

CSE 604

# Artificial Intelligence

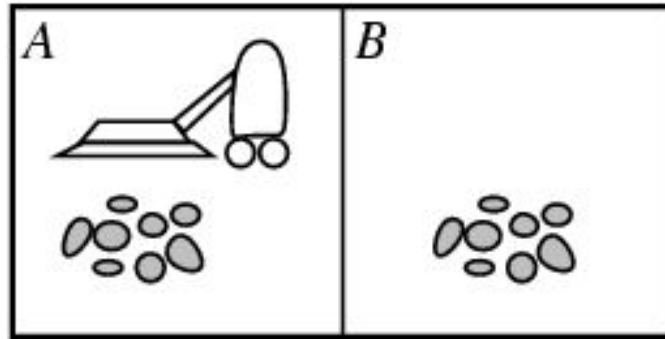
## Chapter 3: Solving Problems by Searching

Adapted from slides available in Russell & Norvig's textbook webpage

**Dr. Ahmedul Kabir**

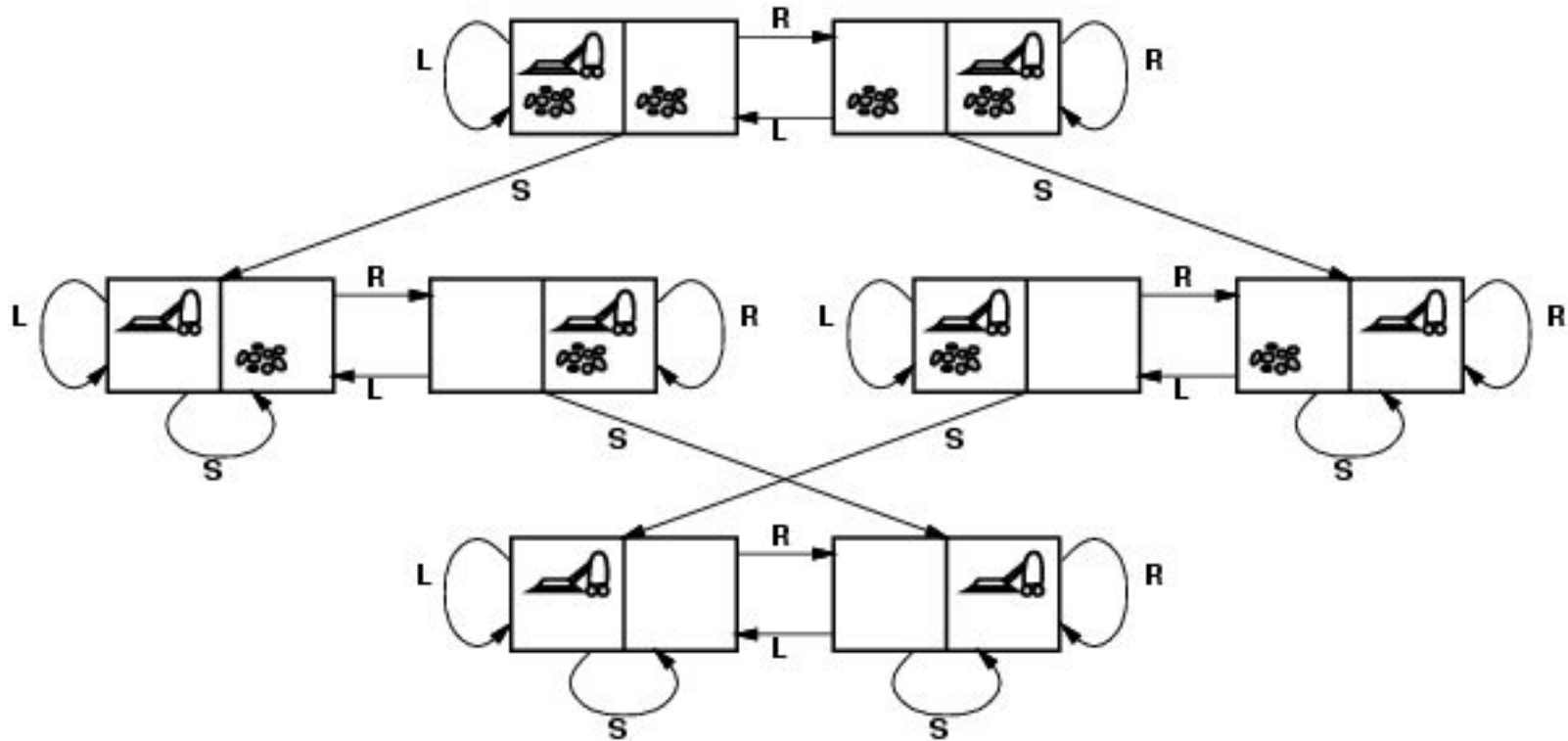


# Remember the Vacuum-cleaner world?



- **Percepts:** location and contents, e.g., [A, Dirty]
- **Actions:** *Left, Right, Suck*

# Vacuum world state space graph

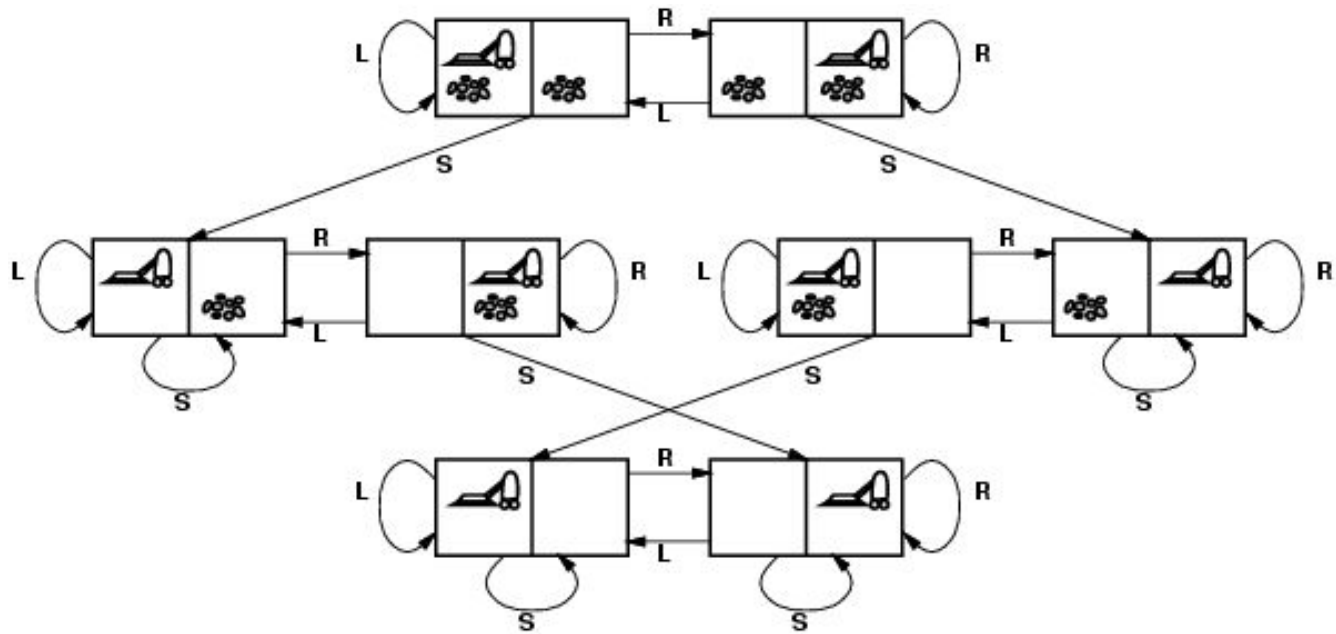


**State space:** Set of all reachable states. In state space graph,  
nodes/vertices = states, links/edges = actions

# Formulation of a Problem

- A **Problem** is defined by the following items:
  - Set of **states** the agent can be in, with a designated **initial state**
  - Set of **actions** available to the agent
  - **Transition model** describing what each action does (maps a  $\langle \text{state}, \text{action} \rangle$  pair to a state)
  - **Goal test** which determines if a given state is a goal state
  - A **path cost** function that assigns a numeric cost to each path

