CS440/ECE 448 Lecture 4: Search Intro

Slides by Svetlana Lazebnik, 9/2016

Modified by Mark Hasegawa-Johnson, 1/2019



Types of agents

Reflex agent



- Consider how the world IS
- Choose action based on current percept
- Do not consider the future consequences of actions

Goal-directed agent



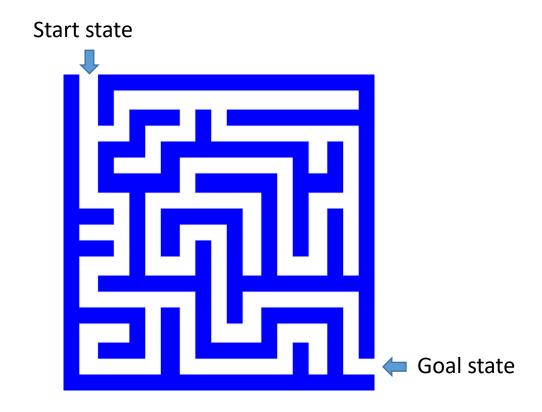
- Consider how the world WOULD BE
- Decisions based on (hypothesized) consequences of actions
- Must have a model of how the world evolves in response to actions
- Must formulate a goal

Outline of today's lecture

- 1. How to define search problems:
 - 1. Initial state, goal state, transition model
 - 2. Actions, path cost
- 2. General algorithm for solving search problems
 - 1. First data structure: a frontier list
 - 2. Second data structure: a search tree
 - 3. Third data structure: a "visited states" list
- 3. Depth-first search: very fast, but not guaranteed
- 4. Breadth-first search: guaranteed optimal
- 5. Uniform cost search = Dijkstra's algorithm = BFS with variable costs

Search

We will consider the problem of designing goal-based agents in fully observable, deterministic, discrete, static, known environments



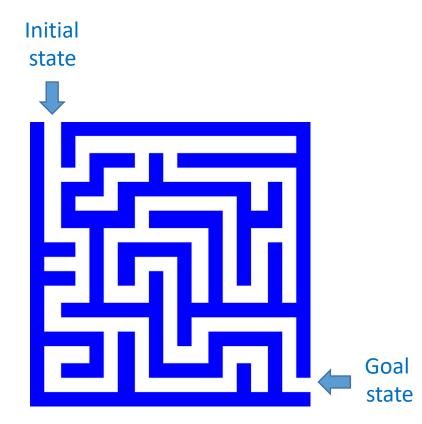
Search

We will consider the problem of designing goal-based agents in fully observable, deterministic, discrete, static, known environments

- The agent must find a *sequence of actions* that reaches the goal
- The performance measure is defined by (a) reaching the goal and (b) how "expensive" the path to the goal is
 - The **agent doesn't know** the performance measure. This is a goal-directed agent, not a utility-directed agent
 - The **programmer (you) DOES know** the performance measure. So you design **a goal-seeking strategy** that minimizes cost.
- We are focused on the process of finding the solution;
 we assume that the agent can safely ignore its percepts while executing the solution (static environment, open-loop system)

Search problem components

- Initial state
- Actions
- Transition model
 - What state results from performing a given action in a given state?
- Goal state
- Path cost
 - Assume that this is a sum of nonnegative step costs



• The **optimal solution** is the sequence of actions that gives the *lowest* path cost for reaching the goal

Knowledge Representation: State

- State = description of the world
 - Must have enough detail to decide whether or not you're currently in the **initial state**
 - Must have enough detail to decide whether or not you've reached the goal state
 - Often but not always: "defining the state" and "defining the transition model" are the same thing

Example: Romania

- On vacation in Romania; currently in Arad
- Flight leaves tomorrow from Bucharest

Initial state

Arad

Actions

Go from one city to another

Transition model

 If you go from city A to city B, you end up in city B

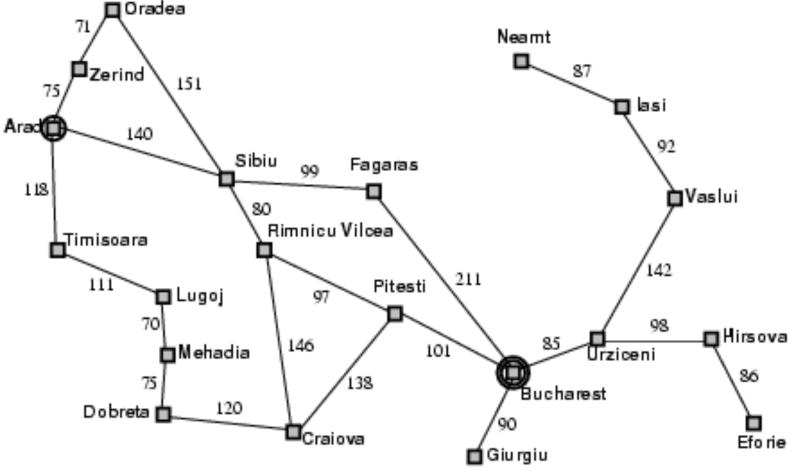
Goal state

Bucharest

Path cost

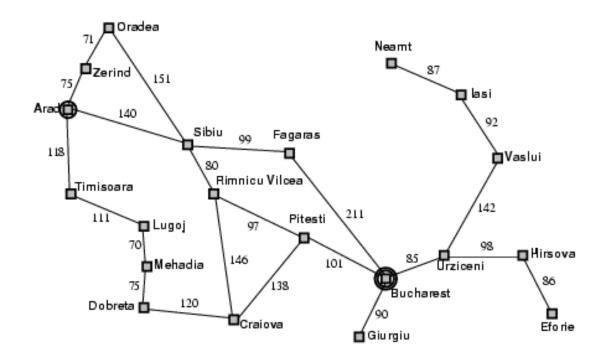
 Sum of edge costs (total distance traveled)





State space

- The initial state, actions, and transition model define the state space of the problem
 - The set of all states reachable from the initial state by any sequence of actions
 - Can be represented as a directed graph where the nodes are states and links between nodes are actions
- What is the state space for the Romania problem?
 - State Space = O{# cities}



Traveling Salesman Problem

• Goal:

Visit every city in US

• Path cost:

Total miles traveled

Initial state:

Champaign, IL

• Actions:

Travel from one city to another

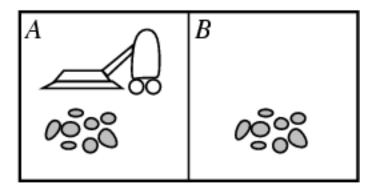
Transition model:

When you visit a city, mark it as "visited."

