

# Introduction to Artificial Intelligence

**DA 221**

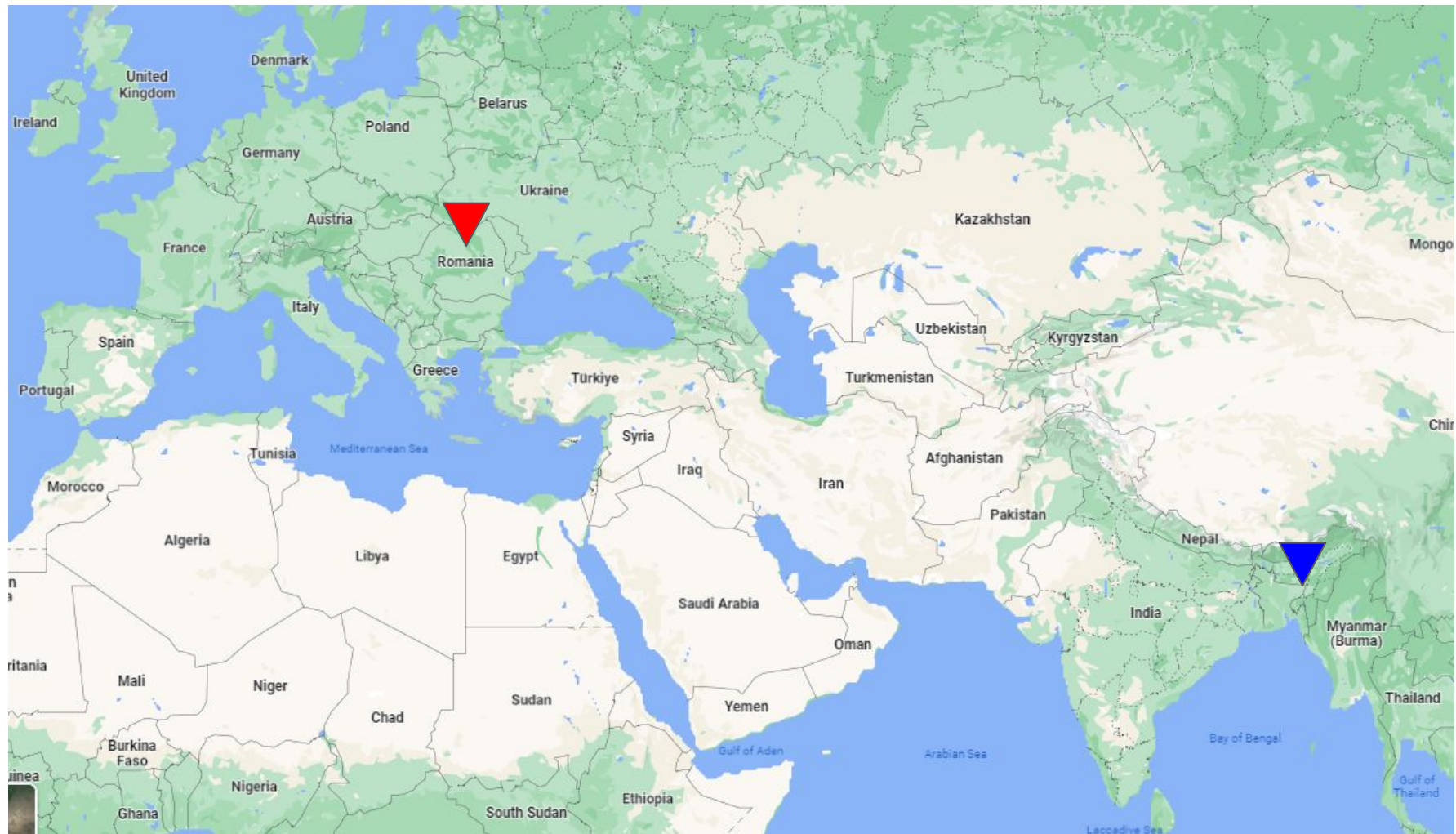
Jan - May 2023

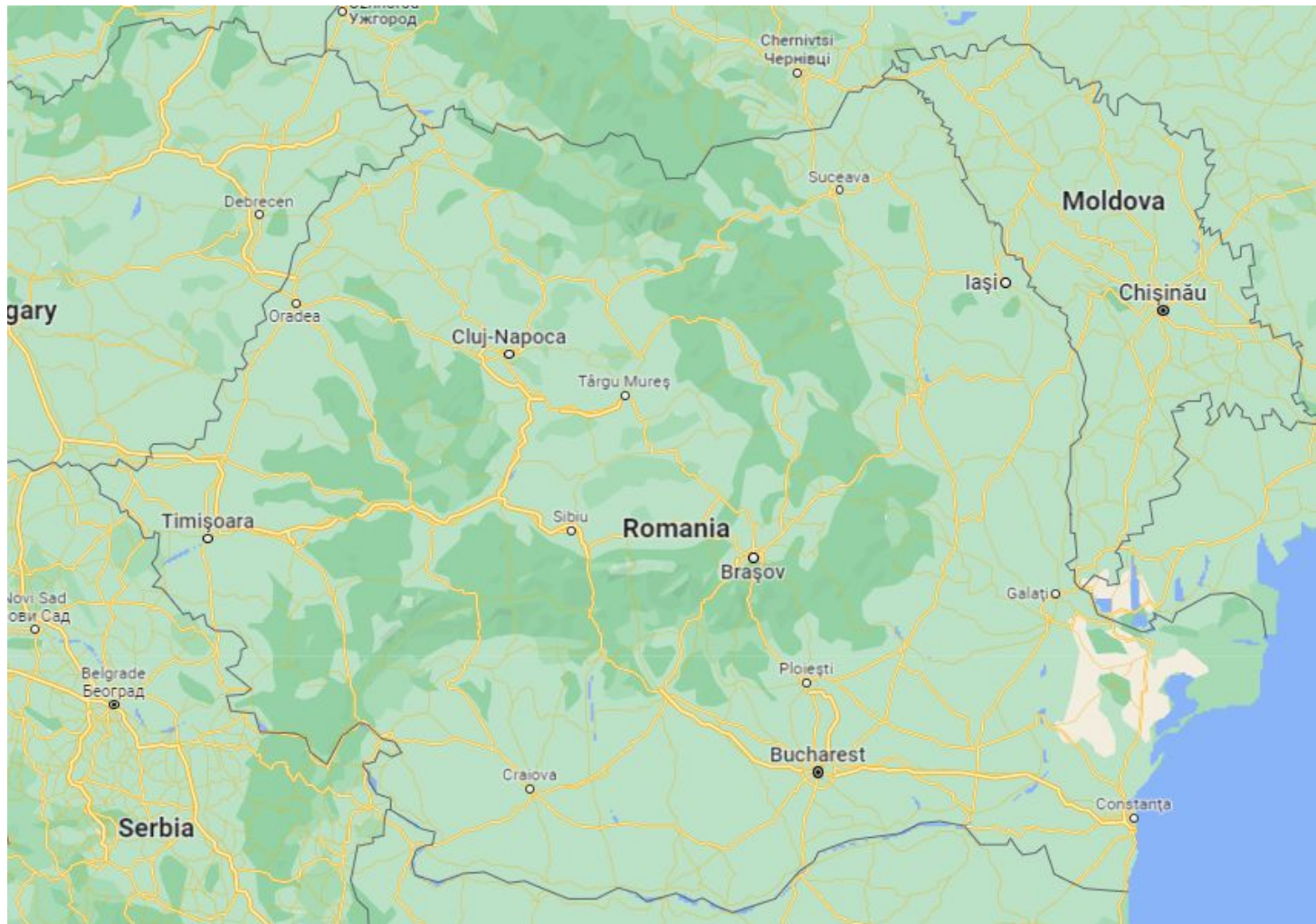
IIT Guwahati

Instructors: Neeraj Sharma (& Arghyadip Roy)

