

# **Introduction to Artificial Intelligence**

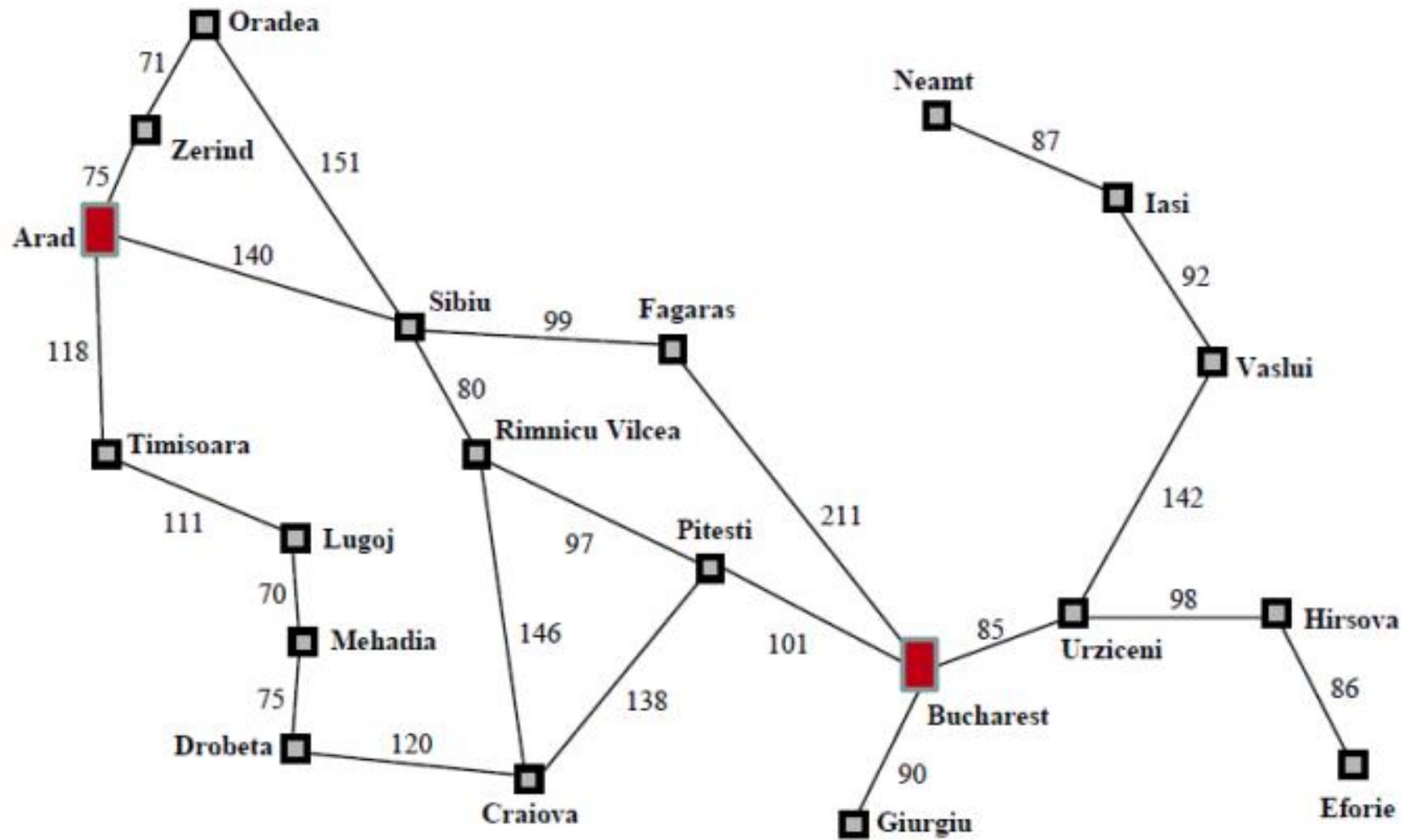
**Lecture: Problem Solving Using Searching**

# Outline

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- Problem-Solving Agents
- Example Problems
- Searching for Solutions

