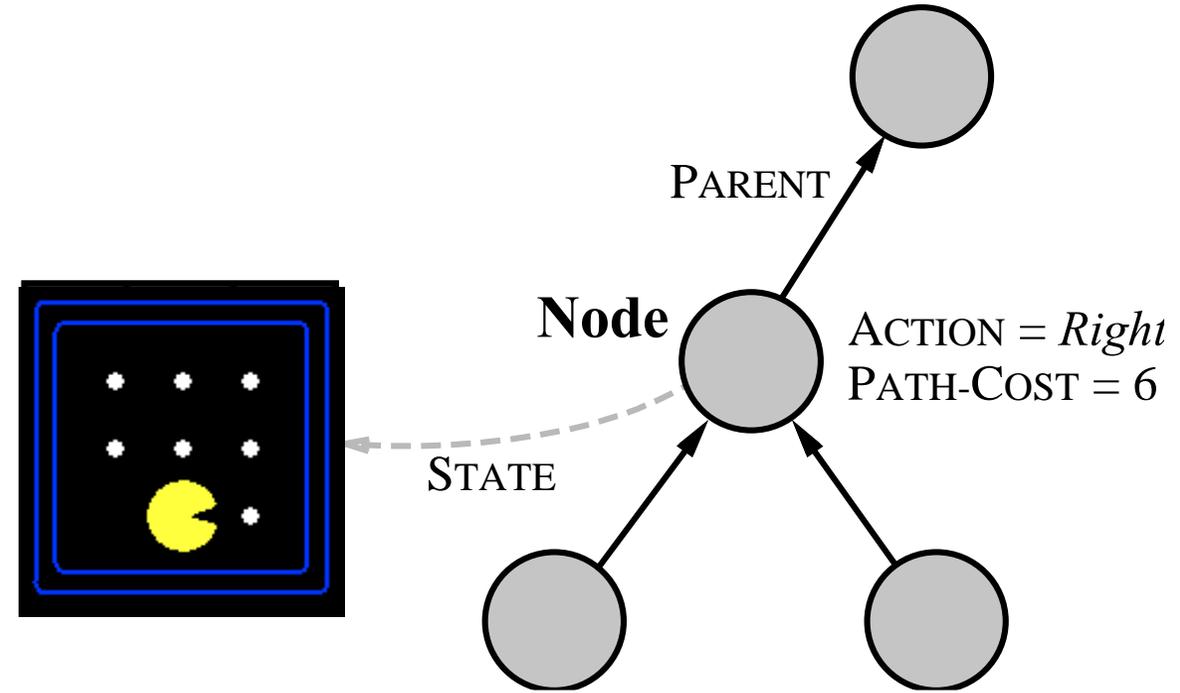
A child of `parent_node` by action a has:

`state` = `result(parent_node.state, a)`

`parent` = `parent_node`

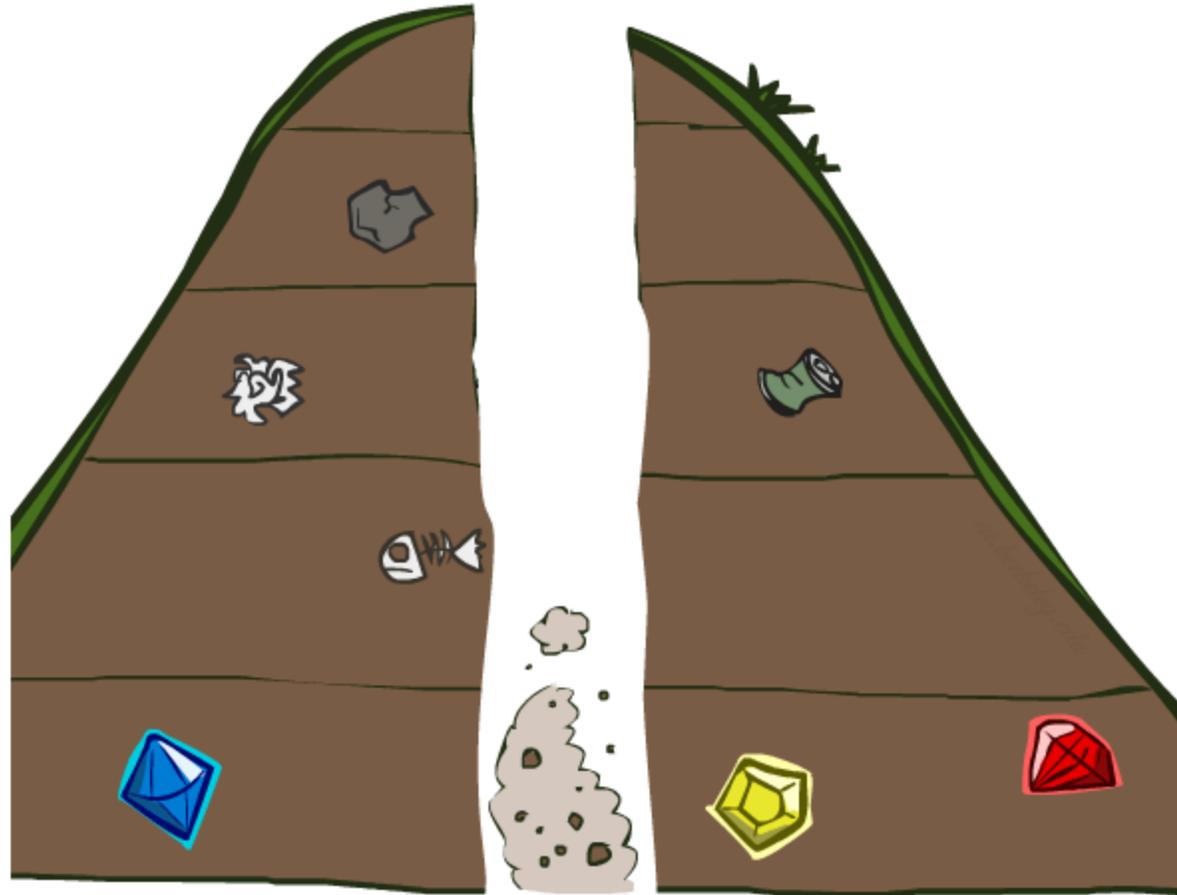
`action` = a

`path-cost` = `parent_node.path_cost` +
`step_cost(parent_node.state, a, self.state)`

