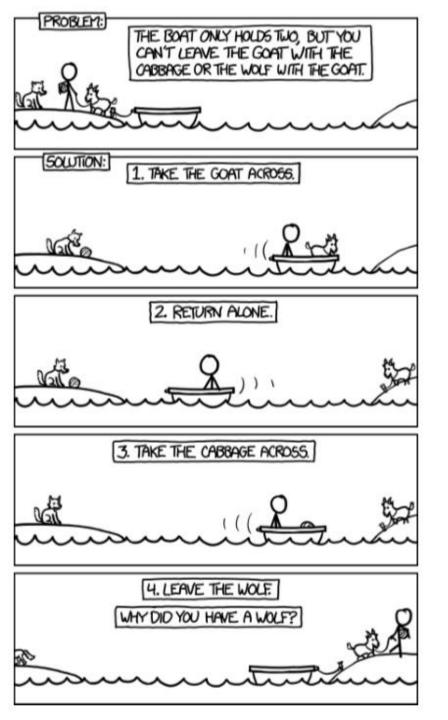
# Introduction to Search

1

Fundamentals of Artificial Intelligence

Slides are mostly adapted from AIMA, Svetlana Lazebnik (UIUC) and Percy Liang (Stanford) A farmer wants to get his cabbage, goat, and wolf across a river. He has a boat that only holds two. He cannot leave the cabbage and goat alone or the goat and wolf alone. How many river crossings does he need When you solve this problem, try to think about how you did it. You probably simulated the scenario in your head, trying to send the farmer over with the goat, observing the consequences. If nothing got eaten, you might continue with the next action. Otherwise, you undo that move and try something else.

How can we get a machine to do this automatically? One of the things we need is a systematic approach that considers all the possibilities. We will see that search problems define the possibilities, and search algorithms explore these possibilities.



Sometimes you can do better if you change the model (perhaps the value of having a wolf is zero) instead of focusing on the algorithm.

	Search problems		
	Markov decision processes	Constraint satisfaction problems	
	Adversarial games	Bayesian networks	
Reflex	States	Variables	Logic
"Low-level intelligence"		17	High-level intelligence"

#### **Machine learning**

## Application: route finding



Objective: shortest? fastest? most scenic?

Actions: go straight, turn left, turn right

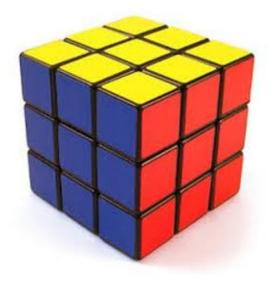
#### Application: robot motion planning

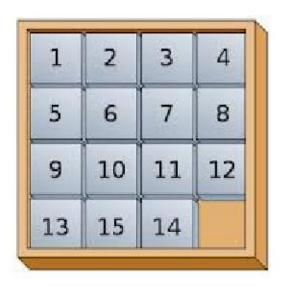


Objective: fastest? most energy efficient? safest?

Actions: translate and rotate joints

## Application: solving puzzles





Objective: reach a certain configuration

Actions: move pieces (e.g., Move12Down)

### Types of agents

#### **Reflex agent**



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

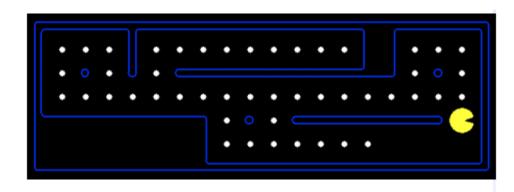
#### **Planning agent**

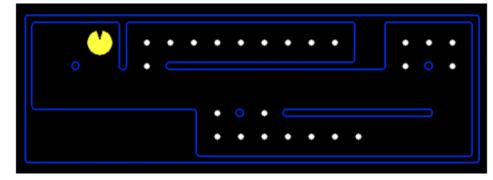


- Consider how the world WOULD BE
- Decisions based on (hypothesized) consequences of actions
- Must have a model of how the world evolves in response to actions
- Must formulate a goal

# **Reflex Agents**

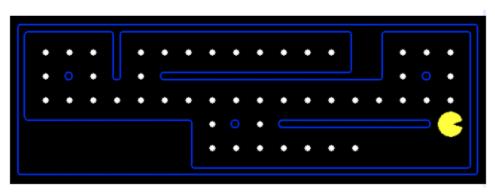
- Reflex Agents:
  - Choose action based on current percept (and maybe memory)
  - May have memory or a model of the world's current state
  - Do not consider the future consequences of their actions
  - Act on how the world IS
- Can a reflex agent be rational?