• State Space = O(2#cities)



Example: Vacuum world



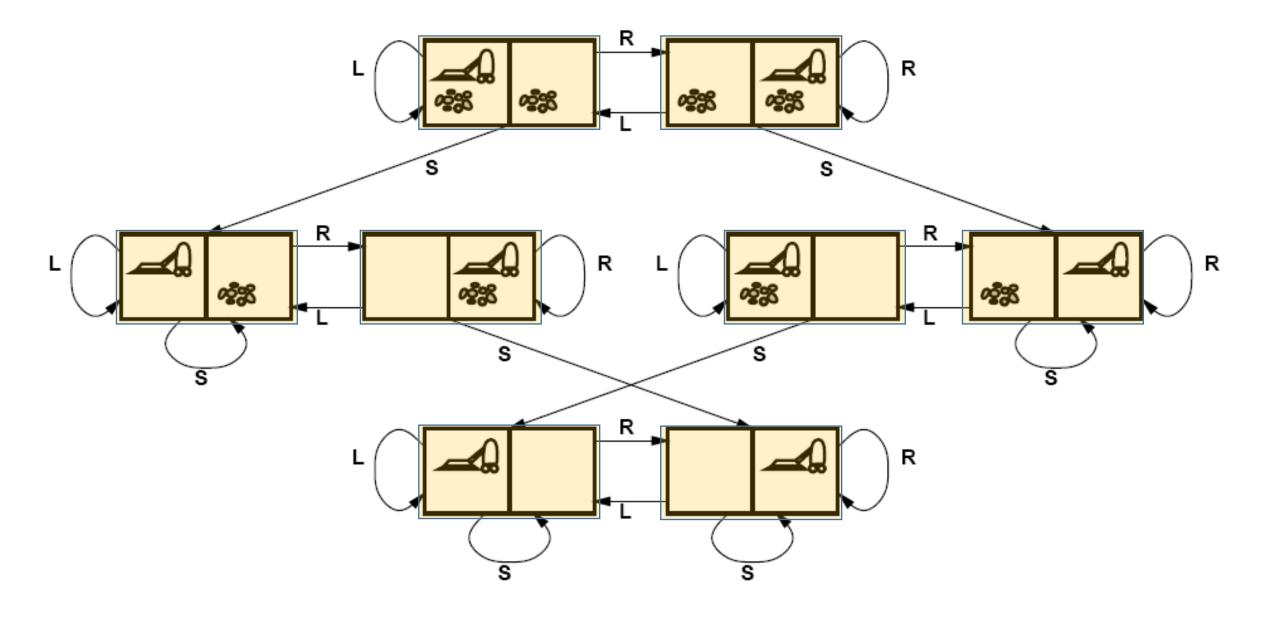
States

- Agent location and dirt location
- How many possible states?
- What if there are *n* possible locations?
 - The size of the state space grows exponentially with the "size" of the world!

Actions

- Left, right, suck
- Transition model

Vacuum world state space graph



Complexity of the State Space

- Many "video game" style problems can be subdivided:
 - If there are M different things your character needs to pick up: 2^M different world states (one for each subset of things that you've picked up)
 - If there are N different locations you can be in while carrying any subset of those M objects: Total number of world states = $O(2^M N)$
- Why a maze is nice: you don't need to pick anything up
 - Only N different world states to consider

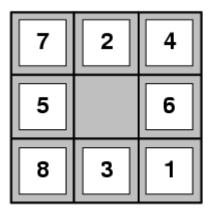
Example: The 8-puzzle

States

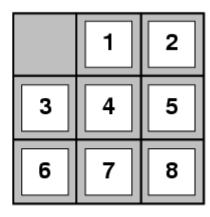
- Locations of tiles
 - 8-puzzle: 181,440 states (9!/2)
 - 15-puzzle: ~10 trillion states
 - 24-puzzle: ~10²⁵ states

Actions

- Move blank left, right, up, down
- Path cost
 - 1 per move



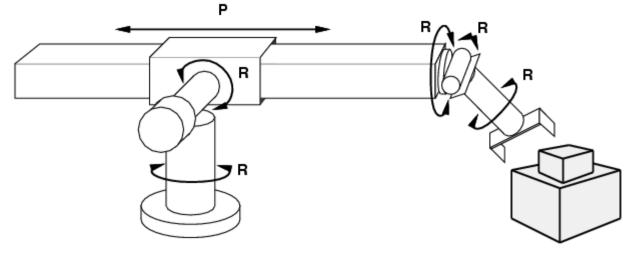
Start State



Goal State

• Finding the optimal solution of n-Puzzle is NP-hard

Example: Robot motion planning



States

- Real-valued joint parameters (angles, displacements)
- Actions
 - Continuous motions of robot joints
- Goal state
 - Configuration in which object is grasped
- Path cost
 - Time to execute, smoothness of path, etc.

Outline of today's lecture

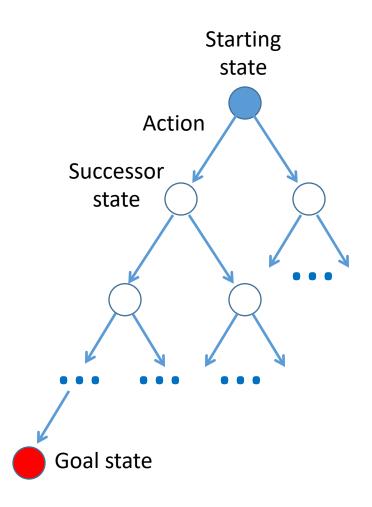
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First data structure: a frontier list

- Let's begin at the start state and expand it by making a list of all possible (immediate) successor states
- Maintain a frontier, i.e. a list of unexpanded states
- At each step, pick a state from the frontier to expand:
 - Check to see if it's a goal state
 - If not, find the other states that can be reached from this state, and add them to the frontier, if they're not already there
- Keep going until you reach a goal state

Second data structure: a search tree

- "What if" tree of sequences of actions and outcomes
- The root node corresponds to the starting state
- The children of a node correspond to the successor states of that node's state
- A path through the tree corresponds to a sequence of actions
 - A solution is a path ending in the goal state