# Holiday in Romania

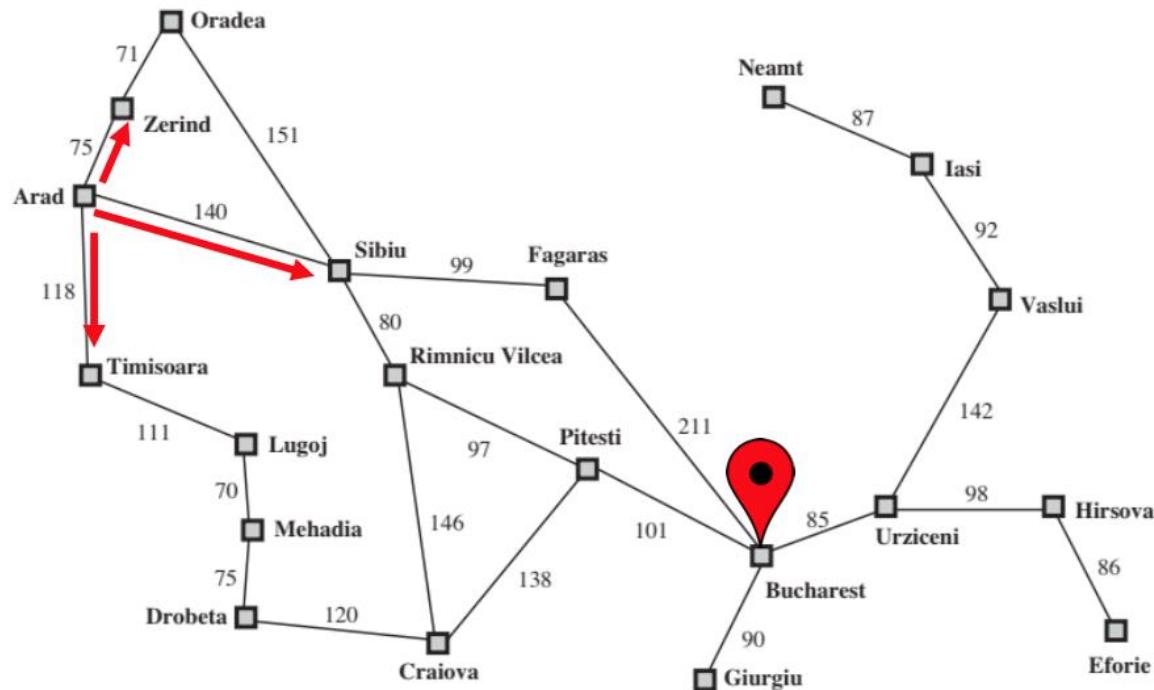


# Goal-based Agents

- Intelligent agents maximize their performance measure.
- Goals help organize behavior by limiting the objectives that the agent is trying to achieve and the actions it considers.

# Problem Formulation

- Consider a goal to be a set of world states in which the objective is satisfied.
- Problem formulation is the process of deciding what actions and states to consider, given a goal.



# Properties of the Romania environment

- Observable
  - Each city has a sign indicating its presence for arriving drivers.
  - The agent always knows the current state.
- Discrete
  - Each city is connected to a small number of other cities.
  - There are only finitely many actions to choose from any given state.
- Known
  - The agent knows which states are reached by each action.
- Deterministic
  - Each action has exactly one outcome.

# Solving problem by searching

- Search: the process of looking for a sequence of actions that reaches the goal
- A search algorithm takes a problem as input and returns a solution in the form of an action sequence.
- Execution phase: once a solution is found, the recommended actions are carried out.

# Solving problem by searching

```
function PROBLEM-SOLVING-AGENT(percept) returns an action
  persistent: seq, an action sequence, initially empty
              state, some description of the current world state
              goal, a goal, initially null
              problem, a problem formulation
  state ← UPDATE-STATE(state, percept)
  if seq is empty then
    goal ← FORMULATE-GOAL(state)
    problem ← FORMULATE-PROBLEM(state, goal)
    seq ← SEARCH(problem)
    if seq = failure then return a null action
  action ← FIRST(seq)
  seq ← REST(seq)
  return action
```

# Well-defined problems and solutions

- A problem can be defined formally by five components.
  - Initial state: in which the agent starts
    - E.g., the agent in Romania has its initial state described as  $In(Arad)$
  - Actions: the possible actions available to the agent
    - E.g.,  $ACTION(Arad) = \{$   
  
 $Go(Sibiu), Go(Timisoara), Go(Zerind)\}$
  - Transition model: what each action does
    - E.g.,  $Result(In(Arad), Go(Zerind)) = In(Zerind)$
  - Successor: a state reachable from a given state by a single action

# Well-defined problems and solutions

- Goal test: determine whether a given state is a goal state
  - The goal is specified by either an explicit set of possible goal states or an abstract property.
  - E.g.,  $In(Bucharest)$ , checkmate
- Path cost: a function that sets a numeric cost to each path
  - Non-negative, reflecting the agent's performance measure
  - E.g.,  $c(In(Arad), Go(Zerind), In(Zerind)) = 75$
- An optimal solution has the lowest path cost.

