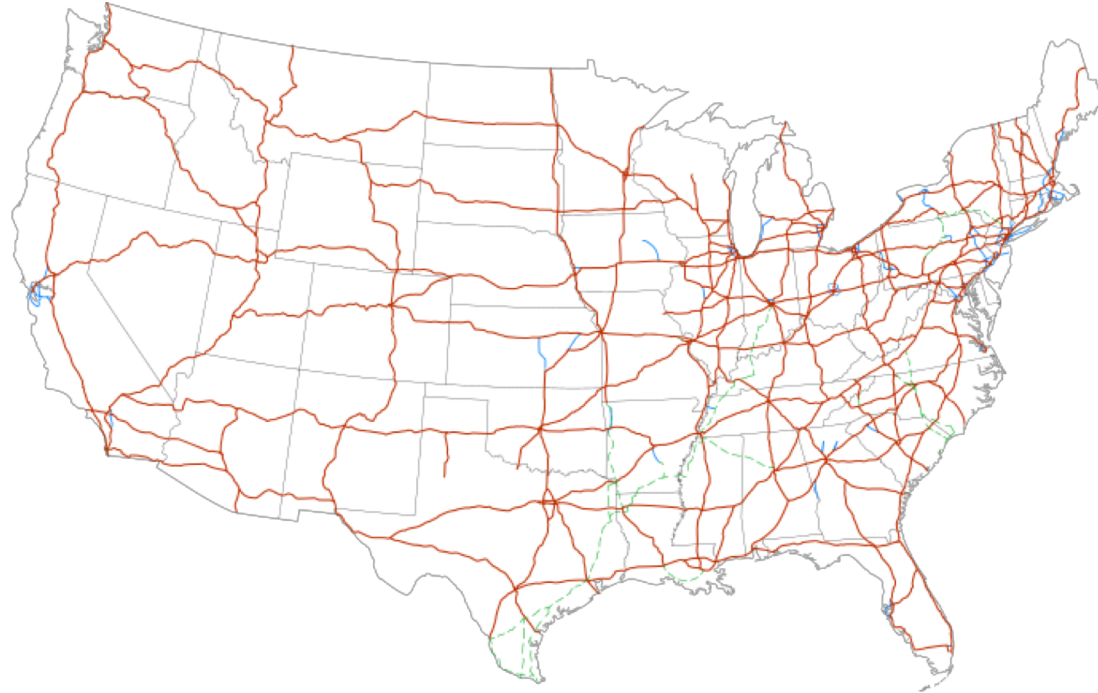


# CS440/ECE 448 Lecture 4: Search Intro

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



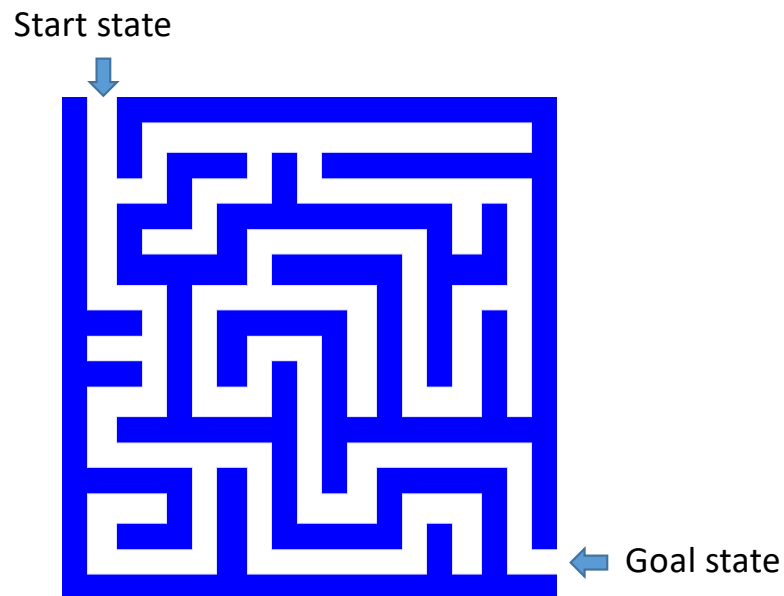
- Consider how the world **WOULD BE**
- Decisions based on (hypothesized) consequences of actions
- Must have a model of how the world evolves in response to actions
- Must formulate a goal

# Outline of today's lecture

1. How to turn ANY problem into a SEARCH problem:
  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



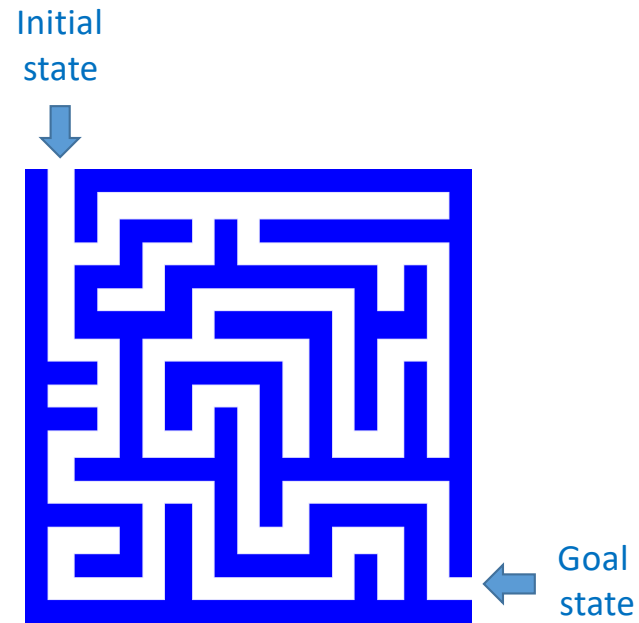
# Search

We will consider the problem of designing **goal-based agents** in **fully observable, deterministic, discrete, 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; while executing the solution, we assume that the agent can safely ignore its percepts (**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 it 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**

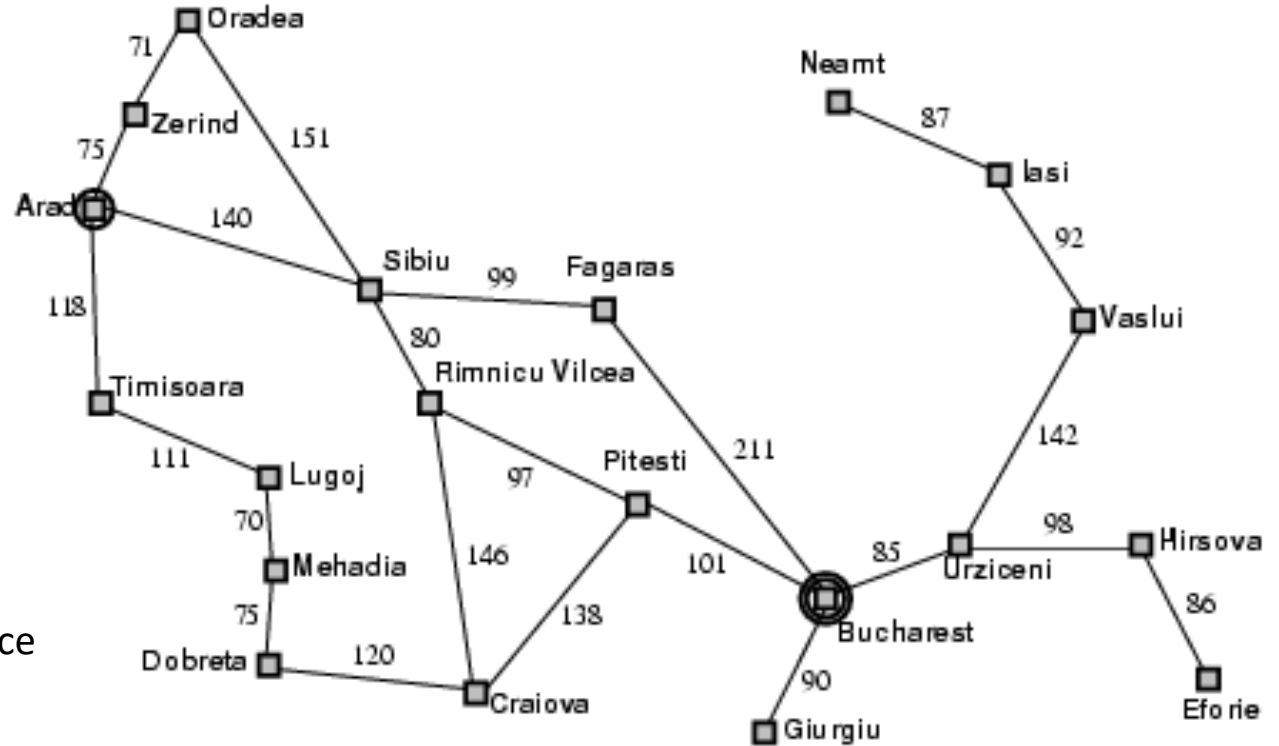
- If you go from city A to city B, you end up in city B

- **Goal state**

- Bucharest

- **Path cost**

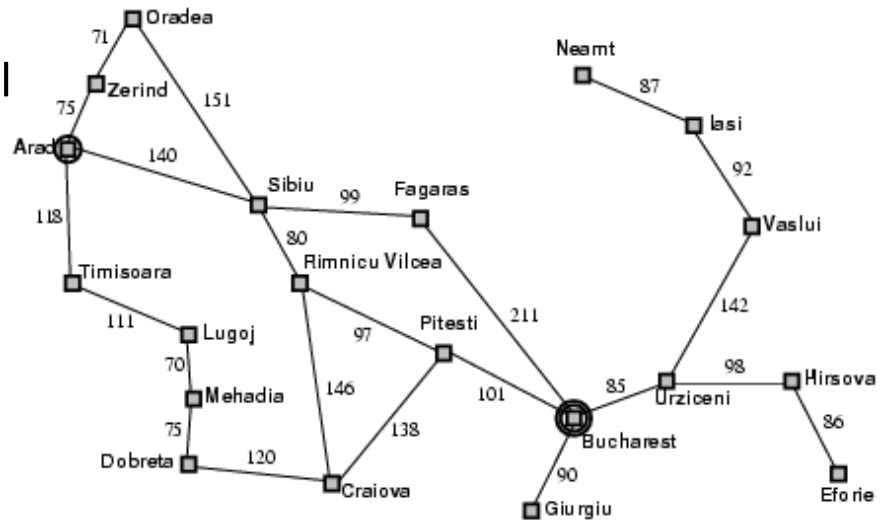
- Sum of edge costs (total distance traveled)





