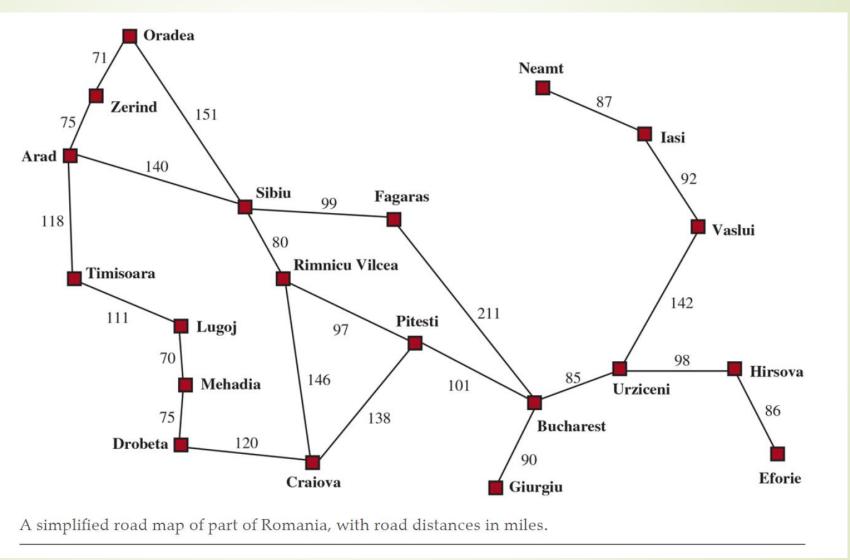
Solving Problems by Searching

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Problem-solving Agent

- When the correct action to take is not immediately obvious,
 - an agent may need to to plan ahead:
 - to consider a sequence of actions that form a path to a goal state
 - The computational process it undertakes is called **search**.

A Touring Vacation in Romania



Map: The Information About The World

- With that information, the agent can follow this four-phase problem-solving process:
 - GOAL FORMULATION: e.g. Bucarest
 - PROBLEM FORMULATION :
 - considers the actions of traveling from one city to an adjacent city
 - The only fact about the state of the world (that will change due to an action) is the current city
 - SEARCH:
 - simulates sequences of actions in its model
 - searching until it finds a sequence of actions that reaches the goal (a solution)
 - EXECUTION (the actions the solution)

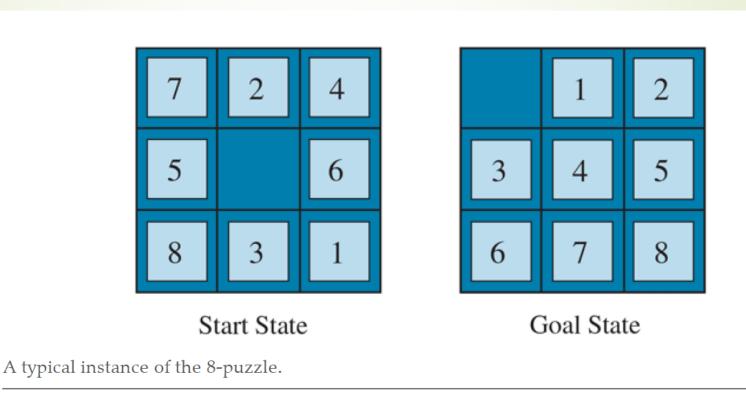
A Search Problem

- State space: A set of possible states that the environment can be in
- The initial state that the agent starts in
- A set of one or more goal states
- ACTION(s) returns a finite set of actions that can be executed in s
- A transition model: RESULT(Arad, ToZerind) = Zerind
- An action cost function
 - reflects performance measure (e.g. distance, time)
- An optimal solution has the lowest cost among all solutions

Formulating problems

- The problem of getting to Bucharest is a model
 - not the real thing.
 - The process of removing detail from a representation is called abstraction

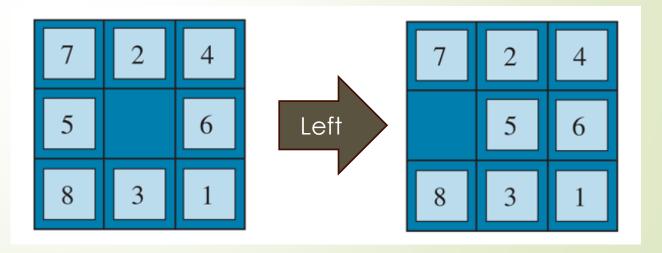
8-puzzle



8-puzzle

- **STATES:** A state description specifies the location of each of the tiles.
- INITIAL STATE: Any state can be designated as the initial state
 - State space is partitioned into two halves.
- ACTIONS: blank space moving Left, Right, Up, or Down.
 - If the blank is at an edge or corner then not all actions will be applicable.
- TRANSITION MODEL: Maps a state and action to a resulting state

- /GOAL STATE
- **ACTION COST:** Each action costs 1.



