

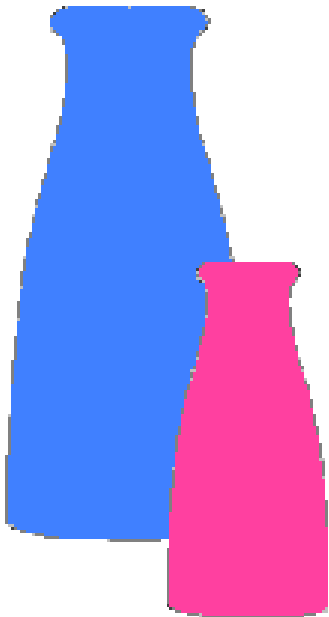
# State-Space Searches

# State spaces

- A state space consists of
  - A (possibly infinite) set of states
    - The start state represents the initial problem
    - Each state represents some configuration reachable from the start state
    - Some states may be goal states (solutions)
  - A set of operators
    - Applying an operator to a state transforms it to another state in the state space
    - Not all operators are applicable to all states
- State spaces are used extensively in Artificial Intelligence (AI)

# State Space: Water Jug Problem

“You are given two jugs, a 4-litre one and a 3-litre one. Neither has any measuring markers on it. There is a pump that can be used to fill the jugs with water. How can you get exactly 2 litres of water into 4-litre jug.”



# State Space: Water Jug Problem

- State:  $(x, y)$

$x = 0, 1, 2, 3, \text{ or } 4$

$y = 0, 1, 2, 3$

- Start state:  $(0, 0)$ .
- Goal state:  $(2, n)$  for any  $n$ .
- Attempting to end up in a goal state.

# State Space: Water Jug Problem

1.  $(x, y) \rightarrow (4, y)$   
if  $x < 4$
2.  $(x, y) \rightarrow (x, 3)$   
if  $y < 3$
3.  $(x, y) \rightarrow (x - d, y)$   
if  $x > 0$
4.  $(x, y) \rightarrow (x, y - d)$   
if  $y > 0$

# State Space: Water Jug Problem

- 5.  $(x, y) \rightarrow (0, y)$   
if  $x > 0$
- 6.  $(x, y) \rightarrow (x, 0)$   
if  $y > 0$
- 7.  $(x, y) \rightarrow (4, y - (4 - x))$   
if  $x + y \geq 4, y > 0$
- 8.  $(x, y) \rightarrow (x - (3 - y), 3)$   
if  $x + y \geq 3, x > 0$

# State Space: Water Jug Problem

9.  $(x, y) \rightarrow (x + y, 0)$

if  $x + y \leq 4, y > 0$

10.  $(x, y) \rightarrow (0, x + y)$

if  $x + y \leq 3, x > 0$

11.  $(0, 2) \rightarrow (2, 0)$

12.  $(2, y) \rightarrow (0, y)$

# State Space Search: Water Jug Problem

1. current state =  $(0, 0)$
2. Loop until reaching the goal state  $(2, 0)$ 
  - Apply a rule whose left side matches the current state
  - Set the new current state to be the resulting state

$(0, 0)$

$(0, 3)$

$(3, 0)$

$(3, 3)$

$(4, 2)$


$(0, 2)$

$(2, 0)$




# Example 2: The 15-puzzle

Start state:

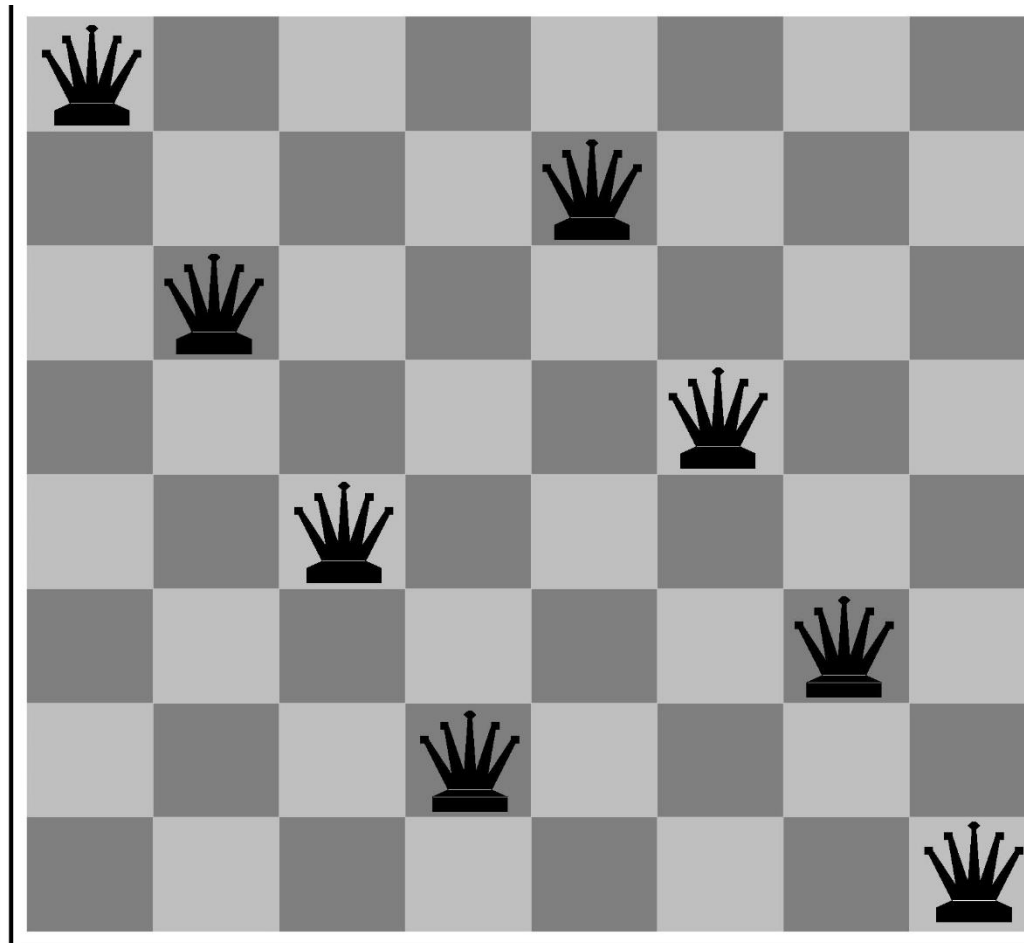
|    |   |    |    |
|----|---|----|----|
| 3  | 10  | 13 | 7  |
| 9  | 14  | 6  | 1  |
| 4  |  | 15 | 2  |
| 11 | 8   | 5  | 12 |

Goal state:

|    |    |    |   |
|----|----|----|---|
| 1  | 2  | 3  | 4   |
| 5  | 6  | 7  | 8   |
| 9  | 10 | 11 | 12  |
| 13 | 14 | 15 |  |

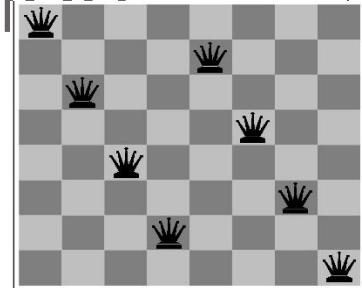
- The start state is some (almost) random configuration of the tiles
- The goal state is as shown
- Operators are
  - Move empty space up
  - Move empty space down
  - Move empty space right
  - Move empty space left
- Operators apply if not against edge

# 8 Queen Problem



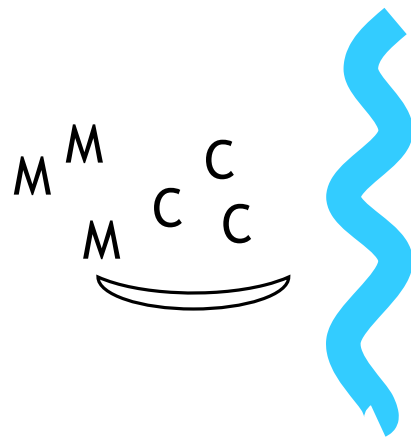
# State-Space problem formulation

- states? -any arrangement of  $n \leq 8$  queens  
-or arrangements of  $n \leq 8$  queens in leftmost  $n$  columns, 1 per column, such that no queen attacks any other.
- initial state? no queens on the board
- actions? -add queen to any empty square  
-or add queen to leftmost empty square such that it is not attacked by other queens.
- goal test? 8 queens on the board, none attacked.
- path cost? 1 per move

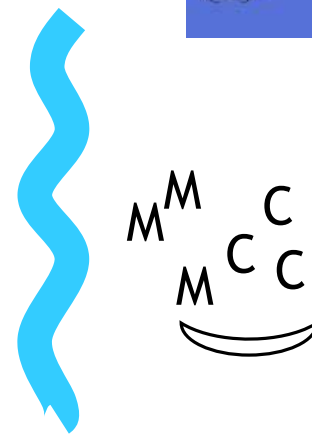


# Example 3: Missionaries and cannibals

- An old puzzle is the “Missionaries and cannibals” problem (in various guises)
- The missionaries and cannibals wish to cross a river
- They have a canoe that can hold two people
- It is unsafe to have cannibals outnumber missionaries