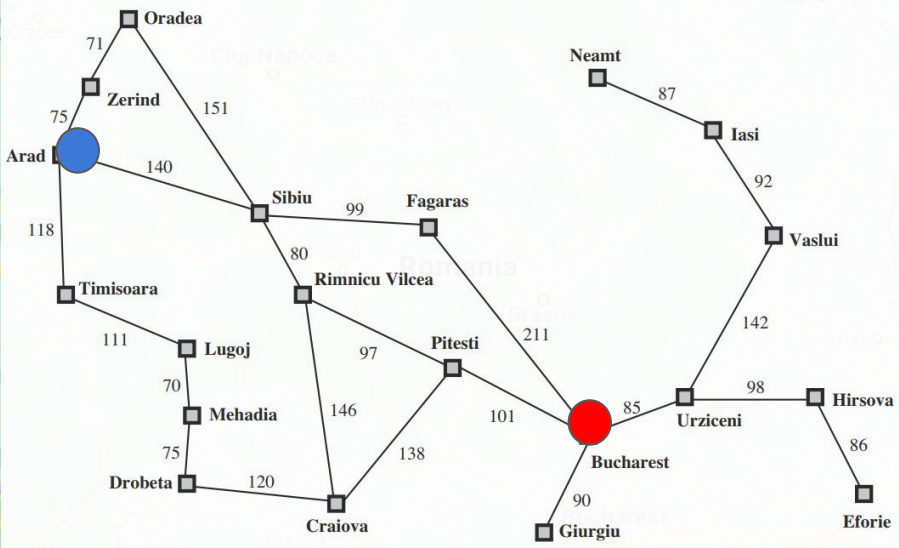
Lecture 07: Neeraj Sharma

Uninformed Search Strategies ... bringing more context

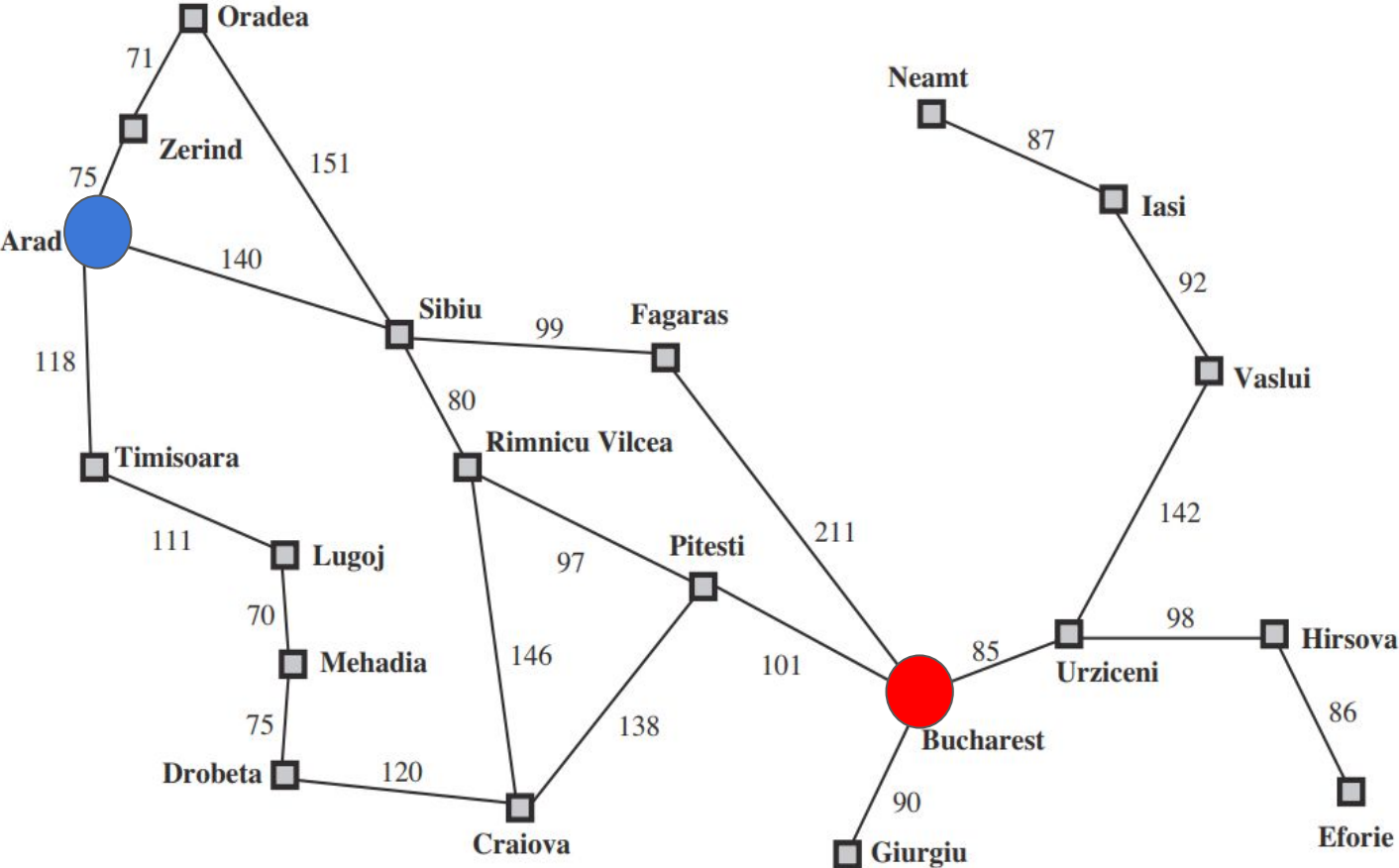




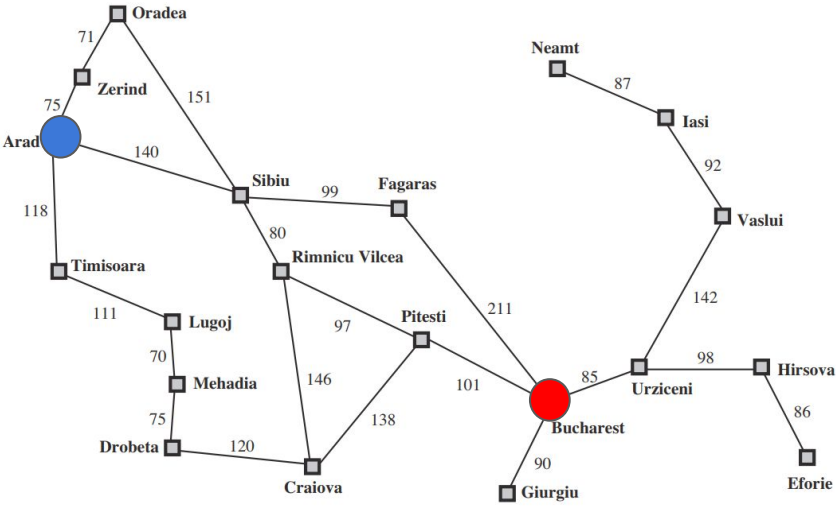
# Go from Arad to Bucharest



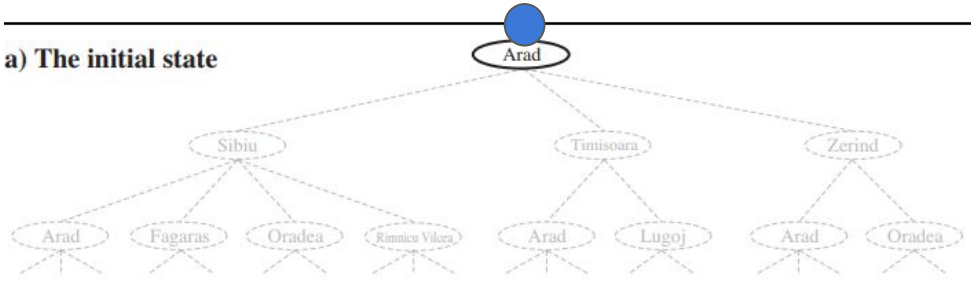
The agent's goal in Romania is the singleton set  $\{In(Bucharest)\}$