# State space

- The initial state, actions, and transition model define the **state space** of the problem
  - The set of all states reachable from 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 =  $O\{\# \text{ cities}\}$

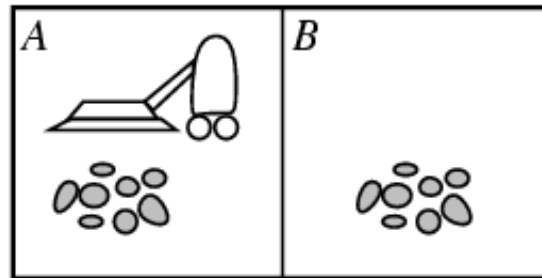


# Traveling Salesman Problem

- Goal: visit every city in the United States
- Path cost: total miles traveled
- Initial state: Champaign
- Action: 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**

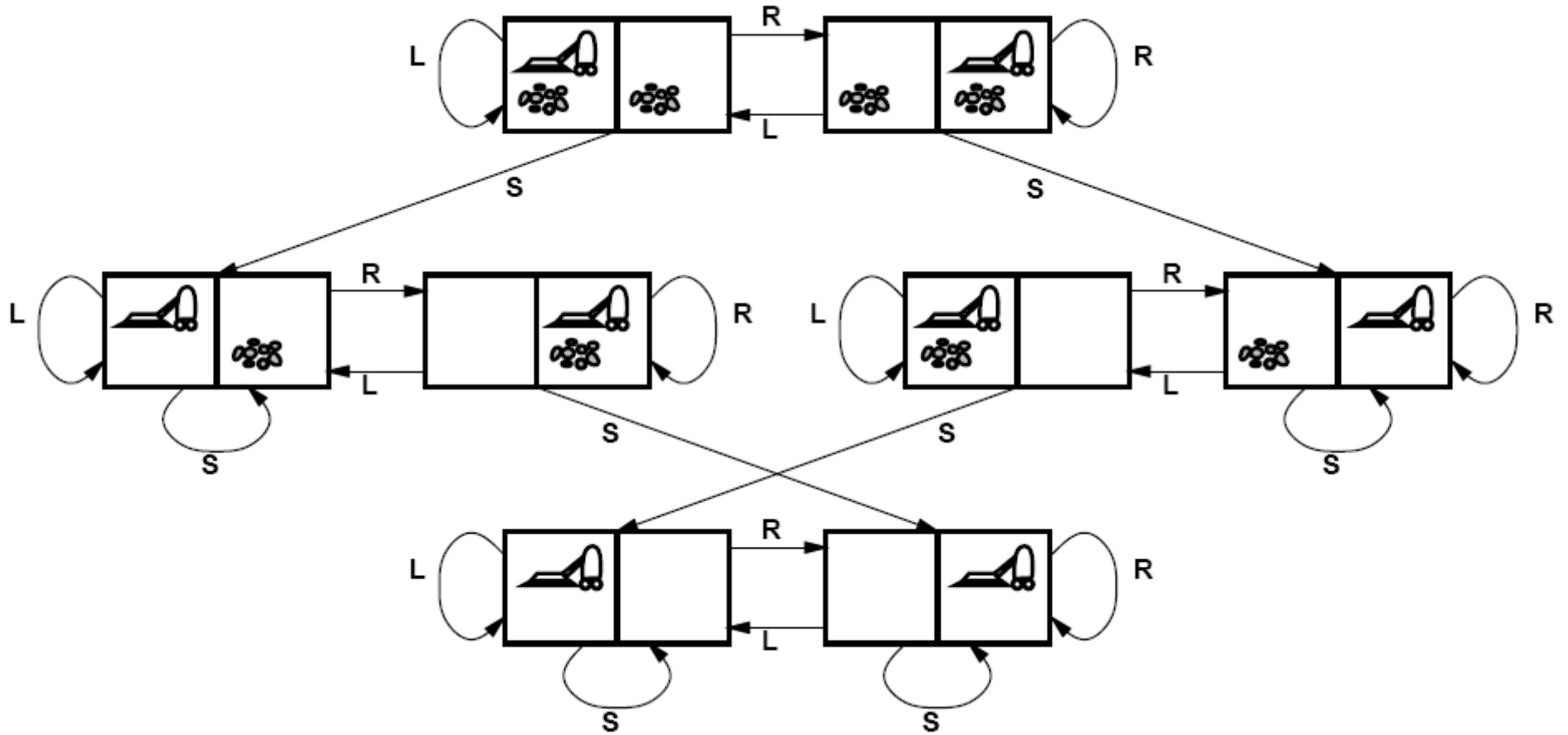
- Agent location and dirt location
- How many possible states?
- What if there are  $n$  possible locations?
  - The size of the state space grows exponentially with the “size” of the world!

- **Actions**

- Left, right, suck

- **Transition model**

# Vacuum world state space graph



# Complexity of the State Space

- Many “video game” style problems can be subdivided:
  - There are M different things your character needs to pick up:  $2^M$  different world states
  - There are N 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:  $\sim 10$  trillion states
  - 24-puzzle:  $\sim 10^{25}$  states

- **Actions**

- Move blank left, right, up, down

- **Path cost**

- 1 per move

- [Finding the optimal solution of n-Puzzle is NP-hard](#)

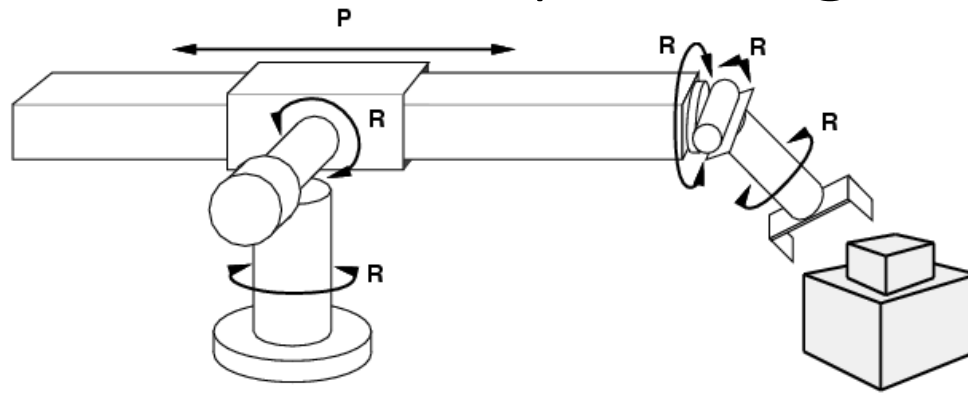
7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

Goal State

# 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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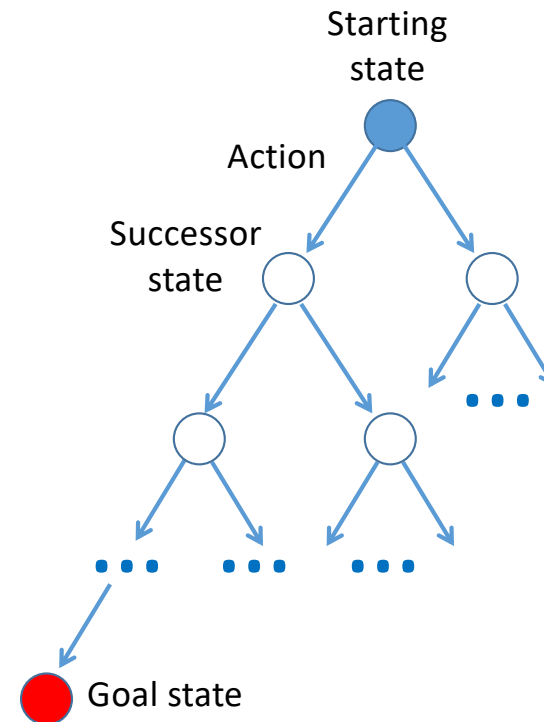


# First data structure: a frontier list

- Let's begin at the start state and **expand** it by making a list of all possible successor states
- Maintain a **frontier** or 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