Initial state



Goal state

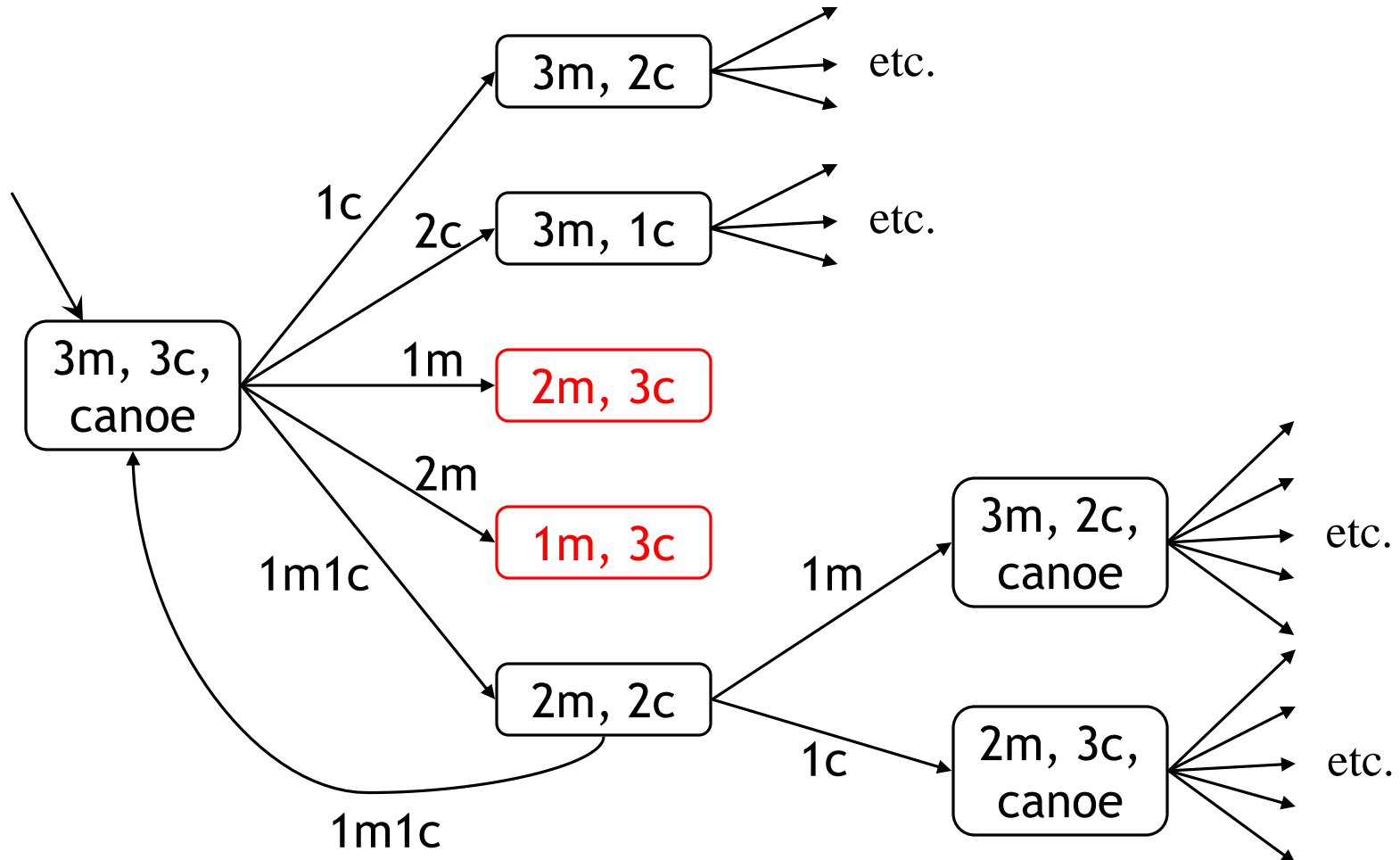
# States

- A *state* can be represented by the number of missionaries and cannibals on each side of the river
  - Initial state: **3m,3c,canoe / 0m,0c**
  - Goal state: **0m,0c / 3m,3c,canoe**
  - We assume that crossing the river is a simple procedure that always works (so we don't have to represent the canoe being in the middle of the river)
- However, this is redundant; we only need to represent how many missionaries/cannibals are on *one* side of the river
  - Initial state: **3m,3c,canoe**
  - Goal state: **0m,0c**

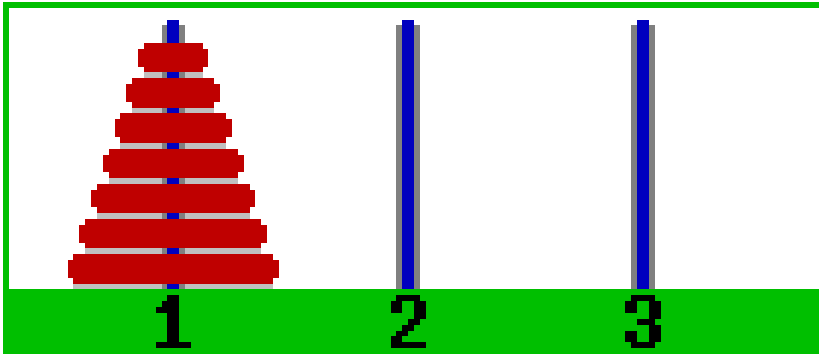
# Operations

- An *operation* takes us from one state to another
- Here are five possible operations:
  - Canoe takes 1 missionary across river (**1m**)
  - Canoe takes 1 cannibal across river (**1c**)
  - Canoe takes 2 missionaries across river (**2m**)
  - Canoe takes 2 cannibals across river (**2c**)
  - Canoe takes 1 missionary and 1 cannibal across river (**1m1c**)
- We don't have to specify “west to east” or “east to west” because only one of these will be possible at any given time

# The state space



# Example Problems – Towers of Hanoi



States: combinations of poles and disks

Operators: move disk x from pole y to pole z subject to constraints

- cannot move disk on top of smaller disk
- cannot move disk if other disks on top

Goal test: disks from largest (at bottom) to smallest on goal pole

Path cost: 1 per move

[Towers of Hanoi applet](#)



# Example Problems – Rubik's Cube



States: list of colors for each cell on each face

Initial state: one specific cube configuration

Operators: rotate row x or column y on face z direction a

Goal: configuration has only one color on each face

Path cost: 1 per move

[Rubik's cube applet](#)

# State-space searching

- Most problems in AI can be cast as searches on a state space
- The space can be tree-shaped or graph-shaped
  - If a graph, need some way to keep track of where you have been, so as to avoid loops
- The state space is often very, very large
- We can minimize the size of the search space by careful choice of operators
- Exhaustive searches don't work—we need *heuristics*

# Sample Search Problems

- Graph coloring
- Protein folding
- Game playing
- Airline travel
- Proving algebraic equalities
- Robot motion planning

# The basic search algorithm

Initialize: put the start node into OPEN

while OPEN is not empty

    take a node N from OPEN

    if N is a goal node, report success

    put the children of N onto OPEN

Report failure

- If **OPEN** is a stack, this is a depth-first search
- If **OPEN** is a queue, this is a breadth-first search
- If **OPEN** is a *priority queue*, sorted according to *most promising first*, we have a best-first search

# Uninformed Search

# A General State-Space Search Algorithm

- Node  $n$ 
  - state description
  - parent (may use a backpointer) (if needed)
  - Operator used to generate  $n$  (optional)
  - Depth of  $n$  (optional)
  - Path cost from  $S$  to  $n$  (if available)
- OPEN list
  - initialization:  $\{S\}$
  - node insertion/removal depends on specific search strategy
- CLOSED list
  - initialization:  $\{\}$
  - organized by backpointers

# A General State-Space Search Algorithm

$\text{open} := \{S\}; \text{closed} := \{\};$

**repeat**

$n := \text{select}(\text{open});$  /\* select one node from open for expansion \*/

**if**  $n$  is a goal

**then exit** with success; /\* delayed goal testing \*/

$\text{expand}(n)$

/\* generate all children of  $n$

put these newly generated nodes in open (check duplicates)

put  $n$  in closed (check duplicates) \*/

**until**  $\text{open} = \{\};$

**exit** with failure

# Some Issues

- Search process constructs a search tree, where
  - **root** is the initial state  $S$ , and
  - **leaf nodes** are nodes
    - not yet been expanded (i.e., they are in OPEN list) or
    - having no successors (i.e., they're "deadends")
- Search tree may be infinite because of loops even if state space is small
- Search strategies mainly differ on *select* (open)
- Each node represents a partial solution path (and cost of the partial solution path) from the start node to the given node.
  - in general, from this node there are many possible paths (and therefore solutions) that have this partial path as a prefix.



# Evaluating Search Strategies

- **Completeness**

- Guarantees finding a solution whenever one exists

- **Time Complexity**

- How long (worst or average case) does it take to find a solution?  
Usually measured in terms of the **number of nodes expanded**

- **Space Complexity**

- How much space is used by the algorithm? Usually measured in terms of the **maximum size that the “OPEN” list** becomes during the search

- **Optimality / Admissibility**

- If a solution is found, is it guaranteed to be an optimal one? For example, is it the one with minimum cost?