Knowledge Representation: States and Nodes

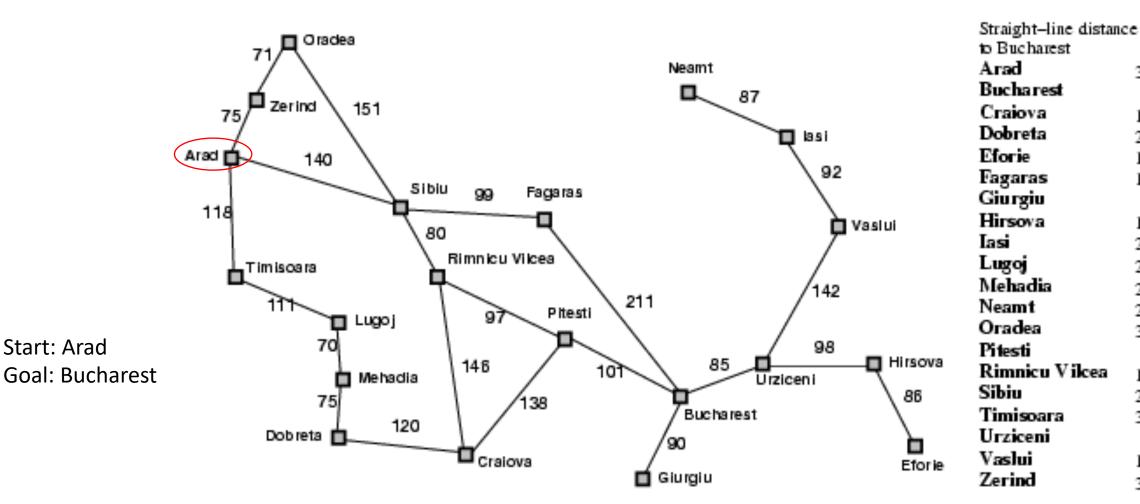
- **State** = description of the world
 - Must have enough detail to decide whether or not you're currently in the <u>initial state</u>
 - Must have enough detail to decide whether or not you've reached the goal state
 - Often but not always: "defining the state" and "defining the transition model" are the same thing
- Node = a point in the search tree
 - Private data: ID of the state reached by this node
 - Private data: the ID of the parent node
 - NB: each state may occur multiple times in the same search tree

Tree Search Algorithm Outline

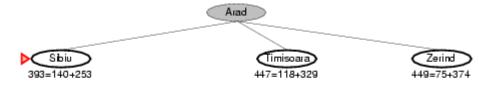
- Initialize the frontier using the starting state
- While the frontier is not empty
 - Choose a frontier node according to search strategy and take it off the frontier
 - If the node contains the goal state, return solution
 - Else expand the node and add its children to the frontier
- Search strategy determines
 - Is this process guaranteed to return an OPTIMAL solution?
 - Is this process guaranteed to return ANY solution?
 - Time complexity: How much time does it take?
 - Space complexity: How much RAM is consumed by the frontier?
- For now: assume that search strategy = random



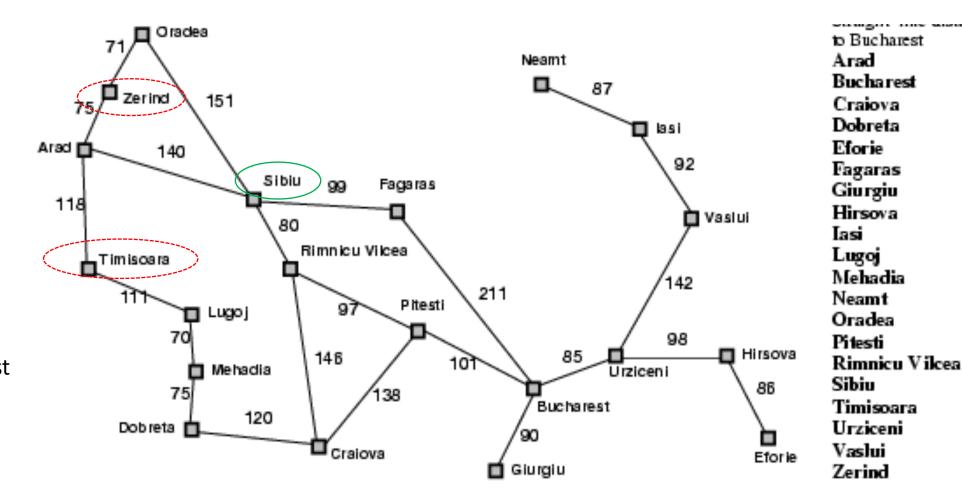
Start: Arad



to Bucharest Arad 366 Bucharest 0 Craiova 160 Dobreta 242Eforie 161 Fagaras 176 Giurgiu 77 Hirsova 151 Iasi 226 Lugoj 244Mehadia 241 Neamt 234 Oradea 380 Pitesti 10 Rimnicu Vilcea 193 Sibiu 253 Timisoara 329 Urziceni 80 Vaslui 199