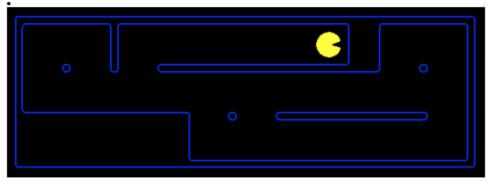




# **Goal Based Agents**

- Goal-based agents:
  - Plan ahead
  - Decisions based on (hypothesized) consequences of actions
  - Must have a model of how the world evolves in response to actions
  - Act on how the world
    WOULD BE



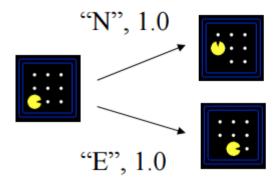


# Search Problems

- A search problem consists of:
  - A state space



A successor function

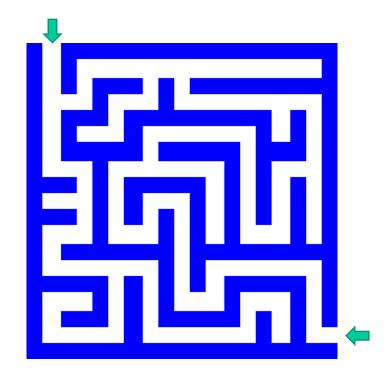


A start state and a goal test

- A solution is a sequence of actions (a plan) which transforms the start state to a goal state
  - The performance measure is defined by (a) reaching the goal and (b) how "expensive" the path to the goal is

#### Search

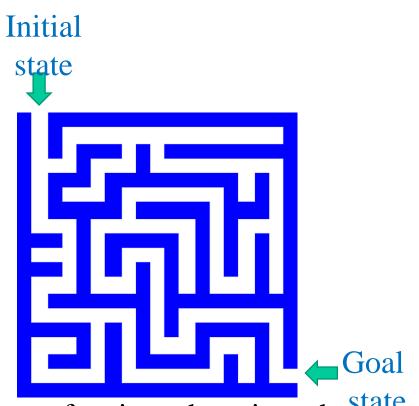
• We will consider the problem of designing goal-based agents in fully observable, deterministic, discrete, known environments



- 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
  - We are focused on the process of finding the solution; while executing the solution, we assume that the agent can safely ignore its percepts (**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*

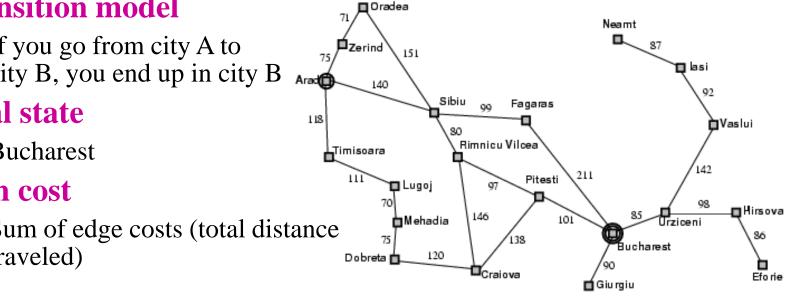


• The **optimal solution** is the sequence of actions that gives the state *lowest* path cost for reaching the goal

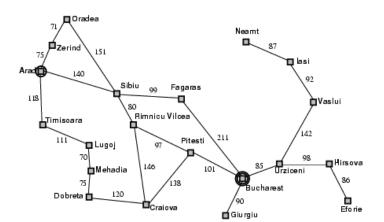
### **Example:** Romania

- 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)

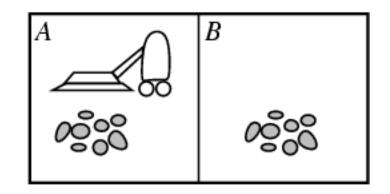




- The initial state, actions, and transition model define the **state space** of the problem
  - The set of all states reachable from initial state by any sequence of actions
  - Can be represented as a directed graph where the nodes are states and links between nodes are actions
- What is the state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