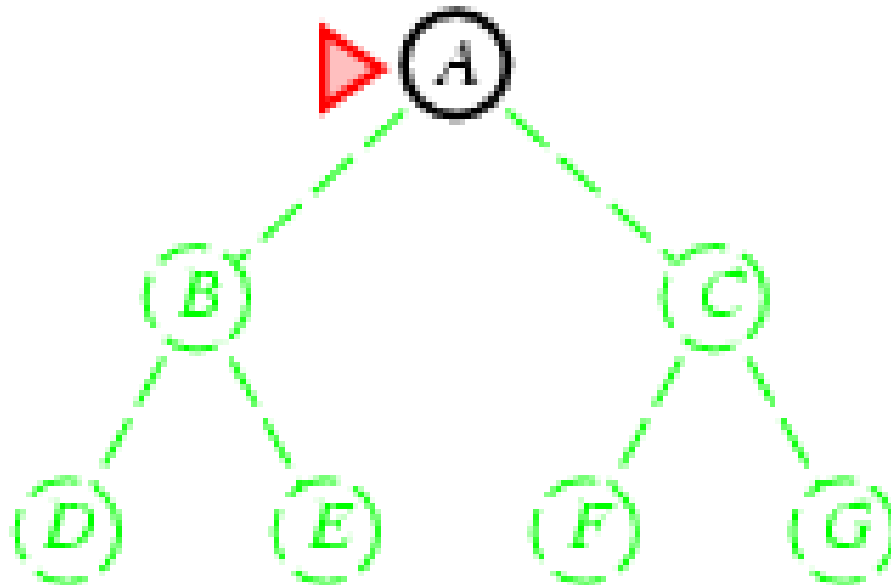
# Uninformed search strategies

- **Uninformed**: While searching you have no clue whether one non-goal state is better than any other. Your search is blind.
- **Various blind strategies:**
  - Breadth-first search
  - Uniform-cost search
  - Depth-first search
  - Iterative deepening search

# Breadth-first search

- Expand shallowest unexpanded node
- **Implementation:**
  - *fringe* is a first-in-first-out (FIFO) queue, i.e., new successors go at end of the queue.

Is A a goal state?

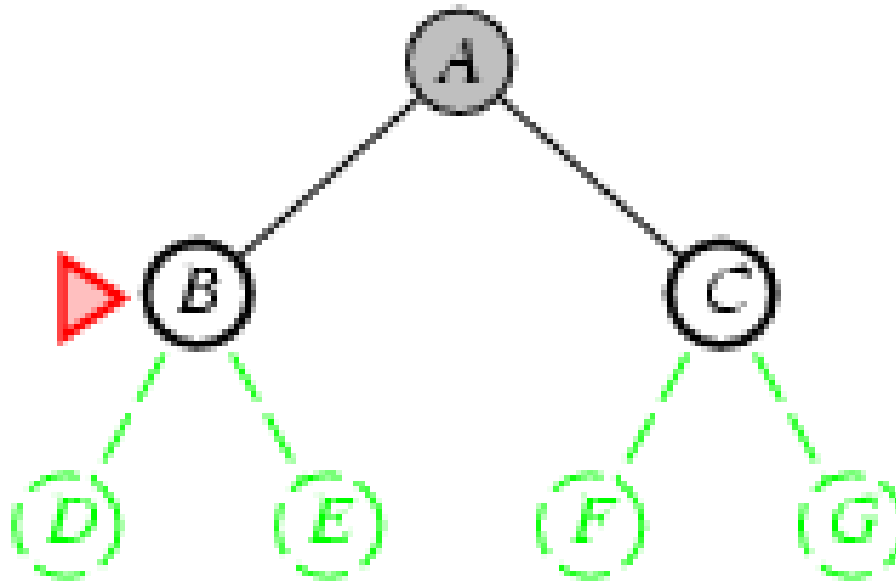


# Breadth-first search

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

Expand:  
fringe = [B,C]

Is B a goal state?

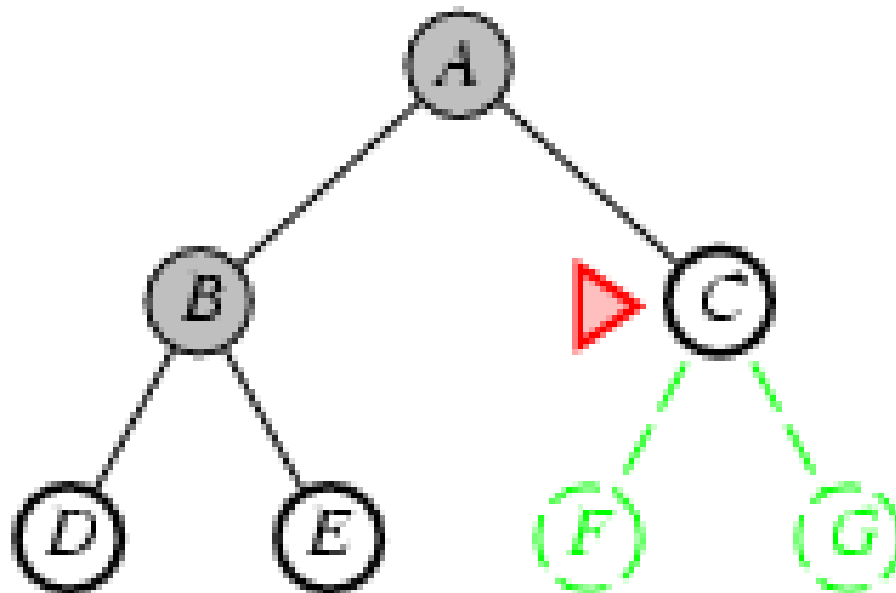


# Breadth-first search

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

Expand:  
fringe=[C,D,E]

Is C a goal state?

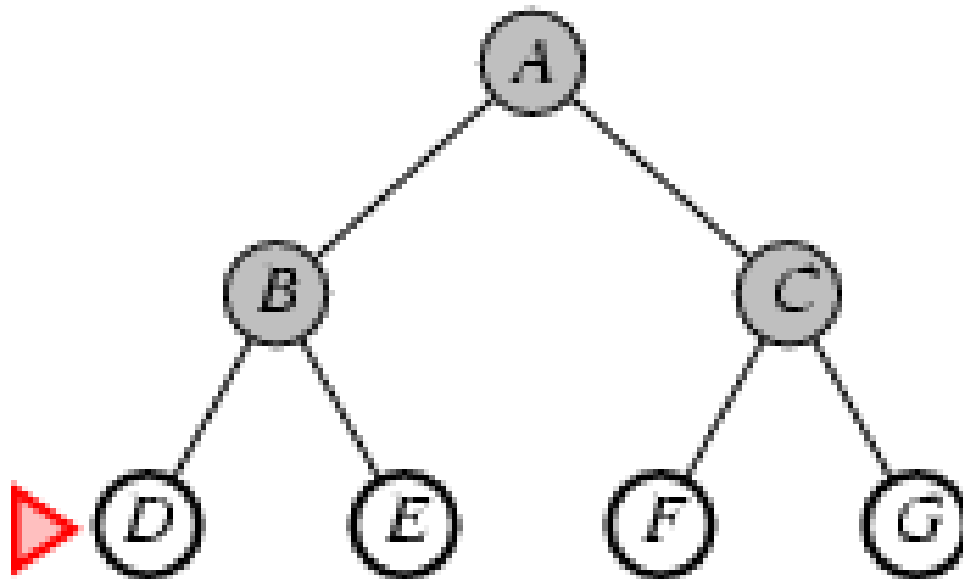


# Breadth-first search

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

Expand:  
fringe=[D,E,F,G]

Is D a goal state?