# Vacuum world state space graph



- states? binary dirt and robot location. Any state can be initial state
- actions? *Left, Right, Suck*
- Transition model? As seen in the state space graph
- goal test? no dirt at all locations
- path cost? 1 per action

# 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
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8	3	1

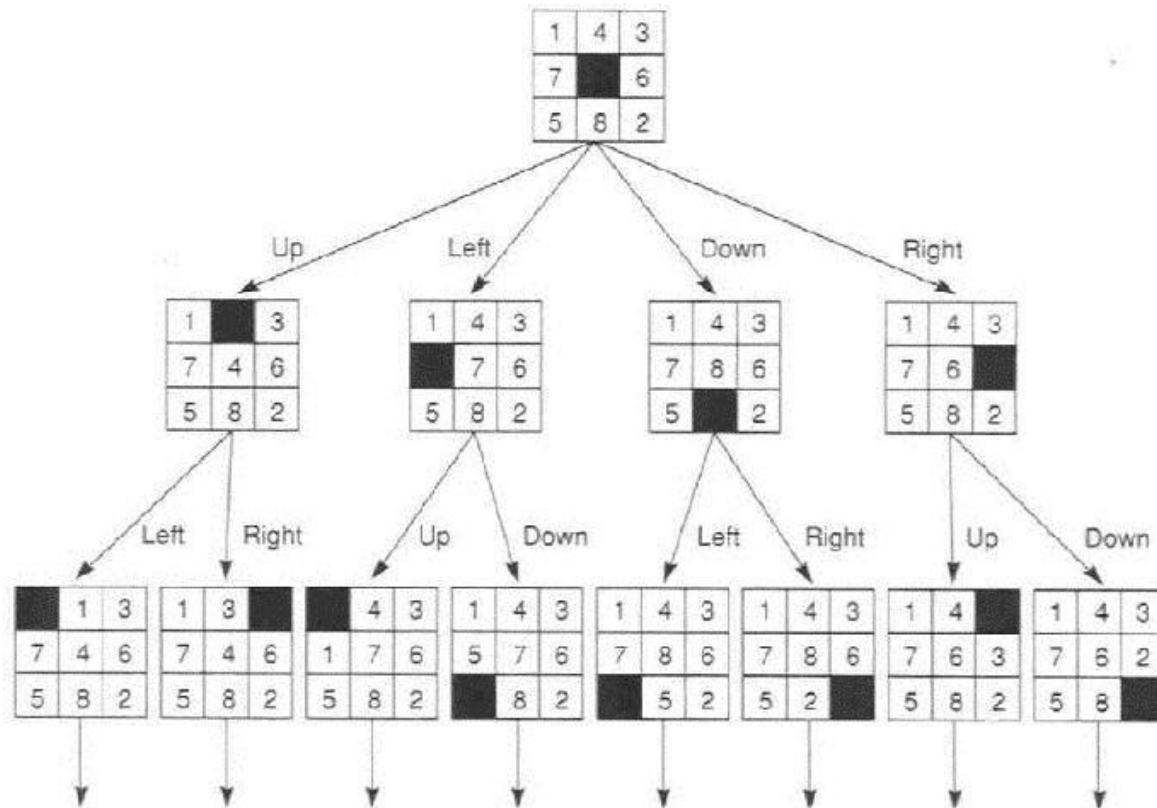
Start State

	1	2
3	4	5
6	7	8

Goal State

- states? locations of tiles
- actions? move blank left, right, up, down
- goal test? = goal state (given)
- path cost? 1 per move