Tree search example

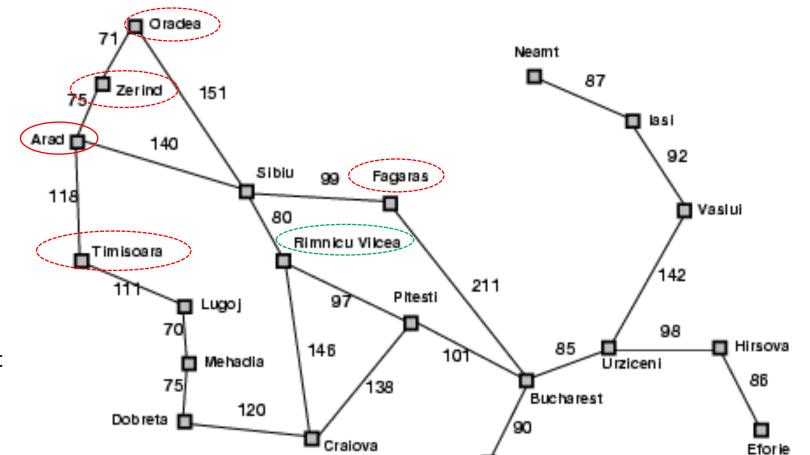


Start: Arad

Sibiu Oradea (Rimniou Vilcea Fagaras







Giurgiu

Start: Arad

Goal: Bucharest

Straight-line distance

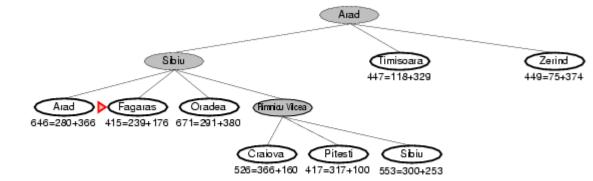
Timisoara

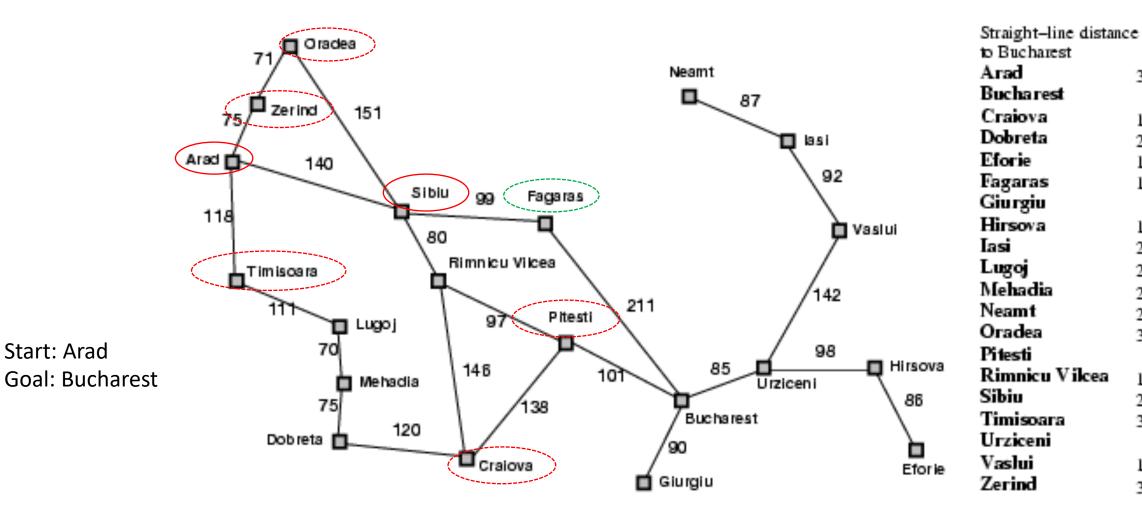
447=118+329

Arad

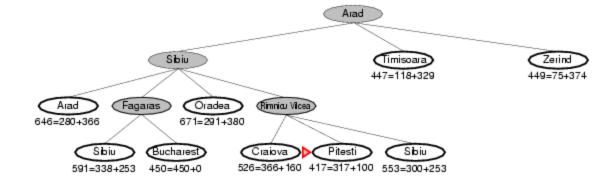
to Bucharest	
Arad	360
Bucharest	(
Craiova	160
Dobreta	242
Eforie	163
Fagaras	176
Giurgiu	77
Hirsova	153
Iasi	226
Lugoj	244
Mehadia	243
Neamt	234
Oradea	380
Pitesti	10
Rimnicu Vilcea	193
Sibiu	253
Timisoara	329
Urziceni	80
Vaslui	199
Zerind	374

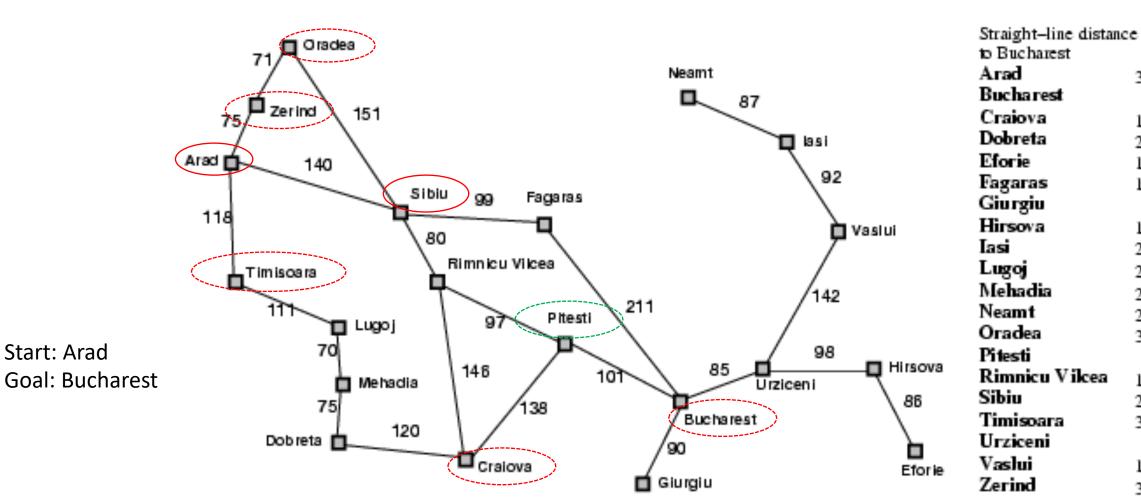
Start: Arad





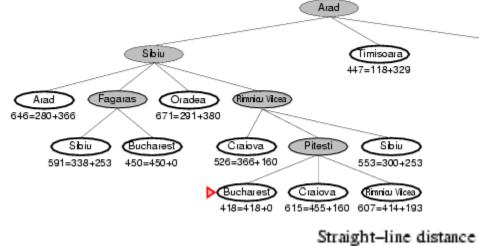
Start: Arad





Start: Arad

Goal: Bucharest



to Bucharest

Bucharest

Craiova

Dobreta

Fagaras

Giurgiu

Hirsova

Iasi

Lugoj

Neamt

Oradea

Pitesti

Sibiu

Timisoara

Urziceni

Vaslui

Zerind

Rimnicu Vilcea

Mehadia

Eforie

Arad

Zerind

449=75+374

366

160

242

161

176

77

151

226

244

241

234

380

193

253

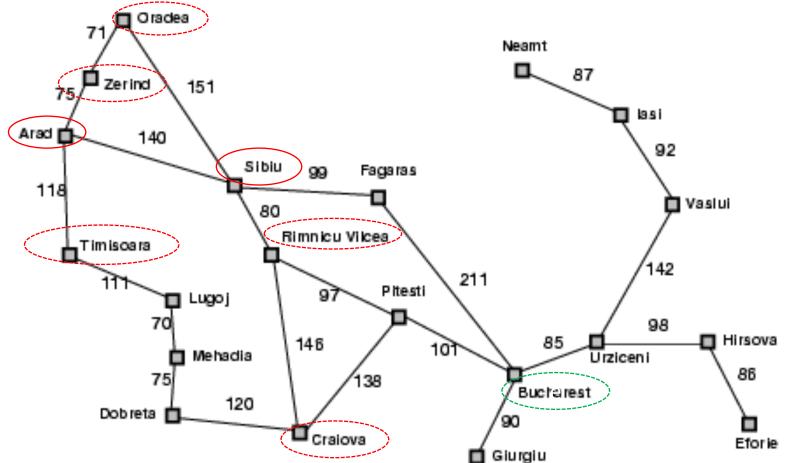
329

199

374

80

10



Handling repeated states

- Initialize the frontier using the starting state
- While the frontier is not empty
 - Choose a frontier node according to search strategy and take it off the frontier
 - If the node contains the goal state, return solution
 - Else expand the node and add its children to the frontier
- To handle repeated states:
 - Every time you expand a node, add that state to the explored set
 - When adding nodes to the frontier, CHECK FIRST to see if they've already been explored