# 8 Puzzel Game

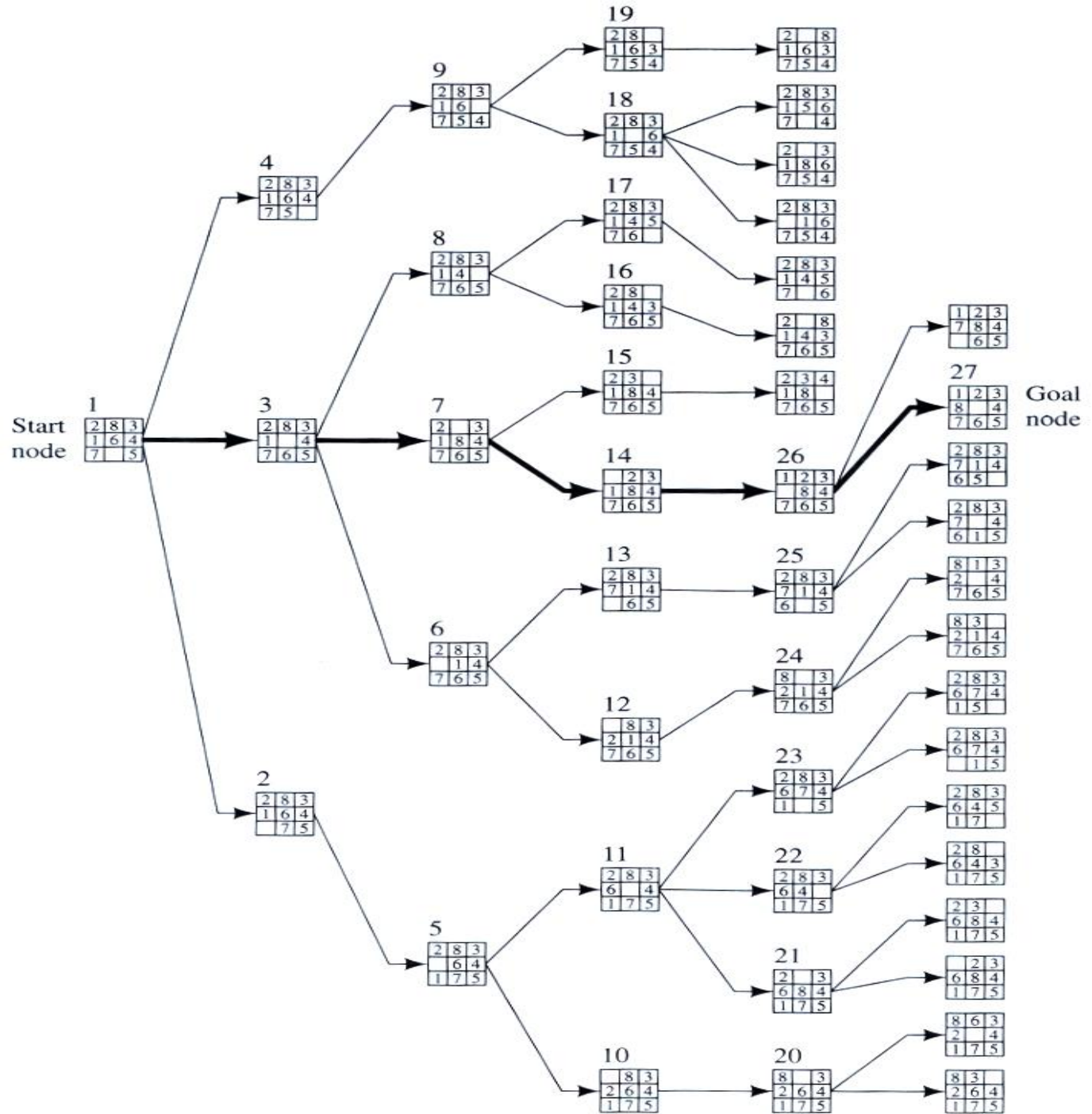
|   |   |   |
|---|---|---|
| 2 | 8 | 3 |
| 1 | 6 | 4 |
| 7 |   | 5 |

|   |   |   |
|---|---|---|
| 2 |   | 8 |
| 1 | 6 | 3 |
| 7 | 5 | 4 |

Initial State

Goal State

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# Analysis

- See what happens with  $b=10$ 
  - expand 10,000 nodes/second
  - 1,000 bytes/node

| Depth | Nodes     | Time        | Memory        |
|-------|-----------|-------------|---------------|
| 2     | 1110      | .11 seconds | 1 megabyte    |
| 4     | 111,100   | 11 seconds  | 106 megabytes |
| 6     | $10^7$    | 19 minutes  | 10 gigabytes  |
| 8     | $10^9$    | 31 hours    | 1 terabyte    |
| 10    | $10^{11}$ | 129 days    | 101 terabytes |
| 12    | $10^{13}$ | 35 years    | 10 petabytes  |
| 15    | $10^{15}$ | 3,523 years | 1 exabyte     |

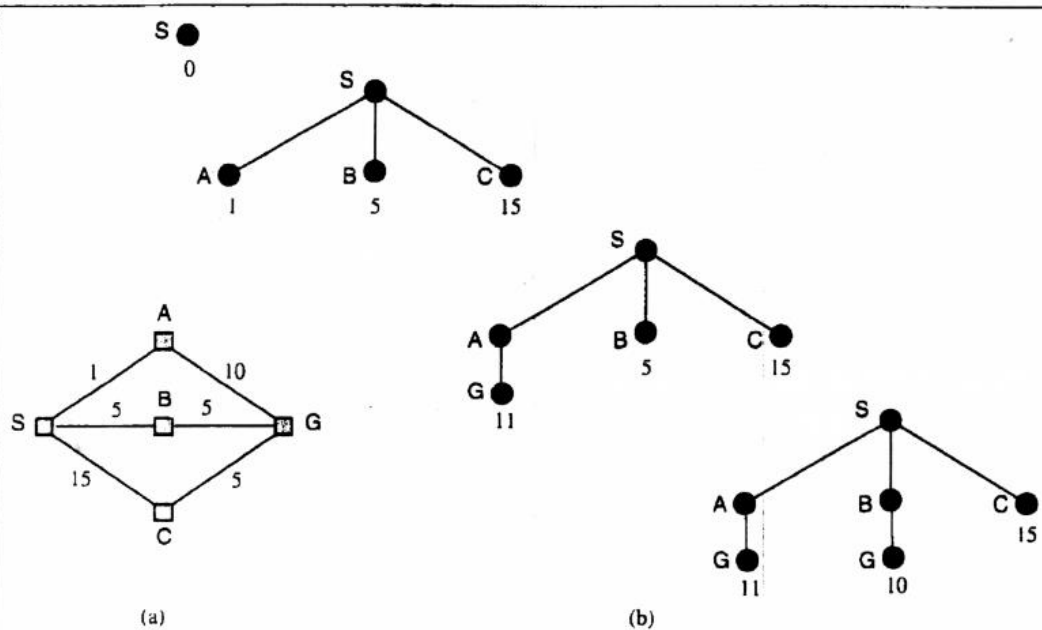
# Properties of breadth-first search

- Complete? Yes it always reaches goal (if  $b$  is finite)
- Time?  $1 + b + b^2 + b^3 + \dots + b^d + (b^{d+1} - b) = O(b^{d+1})$   
(this is the number of nodes we generate)
- Space?  $O(b^{d+1})$  (keeps every node in memory,  
either in fringe or on a path to fringe).
- Optimal? Yes (if we guarantee that deeper solutions are  
less optimal, e.g. step-cost=1).
- **Space** is the bigger problem (more than time)

# Uniform-cost search

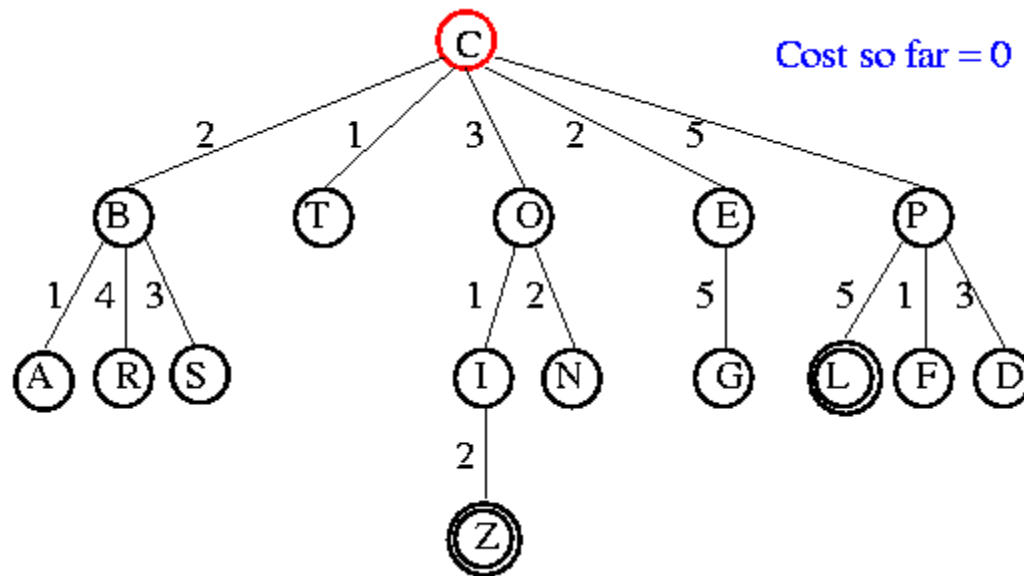
Breadth-first is only optimal if step costs is increasing with depth (e.g. constant). Can we guarantee optimality for any step cost?