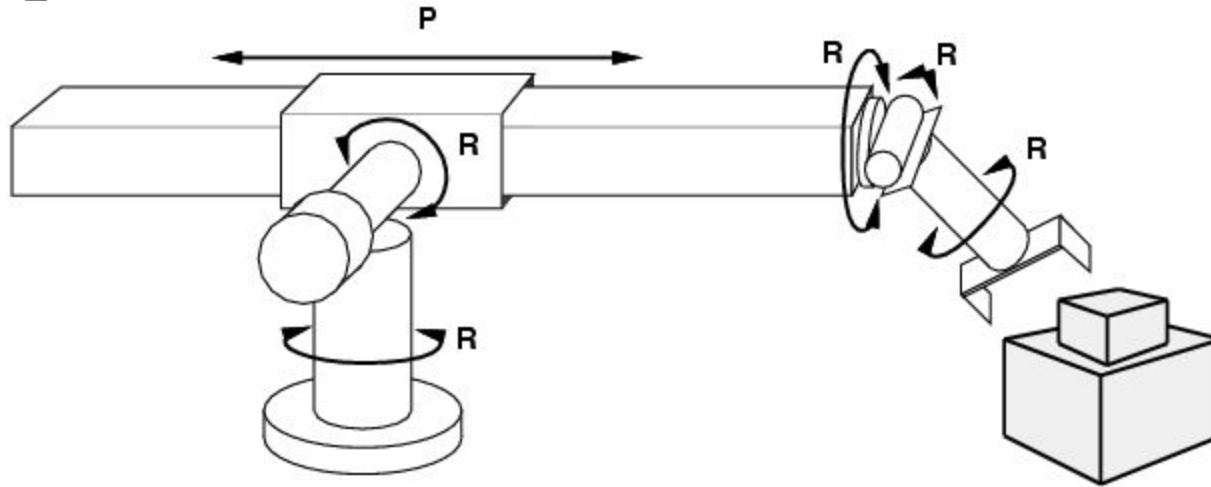
# Example: The 8-puzzle



Partial state space graph

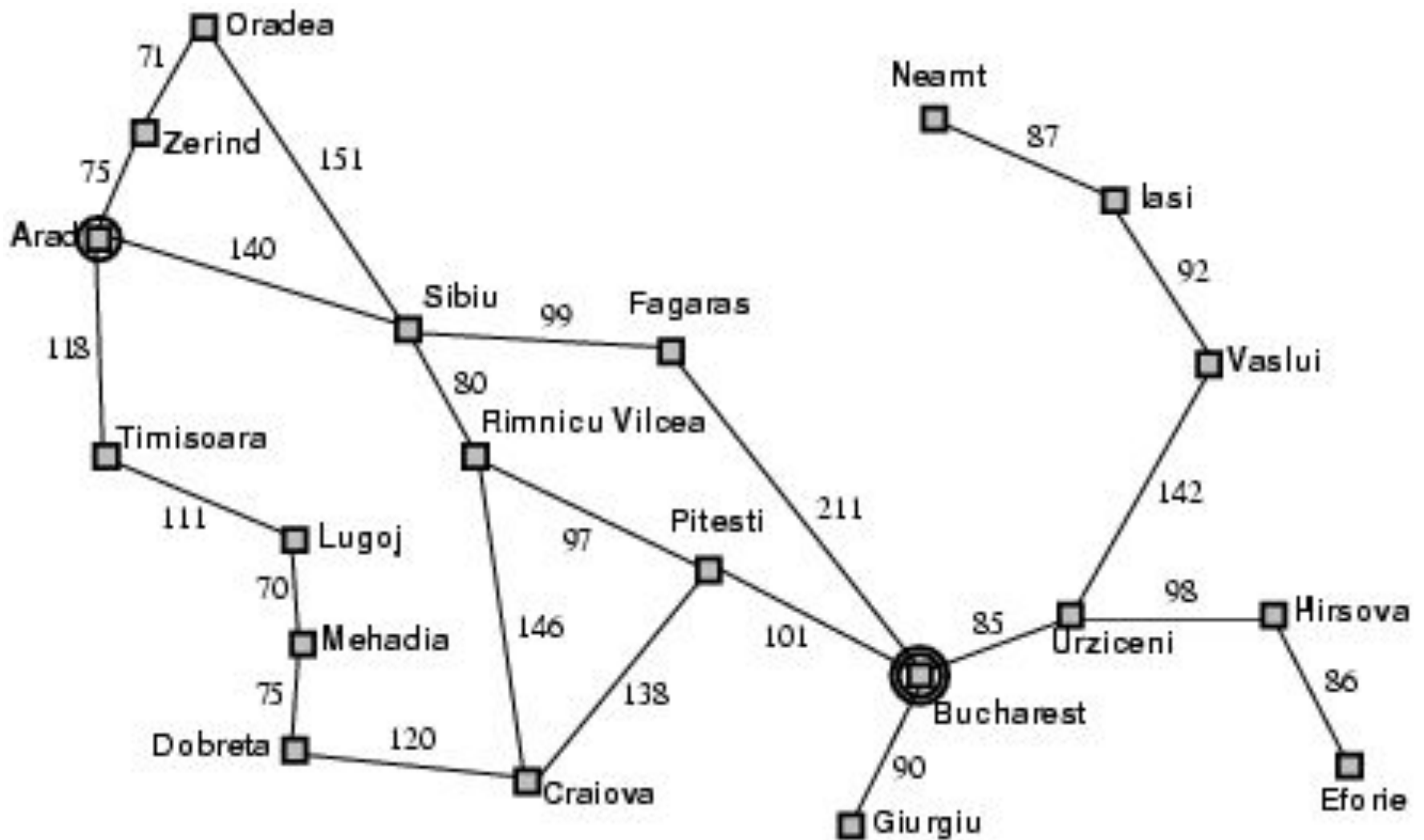


# Example: robotic assembly



- states?: real-valued coordinates of robot joint angles  
parts of the object to be assembled
- actions?: continuous motions of robot joints
- goal test?: complete assembly
- path cost?: time to execute

# Example: Romania



# Search strategies

- A search strategy is defined by picking the **order of node expansion**
- Strategies are evaluated along the following dimensions:
  - **completeness**: does it always find a solution if one exists?
  - **time complexity**: number of nodes generated
  - **space complexity**: maximum number of nodes in memory
  - **optimality**: does it always find a least-cost solution?
- Time and space complexity are measured in terms of
  - $b$ : maximum branching factor of the search tree
  - $d$ : depth of the least-cost solution
  - $m$ : maximum depth of the state space (may be  $\infty$ )

# Uninformed search strategies

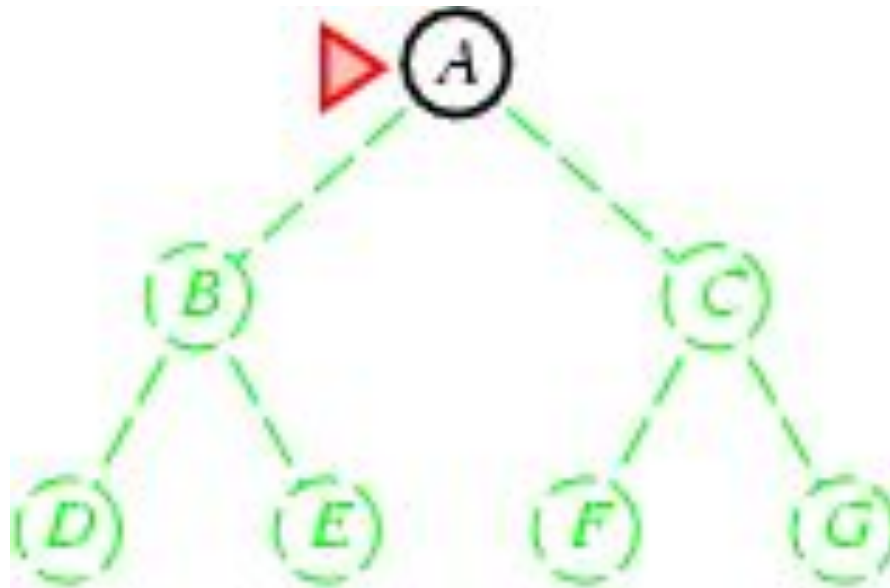
- Uninformed search strategies use only the information available in the problem definition
  - Breadth-first search
  - Uniform-cost search
  - Depth-first search
  - Depth-limited search
  - Iterative deepening search

# Basic concept

- **Frontier** (or fringe): The set of all leaf nodes available for expansion at any given point
- The basics of each algorithm:
  - Start from initial node
  - Expand adjacent nodes and put them in the frontier
  - Choose the next node from the frontier for expansion
  - Repeat until goal is found, or some ending criteria is met
- The algorithms differ in the way they choose the next node from the frontier

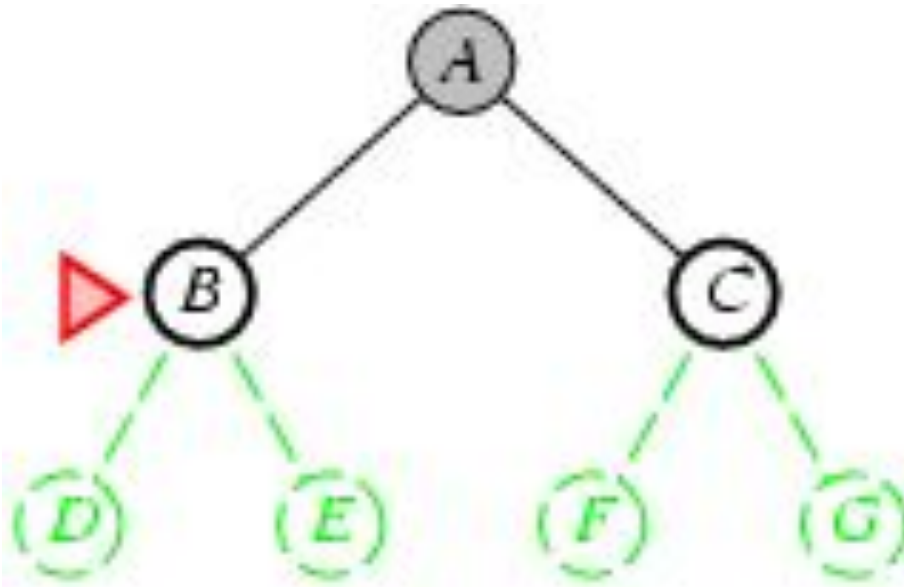
# Breadth-first search

- Expand shallowest unexpanded node
- Implementation:
  - *frontier* is a FIFO queue, i.e., new successors go at end