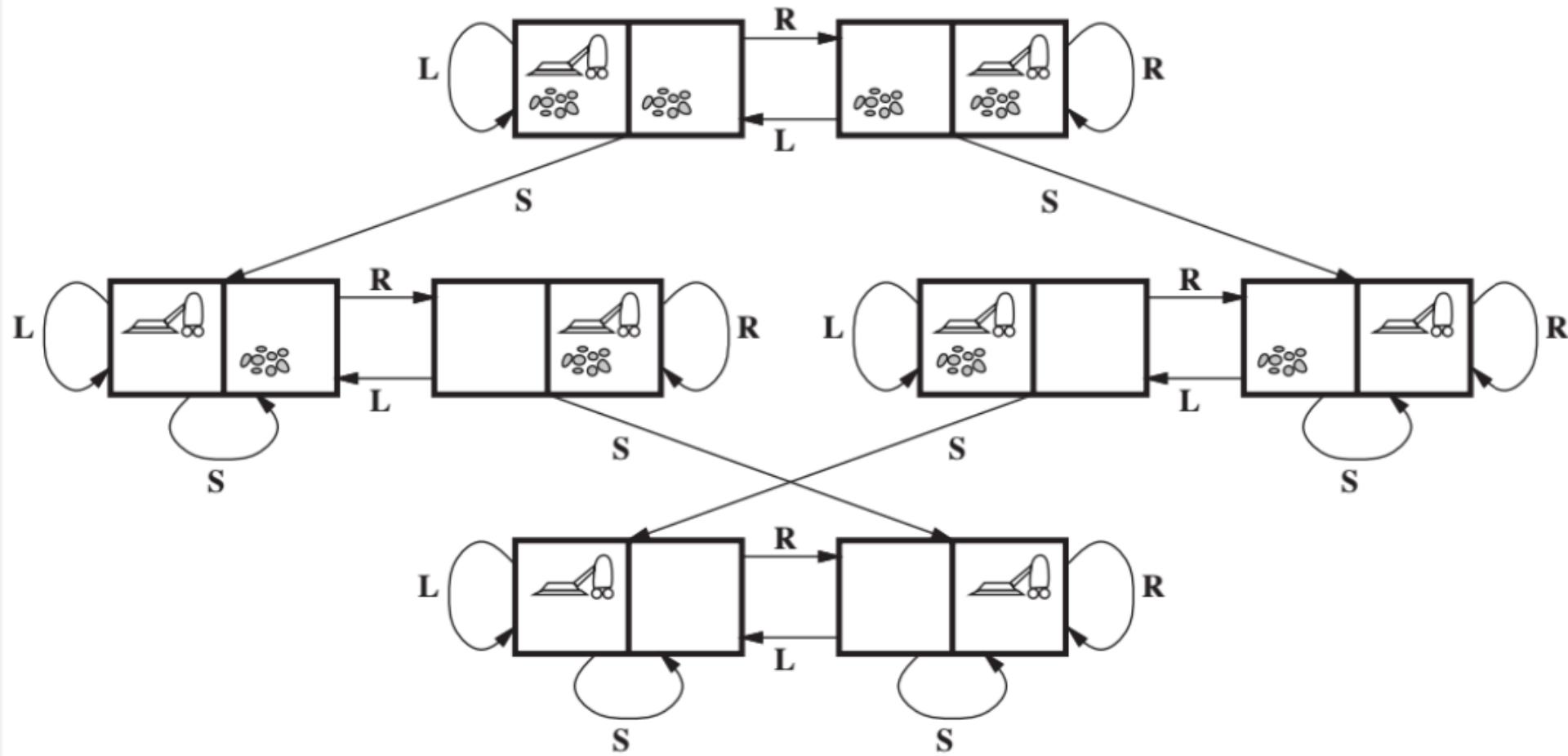
# Formulating problems by abstraction

- Abstraction creates an approximate and simplified model of the real world, which is too detailed for computer.
- This is critical for automated problem solving.
- The choice of a good abstraction involves
  - Remove as much detail as possible while
  - Retain validity and ensure that the abstract actions are easy to be carried out.