**Uniform-cost Search:** Expand node with smallest path cost  $g(n)$ .



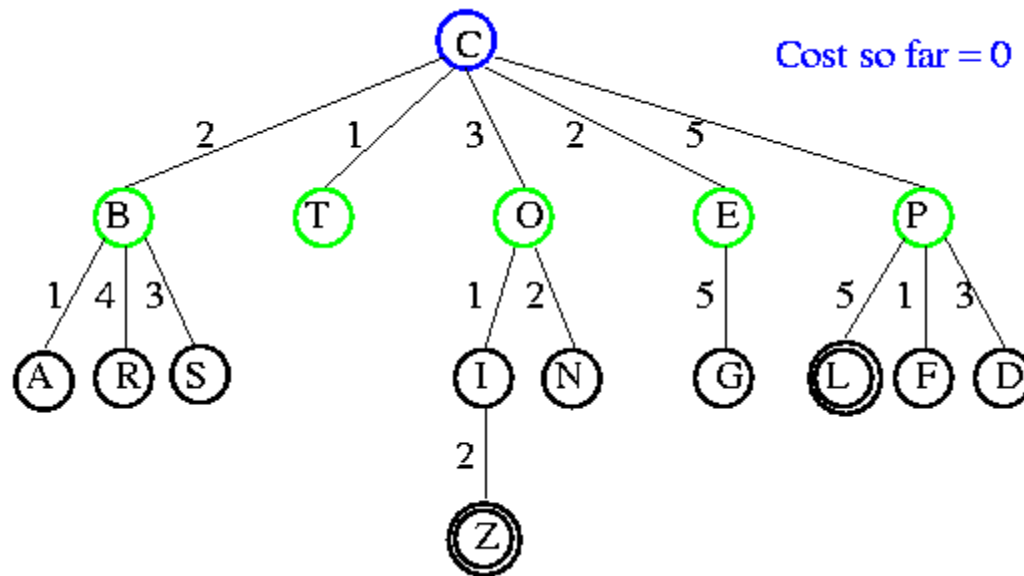
**Figure 3.13** A route-finding problem. (a) The state space, showing the cost for each operator. (b) Progression of the search. Each node is labelled with  $g(n)$ . At the next step, the goal node with  $g = 10$  will be selected.

# UCS Example



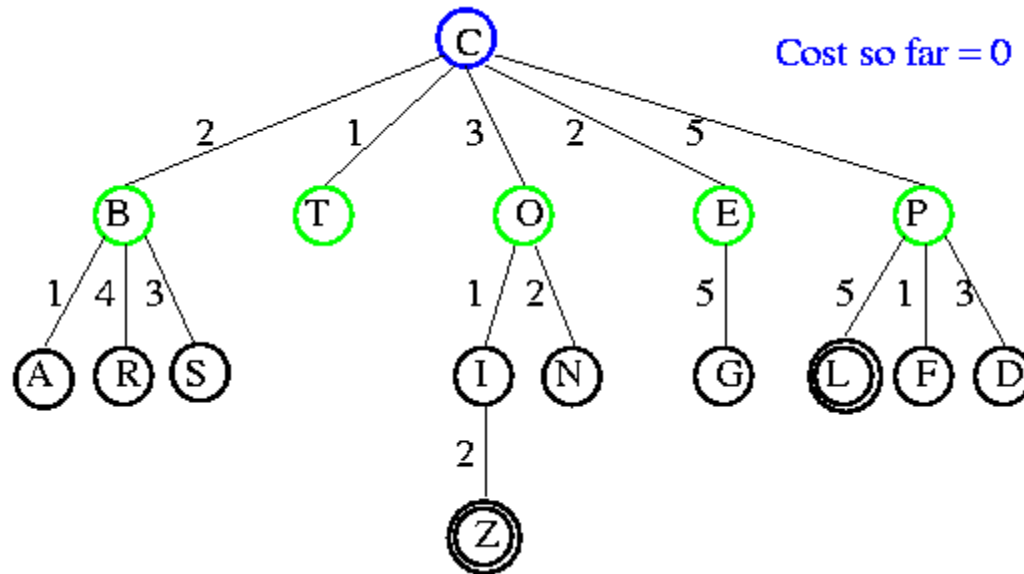
Open list: C

# UCS Example



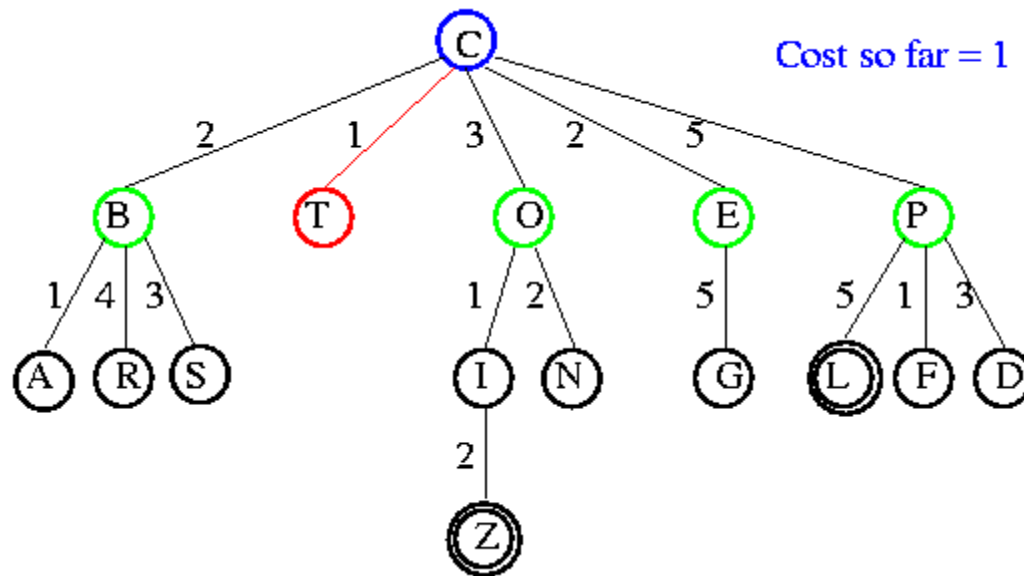
Open list: B(2) T(1) O(3) E(2) P(5)

# UCS Example



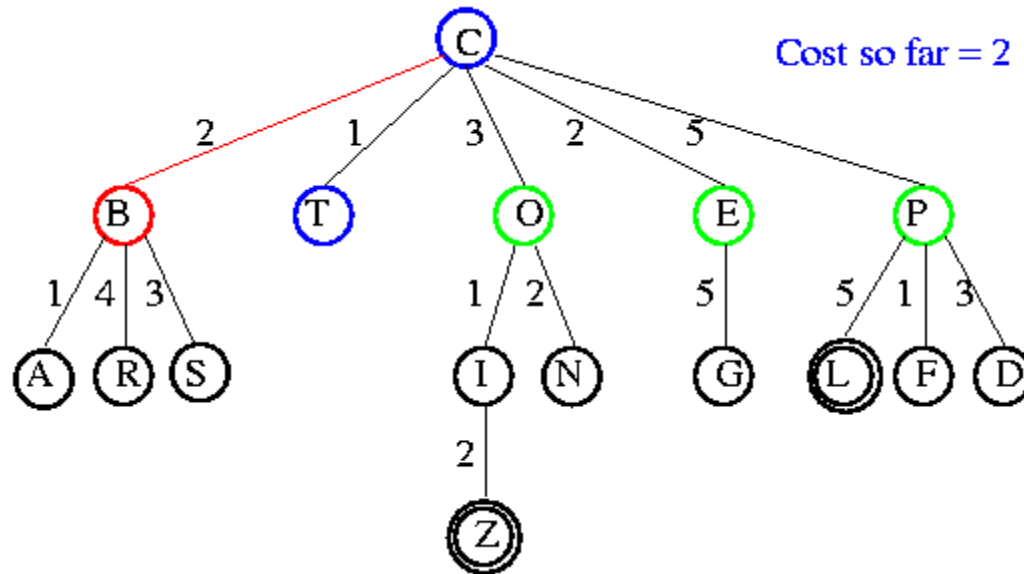
Open list: T(1) B(2) E(2) O(3) P(5)

# UCS Example



Open list: B(2) E(2) O(3) P(5)

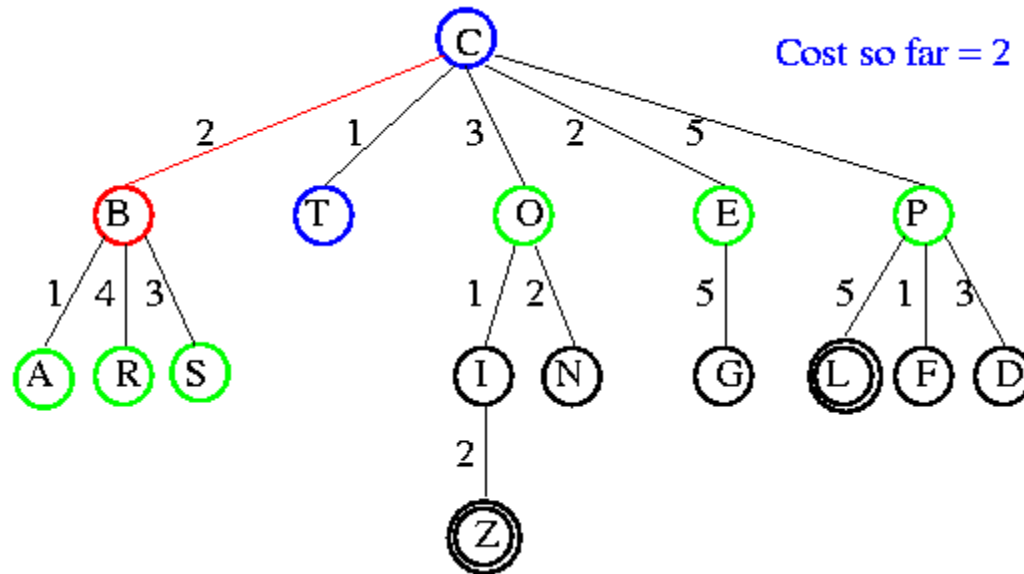
# UCS Example



Open list: E(2) O(3) P(5)

