

# Artificial Intelligence 3e, Lecture 3.1

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Problem-solving agents

Problem Formulation

Example Problems

Directed Graphs

# Learning Objectives

At the end of the class you should be able to:

- define a directed graph
- represent a problem as a state-space graph

# Searching

- Often we are not given an algorithm to solve a problem, but only a specification of what is a solution — we have to search for a solution.
- A typical problem is when the agent is in one state, it has a set of deterministic actions it can carry out, and wants to get to a goal state.
- Many AI problems can be abstracted into the problem of finding a path in a directed graph.
- Often there is more than one way to represent a problem as a graph.
- Searching in this context means finding a path to a goal node in a graph.

# Example: Google Maps-I

The best route could mean

- the shortest (least distance) route
- the quickest route
- the route that uses the least energy
- the lowest-cost route, where the cost takes into account time, money (e.g., fuel and tolls), and the route's attractiveness.

# What is a state?

A **state** needs to include enough information to

- determine what is the next state
- determine whether the goal is achieved
- determine the cost.

Often there are many options for what to include in the state.  
Keep the states as simple as possible but no simpler.

# State-space Problem

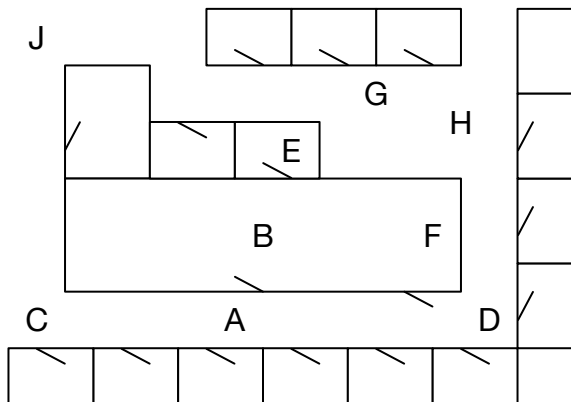
A **state-space problem** consists of

- a set of states
- a subset of states called the **start states**
- a set of actions
- an **action function**: given a state and an action, returns a new state
- a set of goal states, specified as function,  $goal(s)$
- a criterion that specifies the quality of an acceptable solution.

A solution to a search problem is a sequence of actions that will get the agent from its current state to a state that satisfies the goal.

# Example Problem for Delivery Robot

The robot is at A and the goal is to get to G:





# Problem-Solving Agents

Restricted form of general agent:

**function** Simple-Problem-Solving-Agent(*percept*) **returns** an action

**static:** *seq*, an action sequence, initially empty  
*state*, some description of the current world state  
*goal*, a goal, initially null  
*problem*, a problem formulation

*state*  $\leftarrow$  Update-State(*state*, *percept*)

**if** *seq* is empty **then**

*goal*  $\leftarrow$  Formulate-Goal(*state*)

*problem*  $\leftarrow$  Formulate-Problem(*state*, *goal*)

*seq*  $\leftarrow$  Search(*problem*)

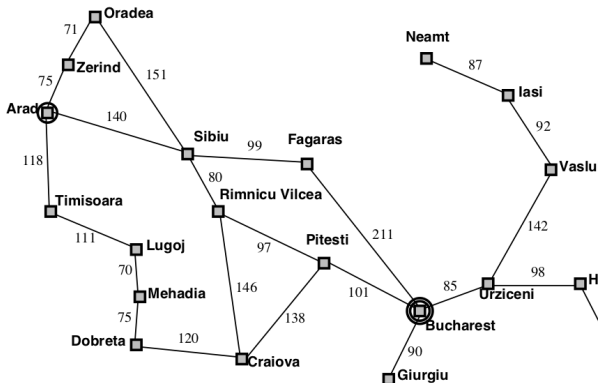
*action*  $\leftarrow$  Recommendation(*seq*, *state*)

*seq*  $\leftarrow$  Remainder(*seq*, *state*)

**return** *action*

# Example: Romania

- **current state:**  
In Arad
- **Formulate goal:**  
be in Bucharest
- **Formulate problem:**  
states: various cities  
actions: drive between cities
- **Find solution:**  
sequence of cities,  
e.g., Arad, Sibiu,  
Fagaras, Bucharest



# Single-state problem formulation

A **problem** is defined by five components:

**initial state** e.g., “at Arad”

**actions** possible actions available to the agent  
e.g.,  $\{Go(Sibiru), Go(Timisoara), Go(Zerind)\}$

**successor function**  $S(x)$  = set of action–state pairs  
e.g.,  $S(Arad) = \{\langle Arad \rightarrow Zerind, Zerind \rangle, \dots\}$