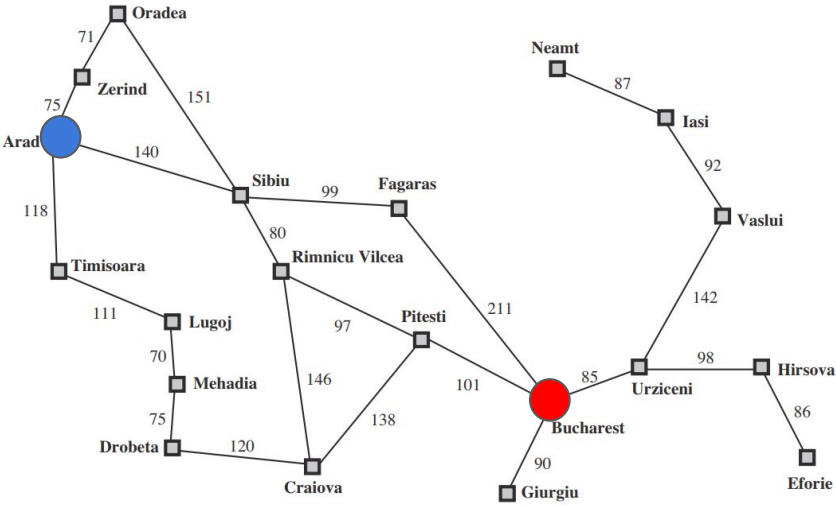
# Creating a Search Tree



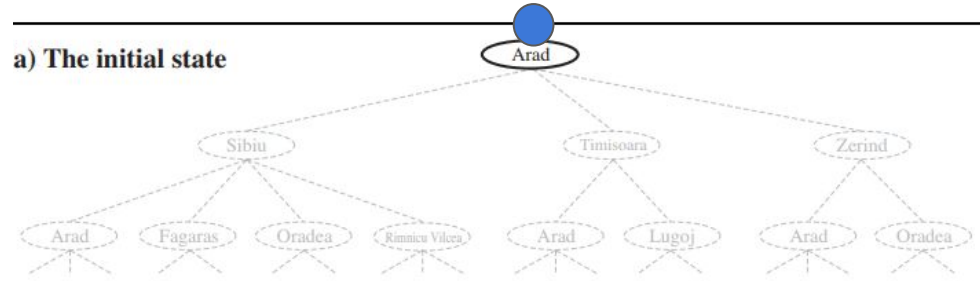
a) The initial state



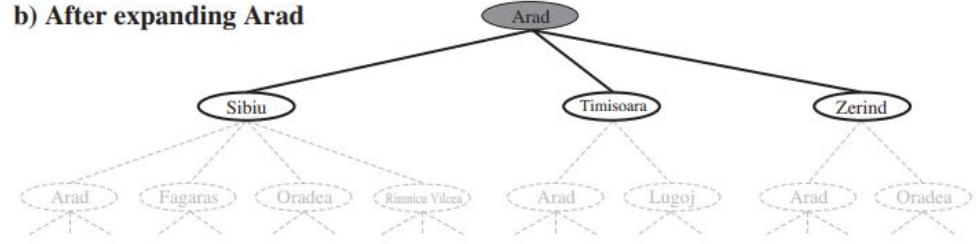
# Creating a Search Tree



a) The initial state

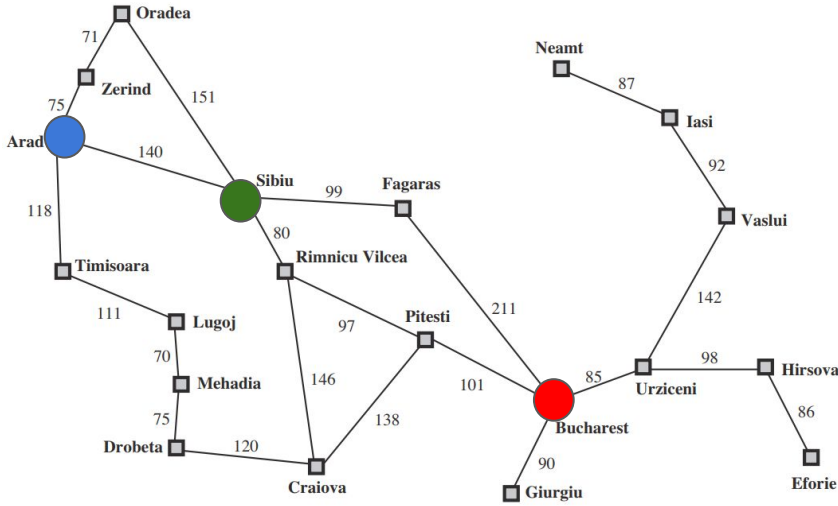


b) After expanding Arad

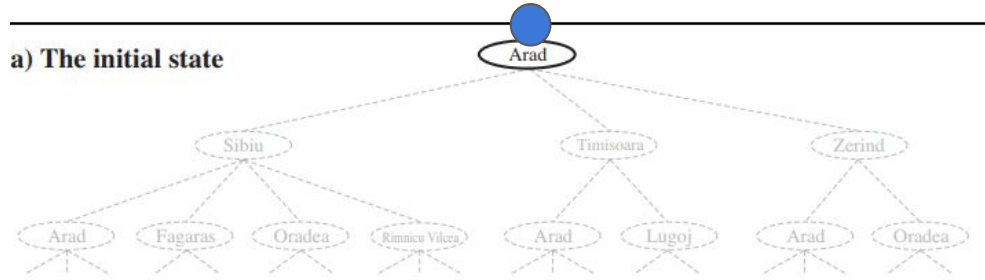




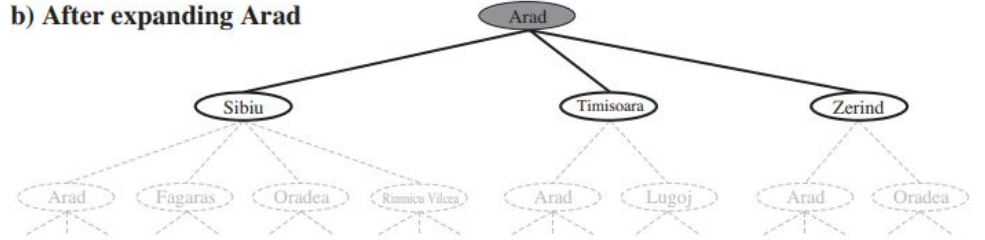