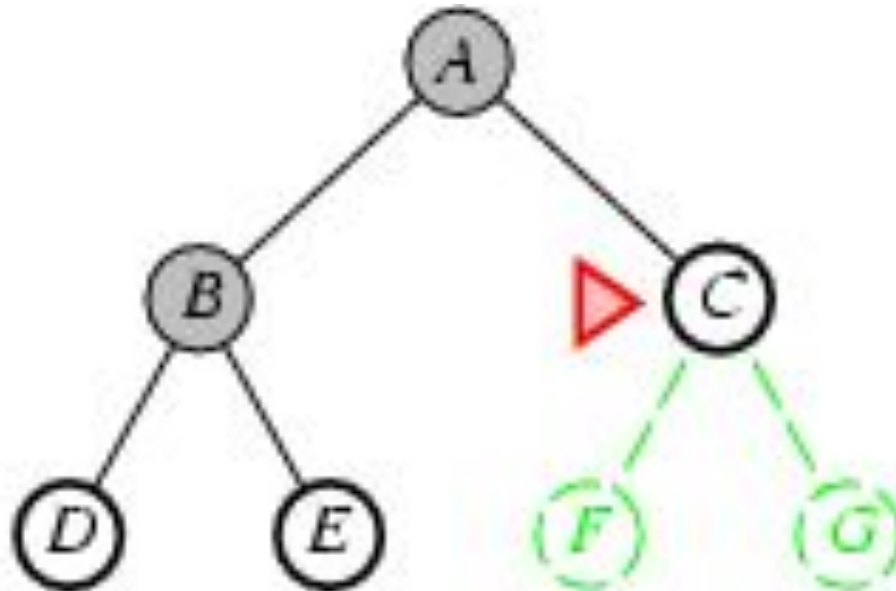
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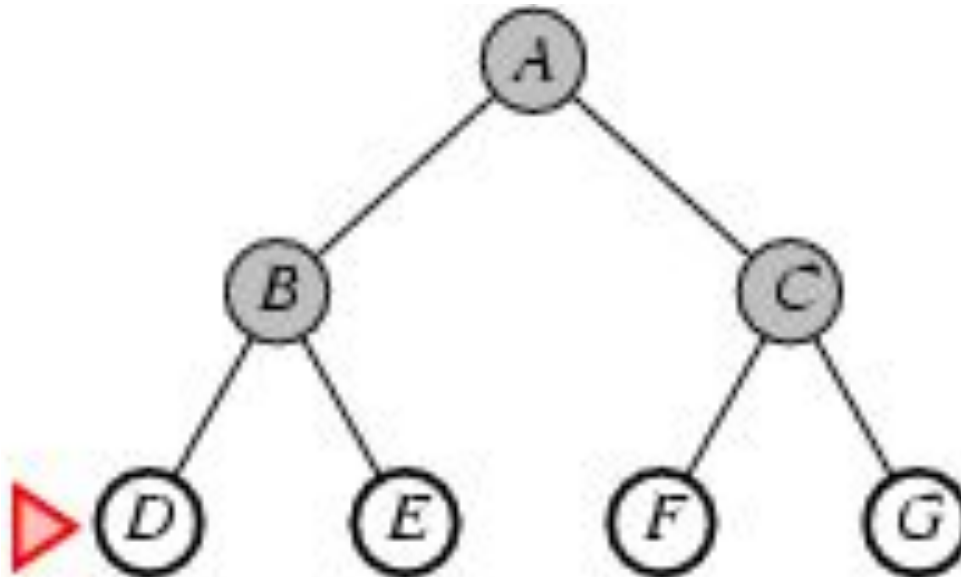
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# Properties of breadth-first search

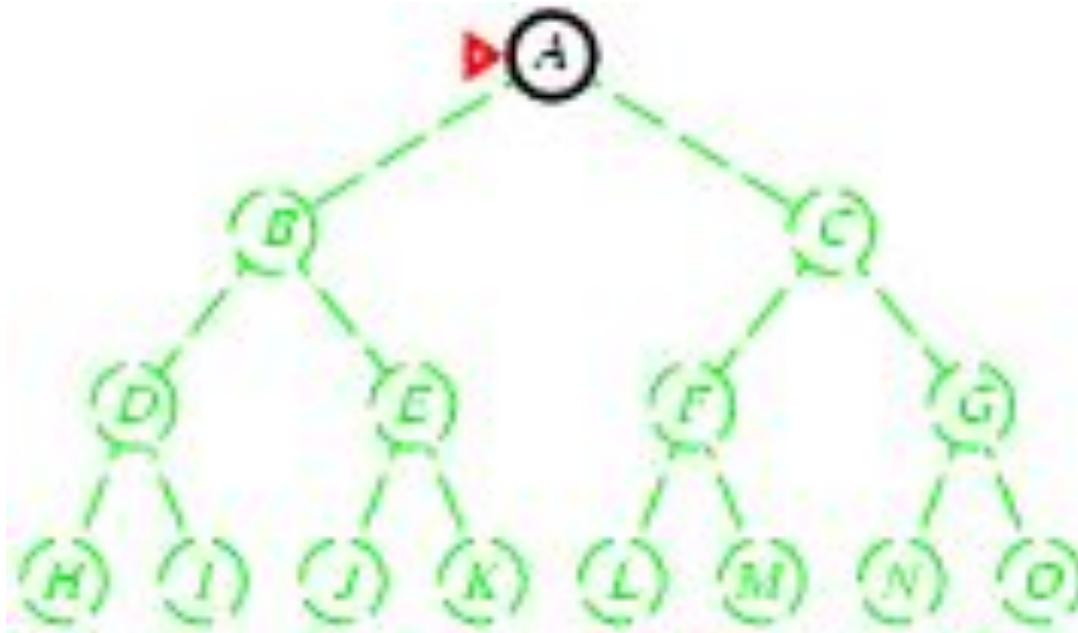
- Complete? Yes (if  $b$  is finite)
- Time?  $1 + b + b^2 + b^3 + \dots + b^d = O(b^d)$
- Space?  $O(b^d)$  (keeps every node in memory)
- Optimal? Yes (if cost = 1 per step)
- Space is the bigger problem (more than time)

# Uniform-cost search

- Expand least-cost unexpanded node
- **Implementation:**
  - *frontier* = queue ordered by path cost
- Equivalent to breadth-first if step costs all equal
- Complete? Yes, if step cost  $\geq \epsilon$
- Time? # of nodes with  $g \leq$  cost of optimal solution,  $O(b^{\lceil C^*/\epsilon \rceil})$  where  $C^*$  is the cost of the optimal solution
- Space? # of nodes with  $g \leq$  cost of optimal solution,  $O(b^{\lceil C^*/\epsilon \rceil})$
- Optimal? Yes – nodes expanded in increasing order of  $g(n)$

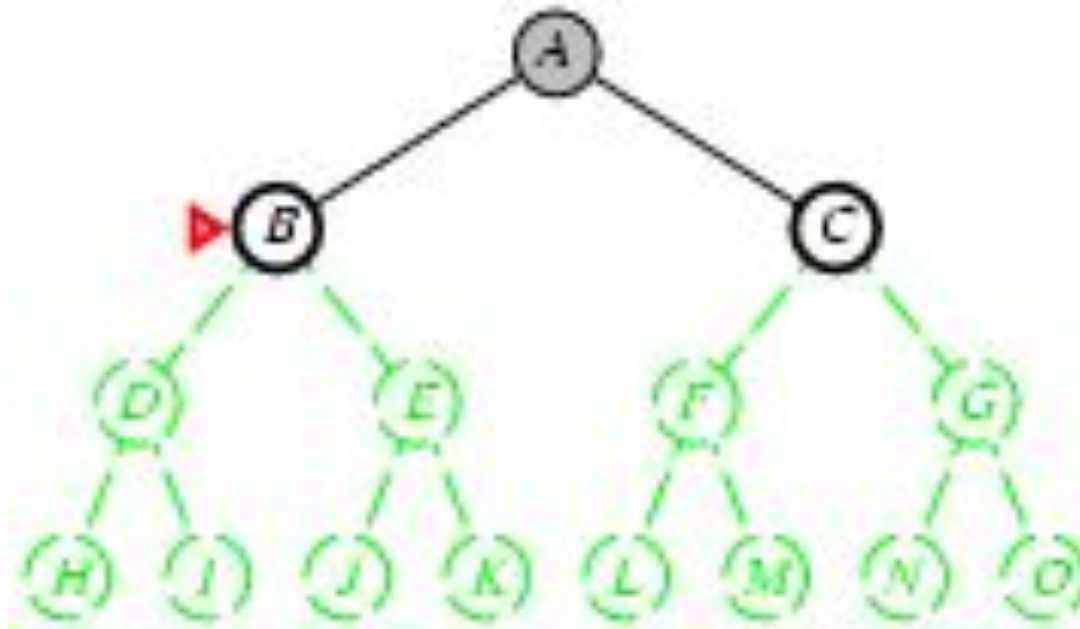
# Depth-first search

- Expand deepest unexpanded node
- **Implementation:**
  - *frontier* = LIFO stack, i.e., put successors at front