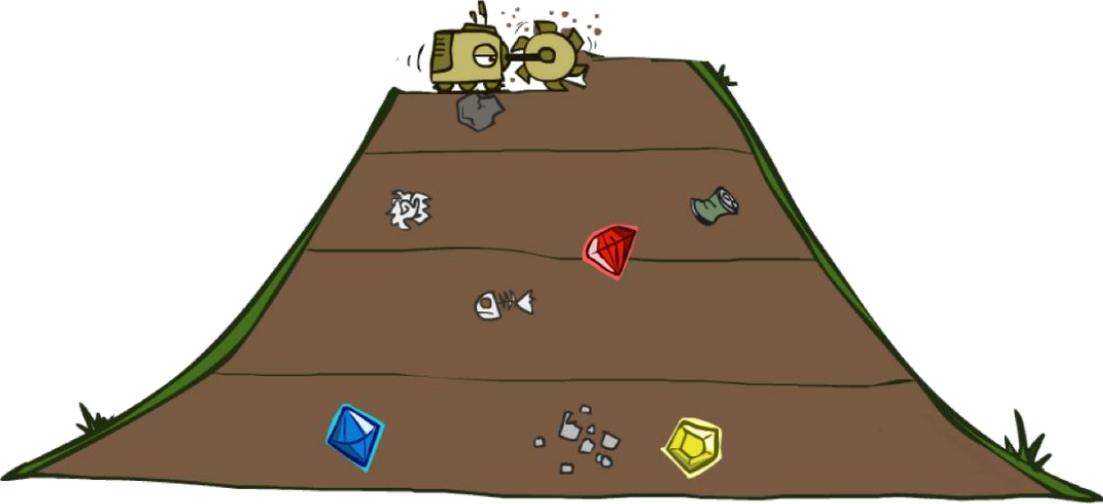


Extract solution by tracing back parent pointers, collecting actions

Search Algorithm Properties

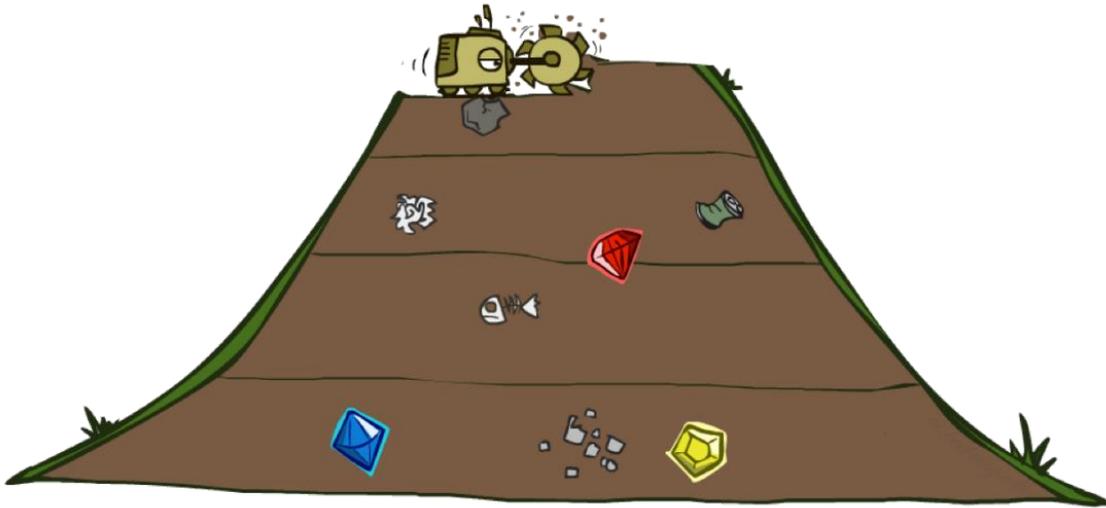


BFS vs DFS

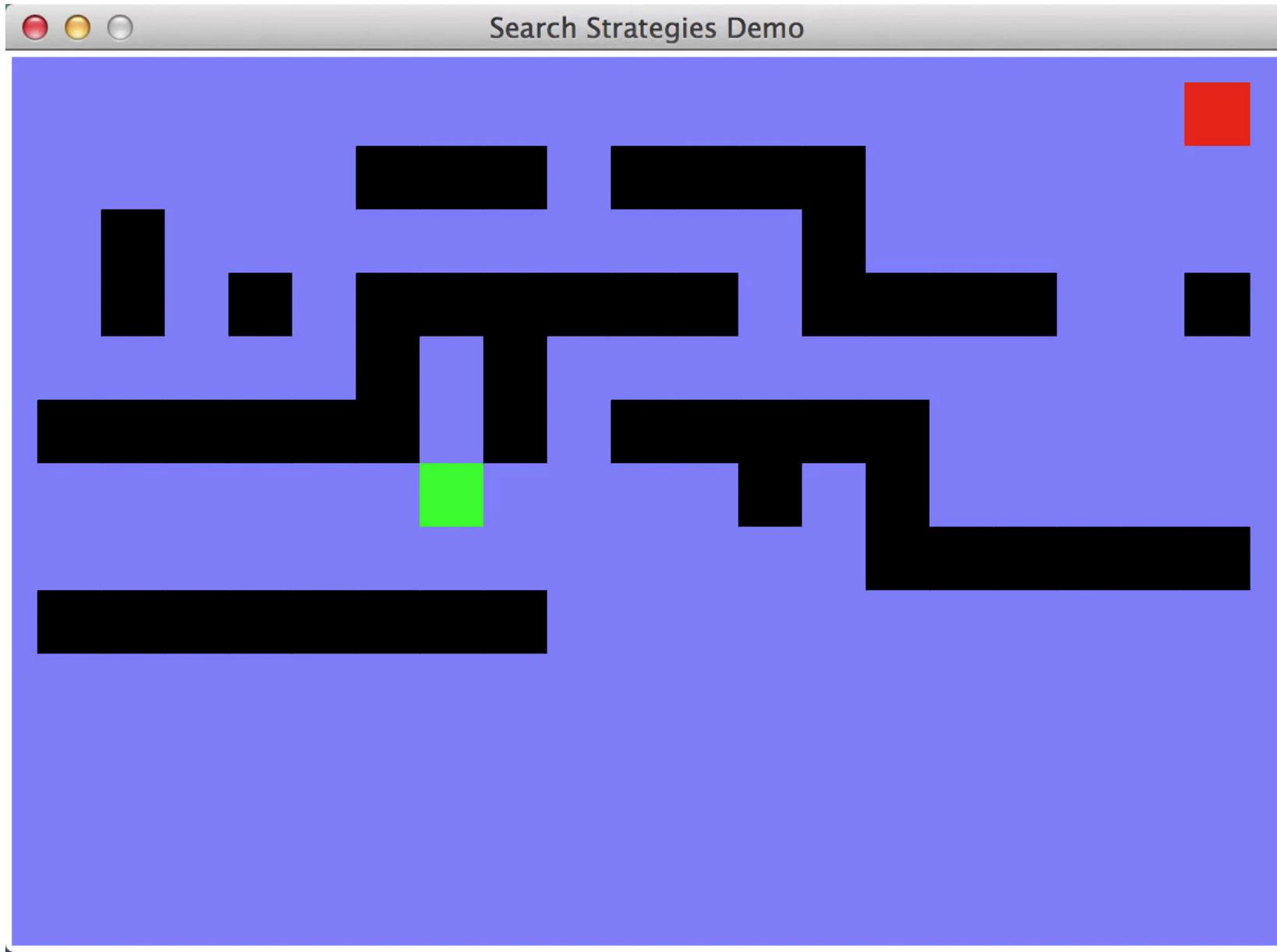


BFS vs DFS

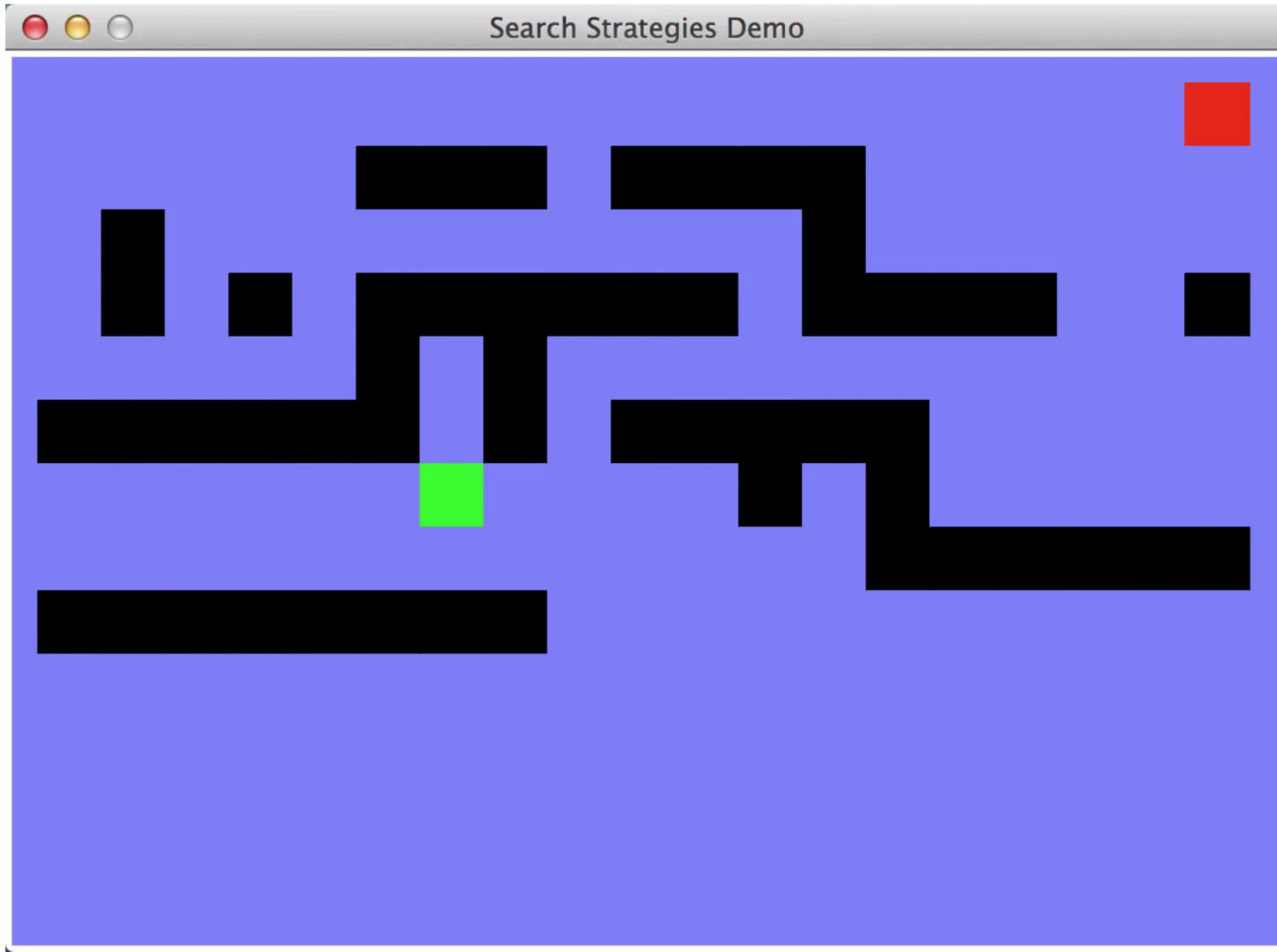
Is the following demo using BFS or DFS



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



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



BFS vs DFS

When will BFS outperform DFS?

When will DFS outperform BFS?

Search Algorithm Properties