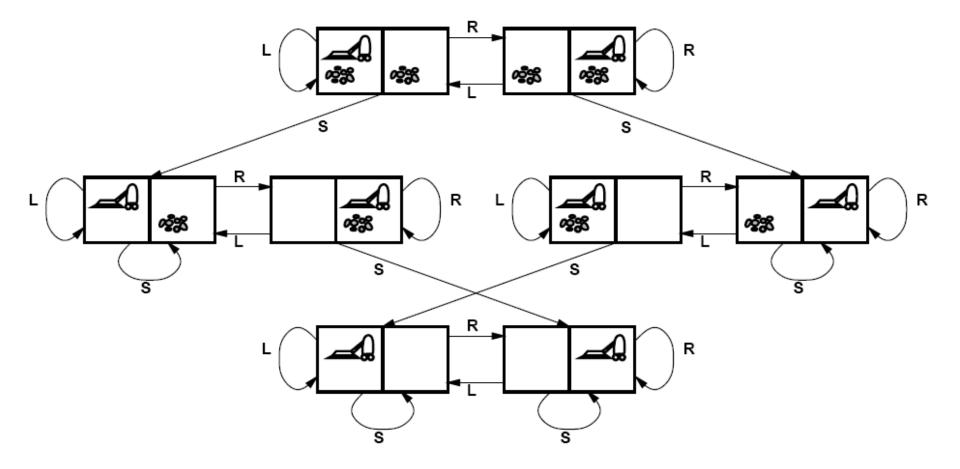


### Example: Vacuum world



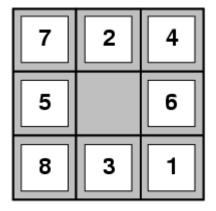
- States
  - Agent location and dirt location
  - How many possible states?
  - What if there are *n* possible locations?
    - The size of the state space grows exponentially with the "size" of the world!
- Actions
  - Left, right, suck
- Transition model

## Vacuum world state space graph

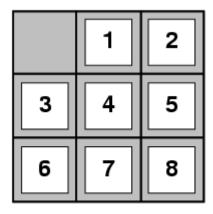


### Example: The 8-puzzle

- States
  - Locations of tiles
    - 8-puzzle: 181,440 states (9!/2)
    - 15-puzzle: ~10 trillion states
    - 24-puzzle:  $\sim 10^{25}$  states
- Actions
  - Move blank left, right, up, down
- Path cost
  - 1 per move



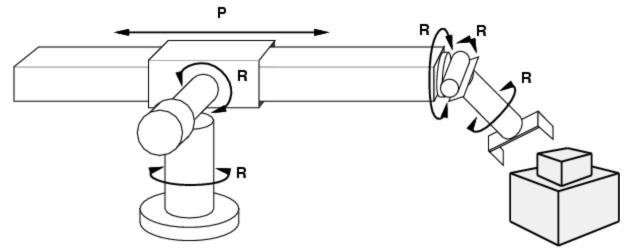
Start State



Goal State

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

#### Example: Robot motion planning

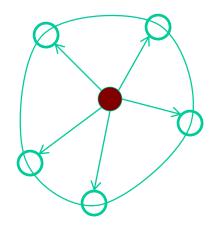


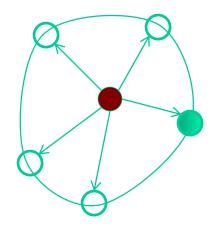
- States
  - Real-valued joint parameters (angles, displacements)
- Actions
  - Continuous motions of robot joints
- Goal state
  - Configuration in which object is grasped
- Path cost
  - Time to execute, smoothness of path, etc.

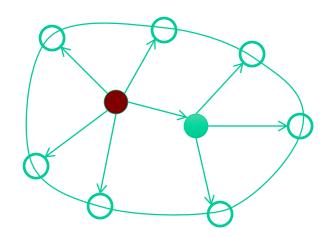
- Given:
  - Initial state
  - Actions
  - Transition model
  - Goal state
  - Path cost
- How do we find the optimal solution?
  - How about building the state space and then using Dijkstra's shortest path algorithm?
    - Complexity of Dijkstra's is  $O(E + V \log V)$ , where V is the size of the state space
    - The state space may be huge!