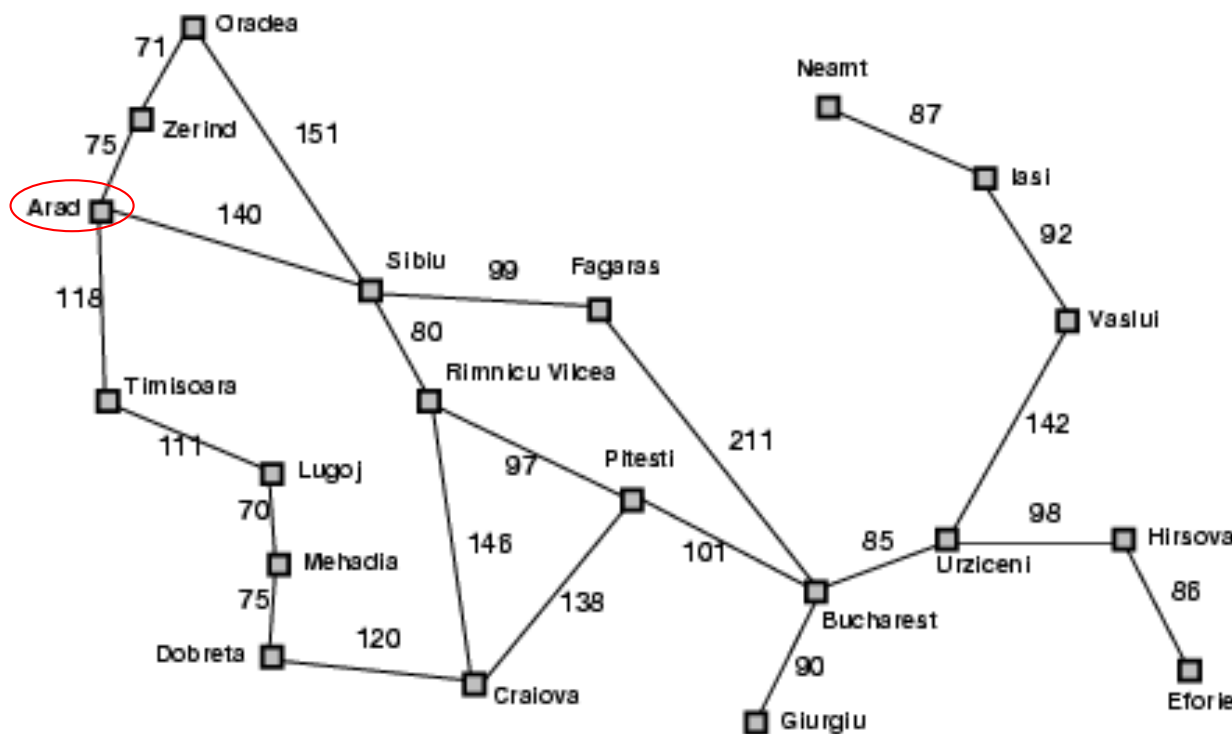
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  - Must have enough detail to decide whether or not you're currently in the initial state
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  - Often but not always: “defining the state” and “defining the transition model” are the same thing
- 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

# 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

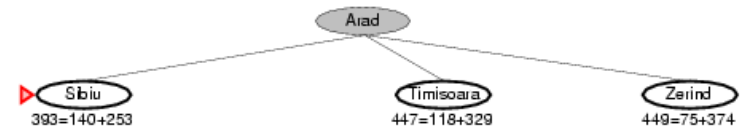


# Tree search example



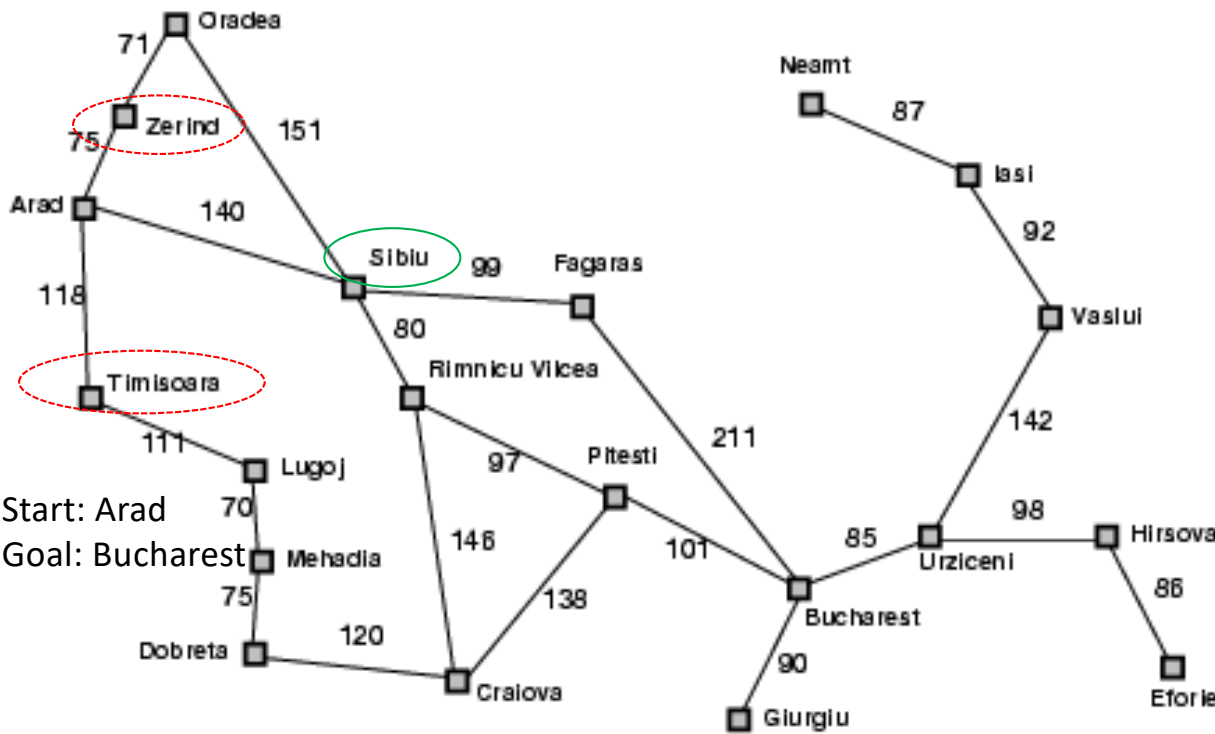
Straight-line distance to Bucharest	
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<b>Zerind</b>	374

# Tree search example



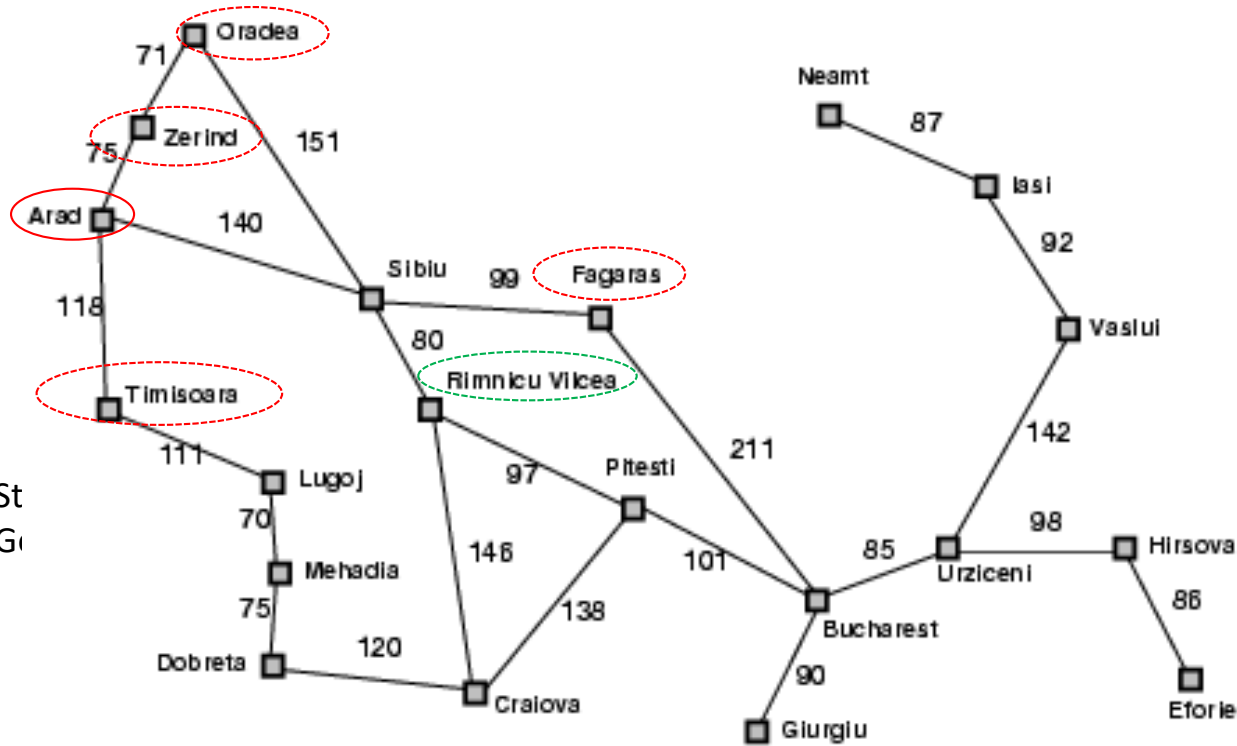
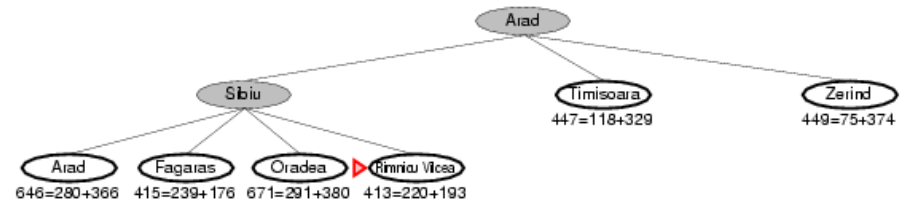
Start: Arad  
Goal: Bucharest

Start: Arad  
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# Tree search example