# The Vacuum-cleaner world

- States: determined by both the agent location and the dirt locations
  - $2 \times 2^2 = 8$  possible world states ( $n \times 2^n$  in general)
- Initial state: Any state can be designated as the initial state.
- Actions: Left, Right, and Suck
- Transition model: The actions have their expected effects.
- Goal test: whether all the squares are clean
- Path cost: each step costs 1

# The Vacuum-cleaner world



# The 8-puzzle

- States: the location of each of the eight tiles and the blank
- Initial state: any state can be designated as the initial state
- Actions: movements of the blank space
  - Left, Right, Up, or Down.
  - Different subsets of these are possible depending on where the blank is
- Transition model: return a resulting state given a state and an action
- Goal test: whether the state matches the goal configuration
- Path cost: each step costs 1

# The 8-puzzle

1	8	2
	4	3
7	6	5

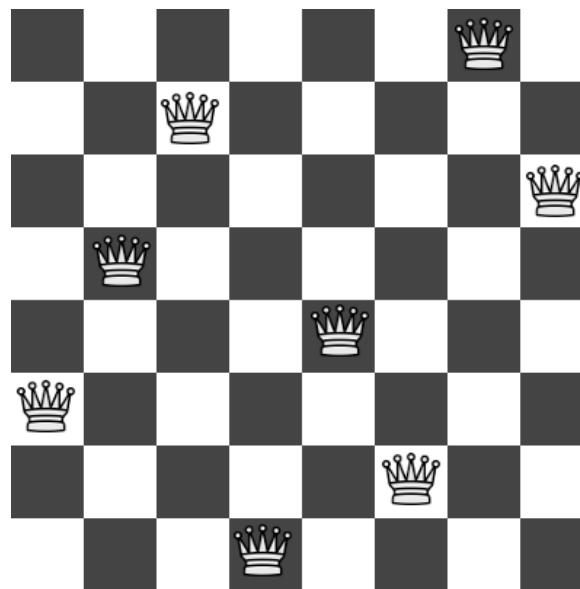
Given State

1	2	3
4	5	6
7	8	

Goal State

# The 8-queens

- Incremental formulation: add a queen step-by-step to the empty initial state
- Complete-state formulation: start with all 8 queens on the board and move them around
- The path cost is trivial because only the final state counts



# The 8-queens

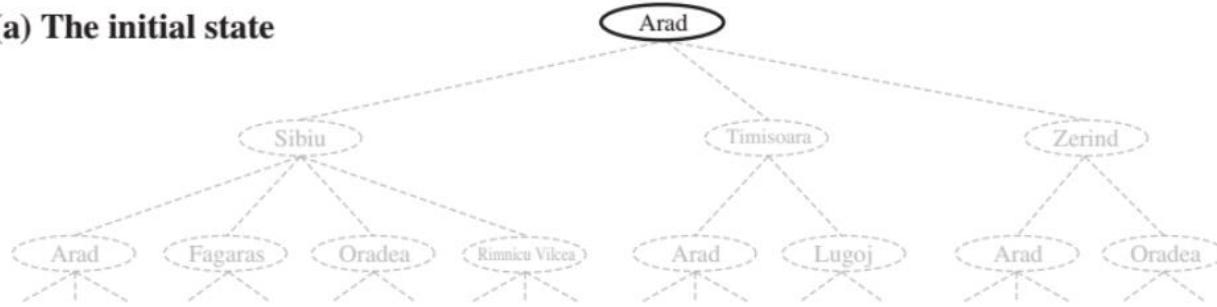
- States: any arrangement of 0 to 8 queens on the board
- Initial state: no queens on the board
- Actions: add a queen to any empty square
- Transition model: returns the board with a queen added to the specified square
- Goal test: 8 queens are on the board, none attacked
- $64 \cdot 63 \cdots 57 \approx 1.8 \times 10^{14}$  possible sequences to investigate

# Search Tree

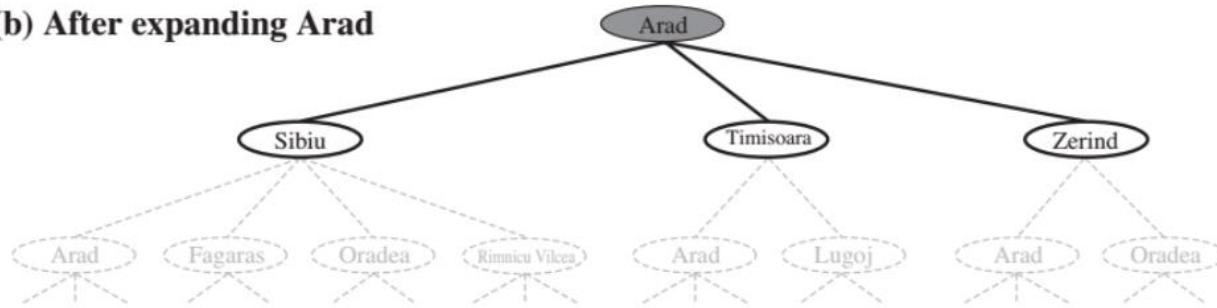
- Search algorithms consider many possible action sequences to find the solution sequence.
- Search tree: the possible action sequences starting at the initial state (root)
  - Branches are actions and nodes are states in the problem's state space
- Frontier: the set of all leaf nodes available for expansion at any given point
- Search algorithms all share the basic structure while vary according to how they choose which state to expand next -- called **search strategy**.