Complete: Guaranteed to find a solution if one exists?

Optimal: Guaranteed to find the least cost path?

Time complexity?

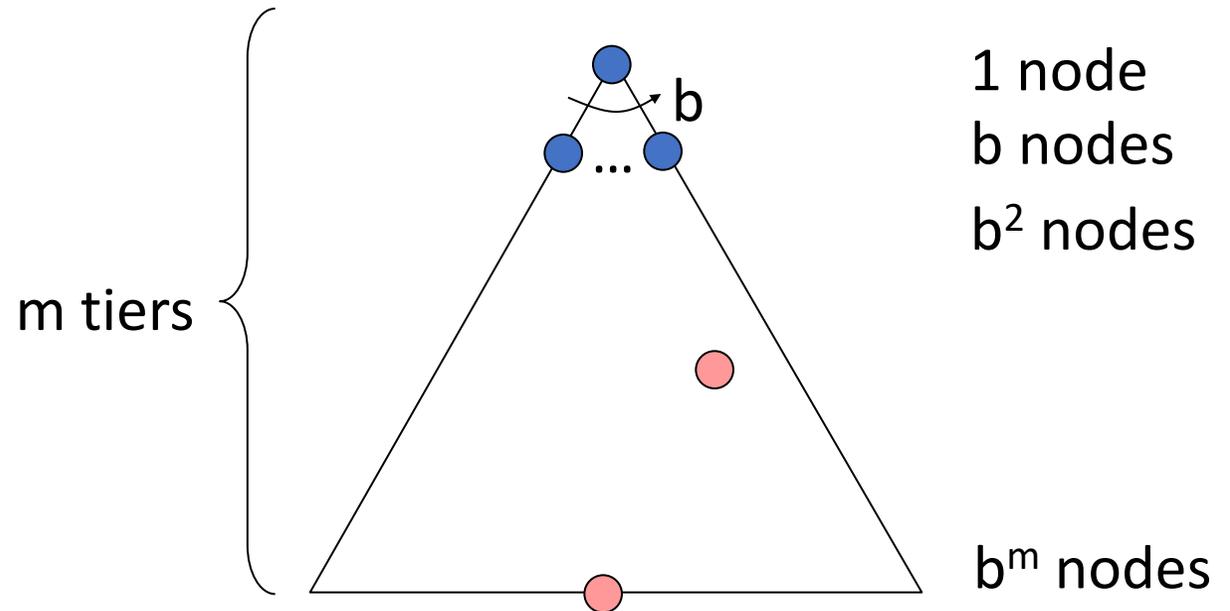
Space complexity?

Cartoon of search tree:

- b is the branching factor
- m is the maximum depth
- solutions at various depths

Number of nodes in entire tree?

- $1 + b + b^2 + \dots + b^m = O(b^m)$



Search Algorithm Properties

Complete: Guaranteed to find a solution if one exists?

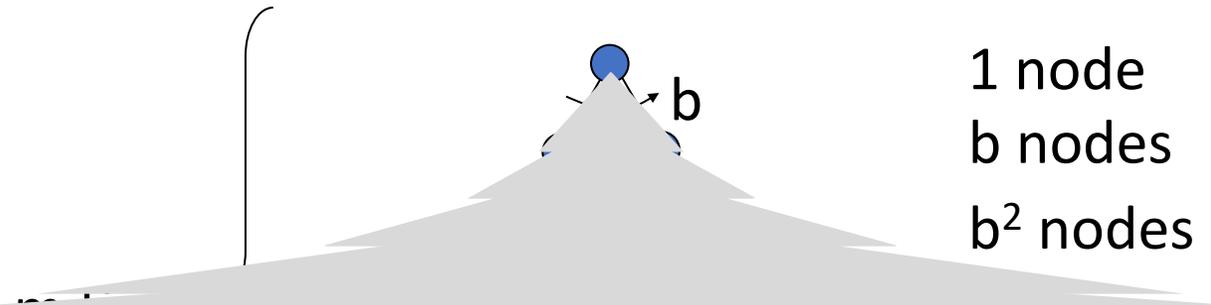
Optimal: Guaranteed to find the least cost path?

Time complexity?

Space complexity?