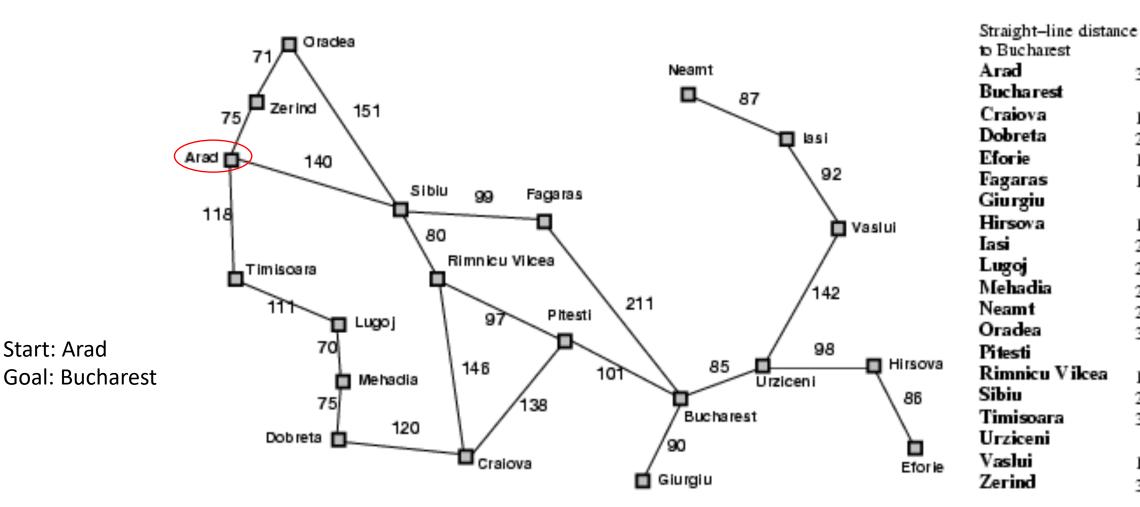
Time Complexity

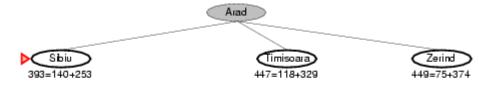
- Without explored set :
 - *0*{1}/node
 - $O\{b^m\}$ = # nodes expanded
 - b = branching factor (number of children each node might have)
 - m = length of the longest possible path
- With explored set:
 - $O\{1\}$ /node using a hash table to see if node is already in explored set
 - $O\{|S|\}$ = # nodes expanded
- Usually, $O\{|S|\} < O\{b^m\}$. I'll continue to talk about $O\{b^m\}$, but remember that it's upper-bounded by $O\{|S|\}$.



Tree search w/o repeats

Start: Arad

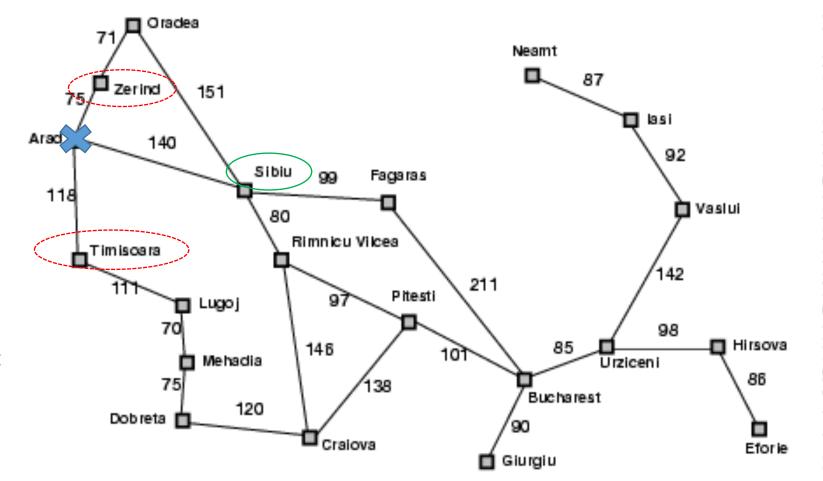




Tree search w/o repeats

Explored:

Arad



Start: Arad

to Bucharest	
Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Eforie	161
Fagaras	176
Giurgiu	77
Hirsova	151
Iasi	226
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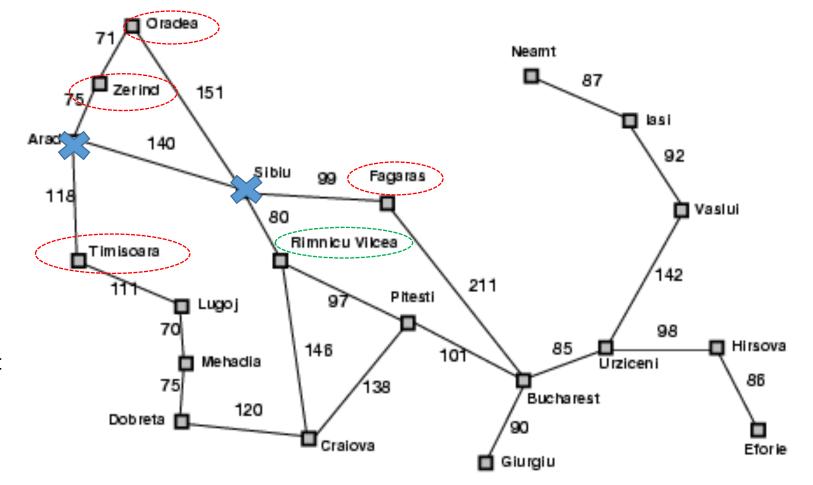
Aiad Shiu Timisoara 447=118+329 Fagaras Oradea Pimriu Vices 46=280+366 415=239+176 671=291+380 413=220+193

Tree search example

Explored:

Arad

Sibiu

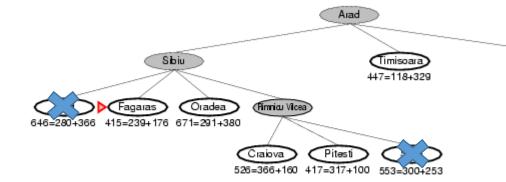


Straight-line distance to Bucharest Arad 366 Bucharest 0 Crajova 160 Dobreta 242 Eforie 161 Fagaras 176 Giurgiu Hirsova 151 Iasi 226 Lugoj 244 Mehadia 241 Neamt 234 Oradea 380 Pitesti 10 Rimnicu Vilcea 193 Sibiu 253 Timisoara 329 Urziceni 80 Vaslui 199