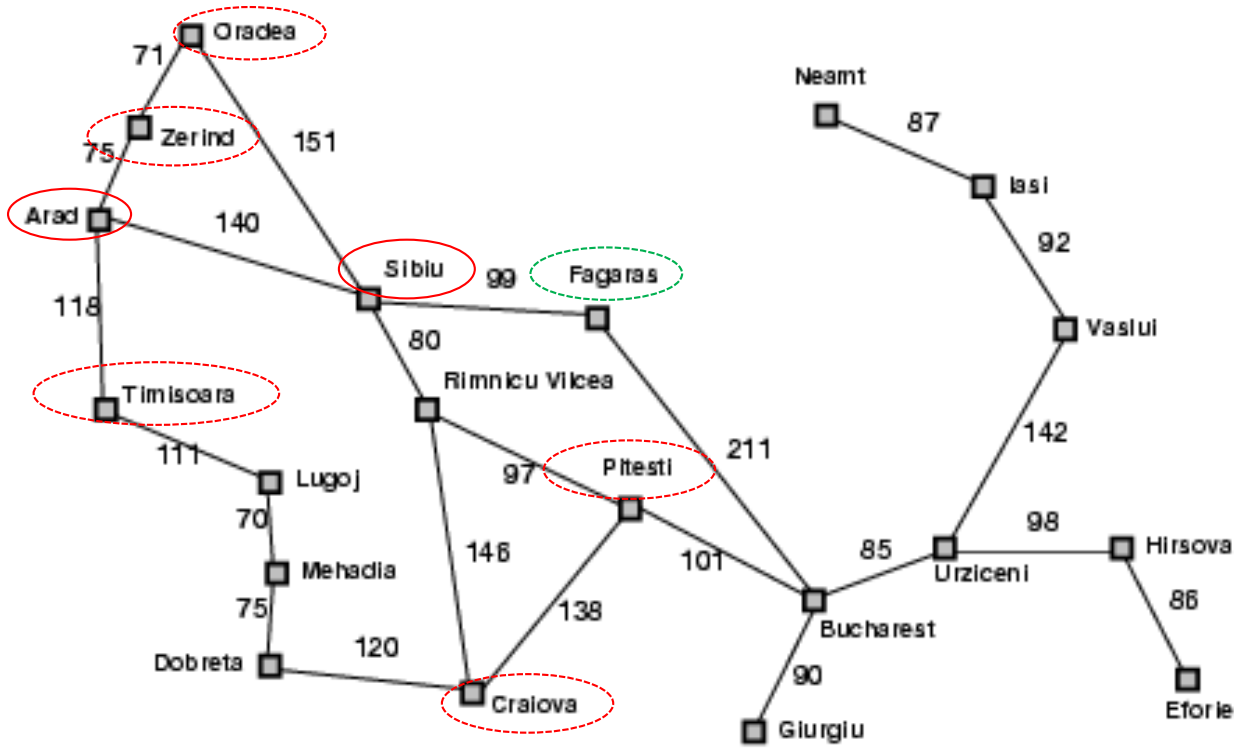
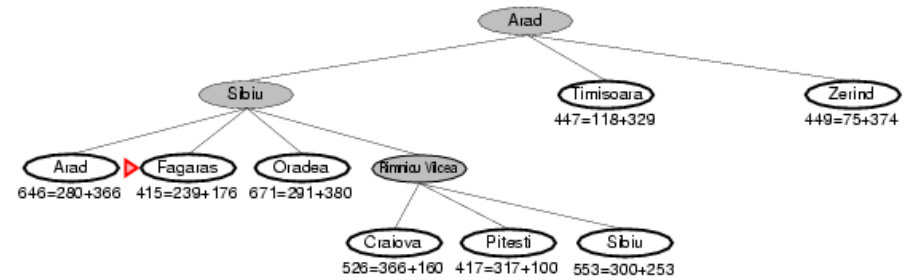
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Start: Arad  
Goal: Bucharest

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# Tree search example



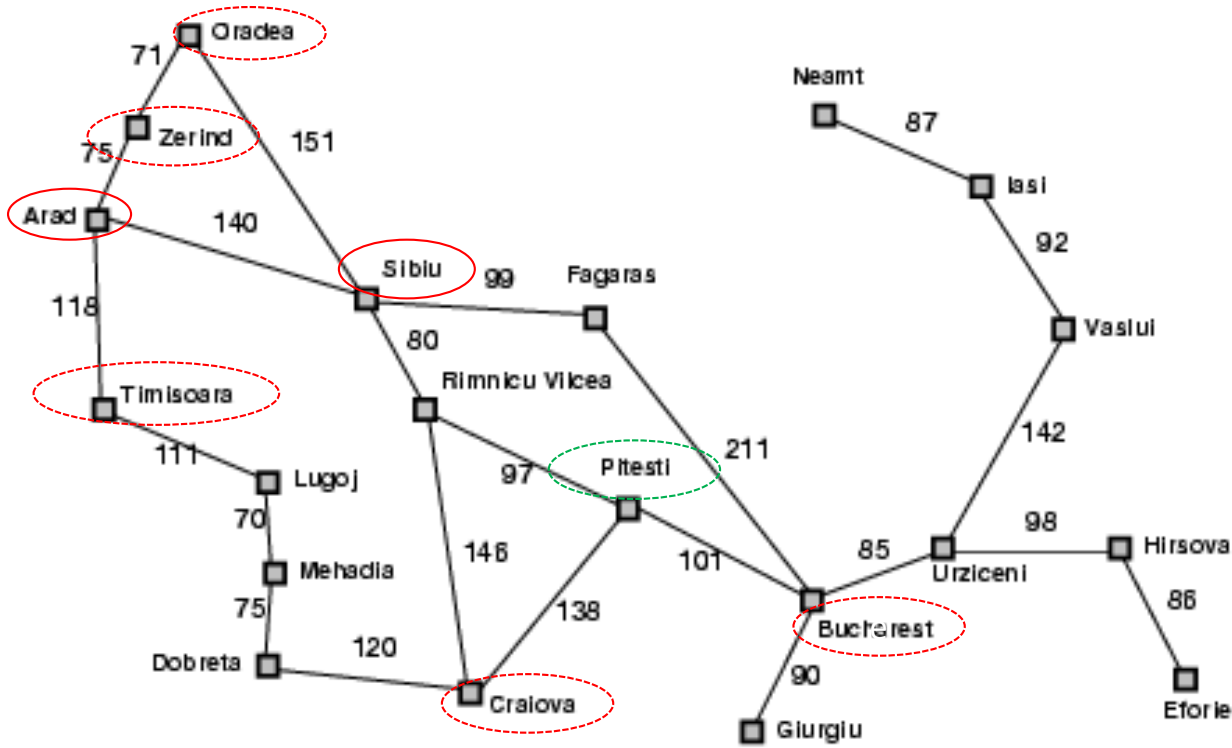
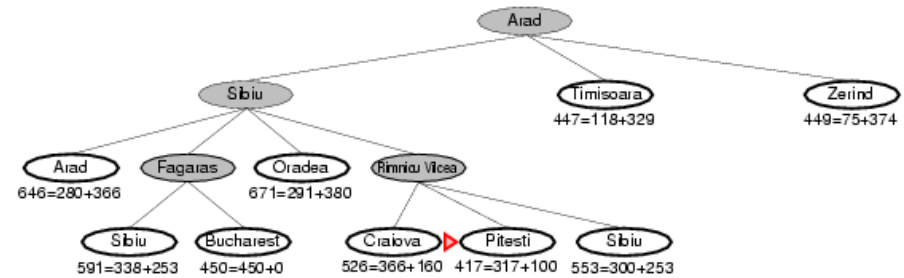
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# Tree search example

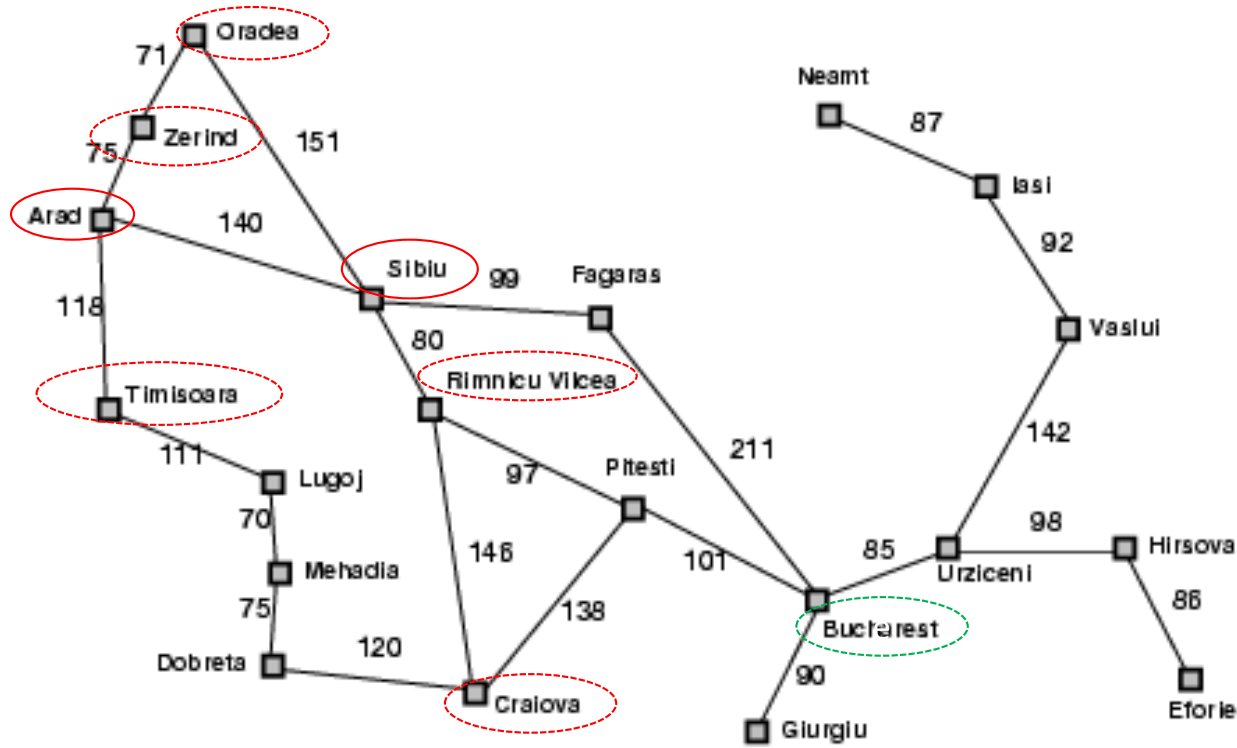
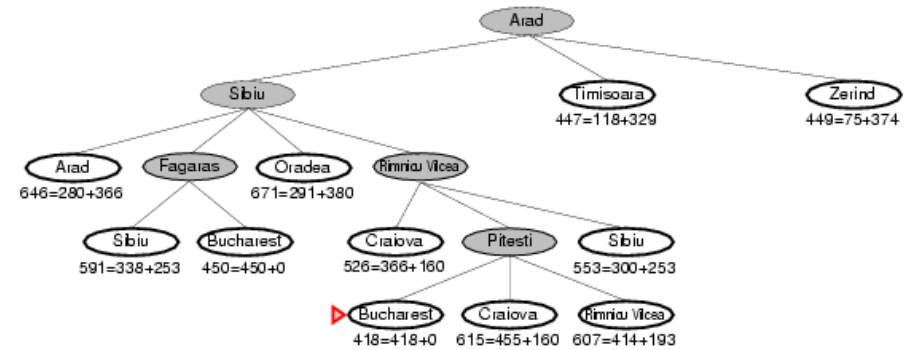