Cartoon of search tree:

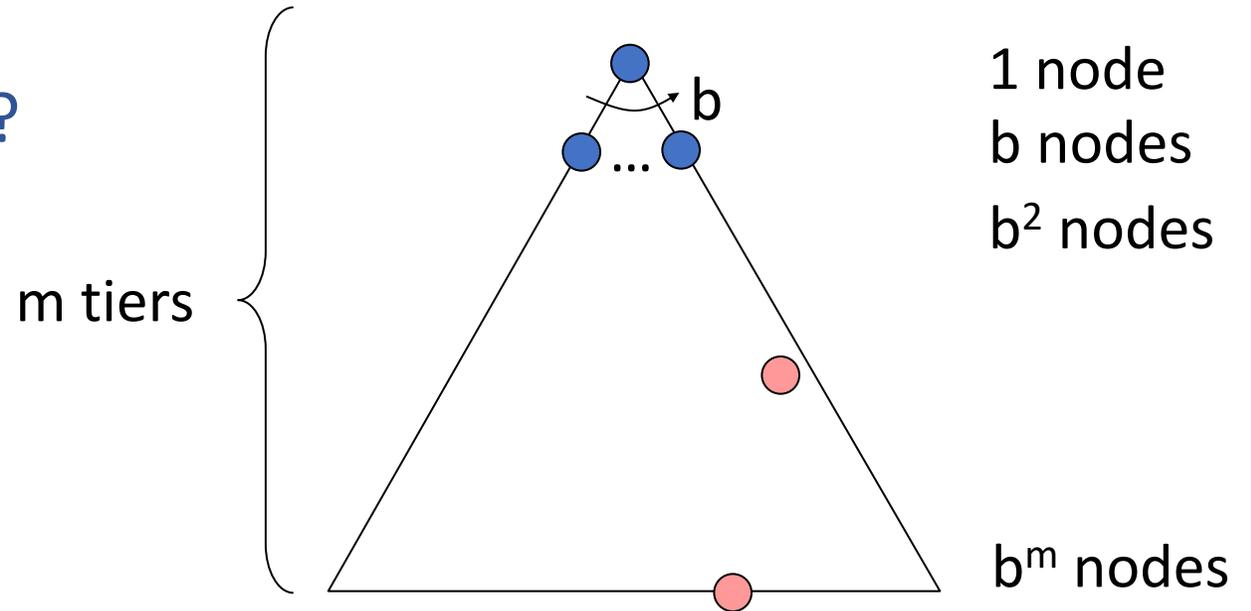
- b is the branching factor



Poll 2

Are these the properties for BFS or DFS?

- Takes $O(b^m)$ time
- Uses $O(bm)$ space on frontier
- Complete with graph search
- Not optimal unless all goals are in the same level (and the same step cost everywhere)



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 frontier take?

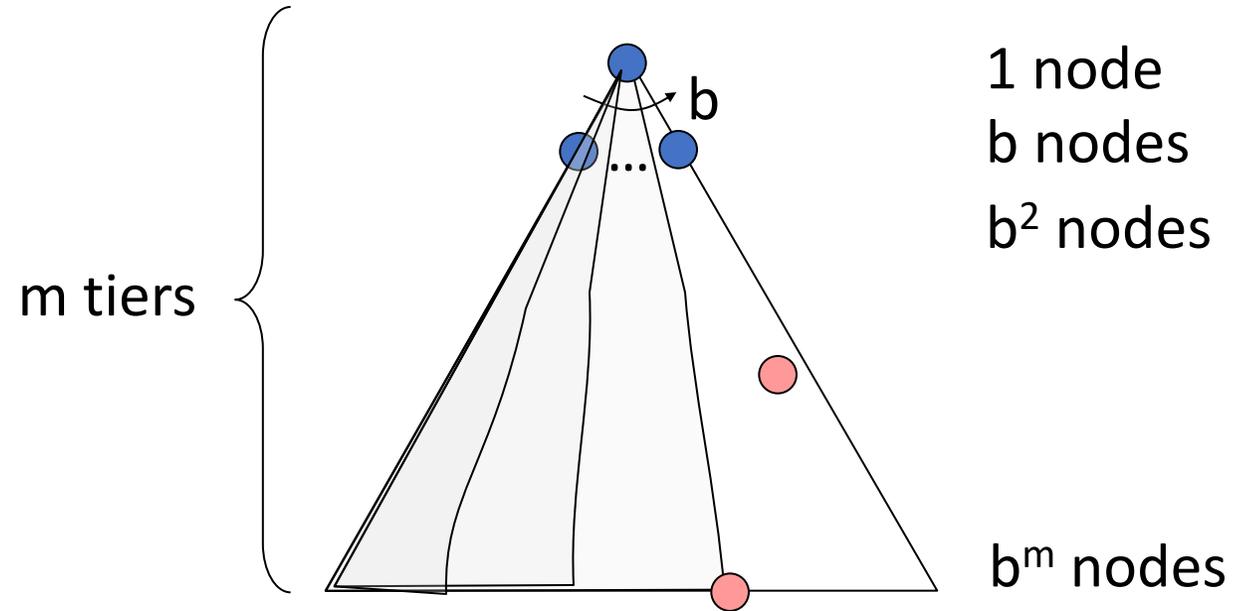
- Only has siblings on path to root, so $O(bm)$

Is it complete?

- m could be infinite, so only if we prevent cycles (graph search)

Is it optimal?

- No, it finds the “leftmost” solution, regardless of depth or cost



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 frontier take?

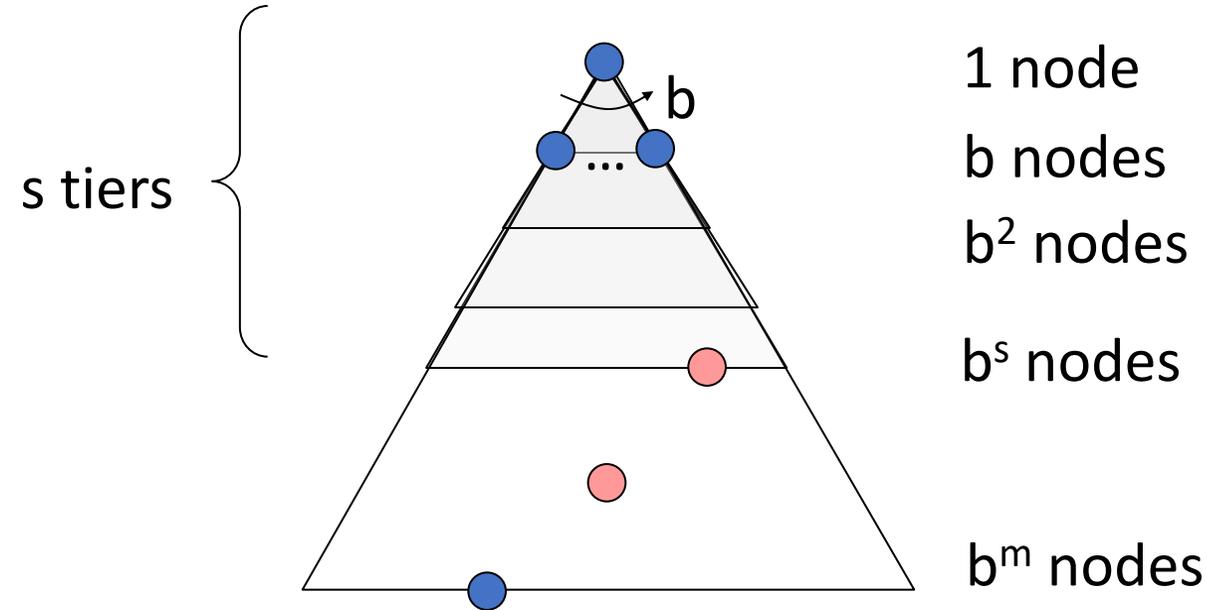
- Has roughly the last tier, so $O(b^s)$

Is it complete?