# Search Tree

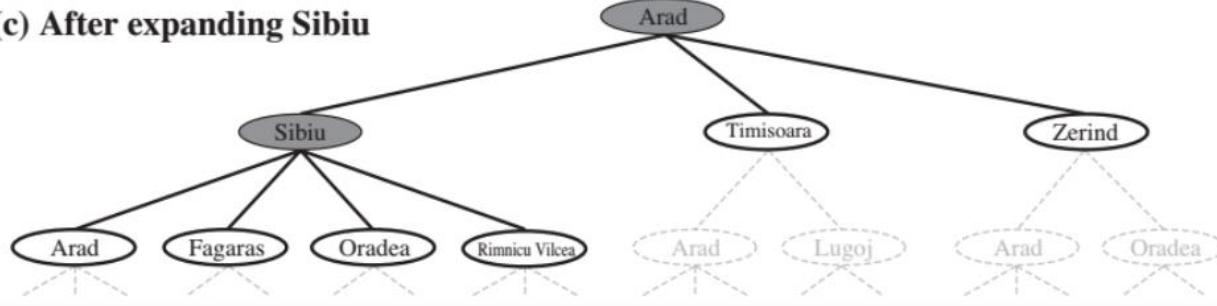
(a) The initial state



(b) After expanding Arad



(c) After expanding Sibiu



# Search Tree

```
function TREE-SEARCH(problem) returns a solution, or failure
    initialize the frontier using the initial state of problem
    loop do
        if the frontier is empty then return failure
        choose a leaf node and remove it from the frontier
        if the node contains a goal state then return the
            corresponding solution
        expand the chosen node, adding the resulting nodes to the
            frontier
```

# Redundant paths

- Redundant paths are unavoidable.
- Following redundant paths may cause a tractable problem to become intractable.
- This is true even for algorithms that know how to avoid infinite loops.

# Graph Search

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

initialize the frontier using the initial state of problem

***initialize the explored set to be empty***

**loop do**

**if** the frontier is empty **then** return failure

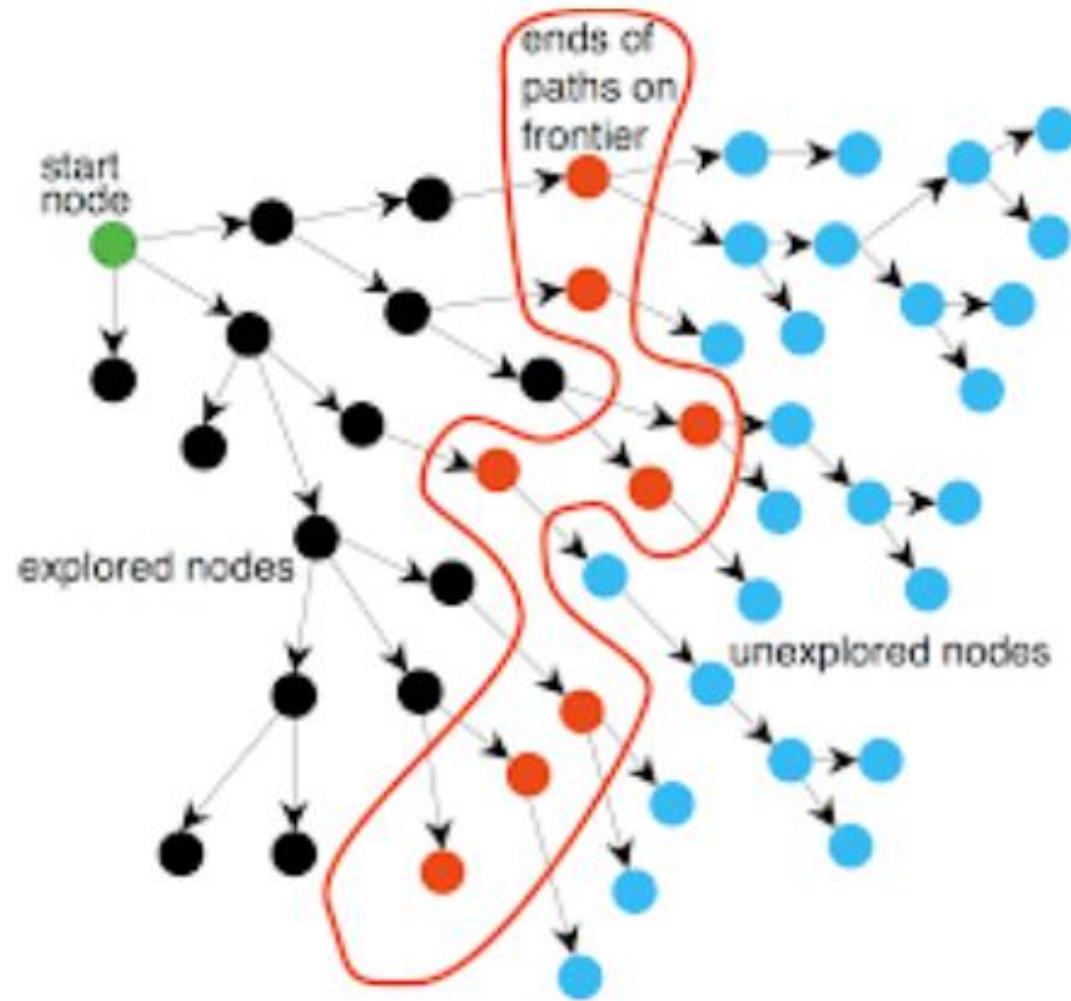
    choose a leaf node and remove it from the frontier