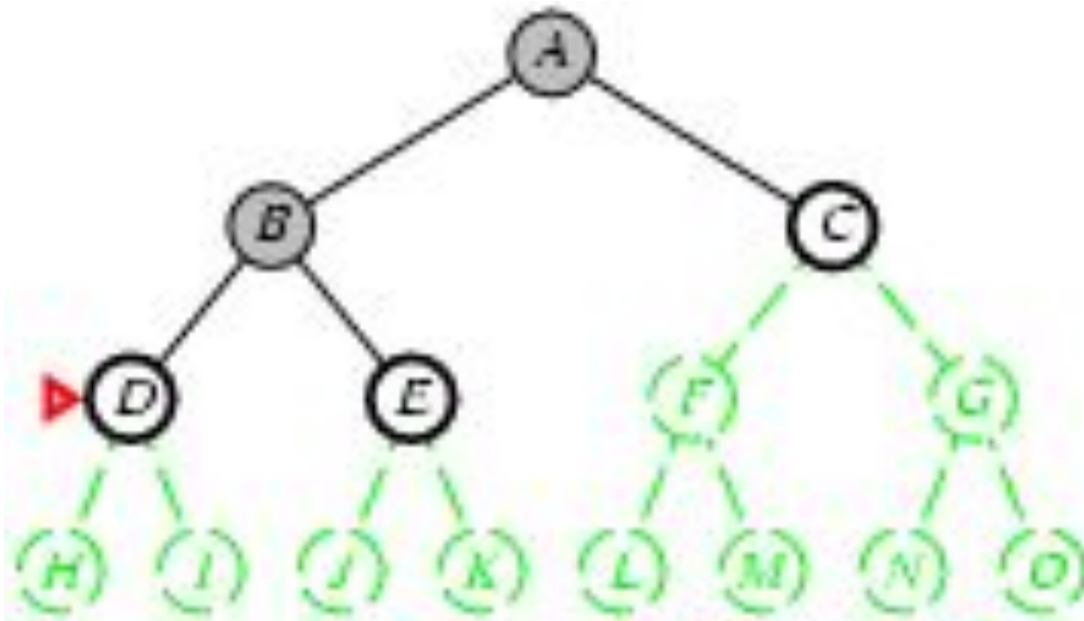
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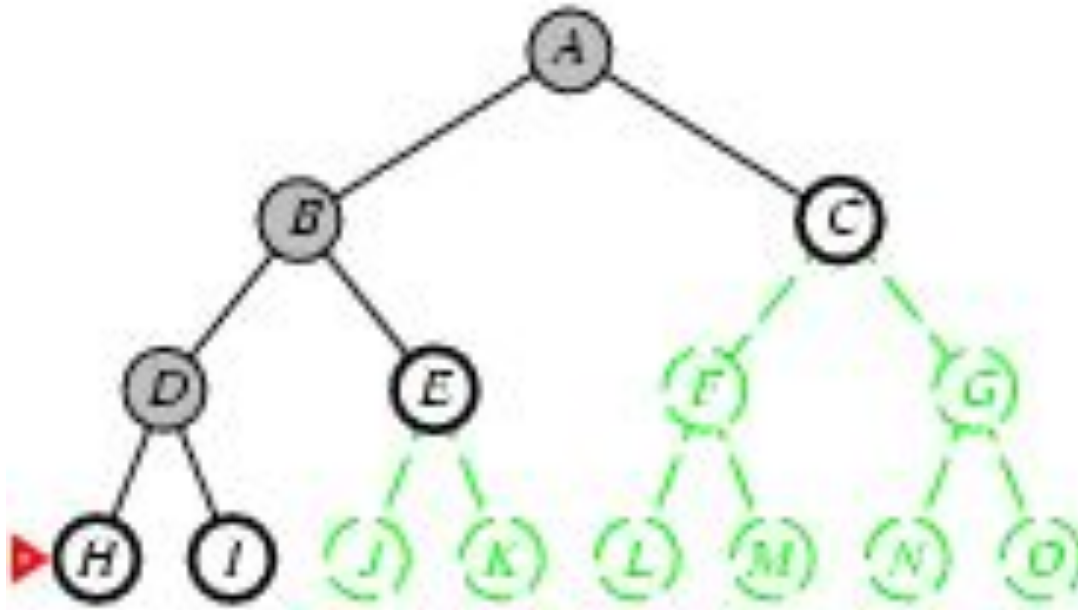
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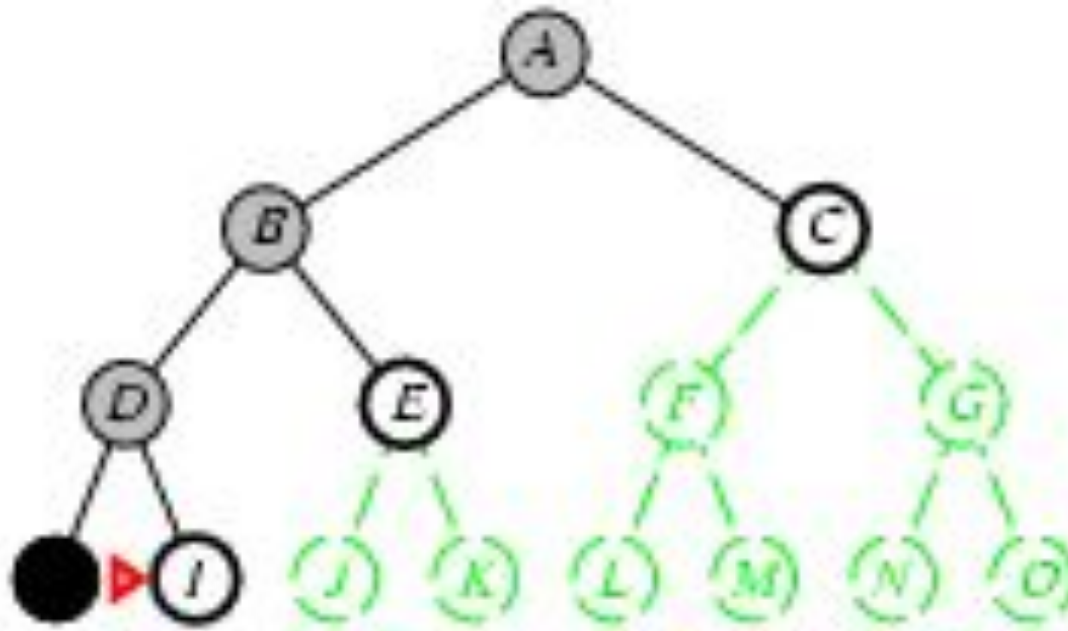
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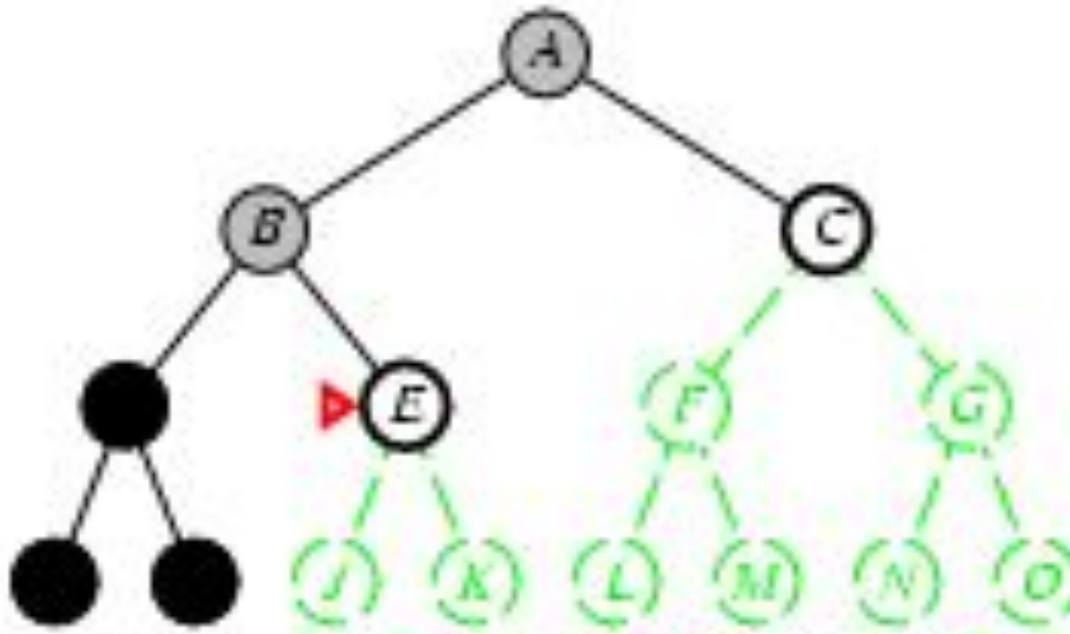