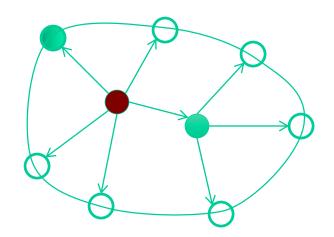
- Let's begin at the start state and **expand** it by making a list of all possible successor states
- Maintain a **frontier** or a list of unexpanded states
- At each step, pick a state from the frontier to expand
- Keep going until you reach a goal state
- Try to expand as few states as possible

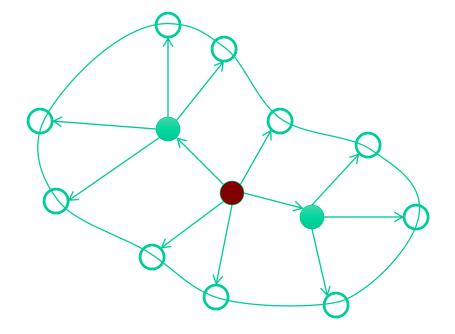


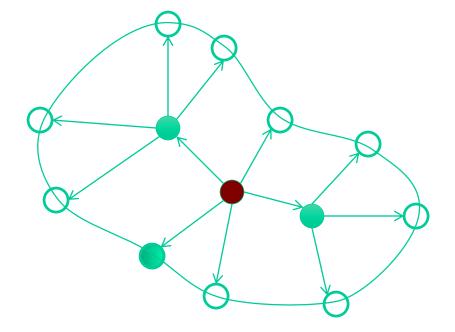


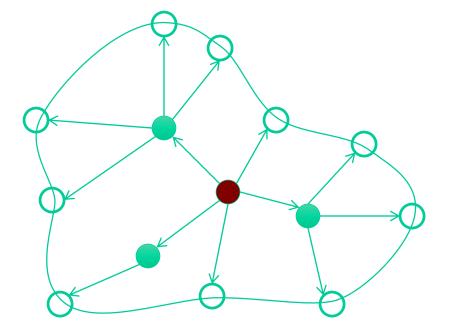


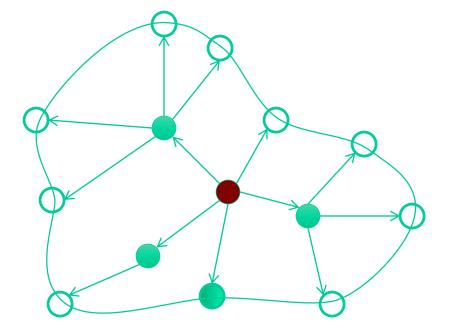


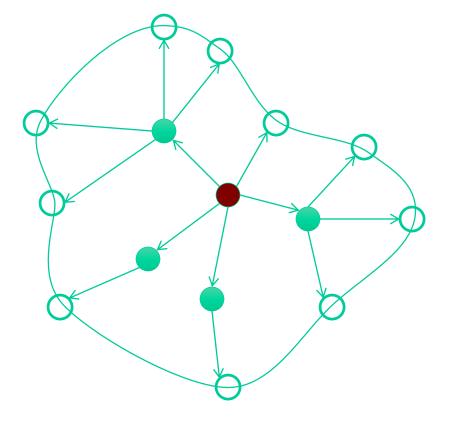


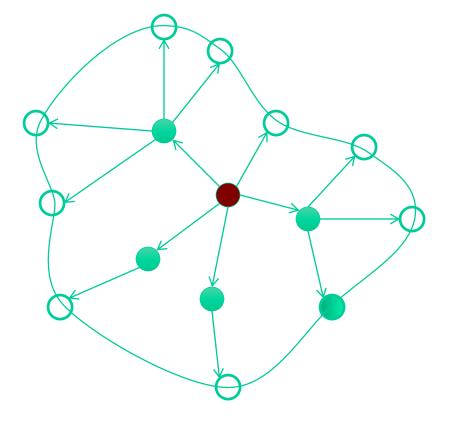


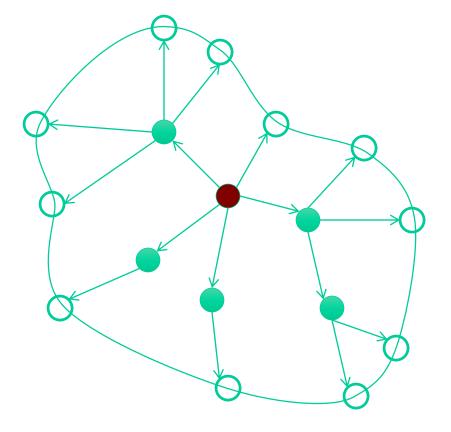




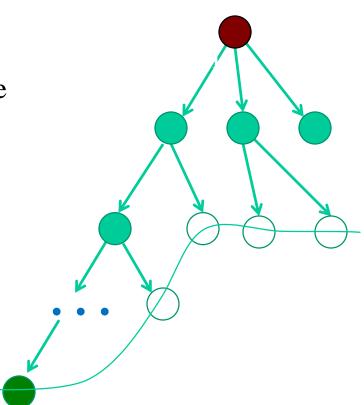