**if** the node contains a goal state **then return** the corresponding solution

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

    expand the chosen node, adding the resulting nodes to the frontier ***only if not in the frontier nor explored set***

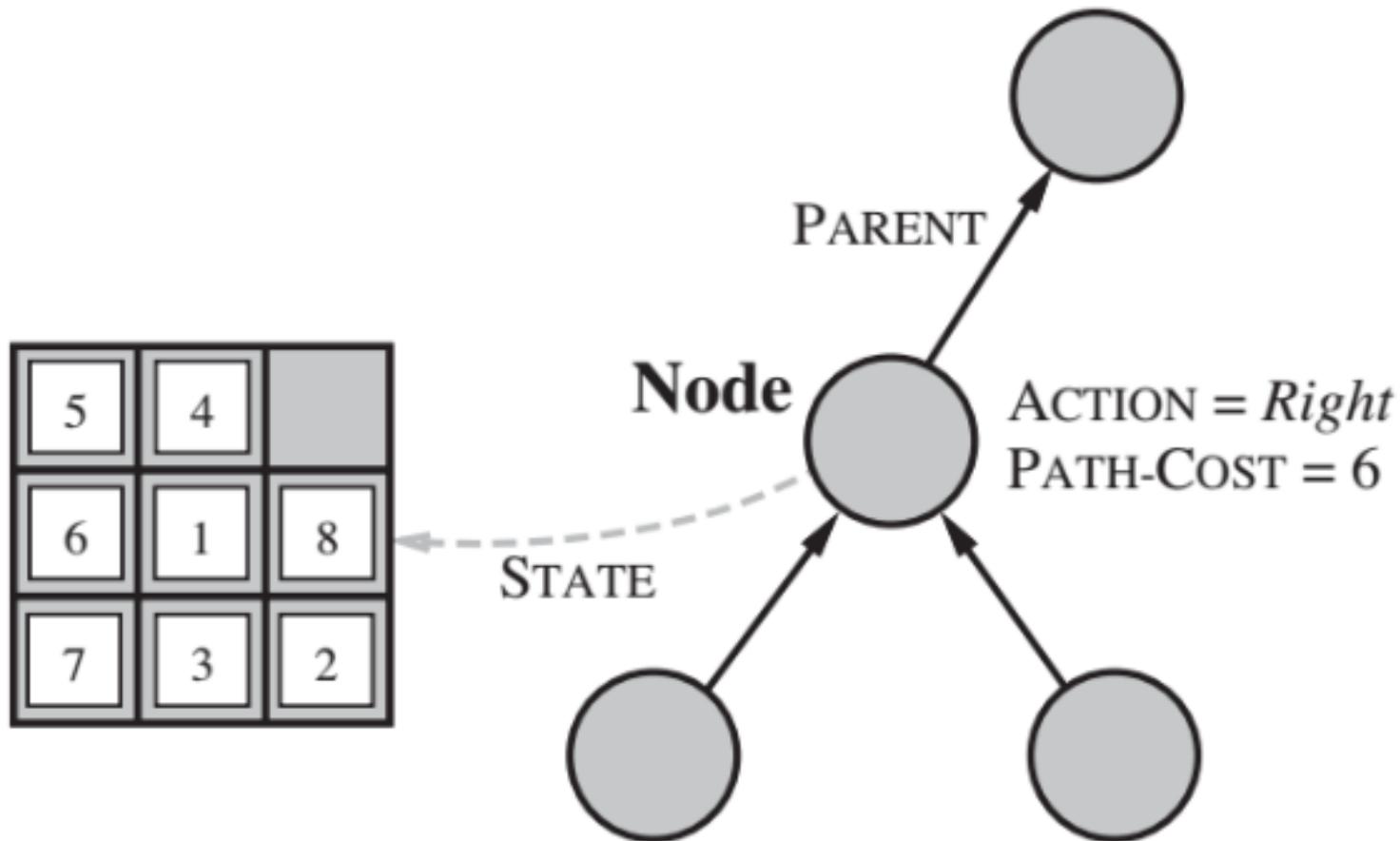
# Graph Search



# Infrastructure for search algorithms

- Each node  $n$  is structuralized by four components.
  - **$n$ . STATE**: the state in the state space to which the node corresponds
  - **$n$ . PARENT**: the node in the search tree that generated the node  $n$
  - **$n$ . ACTION**: the action applied to the parent to generate  $n$
  - **$n$ . PATH – COST** : the cost, denoted by  $g(n)$ , of the path from the initial state to the node, as indicated by the parent pointer
- Frontier can be implemented with a (priority) queue or stack.
- Explored set can be a hash table that allows for efficient checking of repeated states.