# Creating a Search Tree



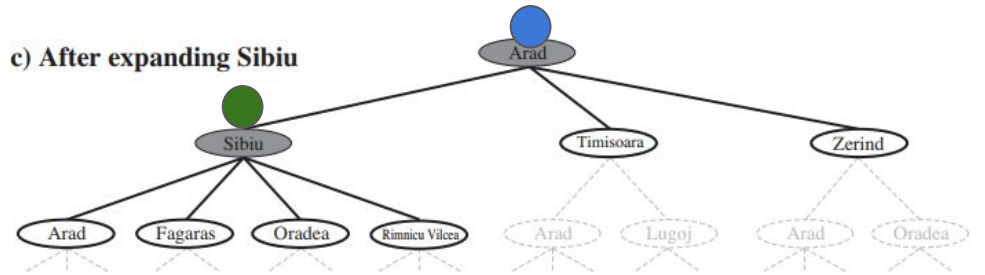
a) The initial state



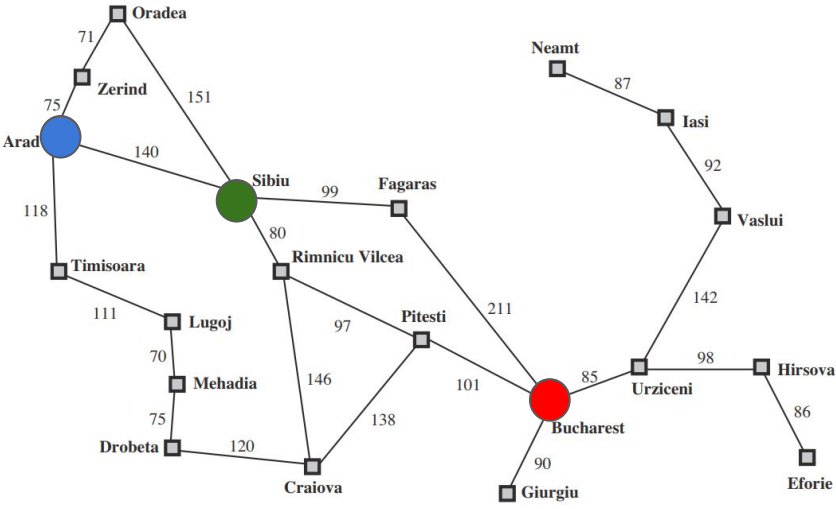
b) After expanding Arad



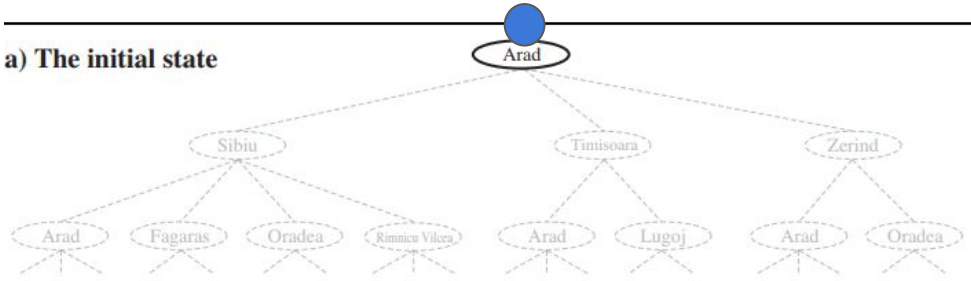
c) After expanding Sibiu



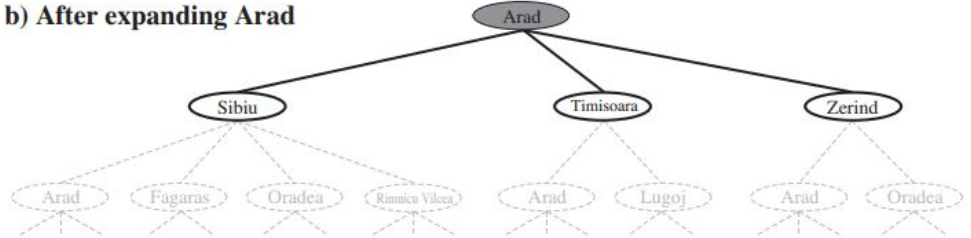
# Creating a Search Tree



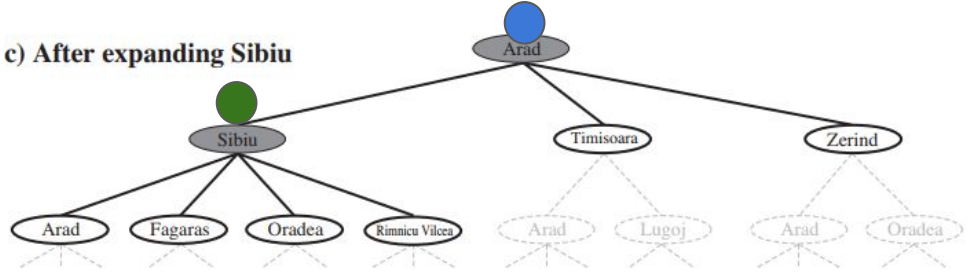
a) The initial state



b) After expanding Arad



c) After expanding Sibiu



What is happening?