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# Tree search example



Straight-line distance to Bucharest

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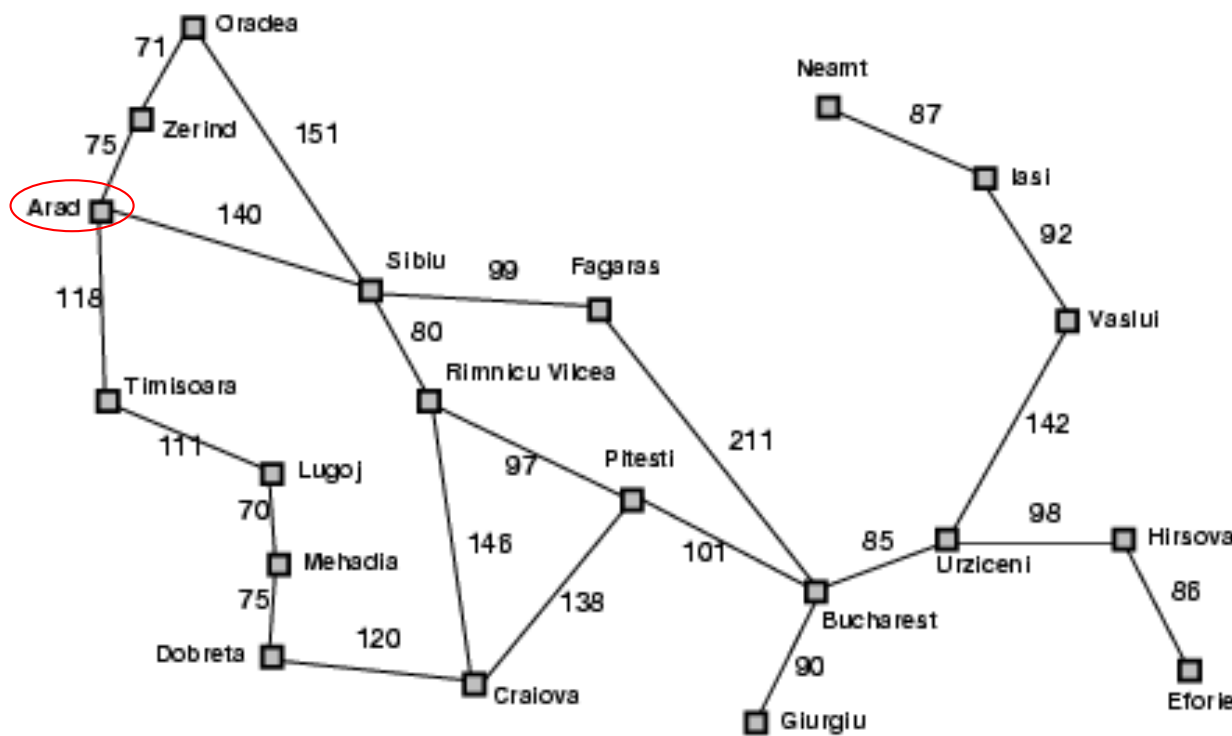
# Handling repeated states

- 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

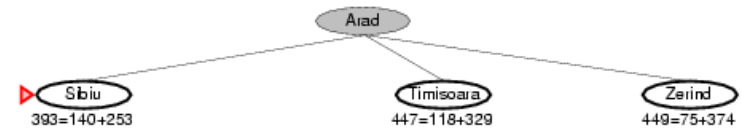
- Without **explored set** :
  - $O\{1\}$ /node
  - $O\{b^m\}$  = # nodes expanded
    - $b$  = branching factor (number of children each node might have)
    - $m$  = length of the longest possible path
- With **explored set** :
  - $O\{1\}$ /node using a hash table to see if node is already in **explored 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

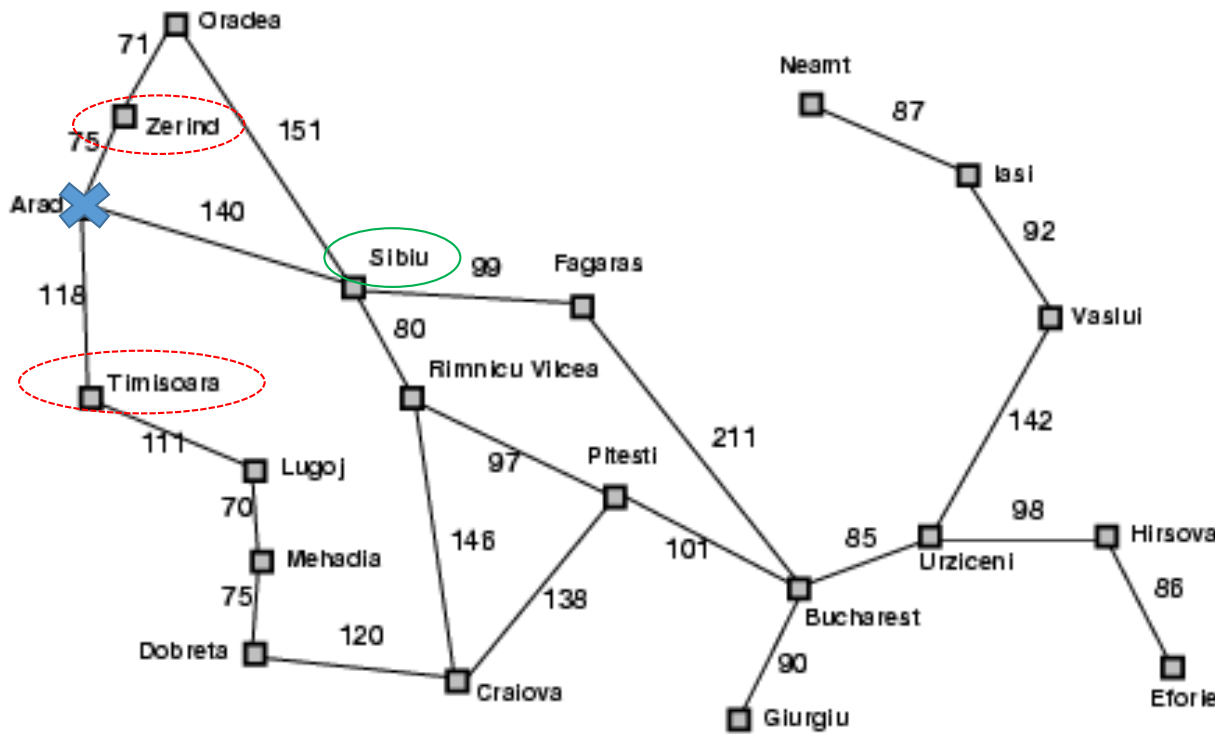


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# Tree search w/o repeats



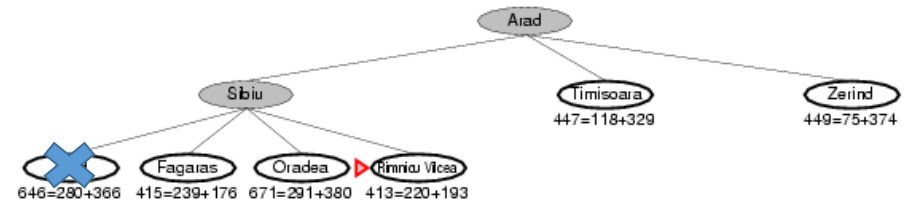
Explored:  
Arad



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Start: Arad  
Goal: Bucharest

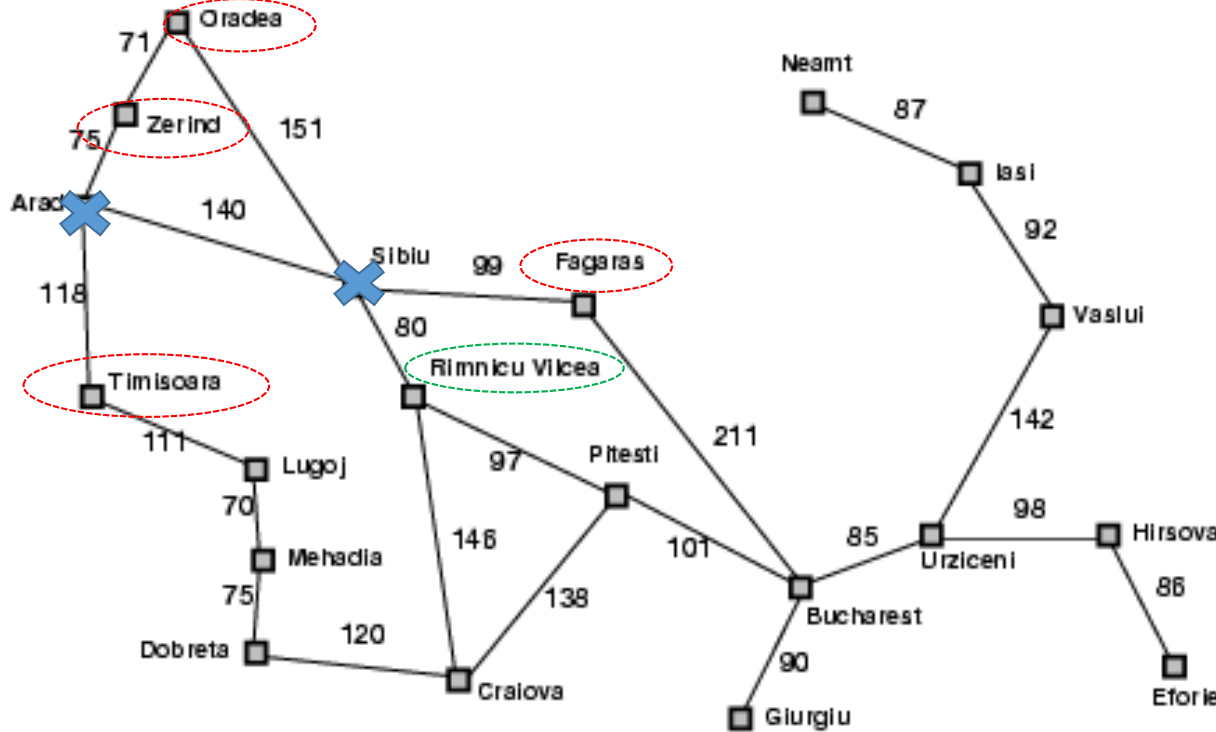
# Tree search example



Explored:

Arad

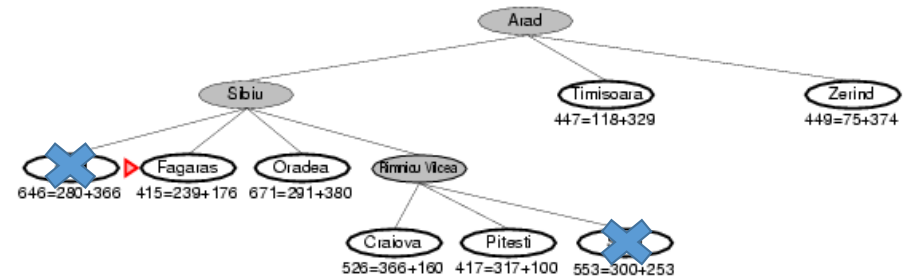
Sibiu



Start: Arad

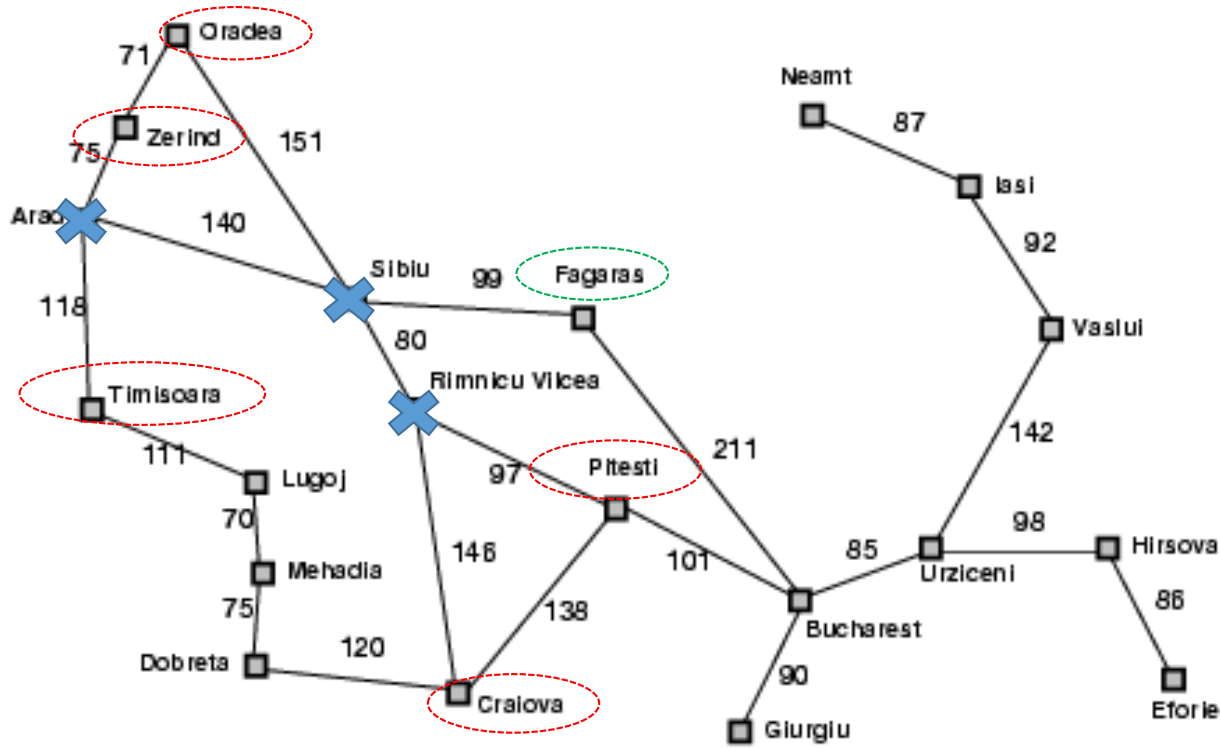
Goal: Bucharest

# Tree search example



Explored:

- Arad
- Sibiu
- Rimnicu Vilcea



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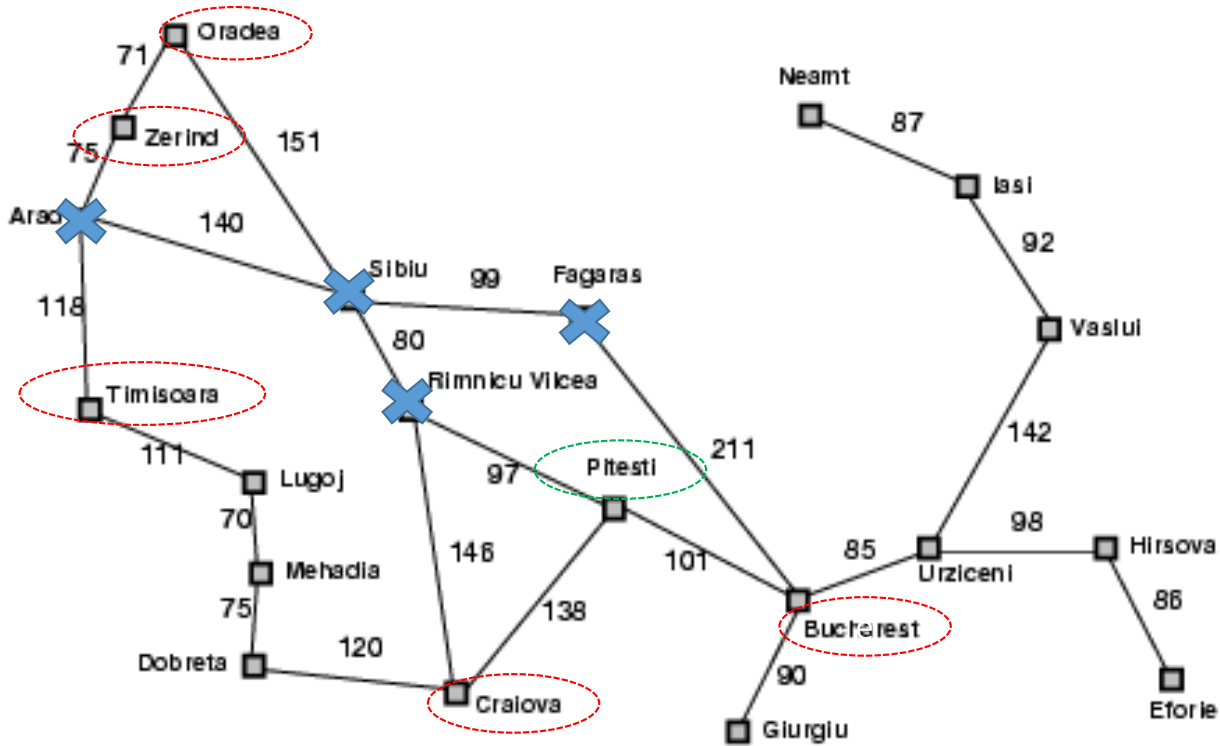


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Explored:

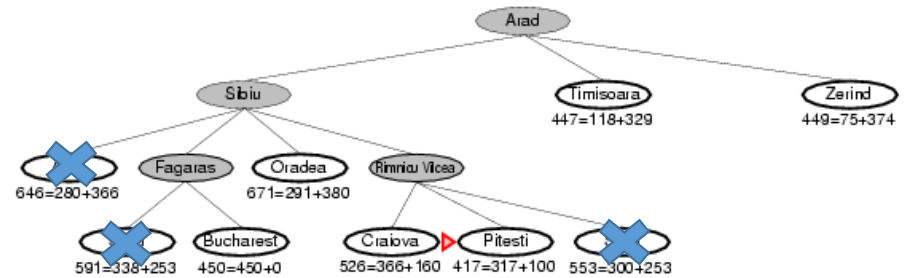
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Sibiu  
Rimnicu Vilces  
Fagaras