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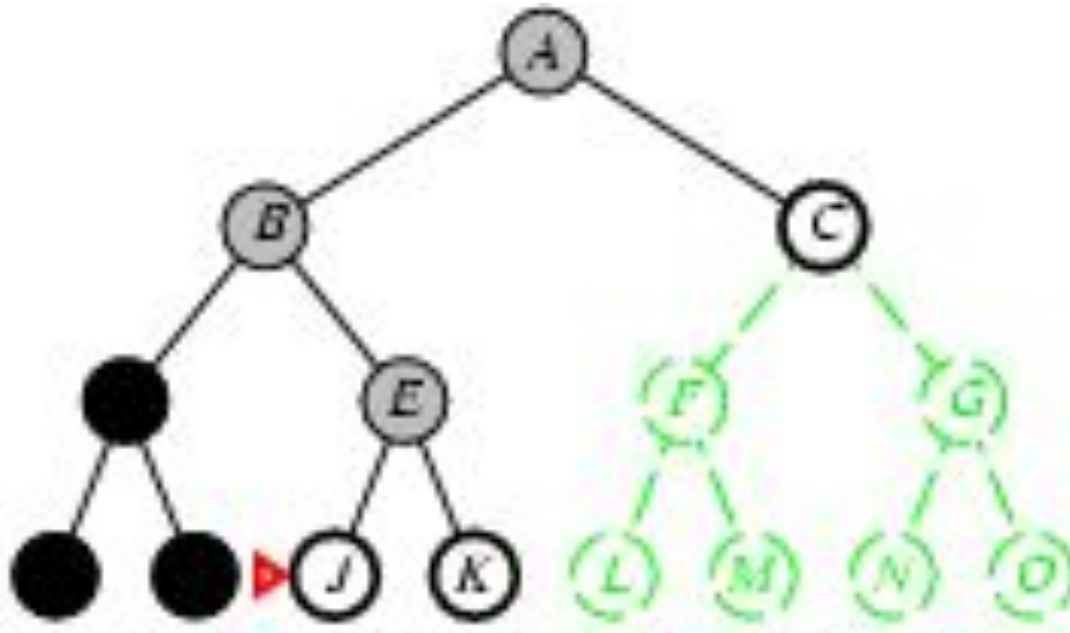
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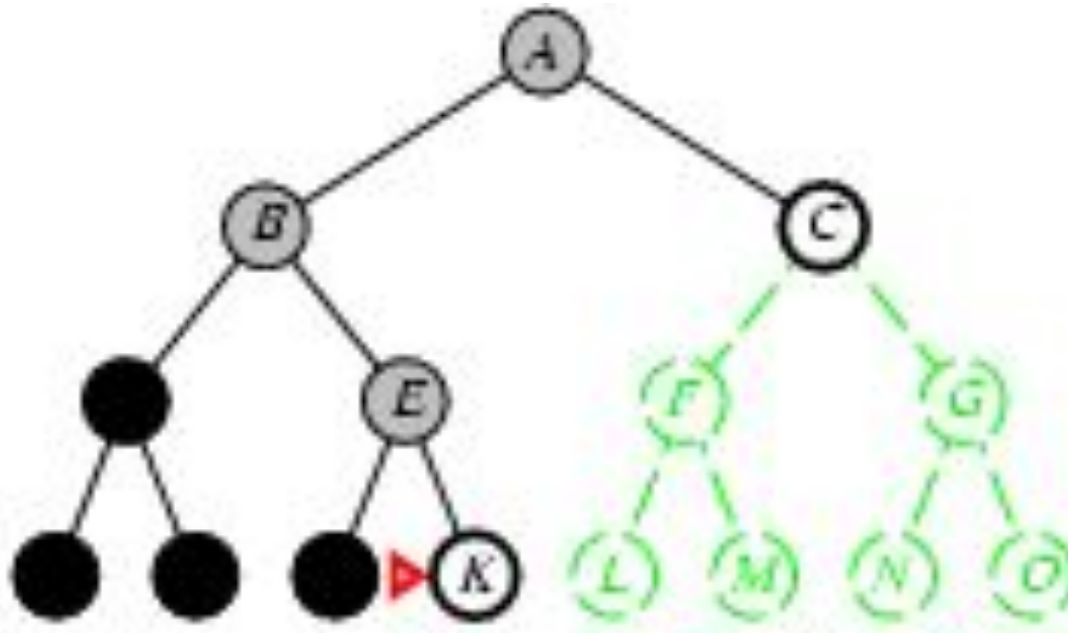
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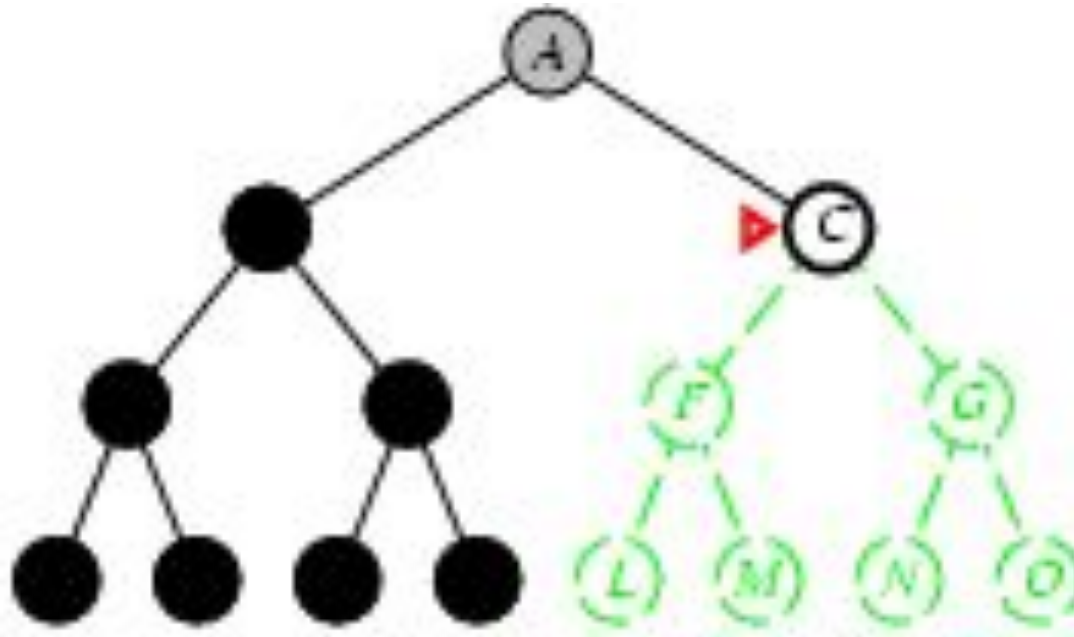