- Generation of nodes
- Expansion of nodes

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

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initialize the frontier using the initial state of *problem*

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**loop do**

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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

```
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
```

```
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
```

“Algorithms that forget their history are doomed to repeat it”: create an explored set to help keep track of node visits.



**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

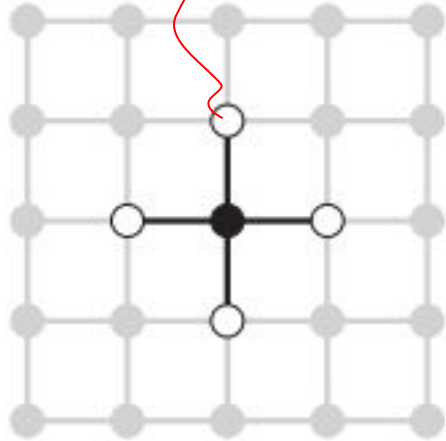
**if** the node contains a goal state **then return** the corresponding solution

*add the node to the explored set*

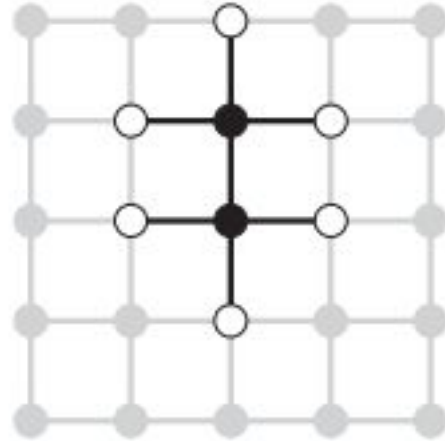
    expand the chosen node, adding the resulting nodes to the frontier

*only if not in the frontier or explored set*

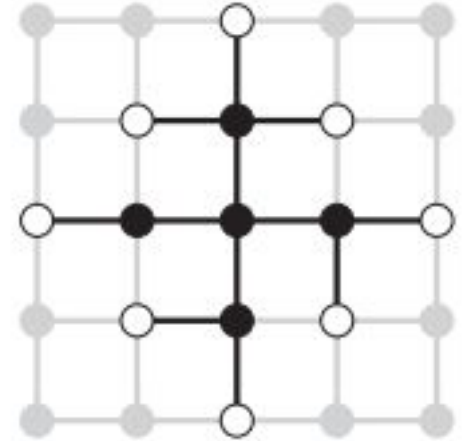
Frontier: separates unexplored from explored



(a)



(b)

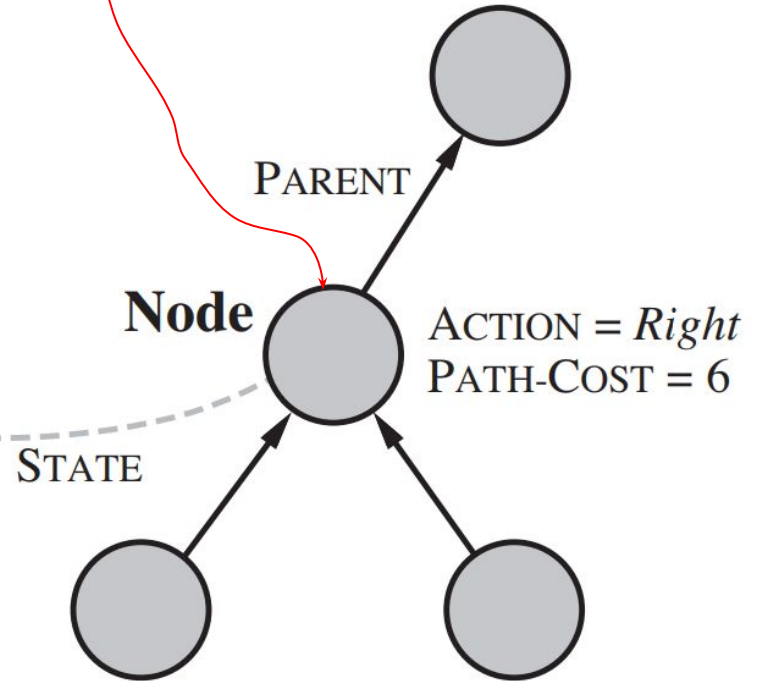


(c)

Programming: What should a node contain?



5	4	
6	1	8
7	3	2



Given the components for a parent node,

CHILD-NODE takes {parent node, action} and returns the resulting child node:

```
function CHILD-NODE(problem, parent, action) returns a node
return a node with
    STATE = problem.RESULT(parent.STATE, action),
    PARENT = parent, ACTION = action,
    PATH-COST = parent.PATH-COST + problem.STEP-COST(parent.STATE, action)
```

## Frontier:

Store such that the search algorithm can easily choose the next node to expand according to its preferred strategy.

The appropriate data structure for this is a queue. The operations on a queue are as follows:

- `EMPTY?(queue)`: returns `TRUE` only if there are no more elements in the queue.
- `POP(queue)`: removes the first element of the queue and returns it.
- `INSERT(element, queue)`: inserts an element and returns the resulting queue.

## Frontier:

Store such that the search algorithm can easily choose the next node to expand according to its preferred strategy.

Queue types based on how elements can be popped out

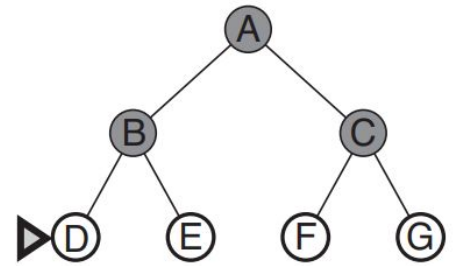
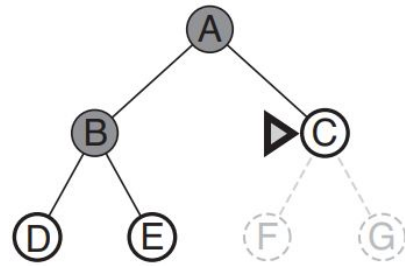
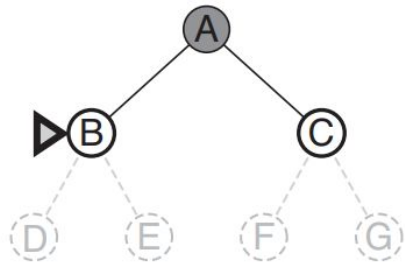
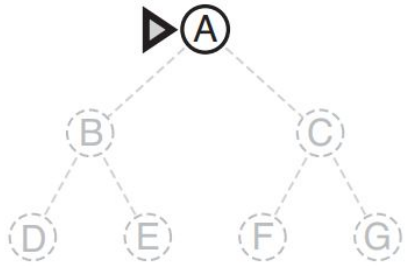
- FIFO Queue
- LIFO Queue
- Priority Queue

## Expansion set:

Needs to be regularly queried for checking if a state has been explored.