Start: Arad  
Goal: Bucharest



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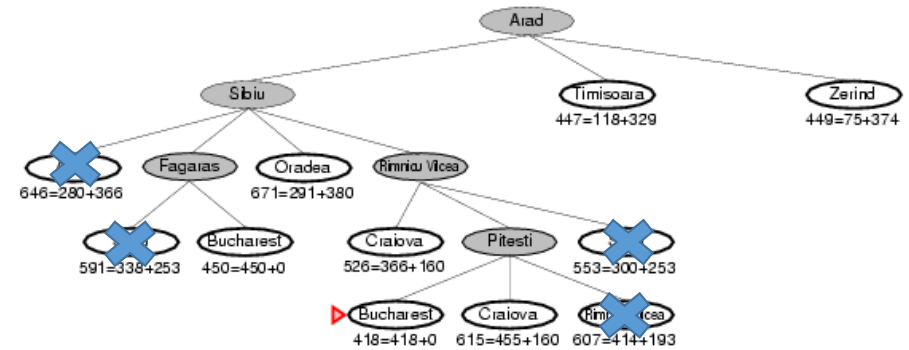
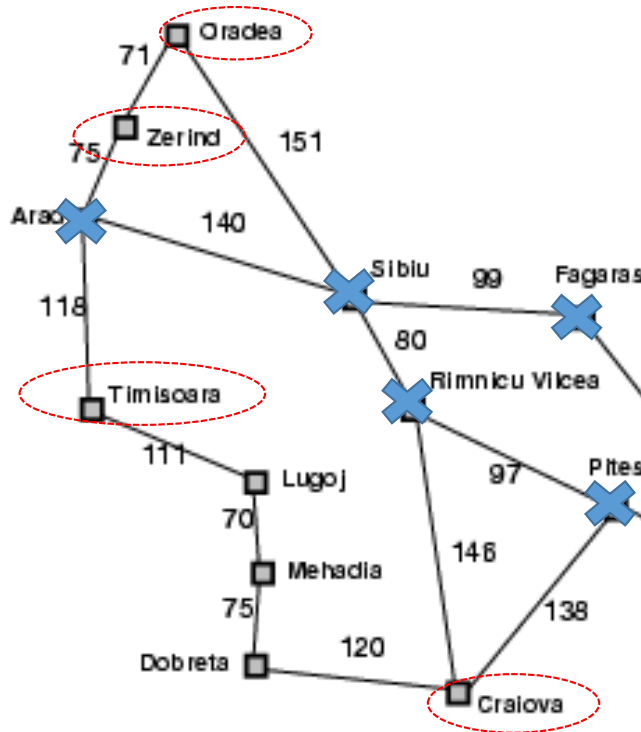


# Tree search example

Explored:

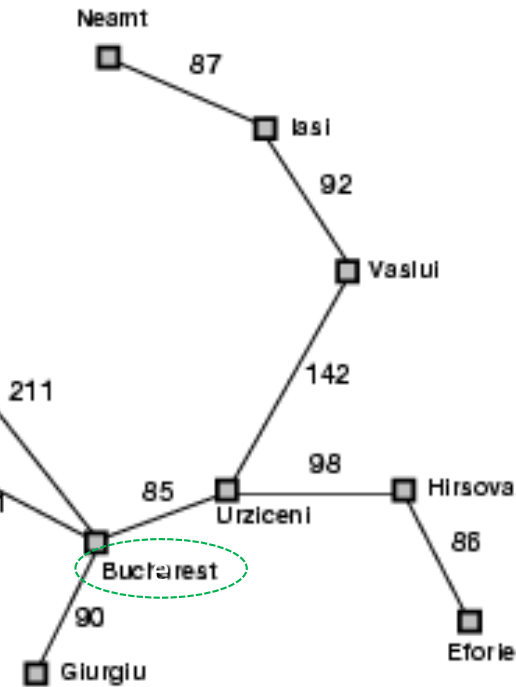
- Arad
- Sibiu
- Rinnicu Vilces
- Fagaras
- Pitesti

Start: Arad  
Goal: Bucharest



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<b>Zerind</b>	374



# Outline of today's lecture

1. How to turn ANY problem into a SEARCH problem:
  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

# 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:

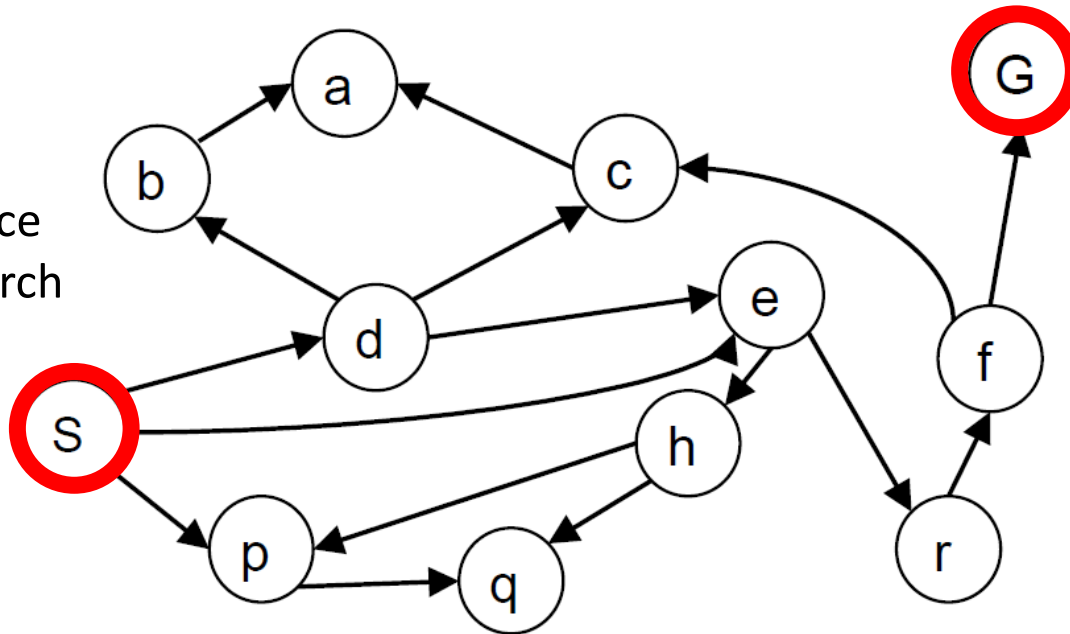
Frontier is a LIFO queue.

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)

Example state space graph for a tiny search problem



# Depth-first search

Expansion order:

(s,d,b,a,

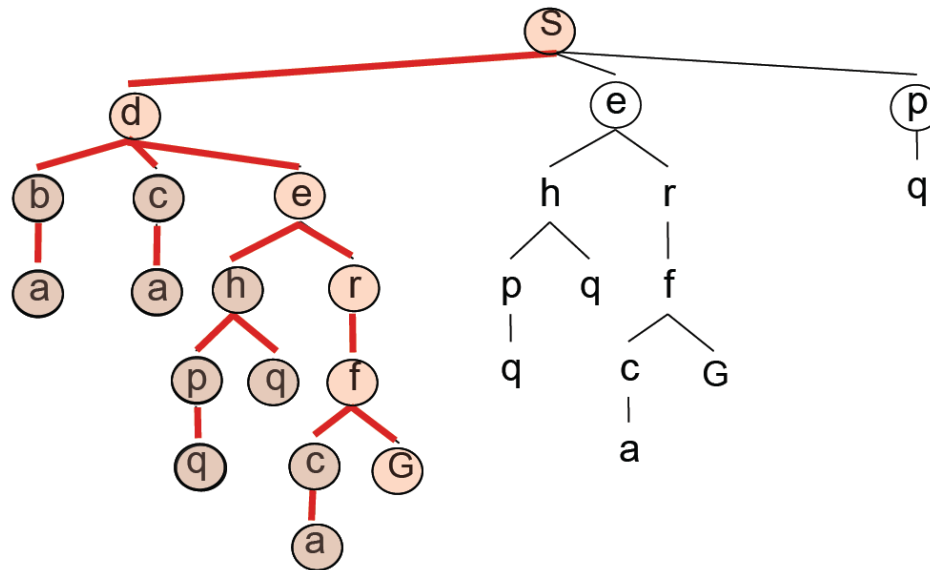
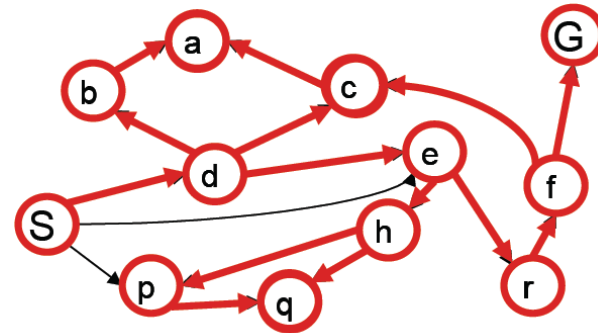
c,a,

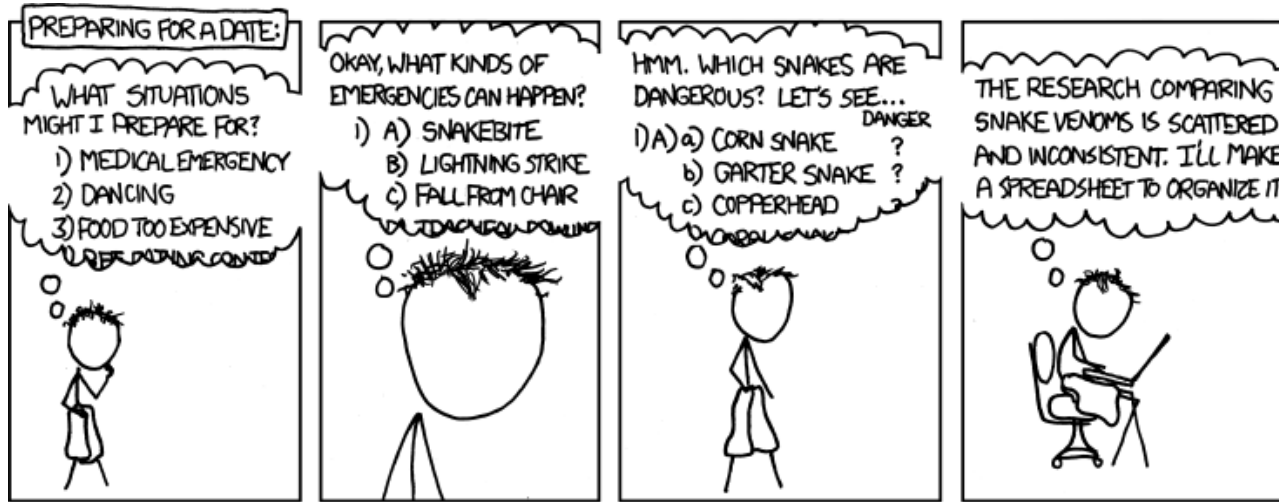
e,h,p,q,

q,

r,f,c,a,

G)