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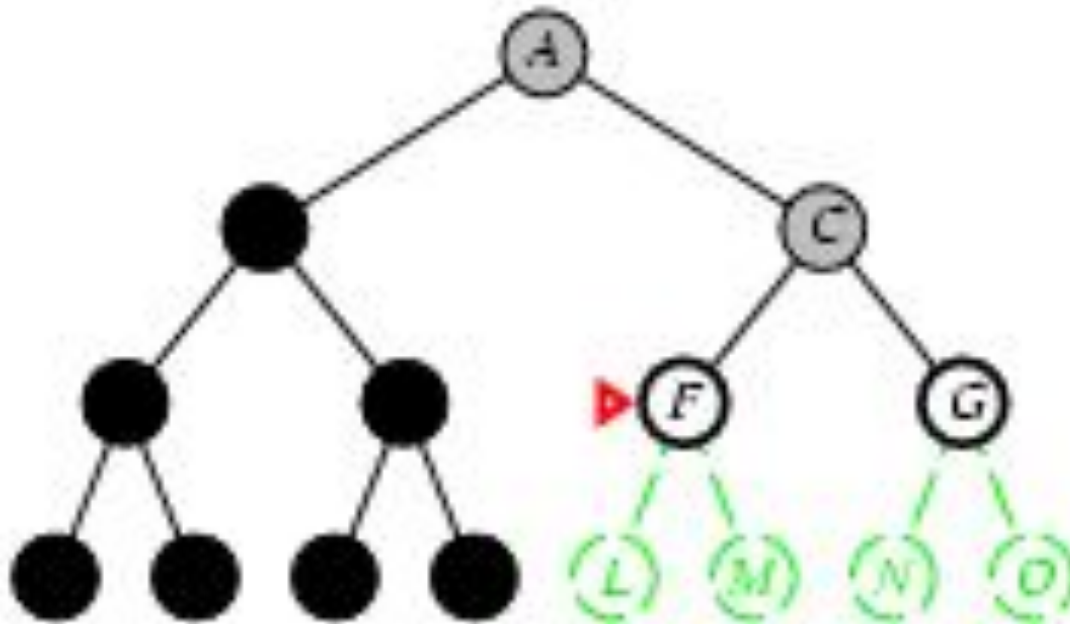
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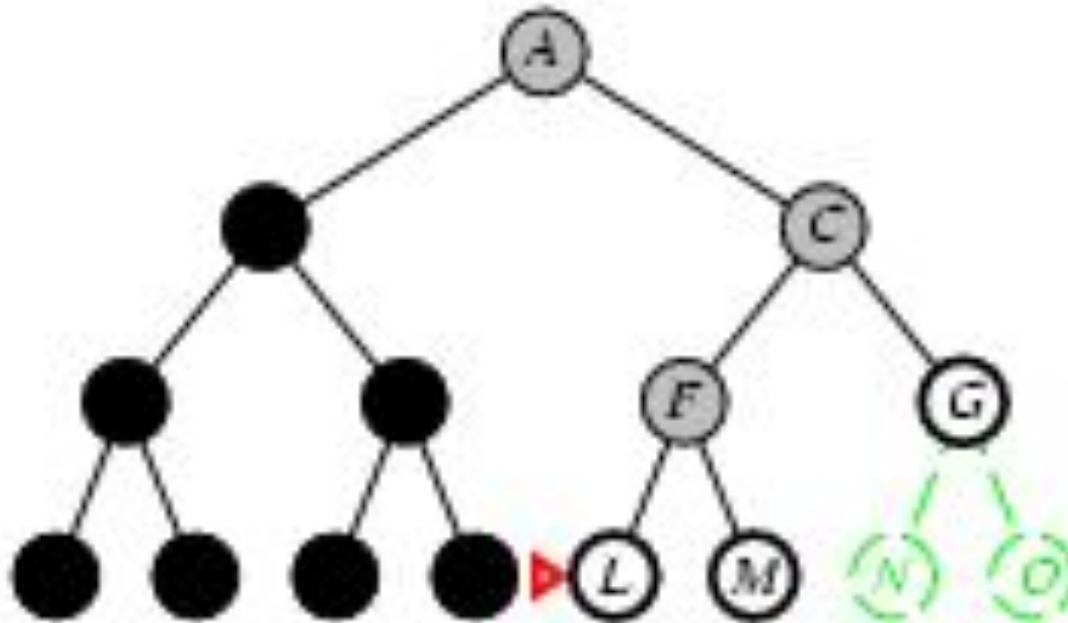
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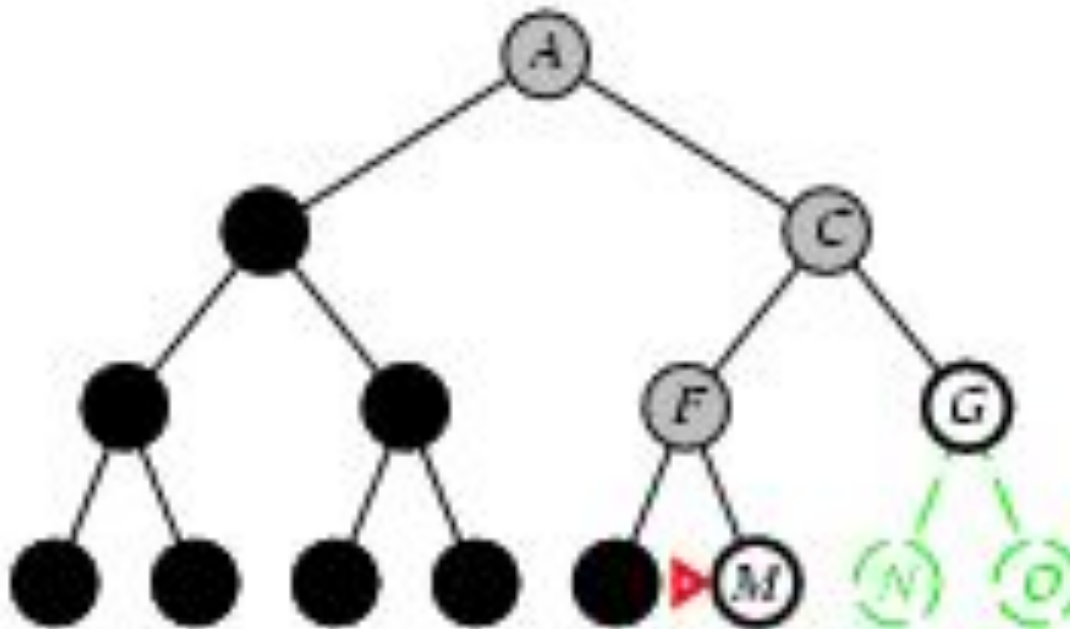
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# Properties of depth-first search