# Infrastructure for search algorithms



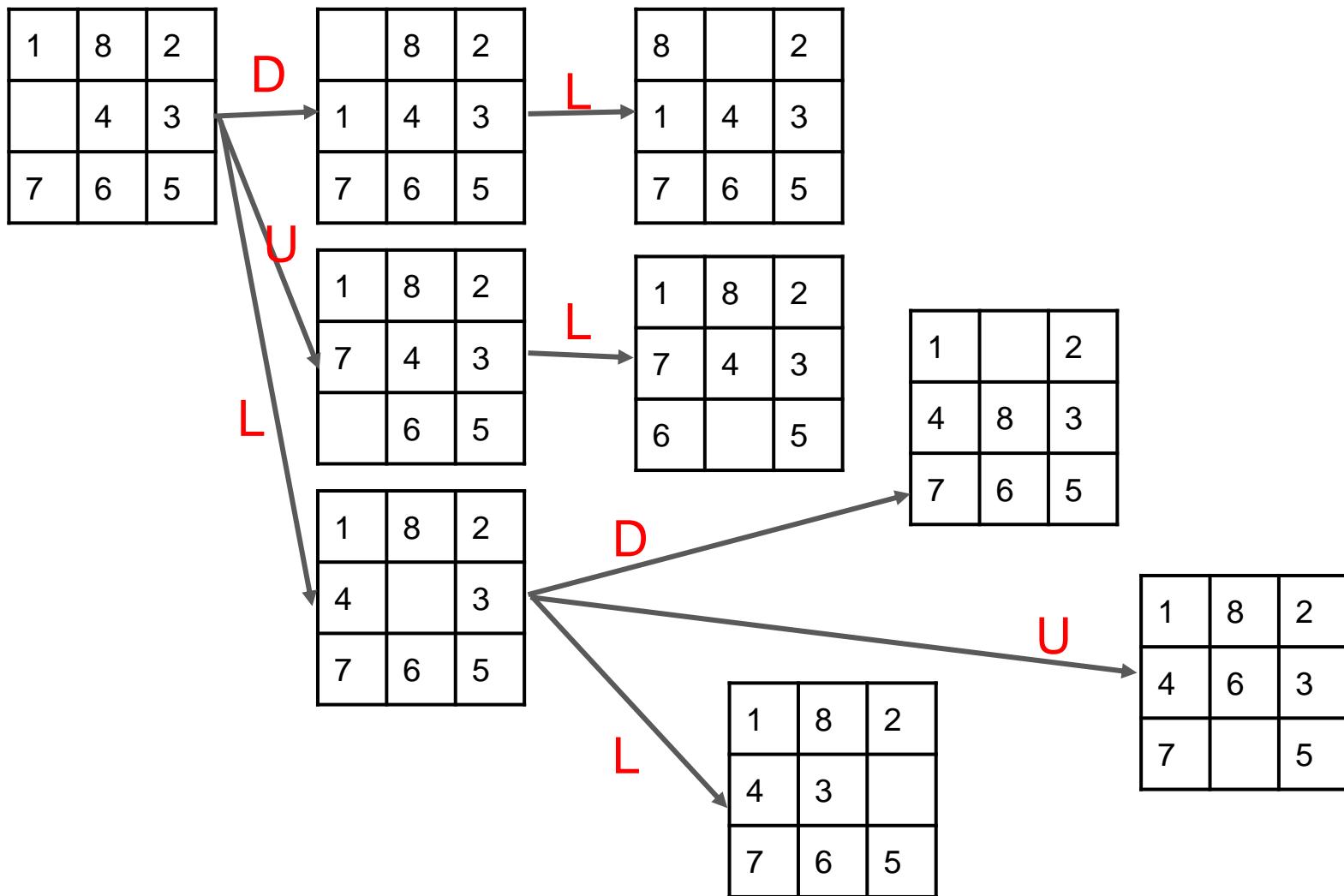
# Problem-solving performance

- **Completeness:** does it always find a solution if one exists?
- **Time complexity:** how long does it take to find a solution?
- **Space complexity:** how much memory is needed to perform the search?
- **Optimality:** does it always find a least-cost solution?

# Homework

- Task 1:
  - Given the start state and the goal one of the 8-puzzle in page 15.
  - Students draw a search tree by the expanding nodes.
  - The expansion stops when the goal state is expanded.
- Task 2:
  - Program Task 1 in Python using Google Colab
  - Use graph search instead of tree search

# Homework

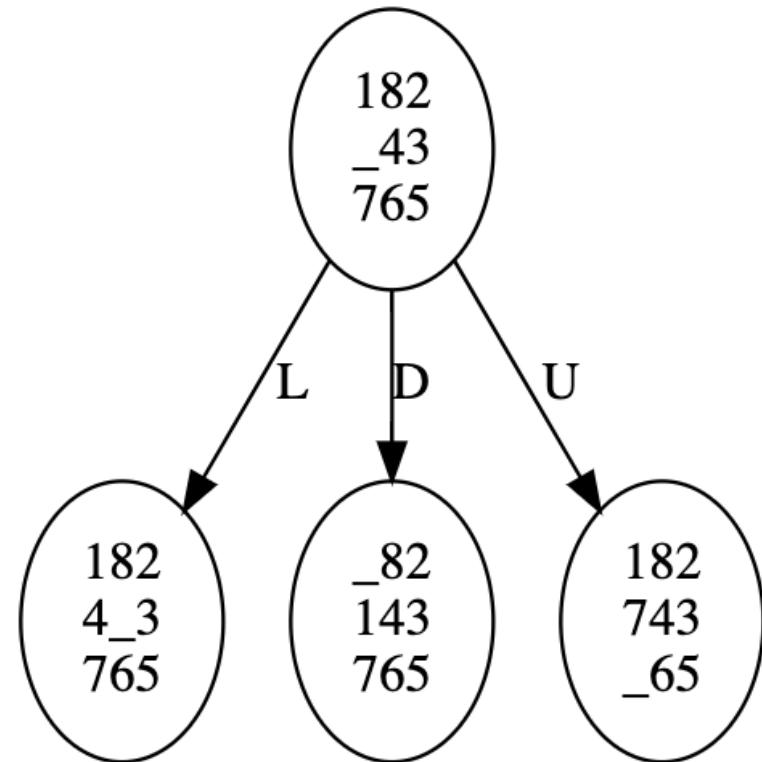


# Homework

```
from graphviz import Digraph

dot = Digraph()
dot.node('0', '182\n4_3\n765')
dot.node('1', '182\n4_3\n765')
dot.node('2', '_82\n143\n765')
dot.node('3', '182\n743\n65')
dot.edge('0', '1', 'L')
dot.edge('0', '2', 'D')
dot.edge('0', '3', 'U')

dot
```



# Homework

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- Conduct homework in the given notebook

# References

- Stuart Russell and Peter Norvig. 2009. Artificial Intelligence: A Modern Approach (3rd ed.). Prentice Hall Press, Upper Saddle River, NJ, USA.
- Lê Hoài Bắc, Tô Hoài Việt. 2014. Giáo trình Cơ sở Trí tuệ nhân tạo. Khoa Công nghệ Thông tin. Trường ĐH Khoa học Tự nhiên, ĐHQG-HCM.
- Nguyễn Ngọc Thảo, Nguyễn Hải Minh. 2020. Bài giảng Cơ sở Trí tuệ Nhân tạo. Khoa Công nghệ Thông tin. Trường ĐH Khoa học Tự nhiên, ĐHQG-HCM.