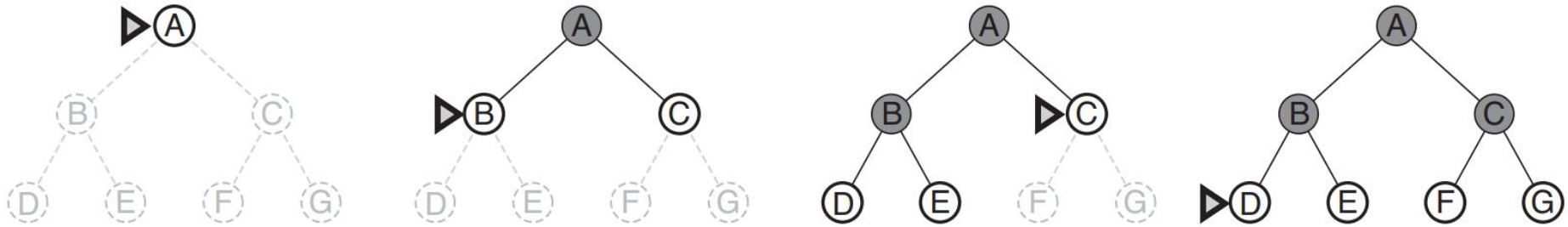
- Efficient storage, lookup and expansion
- Hash tables can be used
- Canonical form:
  - Bit vector representation
  - sorted /ordered list

# Breadth First search





# Breadth First search



Frontier operates as a FIFO queue

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

*node* ← a node with STATE = *problem*.INITIAL-STATE, PATH-COST = 0

**if** *problem*.GOAL-TEST(*node*.STATE) **then return** SOLUTION(*node*)

*frontier* ← a FIFO queue with *node* as the only element

*explored* ← an empty set

**loop do**

**if** EMPTY?(*frontier*) **then return** failure

*node* ← POP(*frontier*) /\* chooses the shallowest node in *frontier* \*/

add *node*.STATE to *explored*

**for each** *action* **in** *problem*.ACTIONS(*node*.STATE) **do**

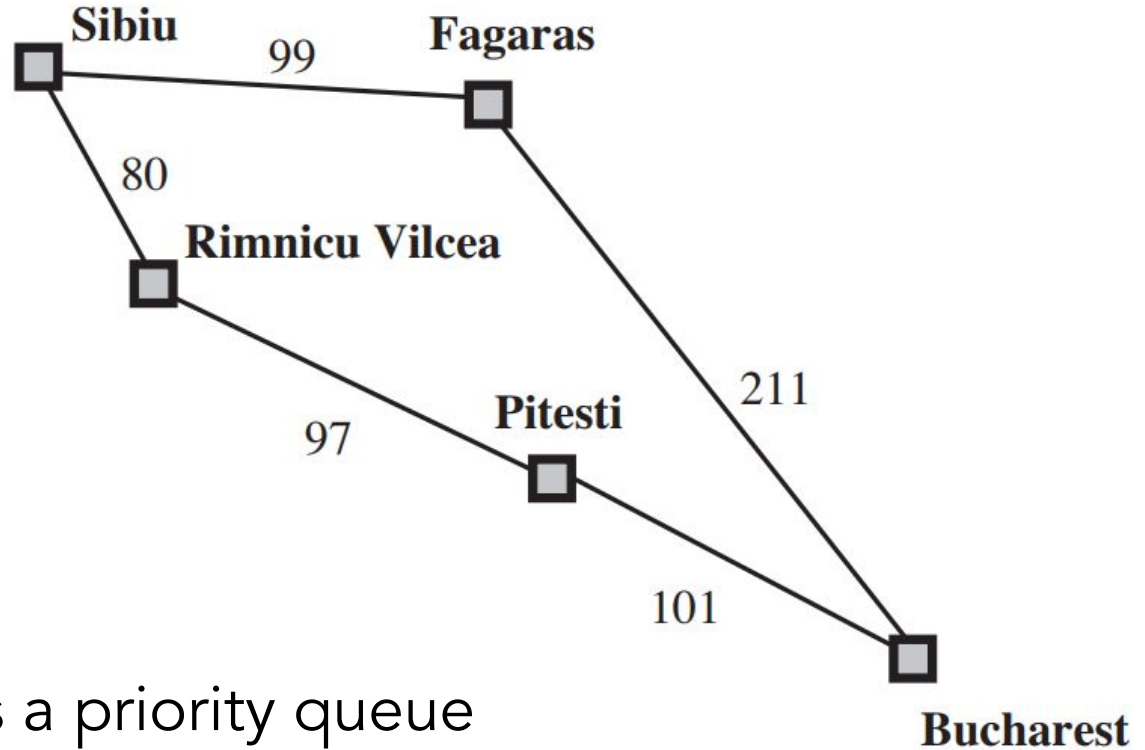
*child* ← CHILD-NODE(*problem*, *node*, *action*)

**if** *child*.STATE is not in *explored* or *frontier* **then**

**if** *problem*.GOAL-TEST(*child*.STATE) **then return** SOLUTION(*child*)

*frontier* ← INSERT(*child*, *frontier*)

## Uniform cost search



Introduces the notion of cost.

Frontier operates as a priority queue

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

*node* ← a node with STATE = *problem*.INITIAL-STATE, PATH-COST = 0

*frontier* ← a priority queue ordered by PATH-COST, with *node* as the only element

*explored* ← an empty set

**loop do**

**if** EMPTY?(*frontier*) **then return** failure

*node* ← POP(*frontier*) /\* chooses the lowest-cost node in *frontier* \*/

**if** *problem*.GOAL-TEST(*node*.STATE) **then return** SOLUTION(*node*)

add *node*.STATE to *explored*

**for each** *action* **in** *problem*.ACTIONS(*node*.STATE) **do**

*child* ← CHILD-NODE(*problem*, *node*, *action*)