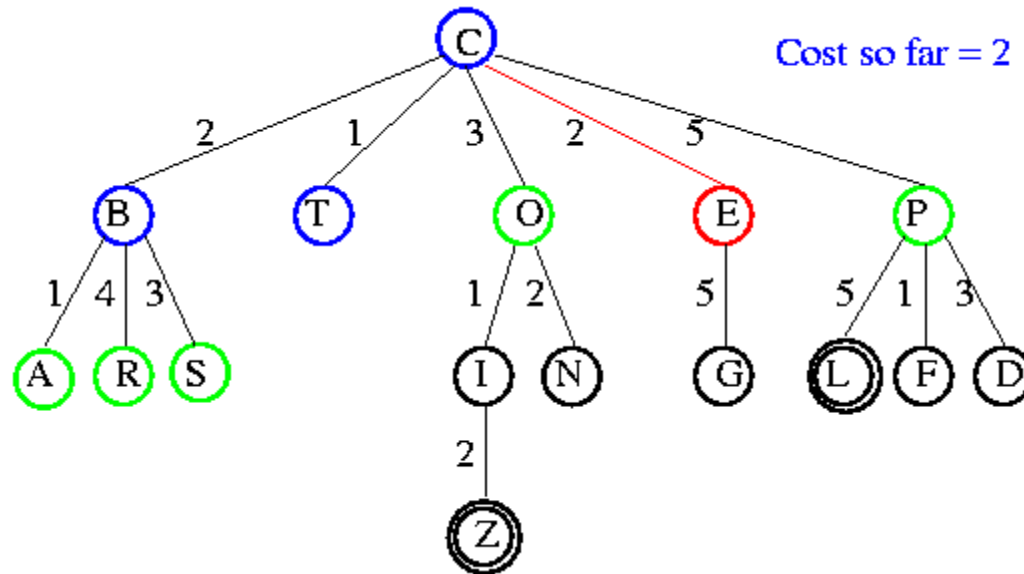


# UCS Example



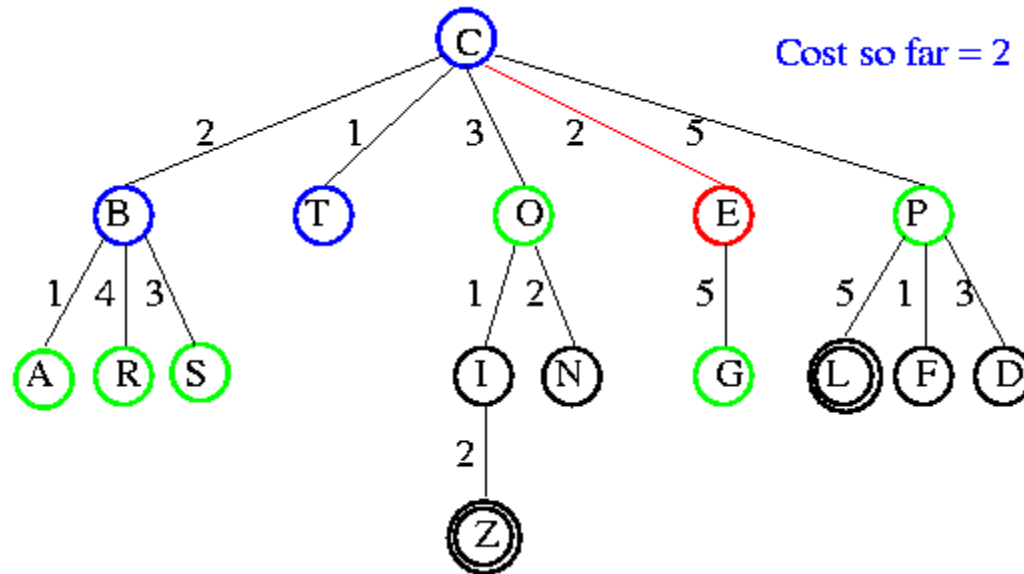
Open list: E(2) O(3) A(3) S(5) P(5) R(6)

# UCS Example



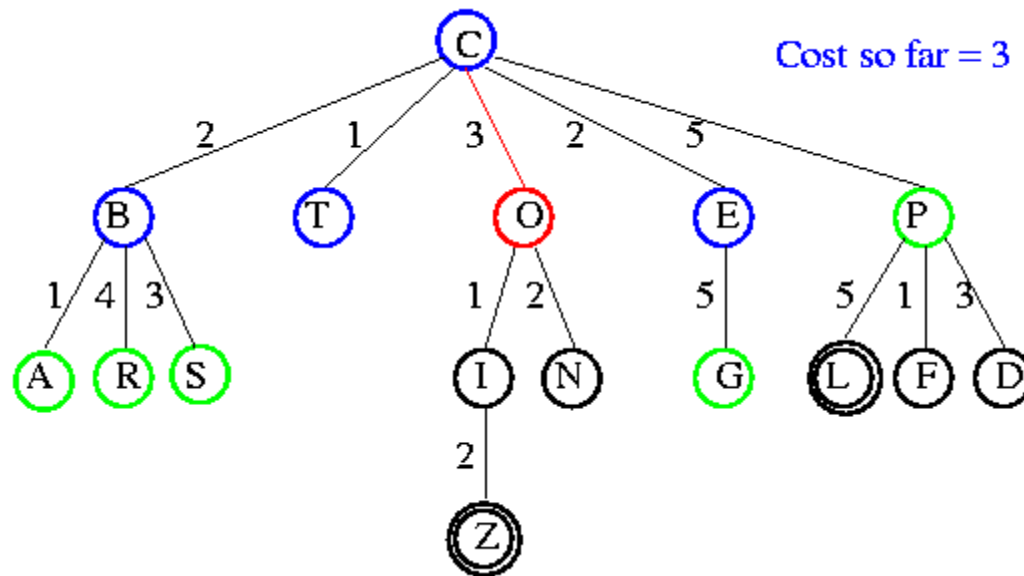
Open list: O(3) A(3) S(5) P(5) R(6)

# UCS Example



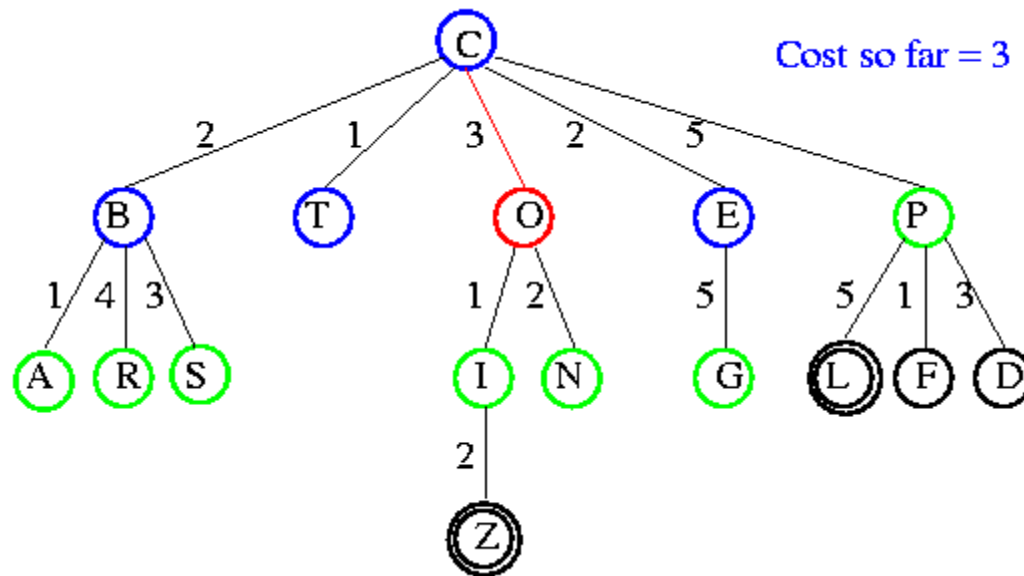
Open list: O(3) A(3) S(5) P(5) R(6) G(10)

# UCS Example



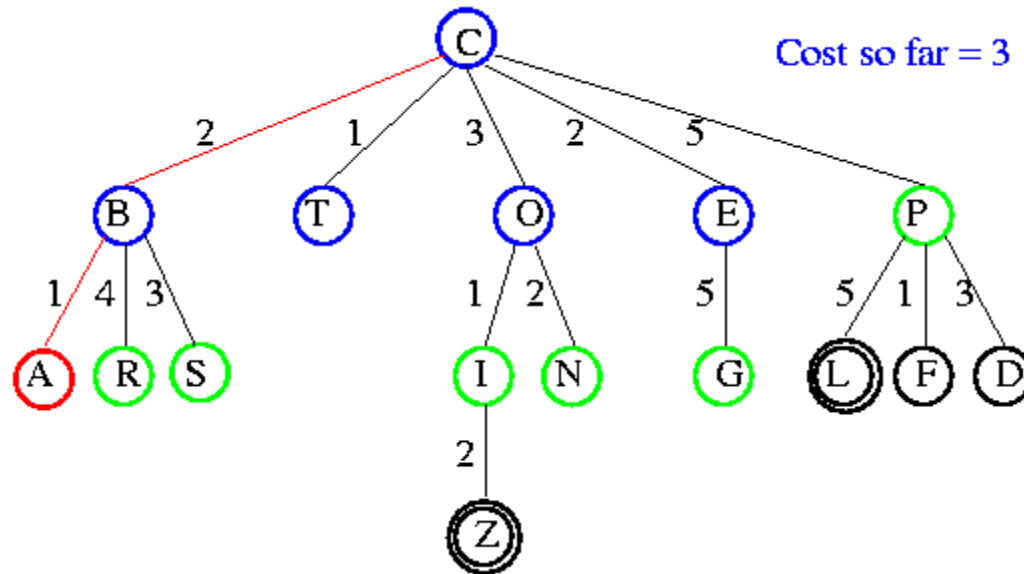
Open list: A(3) S(5) P(5) R(6) G(10)