374

Zerind

Start: Arad

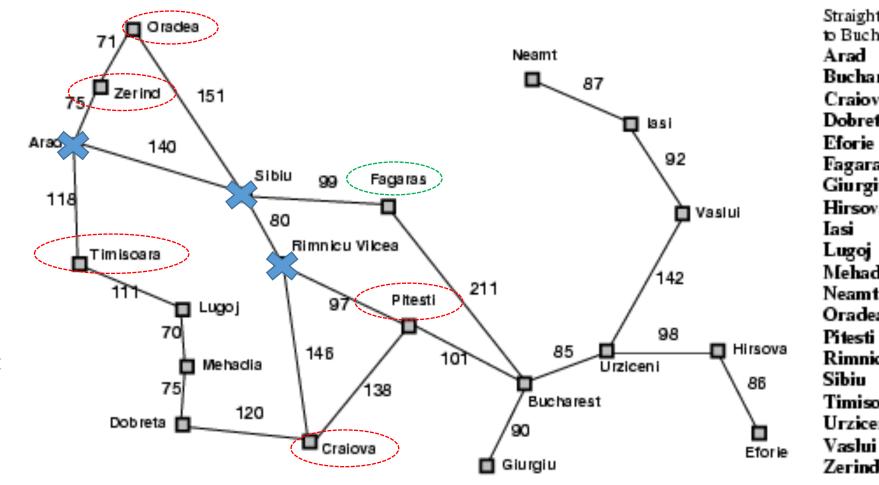


Explored:

Arad

Sibiu

Rimnicu Vilcea



Straight-line distance to Bucharest Arad 366 Bucharest 0 Craiova 160 Dobreta 242 Eforie 161 Fagaras 176 Giurgiu 77 Hirsova 151 Iasi 226 Lugoj 244 Mehadia 241 Neamt 234 Oradea 380 Pitesti 10 Rimnicu Vilcea 193 Sibiu 253 Timisoara 329 Urziceni 80

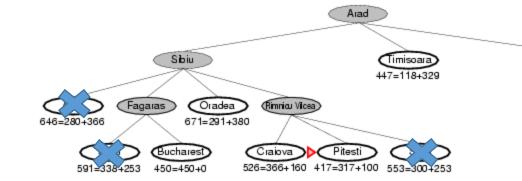
199

374

Zerind

449=75+374

Start: Arad



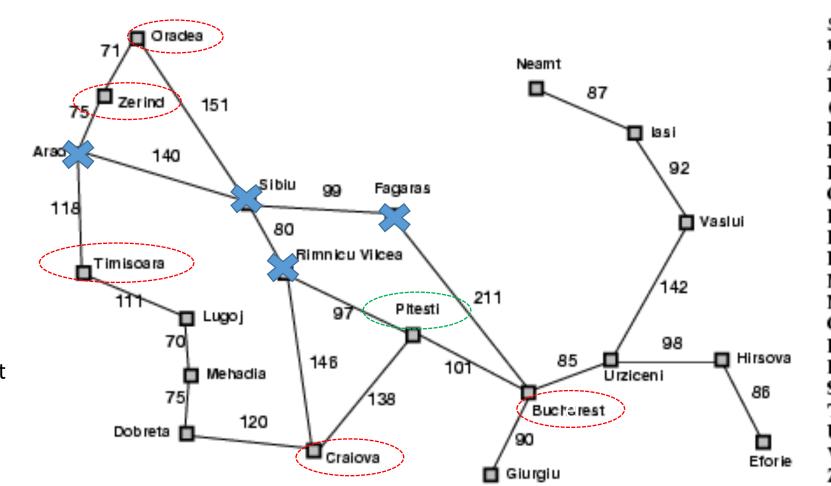
Explored:

Arad

Sibiu

Rimnicu Vilces

Fagaras



Start: Arad

Goal: Bucharest

Straight-line distance to Bucharest Arad 366 Bucharest 0 Craiova 160 Dobreta 242 Eforie 161 Fagaras 176 Giurgiu Hirsova 151 Iasi 226 Lugoj 244 Mehadia 241 Neamt 234 Oradea 380 Pitesti 10 Rimnicu Vilcea 193 Sibiu 253 Timisoara 329 Urziceni 80 Vaslui 199 Zerind 374 Zerind

449=75+374



Explored:

Arad

Sibiu

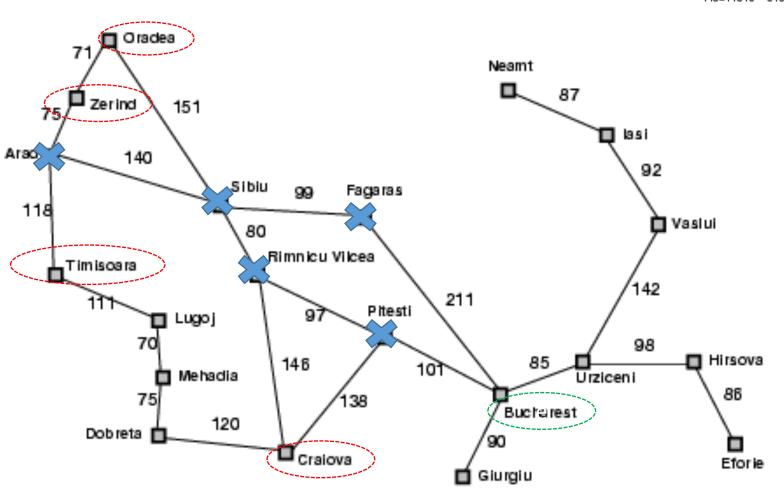
Rinnicu Vilces

Fagaras

Pitesti

Start: Arad

Goal: Bucharest



01:	
Sbiu	(Timisoara) 447=118+329
646=280+366 Fagaras Orades 671=291+	
591=338+253 450=450+0	Craiova Pitesti 553=300+253
	Bucharest Craiova Rimo (88) 418=418+0 615=455+160 607=414+193

Arad

Straight-line distance		
to Bucharest		
Arad	366	
Bucharest	0	
Craiova	160	
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Straight_line distance

Zerind

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Depth-First Search

• Basic idea:

• Try to find a solution as fast as possible

• How:

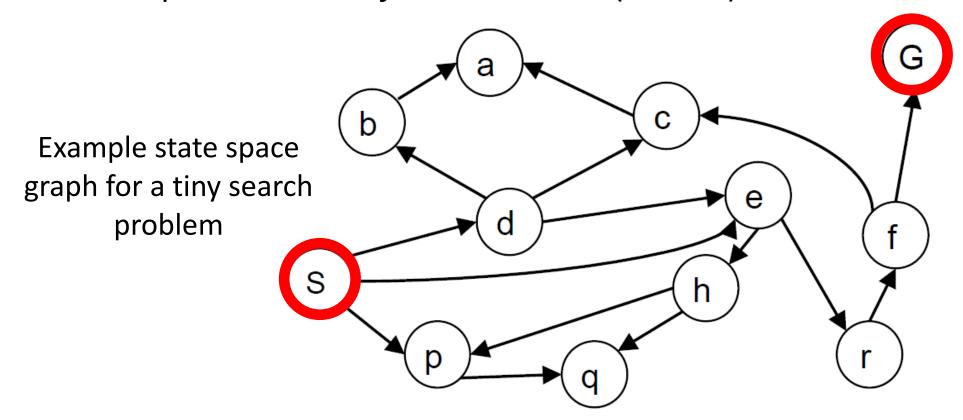
 From the frontier, always choose a node which is AS FAR FROM THE STARTING POINT AS POSSIBLE

• How:

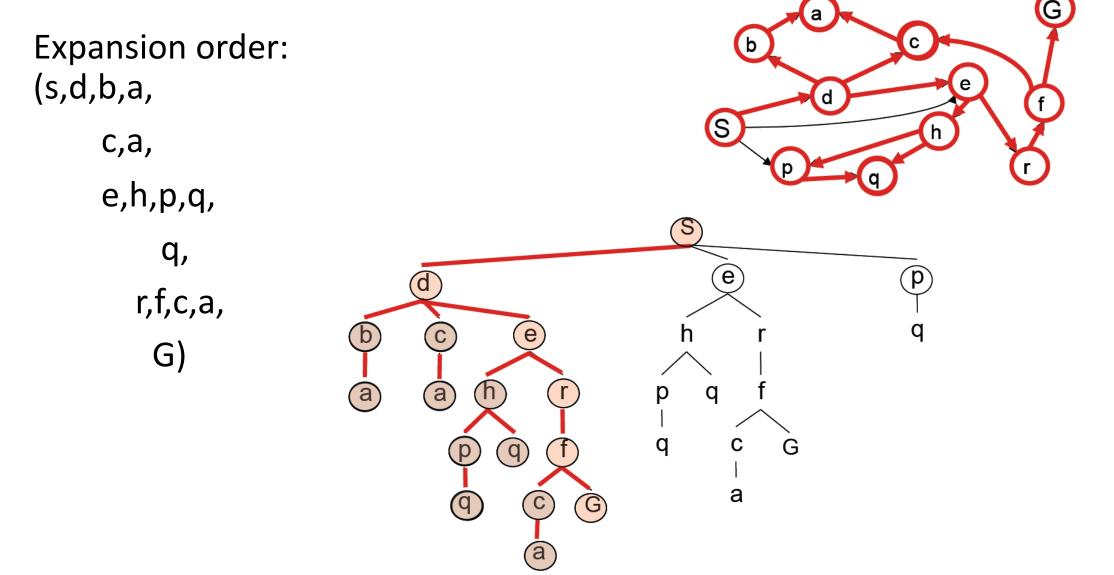
- Frontier is a LIFO (last-in, first-out) stack.
- The node you expand = whichever node has been most recently placed on the queue.

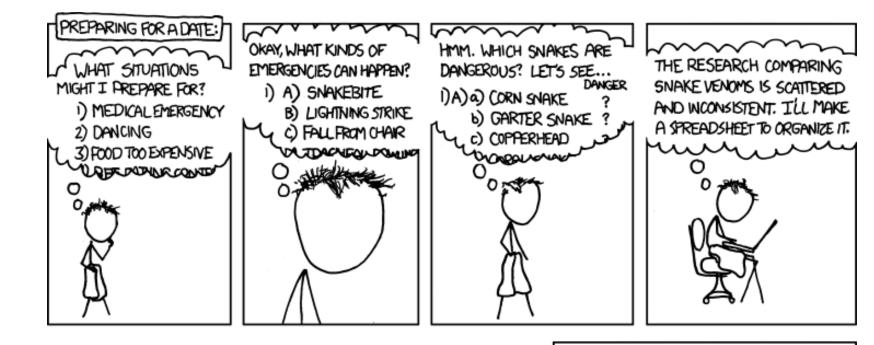
Depth-first search

- Expand deepest unexpanded node
- Implementation: frontier is LIFO (a stack)



Depth-first search







http://xkcd.com/761/

I REALLY NEED TO STOP USING DEPTH-FIRST SEARCHES.

Analysis of search strategies

- Strategies are evaluated along the following criteria:
 - Completeness: does it always find a solution if one exists?
 - Optimality: does it always find a least-cost solution?
 - Time complexity: number of nodes generated
 - Space complexity: maximum number of nodes in memory
- Time and space complexity are measured in terms of
 - b: maximum branching factor of the search tree
 - d: depth of the optimal solution
 - m: maximum length of any path in the state space (may be infinite)
 - |S| : number of distinct states

Properties of depth-first search

Complete? (always finds a solution if one exists?)

Fails in infinite-depth spaces, spaces with loops Modify to avoid repeated states along path → complete in finite spaces

Optimal? (always finds an optimal solution?)

No – returns the first solution it finds

Time? (how long does it take, in terms of b, d, m?)

Could be the time to reach a solution at maximum depth $m: O(b^m)$

Terrible if *m* is much larger than *d*

But VERY FAST if there are LOTS of solutions

Space? (how much storage space, in terms of b, d, m?)

O(bm), i.e., linear space!