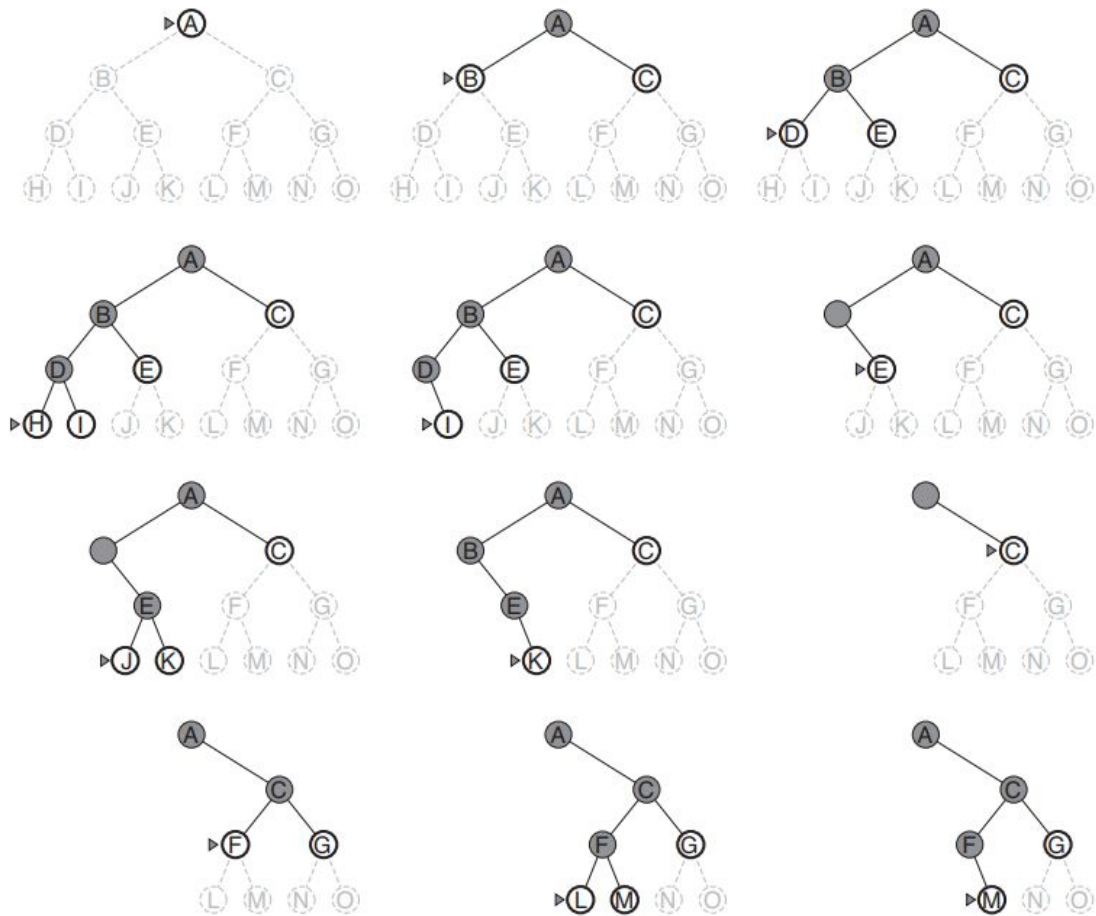
**if** *child*.STATE is not in *explored* or *frontier* **then**

*frontier* ← INSERT(*child*, *frontier*)

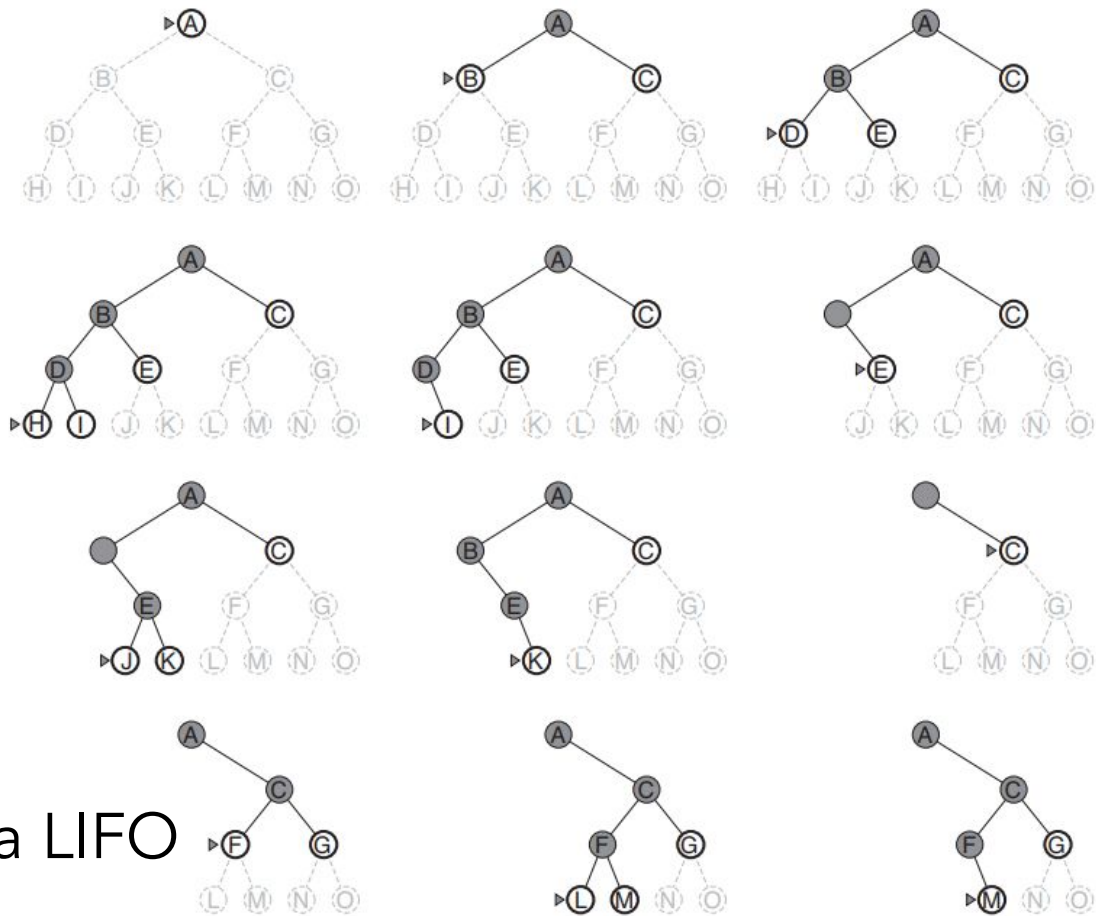
**else if** *child*.STATE is in *frontier* with higher PATH-COST **then**

replace that *frontier* node with *child*

# Depth First Search



# Depth First Search



Frontier operates as a LIFO queue

## Depth Limited Search

```
function DEPTH-LIMITED-SEARCH(problem, limit) returns a solution, or failure/cutoff  
  return RECURSIVE-DLS(MAKE-NODE(problem.INITIAL-STATE), problem, limit)
```

```
function RECURSIVE-DLS(node, problem, limit) returns a solution, or failure/cutoff  
  if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
```

```
  else if limit = 0 then return cutoff
```

```
  else
```

```
    cutoff_occurred? ← false
```

```
    for each action in problem.ACTIONS(node.STATE) do
```

```
      child ← CHILD-NODE(problem, node, action)
```

```
      result ← RECURSIVE-DLS(child, problem, limit - 1)
```

```
      if result = cutoff then cutoff_occurred? ← true
```

```
      else if result ≠ failure then return result
```

```
    if cutoff_occurred? then return cutoff else return failure
```