- Complete? No: fails in infinite-depth spaces, spaces with loops
  - Modify to avoid repeated states along path  
complete in finite spaces
- Time?  $O(b^m)$ : terrible if  $m$  is much larger than  $d$ 
  - but if solutions are dense, may be much faster than breadth-first
- Space?  $O(bm)$ , i.e., linear space!
- Optimal? No



# Depth-limited search

= depth-first search with depth limit  $l$ , i.e., nodes at depth  $l$  have no successors

- Complete? No
- Time?  $O(b^l)$
- Space?  $O(bl)$
- Optimal? No

# Iterative deepening search

= depth-limited search on repeat!

Limit  $l$  is increased at each iteration until goal is found

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

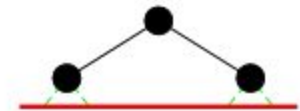
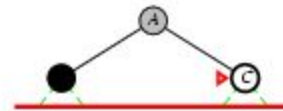
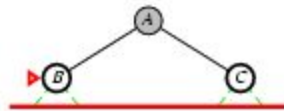
# Iterative deepening search $l = 0$

Limit = 0



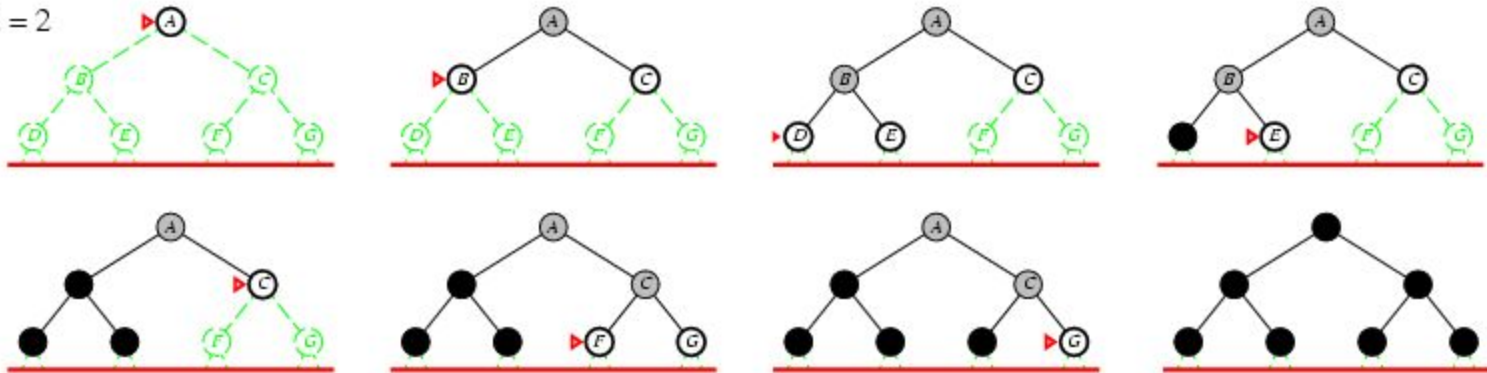
# Iterative deepening search $l = 1$

Limit = 1



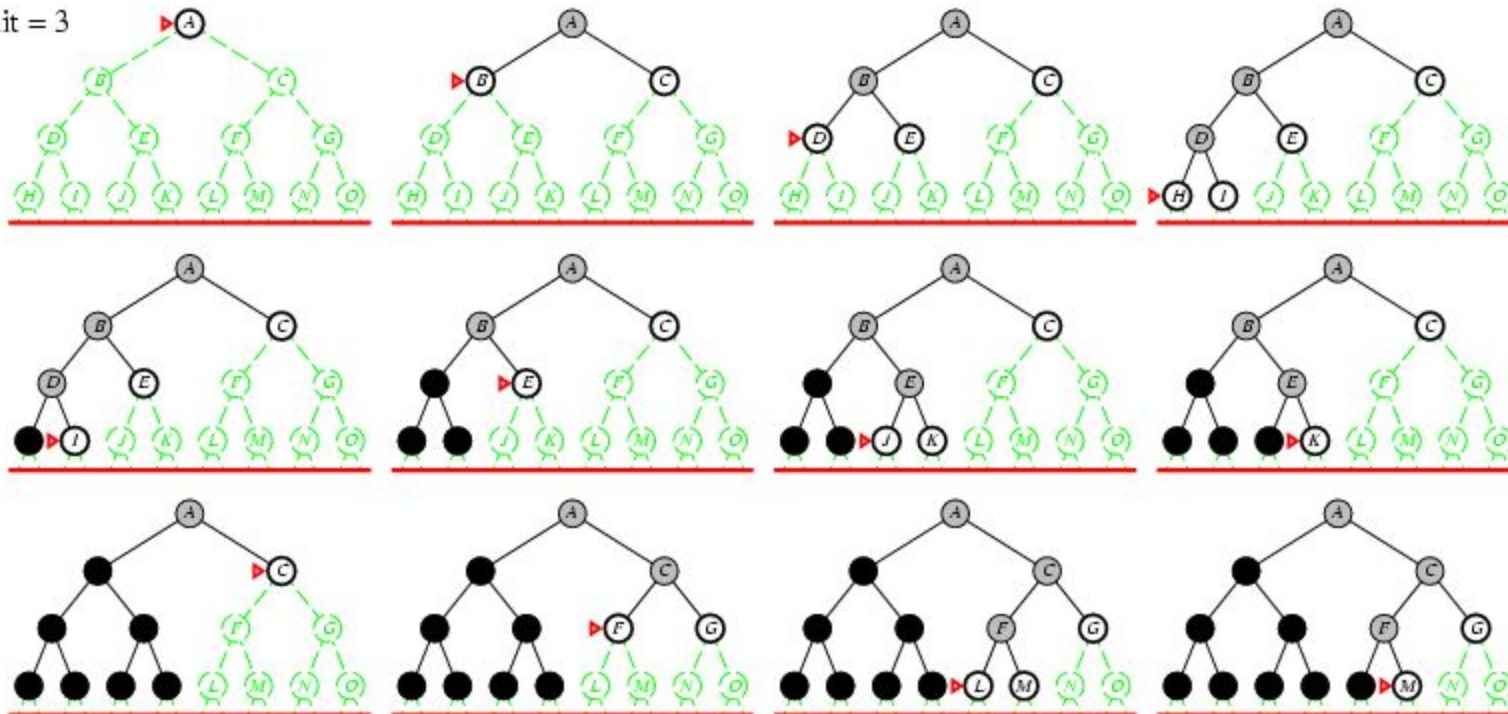
# Iterative deepening search $l=2$

Limit = 2



# Iterative deepening search $l = 3$

Limit = 3



# Properties of iterative deepening search

- Complete? Yes
- Time?  $(d+1)b^0 + d b^1 + (d-1)b^2 + \dots + b^d = O(b^d)$
- Space?  $O(bd)$
- Optimal? Yes, if step cost = 1

# Summary of algorithms

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