Search Algorithms

- Algorithms that superimpose a search tree over the state-space graph
 - forming various paths from the initial state
 - trying to find a path that reaches a goal state

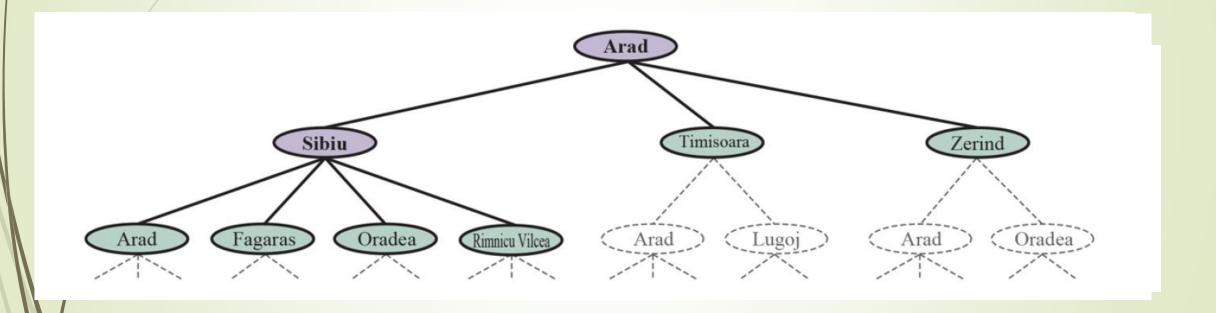
The state space

- describes the (possibly infinite) set of states in the world, and
- the actions that allow transitions from one state to another

The search tree

- describes paths between these states, reaching towards the goal
- may have multiple paths to any given state
- but each node in the tree has a unique path back to the root

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Best-first search: A general approach

```
function BEST-FIRST-SEARCH(problem, f) returns a solution node or failure
  node \leftarrow Node(State=problem.INITIAL)
  frontier \leftarrow a priority queue ordered by f, with node as an element
  reached ← a lookup table, with one entry with key problem. INITIAL and value node
  while not IS-EMPTY(frontier) do
    node \leftarrow Pop(frontier)
    if problem.IS-GOAL(node.STATE) then return node
    for each child in EXPAND(problem, node) do
       s \leftarrow child.STATE
       if s is not in reached or child.PATH-COST < reached[s].PATH-COST then
          reached[s] \leftarrow child
          add child to frontier
  return failure
function EXPAND(problem, node) yields nodes
  s \leftarrow node.STATE
  for each action in problem.ACTIONS(s) do
    s' \leftarrow problem.RESULT(s, action)
    cost \leftarrow node.PATH-COST + problem.ACTION-COST(s, action, s')
    yield NODE(STATE=s', PARENT=node, ACTION=action, PATH-COST=cost)
```

Measuring problem-solving performance

COMPLETENESS:

Is the algorithm guaranteed to find a solution when there is one, and to correctly report failure when there is not?

COST OPTIMALITY:

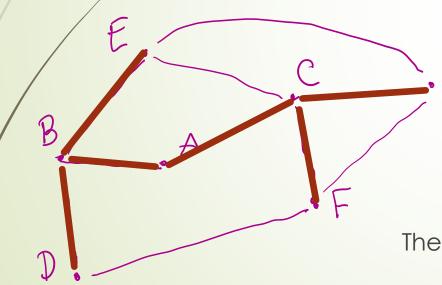
- Does it find a solution with the lowest path cost of all solutions?
- TIME COMPLEXITY
- SPACE COMPLEXITY
- In many Al problems, the state-space graph is represented only implicitly by the initial state, actions, and transition model.
- For an implicit state space, complexity can be measured in terms of
 - the depth or number of actions in an optimal solution; d
 - the maximum number of actions in any path; m
 - and the branching factor or number of successors of a node that need to be considered; b

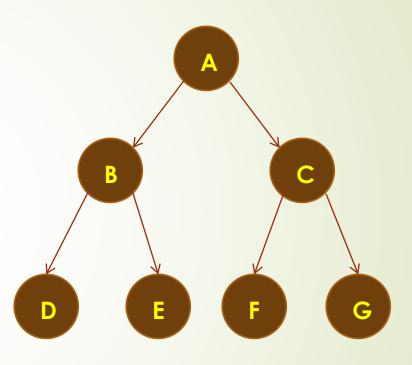
Uninformed Search Strategies

- An uninformed search algorithm is given no clue about how close a state is to the goal(s).
 - For example, consider our agent in Arad with the goal of reaching Bucharest.
 - An uninformed agent with no knowledge of Romanian geography has no clue whether going to Zerind or Sibiu is a better first step.
- In contrast, an **informed agent** who knows the location of each city knows that **Sibiu** is much closer to **Bucharest** and thus more likely to be on the **shortest path**.

Breadth First Search

- When all actions have the same cost
- This is a systematic search strategy that is therefore complete even on infinite state spaces.





The sequence of states being searched

Breadth-first Search

- always finds a solution with a minimal number of actions
 - When it is generating nodes at depth d, it has already generated all the nodes at depth d-1.
 - If one of them were a solution, it would have been found.
- It is complete.
- Suppose that the solution is at depth d, with branching factor = b.
 - The total number of nodes generated is $1 + b + b^2 + b^3 + ... + b^d = O(b^d)$
 - All the nodes remain in memory
 - so both time and space complexity are O(bd)

Breadth-first Search

return failure

```
function Breadth-First-Search(problem) returns a solution node or failure
  node \leftarrow Node(problem.INITIAL)
  if problem.Is-GOAL(node.STATE) then return node
  frontier \leftarrow a FIFO queue, with node as an element
  reached \leftarrow \{problem.INITIAL\}
   while not IS-EMPTY(frontier) do
     node \leftarrow Pop(frontier)
     for each child in Expand(problem, node) do
       s \leftarrow child.STATE
       if problem.Is-GOAL(s) then return child
       if s is not in reached then
                                         avoiding repeated
          add s to reached
                                         states
          add child to frontier
```

Simplest Algorithm Structure of BFS

```
s = initial_state
while not Goal(s)
for each successor_state x of s
    enqueue(x)
s = dequeue()
```

Tracing back for path

```
PRINT-PATH(G, s, v)

1 if v == s

2 print s

2 elseif v . \pi == NIL

4 print "no path from" s "to" v "exists"

5 else PRINT-PATH(G, s, v . \pi)

print v
```

NOTE:

G is state space graph