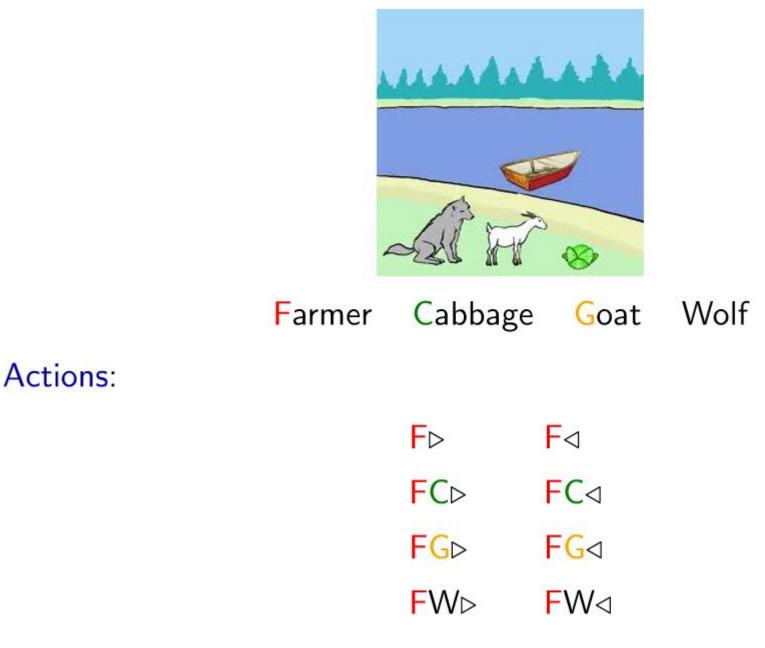




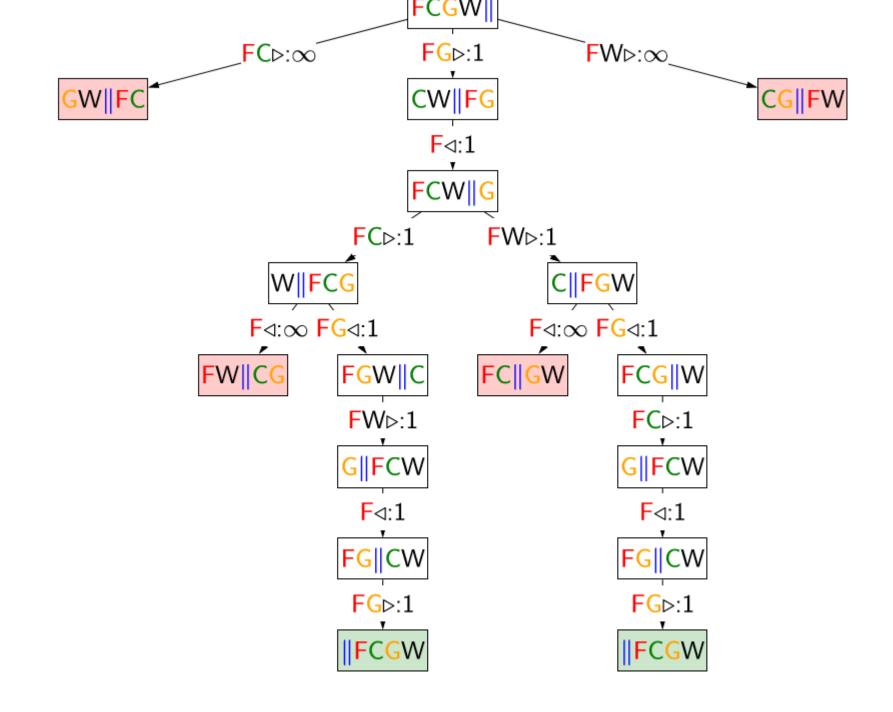
- "What if" tree of sequences of actions and outcomes
  - When we are searching, we are not acting in the world, merely "thinking" about the possibilities
- The root node corresponds to the starting state
- The children of a node correspond to the **successor states** of that node's state
- A path through the tree corresponds to a sequence of actions
  - A solution is a path ending in the goal state
- Nodes vs. states
  - A state is a representation of the world, while a node is a data structure that is part of the search tree
    - Node has to keep pointer to parent, path cost, possibly other info