# UCS Example



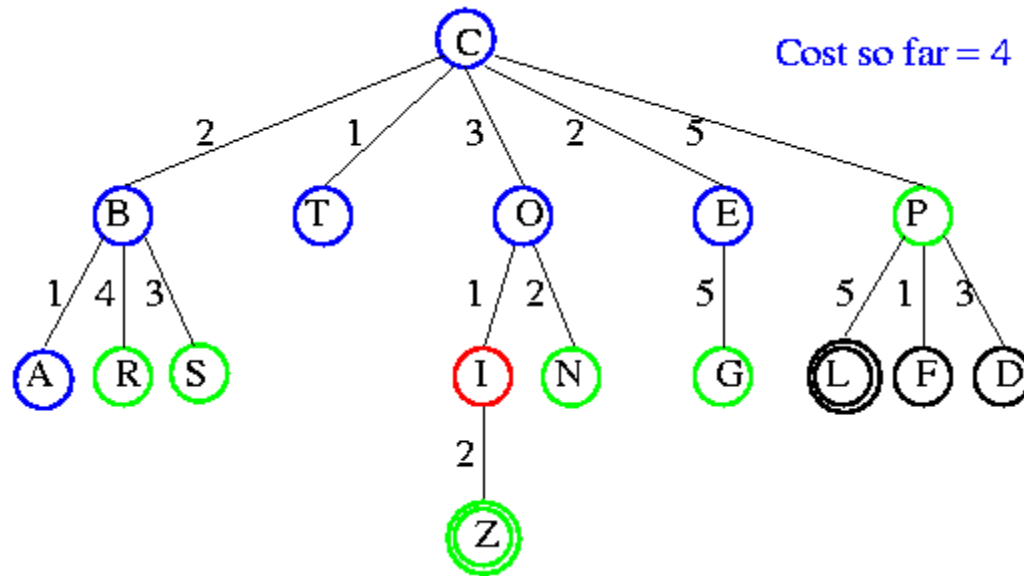
Open list: A(3) I(4) S(5) N(5) P(5) R(6) G(10)

# UCS Example



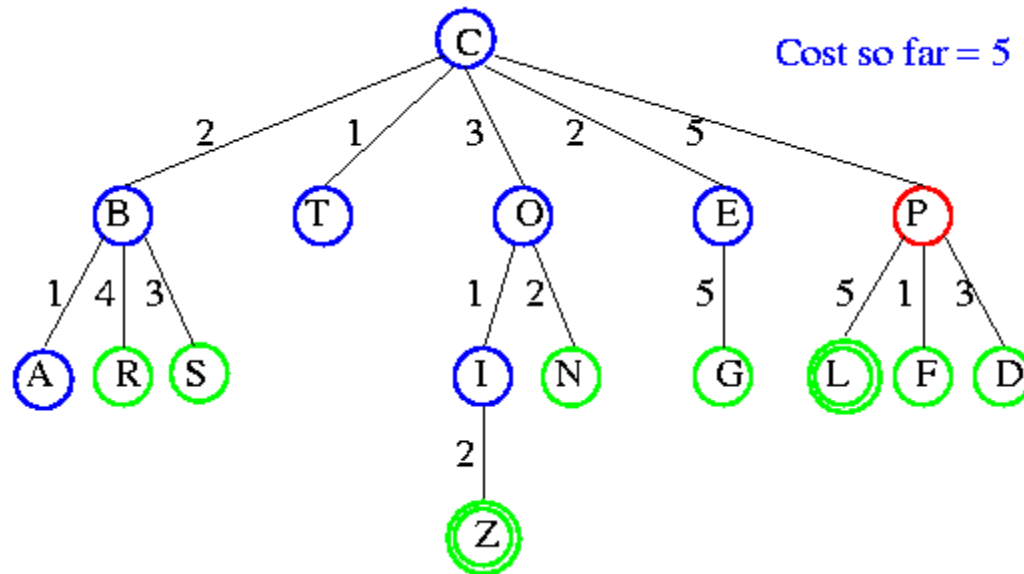
Open list: I(4) P(5) S(5) N(5) R(6) G(10)

# UCS Example



Open list: P(5) S(5) N(5) R(6) Z(6) G(10)

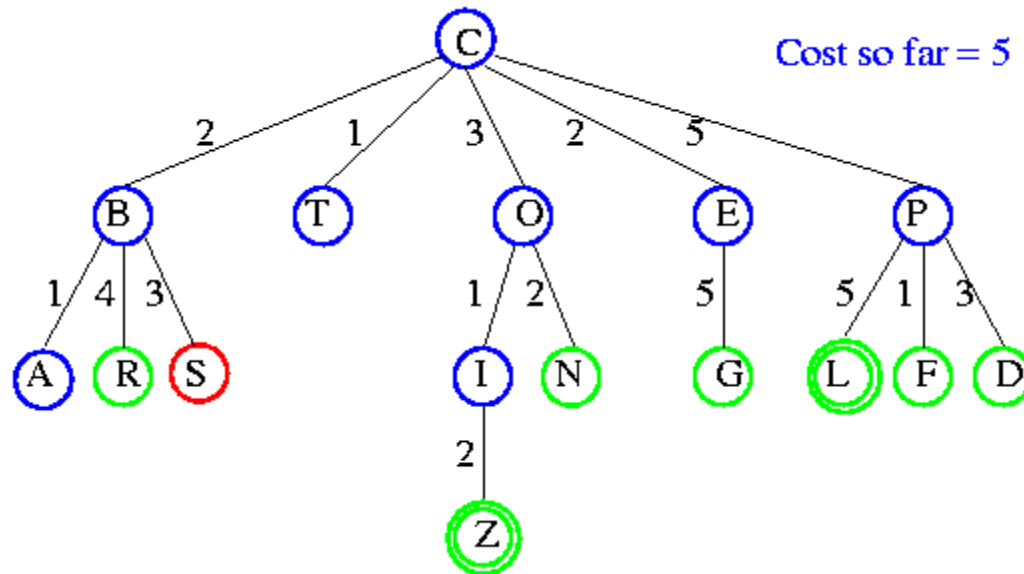
# UCS Example



Open list: S(5) N(5) R(6) Z(6) F(6) D(8) G(10) L(10)