- s must be finite if a solution exists, so yes!

Is it optimal?

- Only if costs are all the same (more on costs later)



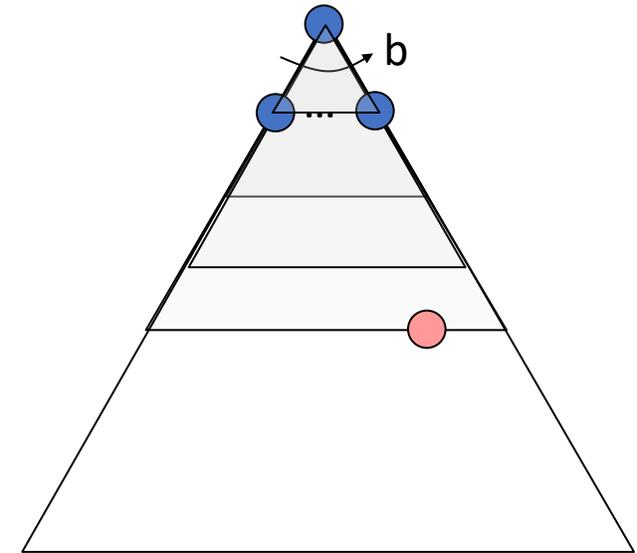
Iterative Deepening

Idea: get DFS's space advantage with BFS's time / shallow-solution advantages

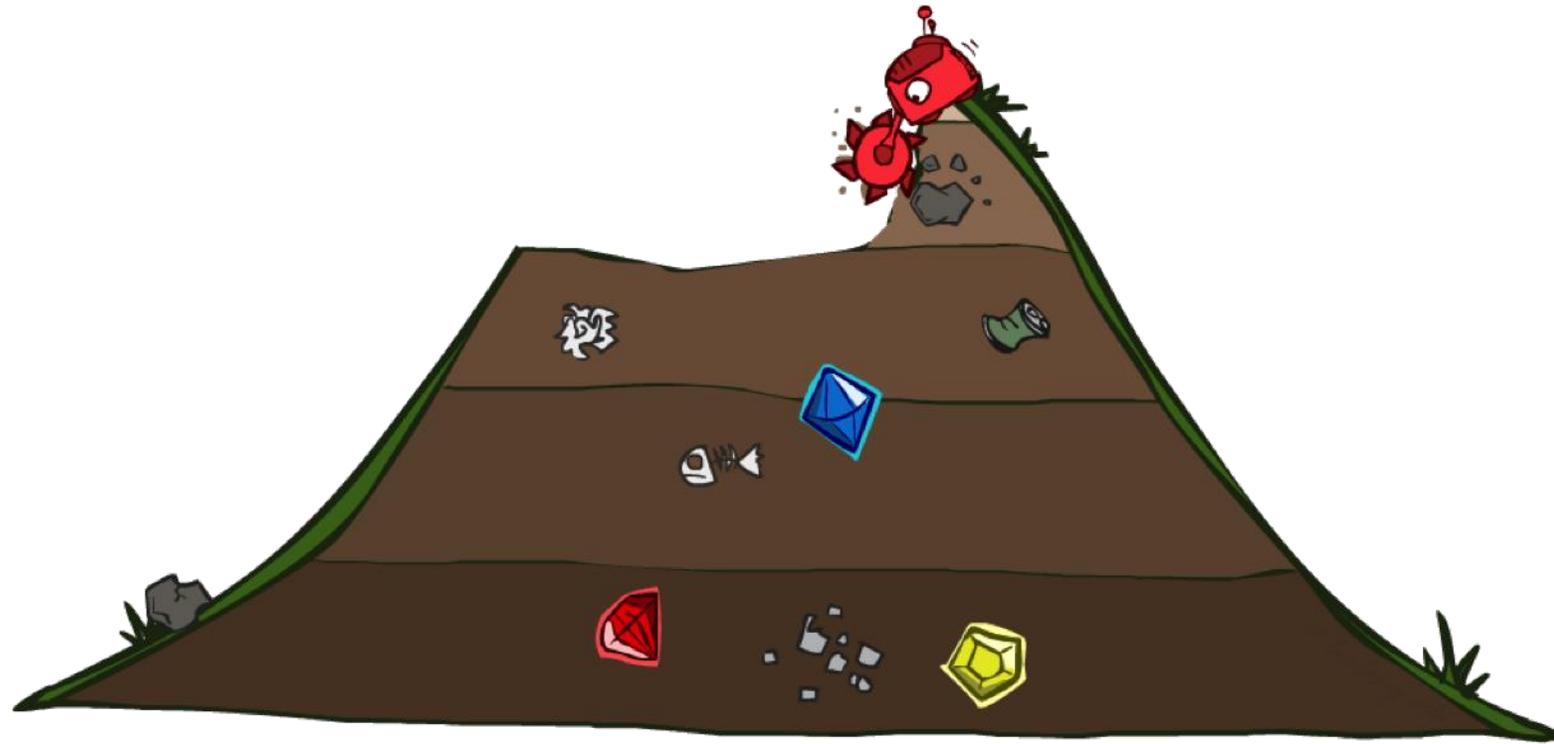
- Run a DFS with depth limit 1. If no solution...
- Run a DFS with depth limit 2. If no solution...
- Run a DFS with depth limit 3.

Isn't that wastefully redundant?