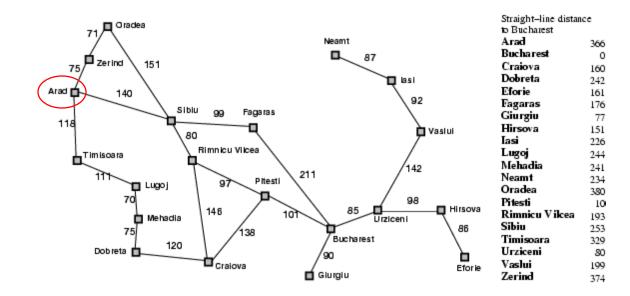
- Initialize the **frontier** using the **starting state**
- While the frontier is not empty
  - Choose a frontier node according to search strategy and take it off the frontier
  - If the node contains the **goal state**, return solution
  - Else **expand** the node and add its children to the frontier



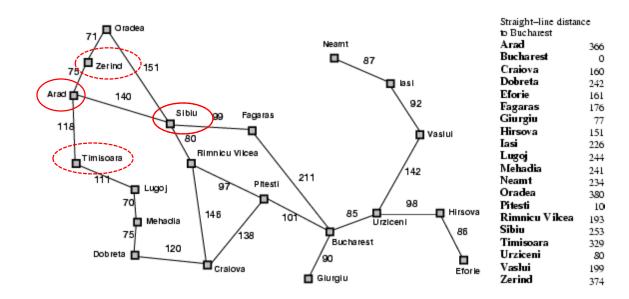
Approach: build a search tree ("what if?")