**goal test**, can be  
**explicit**, e.g.,  $x = \text{“at Bucharest”}$   
**implicit**, e.g.,  $NoDirt(x)$

**path cost** (additive)  
e.g., sum of distances, number of actions executed, etc.  
 $c(x, a, y)$  is the **step cost**, assumed to be  $\geq 0$

A **solution** is a sequence of actions  
leading from the initial state to a goal state

# Selecting a state space

Real world is absurdly complex

⇒ state space must be **abstracted** for problem solving

(Abstract) state = set of real states

(Abstract) action = complex combination of real actions

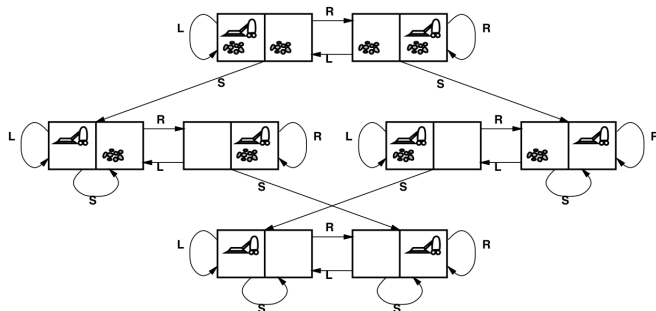
e.g., “Arad → Zerind” represents a complex set  
of possible routes, detours, rest stops, etc.

(Abstract) solution =

set of real paths that are solutions in the real world

Each abstract action should be “easier” than the original problem!

# Example: vacuum world state space graph



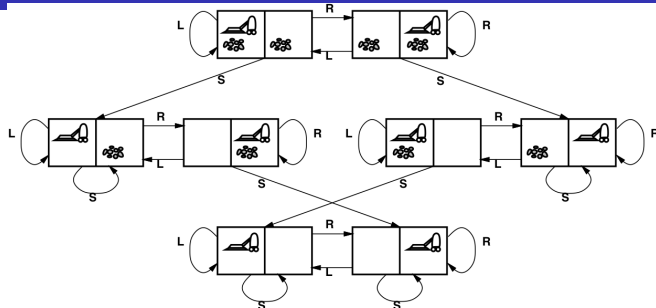
states??

actions??

goal test??

path cost??

# Example: vacuum world state space graph



states??: integer dirt and robot locations (ignore dirt amounts etc.)

actions??: *Left, Right, Suck, NoOp*

goal test??: no dirt

path cost??: 1 per action (0 for *NoOp*)

## Example: The 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

states??

actions??

goal test??

path cost??

# Example: The 8-puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

states??: integer locations of tiles (ignore intermediate positions)

actions??: move blank left, right, up, down (ignore unjamming etc.)

goal test??: : = goal state (given)

path cost??: 1 per move



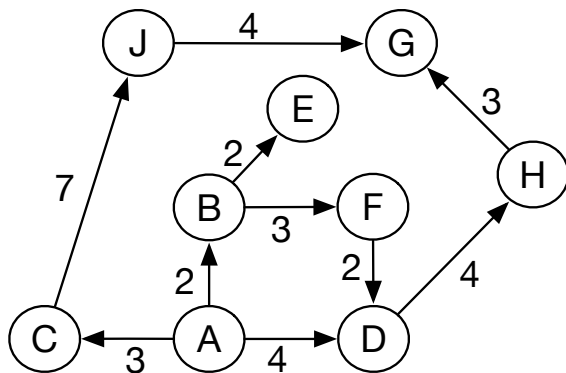
# Directed Graphs

- Searching in graphs provides an appropriate abstract model of problem solving independent of a particular domain.
- A (directed) **graph** consists of a set  $N$  of **nodes** and a set  $A$  of ordered pairs of nodes, called **arcs**.
- Node  $n_2$  is a **neighbor** of  $n_1$  if there is an arc from  $n_1$  to  $n_2$ . That is, if  $\langle n_1, n_2 \rangle \in A$ .
- A **path** is a sequence of nodes  $\langle n_0, n_1, \dots, n_k \rangle$  such that  $\langle n_{i-1}, n_i \rangle \in A$ .
- Given **start nodes** and **goal nodes**, a **solution** is a path from a start node to a goal node.
- When there is a **cost** associated with arcs, the cost of a path is the sum of the costs of the arcs in the path:

$$\text{cost}(\langle n_0, n_1, \dots, n_k \rangle) = \sum_{i=1}^k \text{cost}(\langle n_{i-1}, n_i \rangle)$$

An **optimal solution** is one with minimum cost.