# Iterative Deepening Depth-First Search

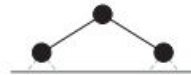
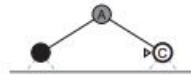
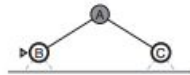
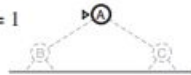
```
function ITERATIVE-DEEPENING-SEARCH(problem) returns a solution, or failure
  for depth = 0 to  $\infty$  do
    result  $\leftarrow$  DEPTH-LIMITED-SEARCH(problem, depth)
    if result  $\neq$  cutoff then return result
```



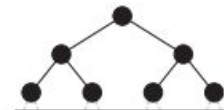
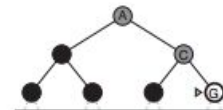
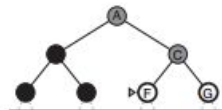
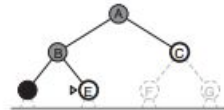
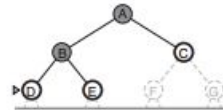
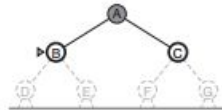
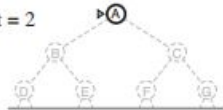
Limit = 0



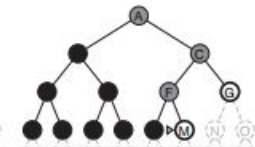
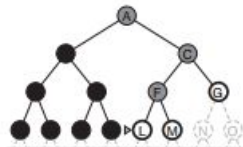
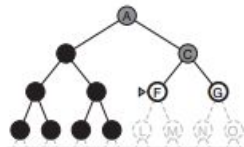
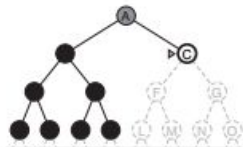
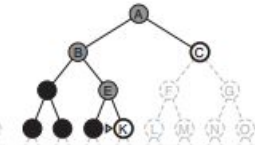
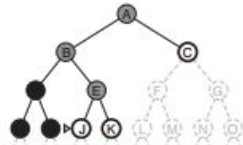
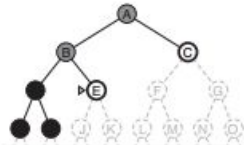
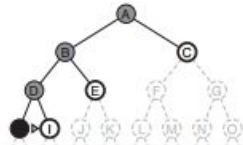
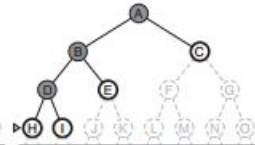
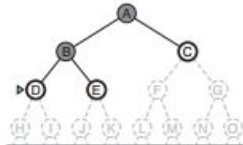
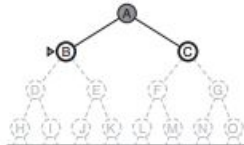
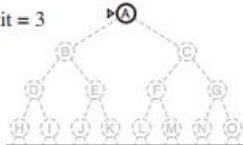
Limit = 1



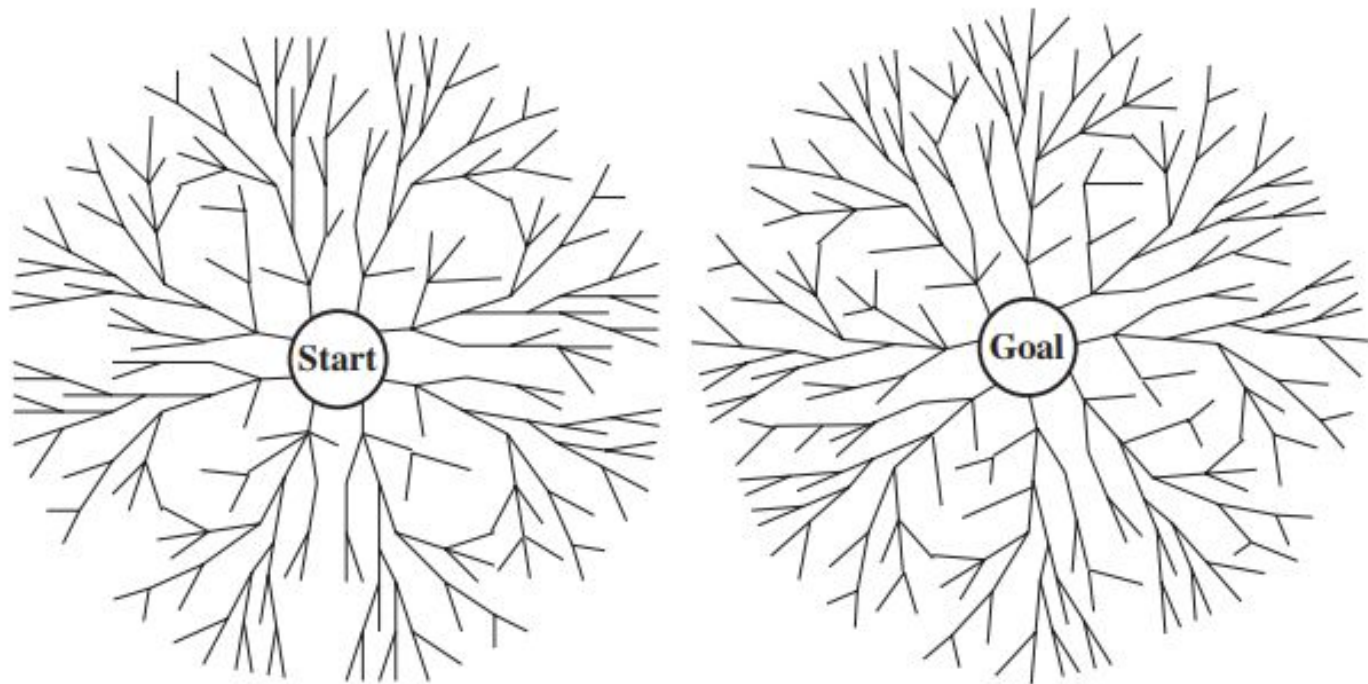
Limit = 2



Limit = 3



## Bi-directional Search



# Comparison of Search Strategies

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening	Bidirectional (if applicable)
Complete?	Yes <sup>a</sup>	Yes <sup>a,b</sup>	No	No	Yes <sup>a</sup>	Yes <sup>a,d</sup>
Time	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^\ell)$	$O(b^d)$	$O(b^{d/2})$
Space	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(b\ell)$	$O(bd)$	$O(b^{d/2})$
Optimal?	Yes <sup>c</sup>	Yes	No	No	Yes <sup>c</sup>	Yes <sup>c,d</sup>

- **Completeness:** Is the algorithm guaranteed to find a solution when there is one?
- **Optimality:** Does the strategy find the optimal solution,
- **Time complexity:** How long does it take to find a solution?
- **Space complexity:** How much memory is needed to perform the search?

b: branching factor or maximum number of successors for any node

d: the depth of the shallowest goal

m: maximum length of any path in the state space