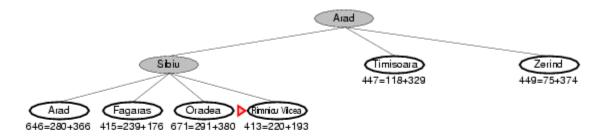


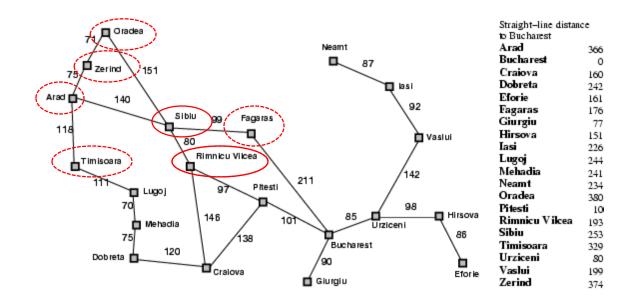


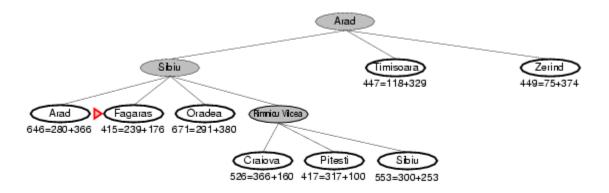


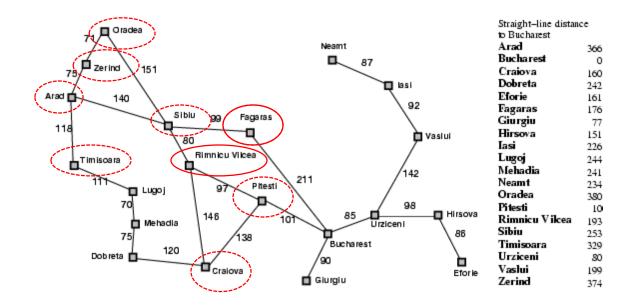


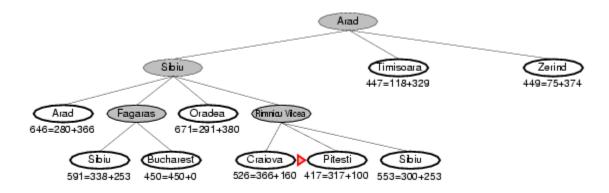


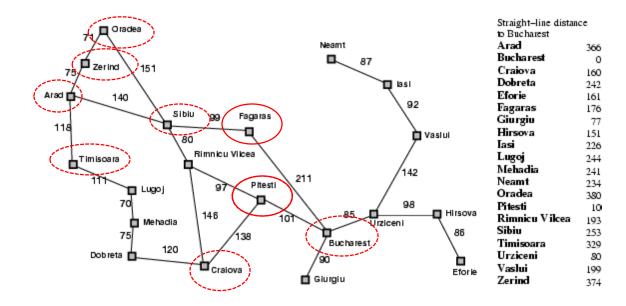


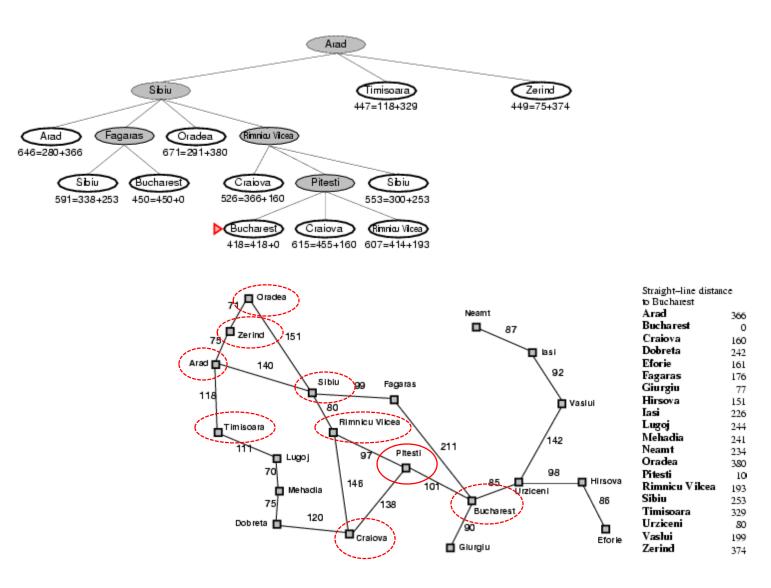


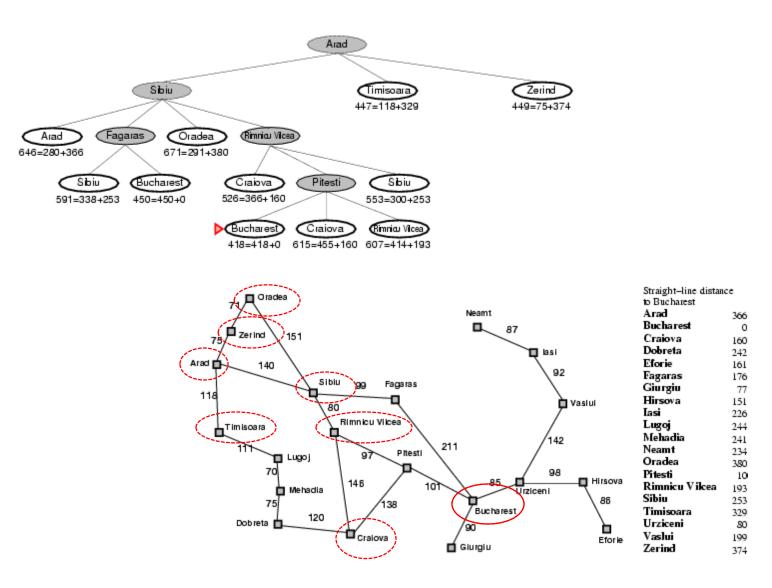












- Initialize the **frontier** using the **starting state**
- While the frontier is not empty
  - Choose a frontier node according to search strategy and take it off the frontier
  - If the node contains the **goal state**, return solution
  - Else **expand** the node and add its children to the frontier
- To handle repeated states:
  - Every time you expand a node, add that state to the explored set; do not put explored states on the frontier again
  - Every time you add a node to the frontier, check whether it already exists in the frontier with a higher path cost, and if yes, replace that node with the new one