# State-Space Graph for the Delivery Robot (Acyclic)

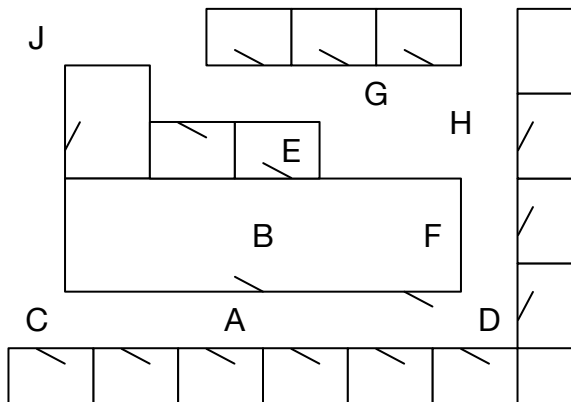


# Directed Acyclic Graph (DAG)

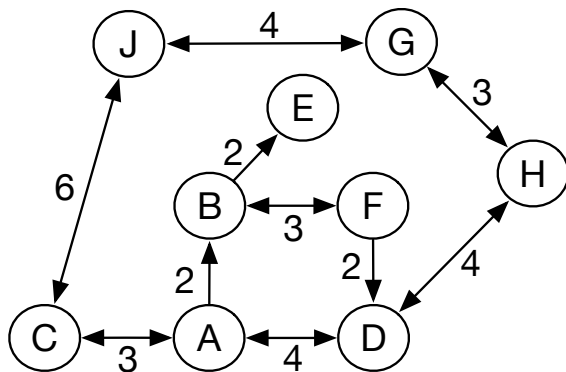
- A **cycle** is a non-empty path where the end node is the same as the start node – that is,  $\langle n_1, n_2, \dots, n_k \rangle$  such that  $n_0 = n_k$
- A (directed) **graph** A directed graph without any cycles is called a **directed acyclic graph (DAG)**.
- A **tree** is a DAG where there is one node with no incoming arcs and every other node has exactly one incoming arc.
- The node with no incoming arcs is called the **root** of the tree. A node with no outgoing arcs is called a **leaf**.
- In a tree, an arc goes from a **parent** to a **child**
- The **forward branching factor** of a node is the number of outgoing arcs from the node.
- The **backward branching factor** of a node is the number of incoming arcs to the node.

# Example Problem for Delivery Robot

The robot is at A and the goal is to get to G:



# State-Space Graph for the Delivery Robot



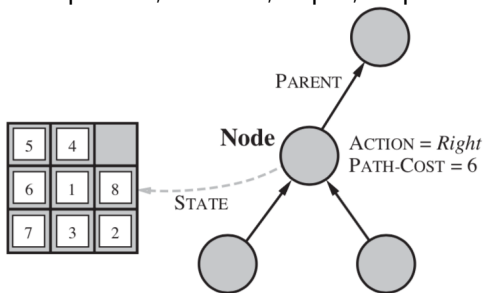
# Implementation: states vs. nodes

A **state** is a (representation of) a physical configuration

A **node** is a data structure constituting part of a search tree

includes **parent**, **children**, **depth**, **path cost**  $g(x)$

States do not have parents, children, depth, or path cost!



The Expand function creates new nodes, filling in the various fields and using the SuccessorFn of the problem to create the corresponding states.

# Tree search algorithms

Basic idea:

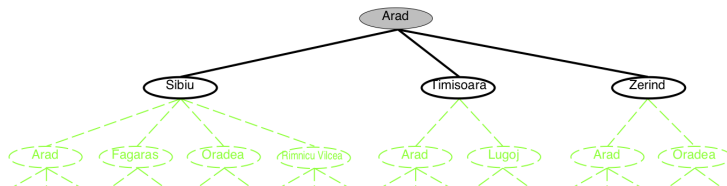
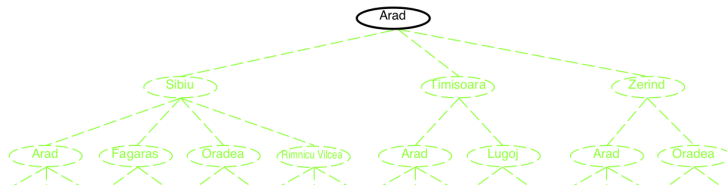
offline, simulated exploration of state space

by generating successors of already-explored states

(a.k.a. **expanding** states)

```
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
```

# Tree search example





# Partial Search Space for a Video Game

Grid game: Rob is on a grid and can move up, down, left or right and needs to collect coins  $C_1$ ,  $C_2$ ,  $C_3$ ,  $C_4$ , without running out of fuel, and end up at location (1, 1):

State:  $\langle X\text{-pos}, Y\text{-pos}, \text{Fuel}, C_1, C_2, C_3, C_4 \rangle$

Goal:  $\langle 1, 1, ?, t, t, t, t \rangle$