<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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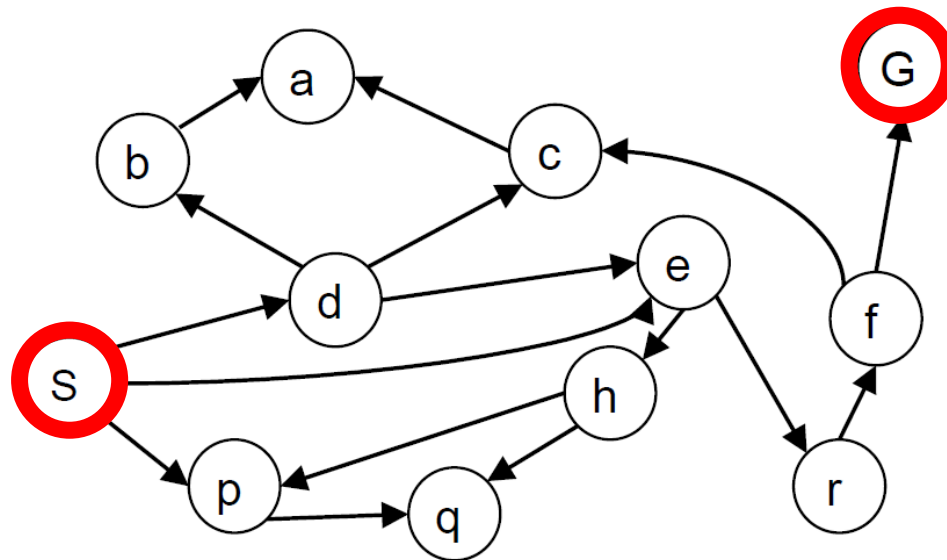
# Breadth-first search

- 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 (a queue)

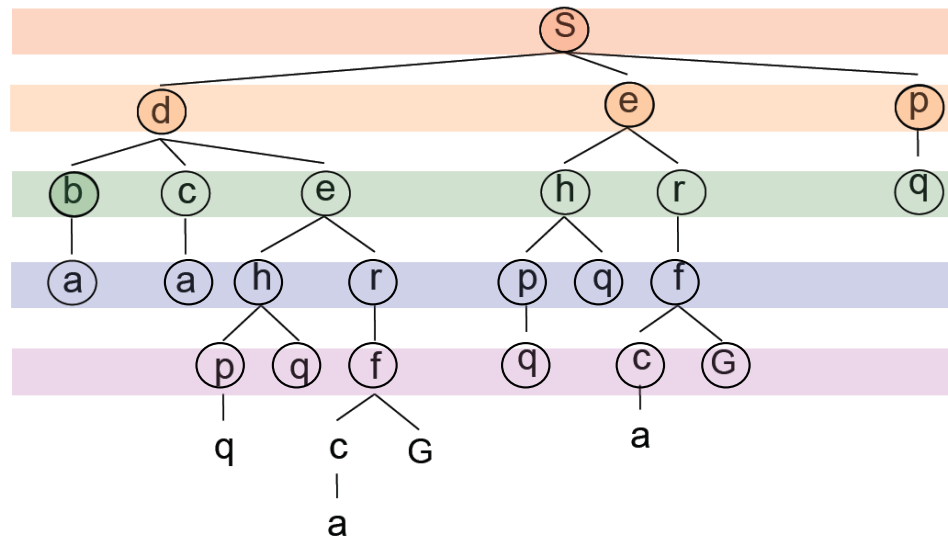
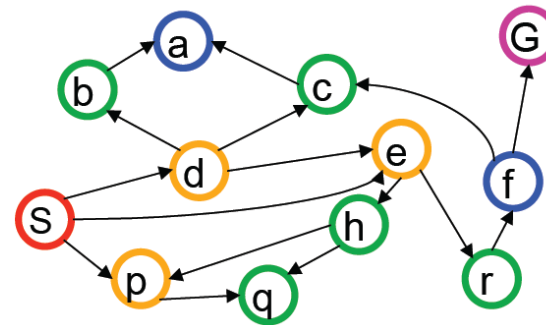


Example from P. Abbeel and D. Klein

# 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!

# Outline of today's lecture

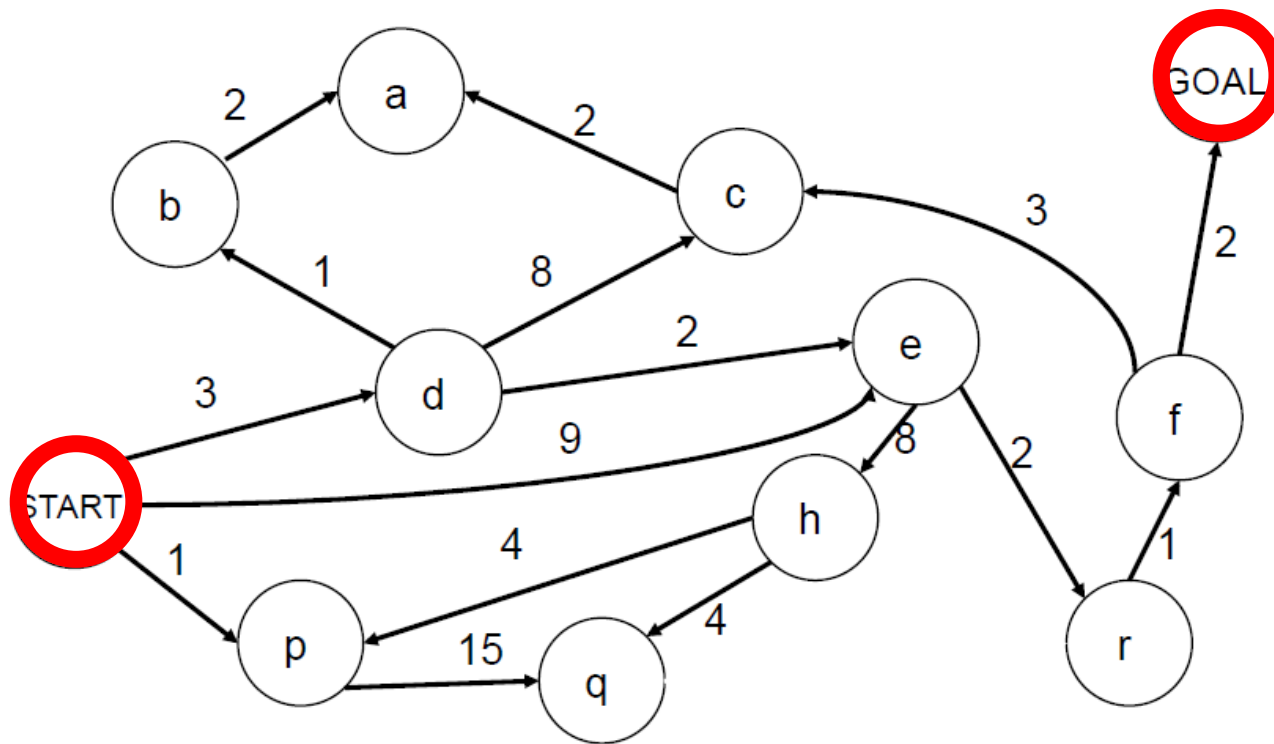
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# Uniform-cost search = Dijkstra's algorithm

- 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:

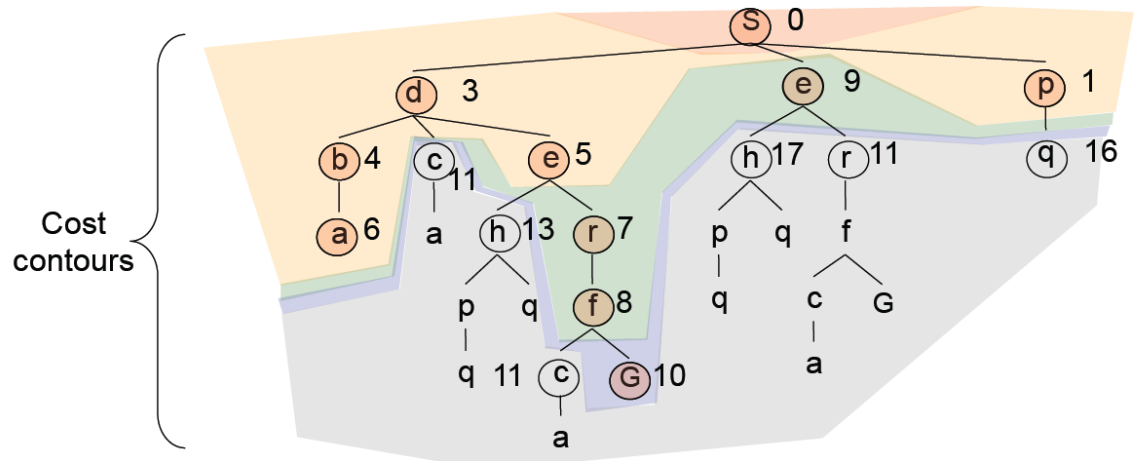
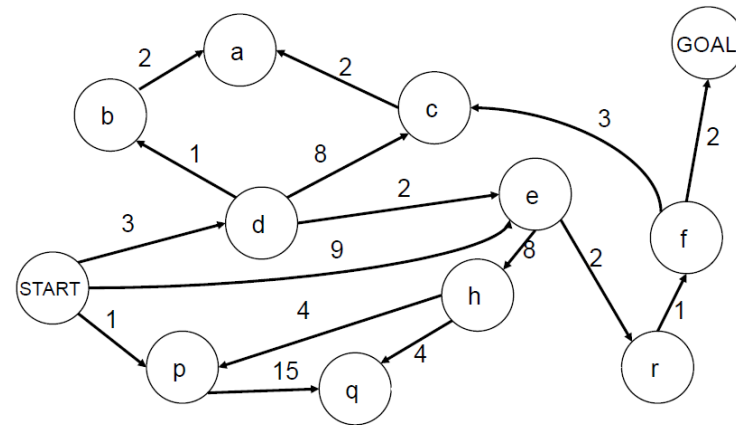
(s,p(1),

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  $O\{\log_2 d\}$ /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

- Already we 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?