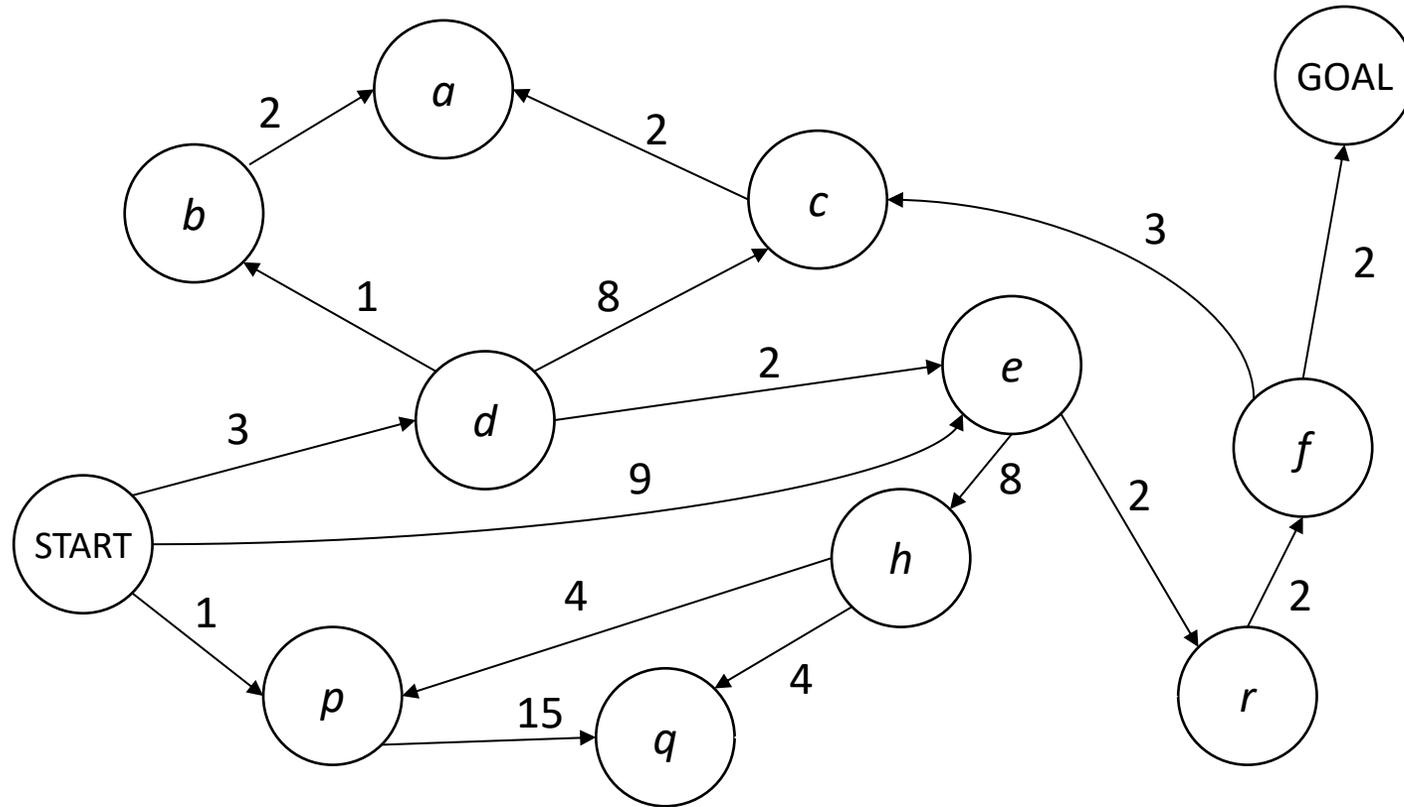
- Generally most work happens in the lowest level searched, so not so bad!



Uniform Cost Search



Finding a Least-Cost Path

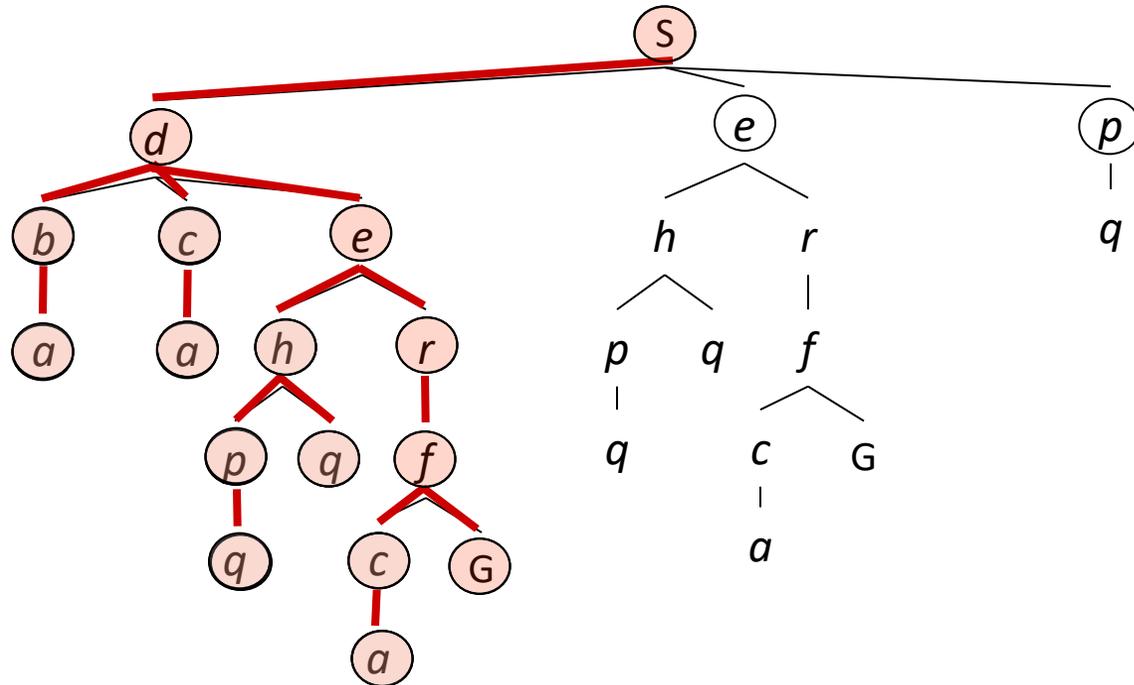
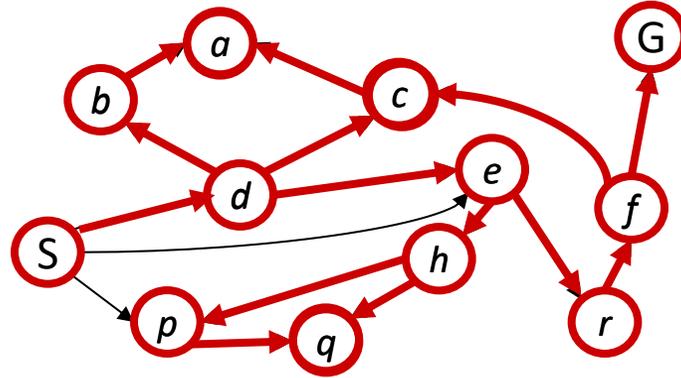


Depth-First (Tree) Search

Strategy: expand a
deepest node first

Implementation:

Frontier is a LIFO stack

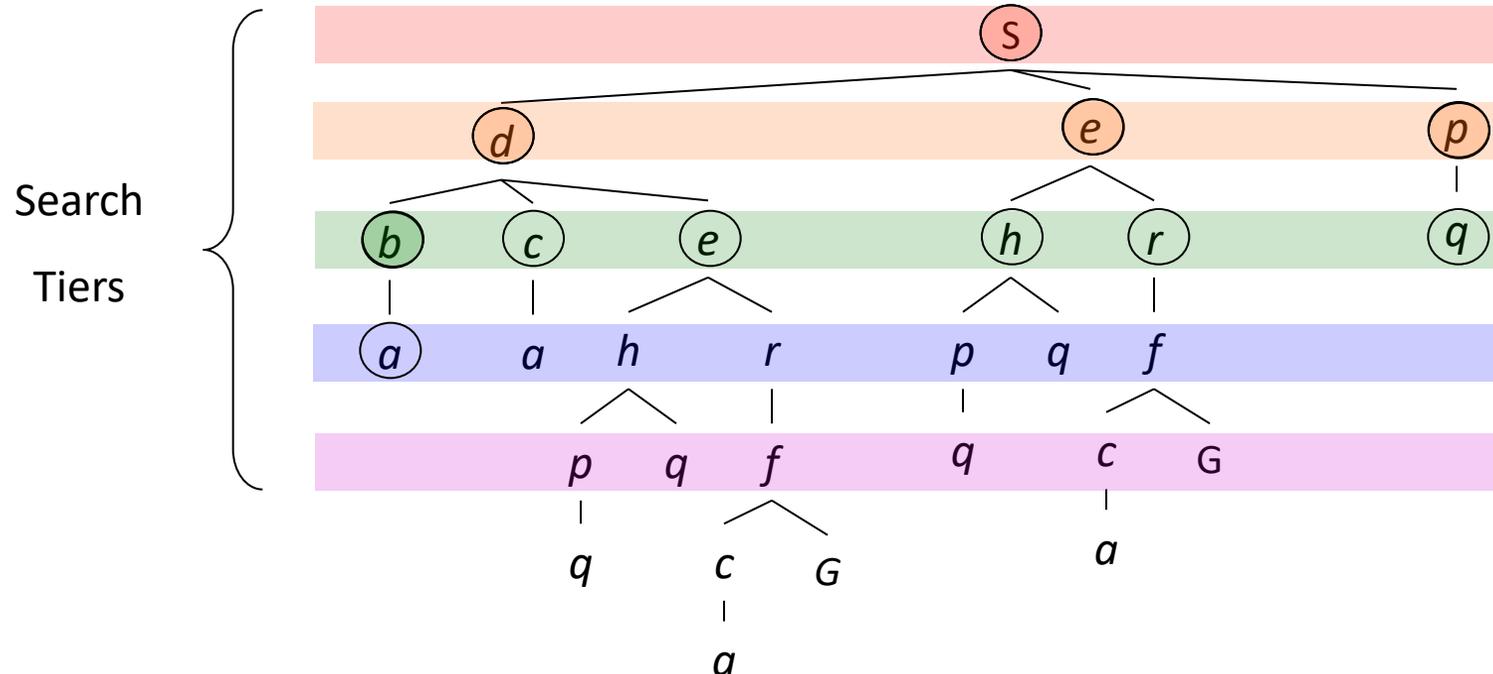
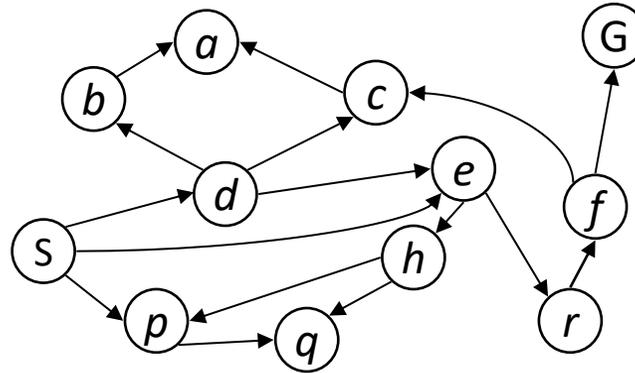


Breadth-First (Tree) Search

Strategy: expand a shallowest node first

Implementation:

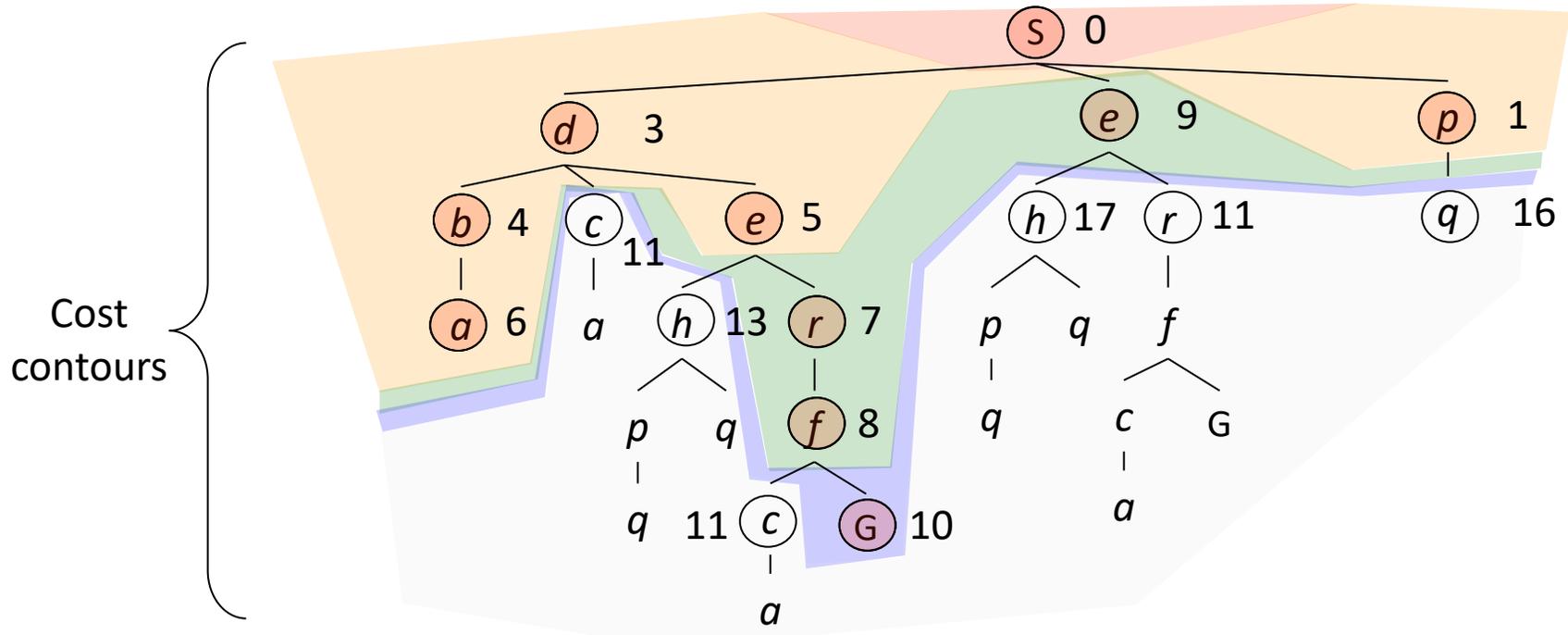
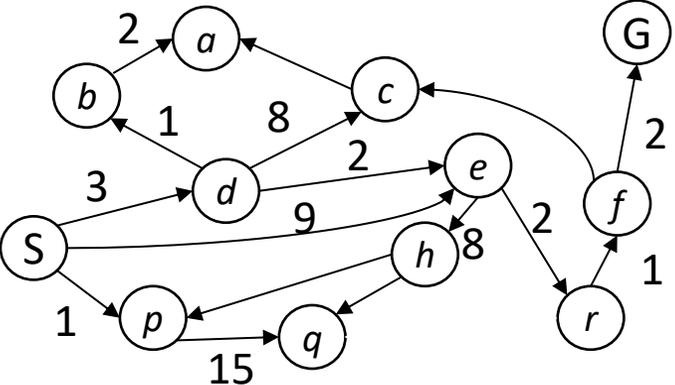
Frontier is a FIFO queue



Uniform Cost (Tree) Search

Strategy: expand a cheapest node first:

Frontier is a priority queue (priority: cumulative cost)



function GRAPH_SEARCH(problem) returns a solution, or failure

initialize the explored set to be empty

initialize the frontier as a specific work list (stack, queue, priority queue)

add initial state of problem to frontier

loop do

if the frontier is empty then

return failure

choose a node and remove it from the frontier

if the node contains a goal state then

return the corresponding solution

add the node state to the explored set

for each resulting child from node

if the child state is not already in the frontier or explored set then

add child to the frontier

function UNIFORM-COST-SEARCH(**problem**) **returns** a solution, or failure

initialize the **explored set** to be empty

initialize the **frontier** as a **priority queue** using **node path_cost** as the **priority**

add initial state of **problem** to **frontier** with **path_cost = 0**

loop do

if the **frontier** is empty **then**

return failure

choose a **node** and remove it from the **frontier**

if the **node** contains a goal state **then**

return the corresponding solution

add the **node** state to the **explored set**

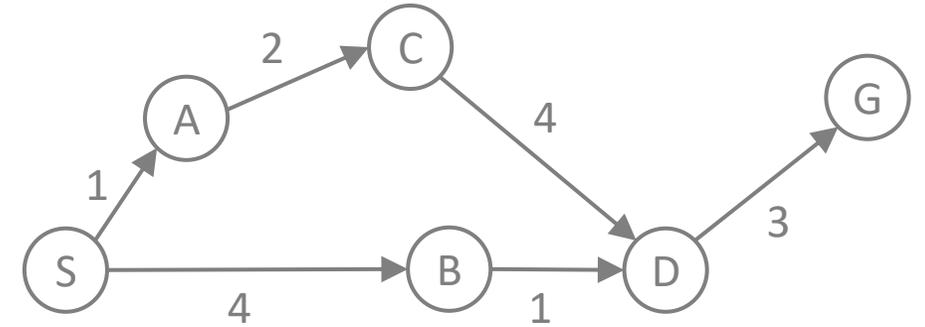
for each resulting **child** from node

if the **child** state is not already in the **frontier** or **explored set** **then**

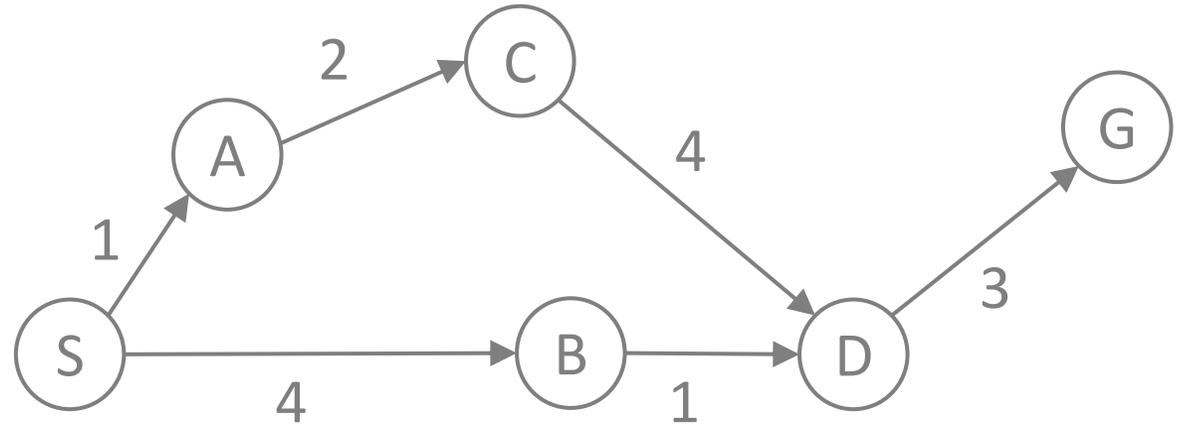
add **child** to the **frontier**

else if the **child** is already in the **frontier** with higher **path_cost** **then**

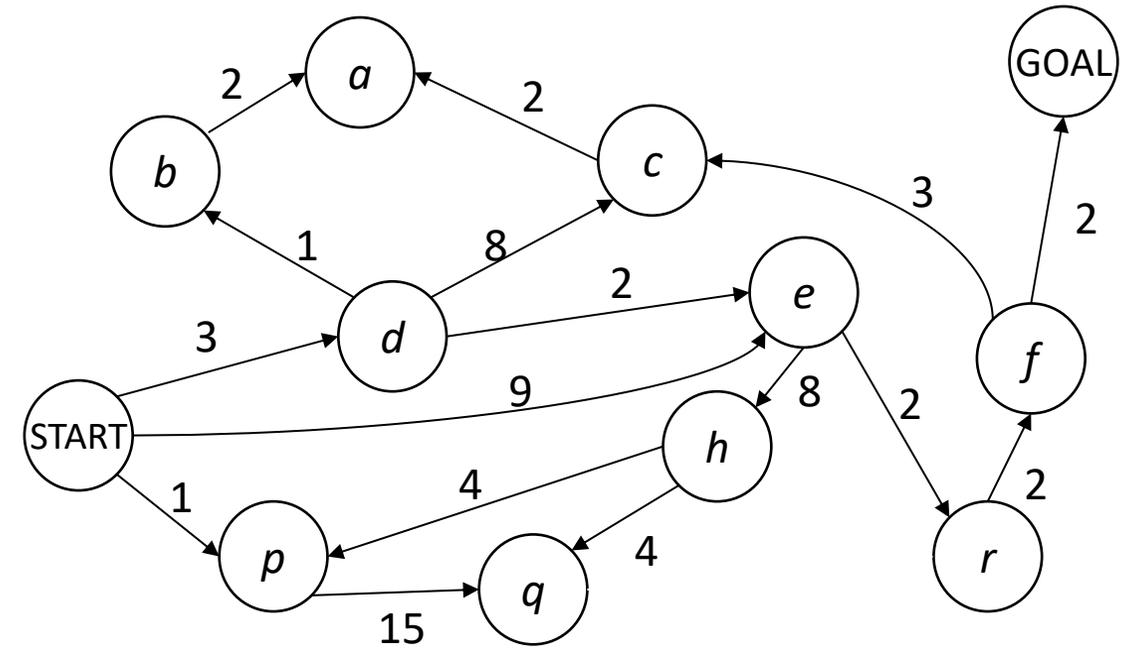
replace that **frontier** node with **child**



Walk-through UCS



UCS: Another Example



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^*/ε
- Takes time $O(b^{C^*/\varepsilon})$ (exponential in effective depth)

How much space does the frontier take?

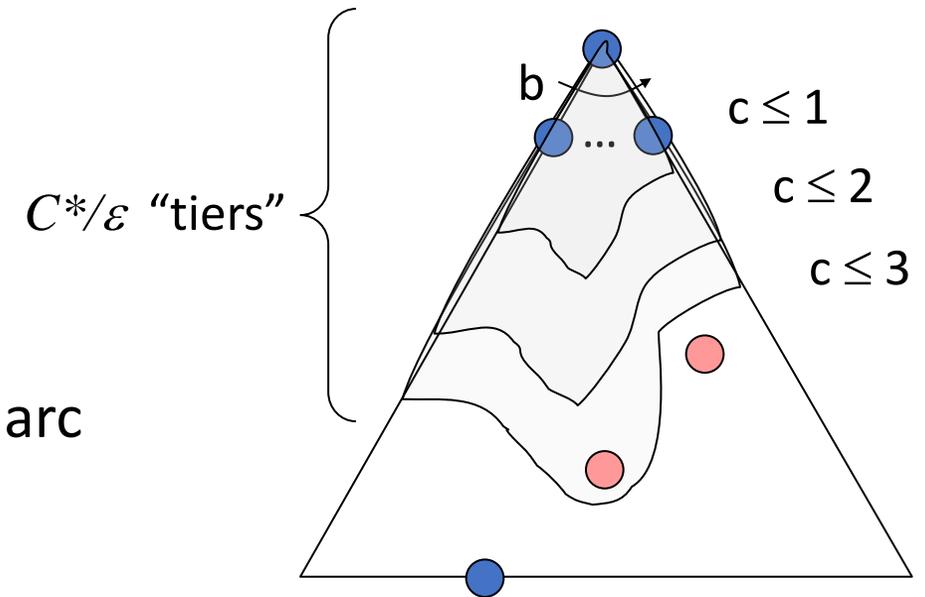
- Has roughly the last tier, so $O(b^{C^*/\varepsilon})$

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!