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Breadth-first search

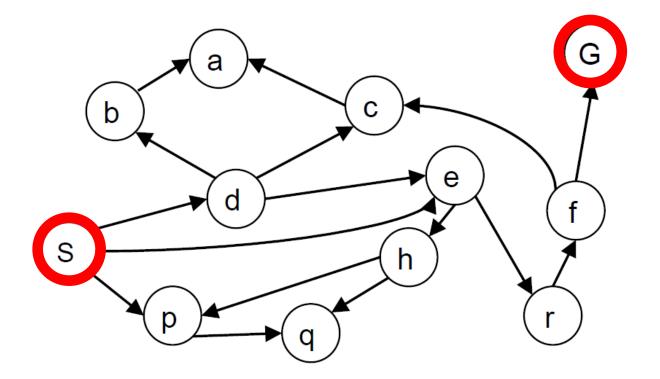
- Initialize the **frontier** using the **starting state**
- While the frontier is not empty
 - Search strategy: choose one of the nodes which is CLOSEST to the starting state
 - If the node contains the **goal state**, return solution
 - Else expand the node and add its children to the frontier

Breadth-first search

Expand shallowest unexpanded node

• Implementation: *frontier* is FIFO (first-in, first out)

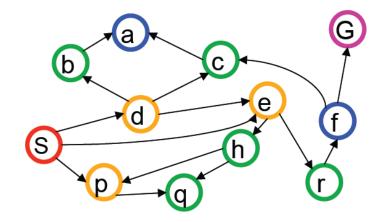
(a queue)

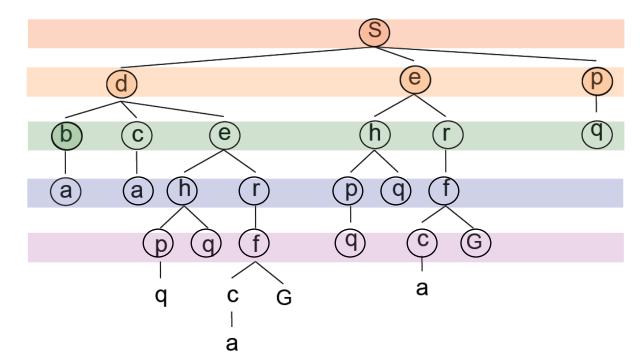


Breadth-first search

Expansion order:

(s, d,e,p, b,c,e,h,r,q, a,a,h,r,p,q,f, p,q,f,q,c,G)





Properties of breadth-first search

Complete?

```
Yes (if branching factor b is finite). Even w/o repeated-state checking, it still works!!!
```

Optimal?

```
Yes – if cost = 1 per step (uniform cost search will fix this)
```

Time?

```
Number of nodes in a b-ary tree of depth d: O(b^d) (d is the depth of the optimal solution)
```

Space?

```
O(b^d). --- much larger than DFS!
```

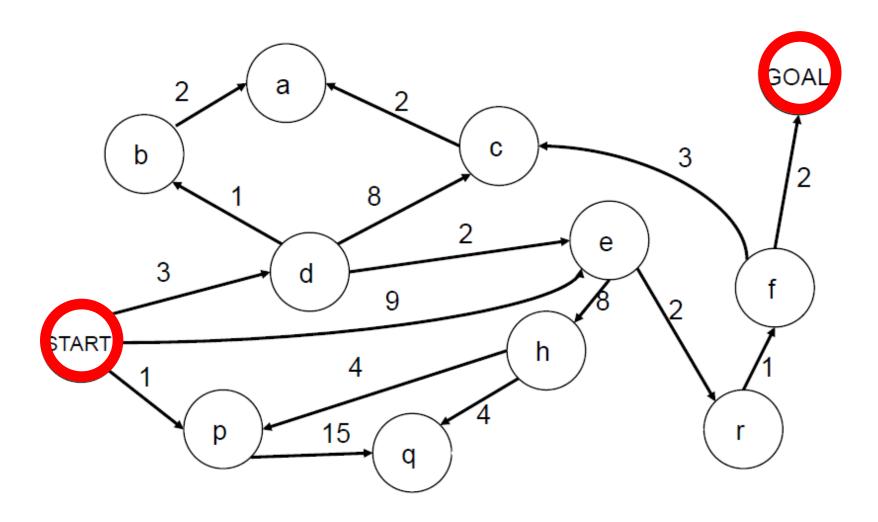
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Uniform-cost search = Dijkstra's algorithm

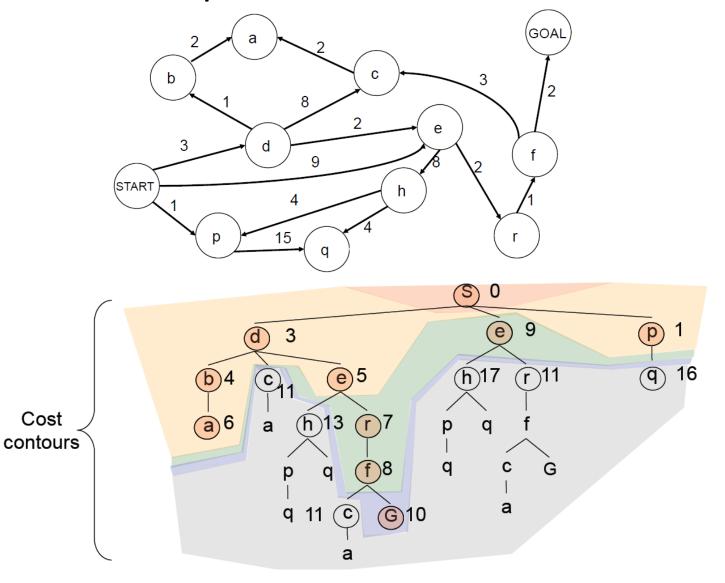
- For each frontier node, save the total cost of the path from the initial state to that node
- Expand the frontier node with the lowest path cost
- Implementation:
 frontier is a priority queue ordered by path cost
- Equivalent to breadth-first if step costs all equal
- Equivalent to Dijkstra's algorithm, if Dijkstra's algorithm is modified so that a node's value is computed only when it becomes nonzero

Uniform-cost search example



Uniform-cost search example

```
Expansion order:
(s,p(1),
d(3),b(4),
e(5),r(7),f(8)
e(9),
G(10))
```



Properties of uniform-cost search

Complete?

```
Yes (if branching factor b is finite). Even w/o repeated-state checking, it still works!!!
```

Optimal?

Yes

Time?

Number of nodes in a b-ary tree of depth d: $O\{b^d\}$ Priority queue is $O\{\log_2 d\}/\text{node}$

Space?

 $O\{b^d\}$ --- much larger than DFS! This might be a reason to use DFS.

Search strategies so far

Algorithm	Complete?	Optimal?	Time complexity	Space complexity	Implement the Frontier as a
BFS	Yes	If all step costs are equal	$O\{b^d\}$	$O\{b^d\}$	Queue
DFS	No	No	$O\{b^m\}$	$O\{bm\}$	Stack
UCS	Yes	Yes	$O\{b^d\log_2 d\}$	$O\{b^d\}$	Priority Queue

Next time

- know how far it is, from the start point, to each node on the frontier.
- What if we also have an ESTIMATE of the distance from each node to the GOAL?