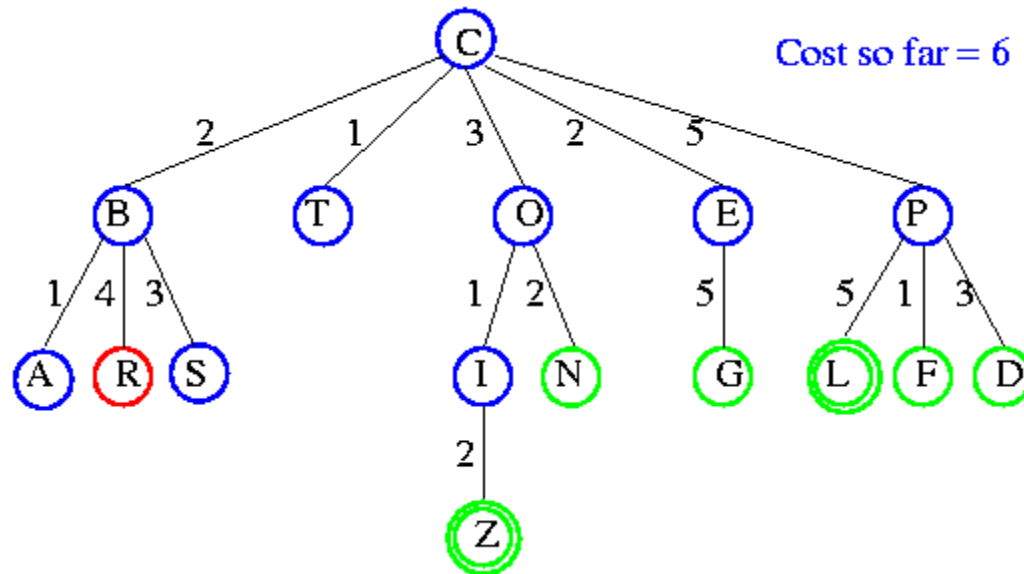


# UCS Example



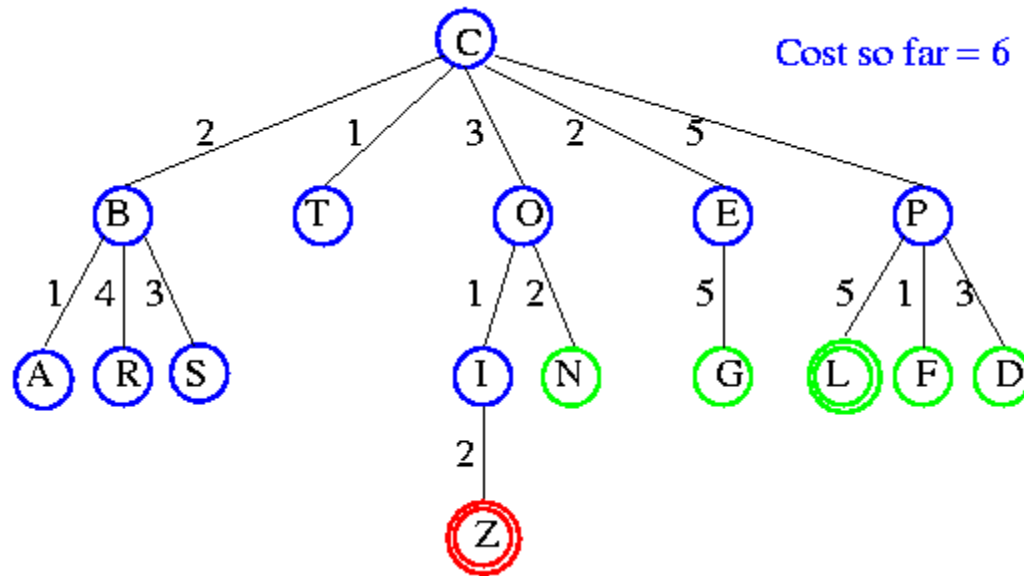
Open list: N(5) R(6) Z(6) F(6) D(8) G(10) L(10)

# UCS Example



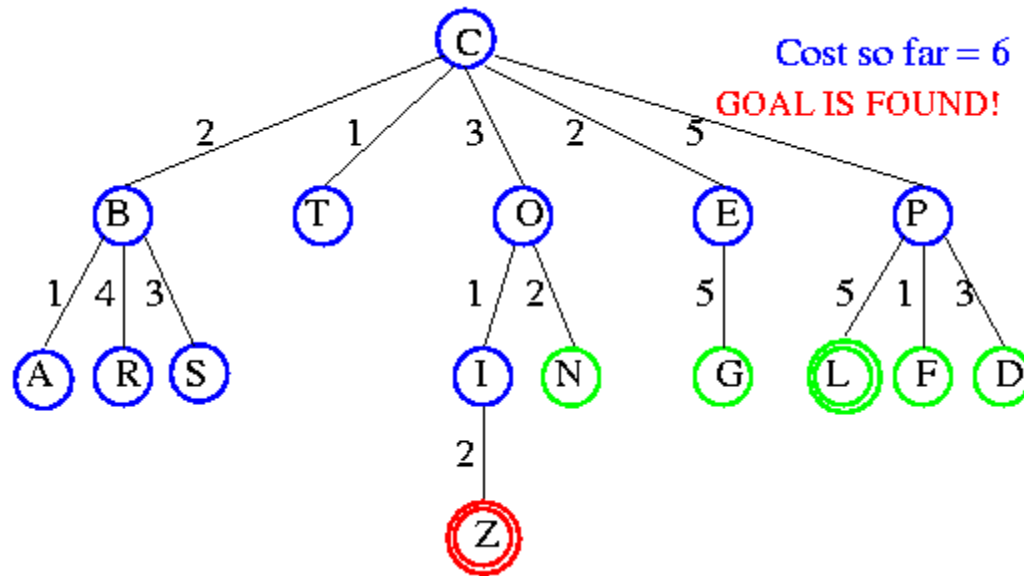
Open list: Z(6) F(6) D(8) G(10) L(10)

# UCS Example



Open list: F(6) D(8) G(10) L(10)

# UCS Example



# Uniform-cost search

**Implementation:** *fringe* = queue ordered by path cost  
Equivalent to breadth-first if all step costs all equal.

Complete? Yes, if step cost  $\geq \epsilon$   
(otherwise it can get stuck in infinite loops)

Time? # of nodes with *path cost*  $\leq$  cost of optimal solution.

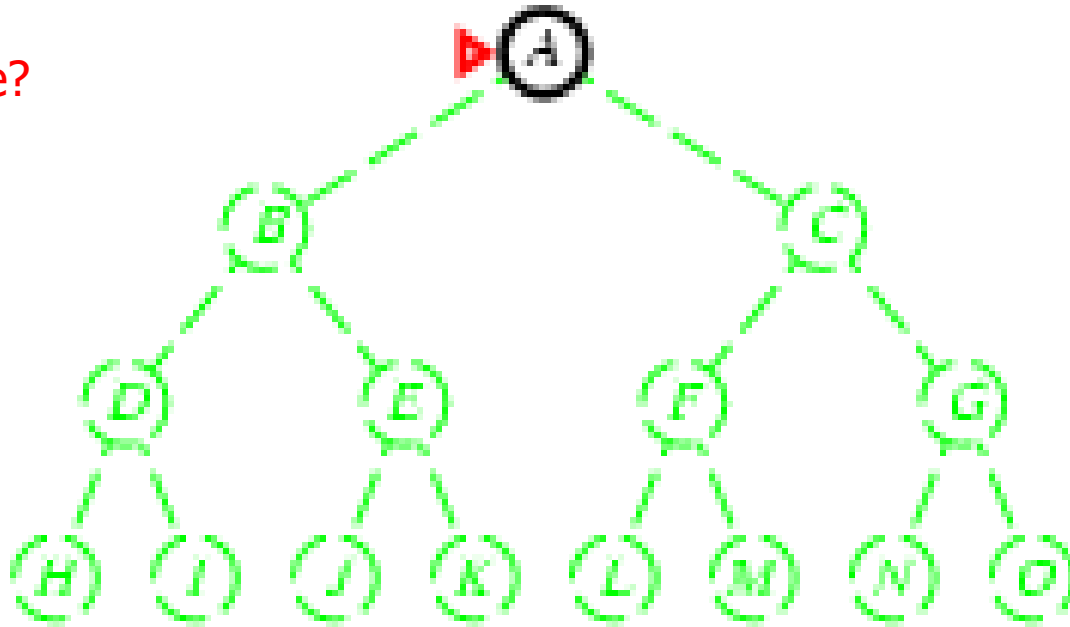
Space? # of nodes on paths with path cost  $\leq$  cost of optimal solution.

Optimal? Yes, for any step cost.

# Depth-first search

- Expand *deepest* unexpanded node
- **Implementation:**
  - *fringe* = Last In First Out (LIPO) queue, i.e., put successors at front

Is A a goal state?

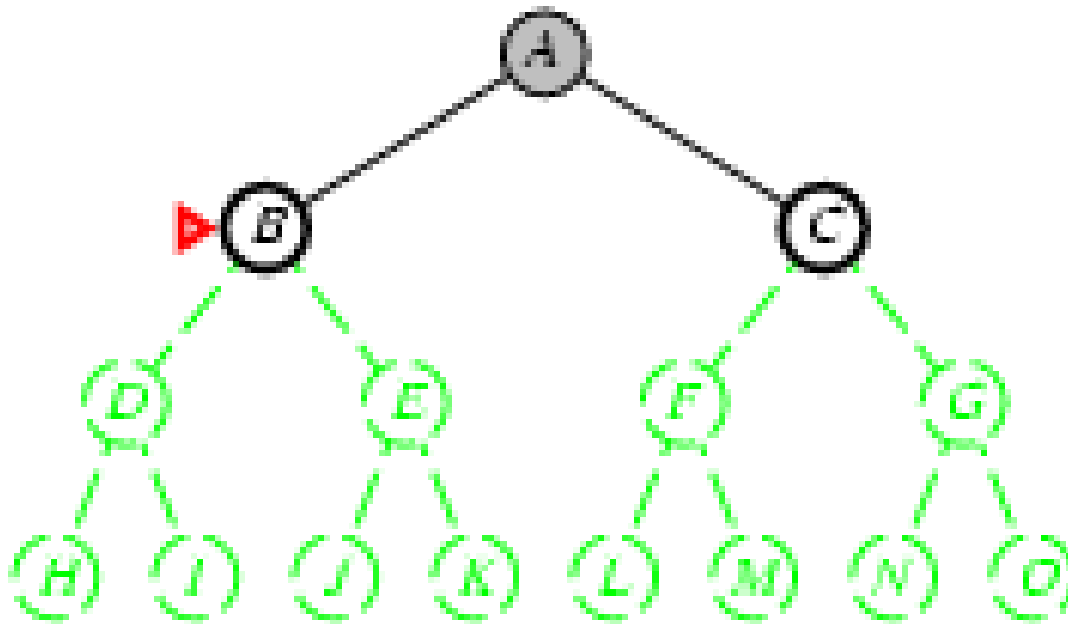


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[B,C]

Is B a goal state?

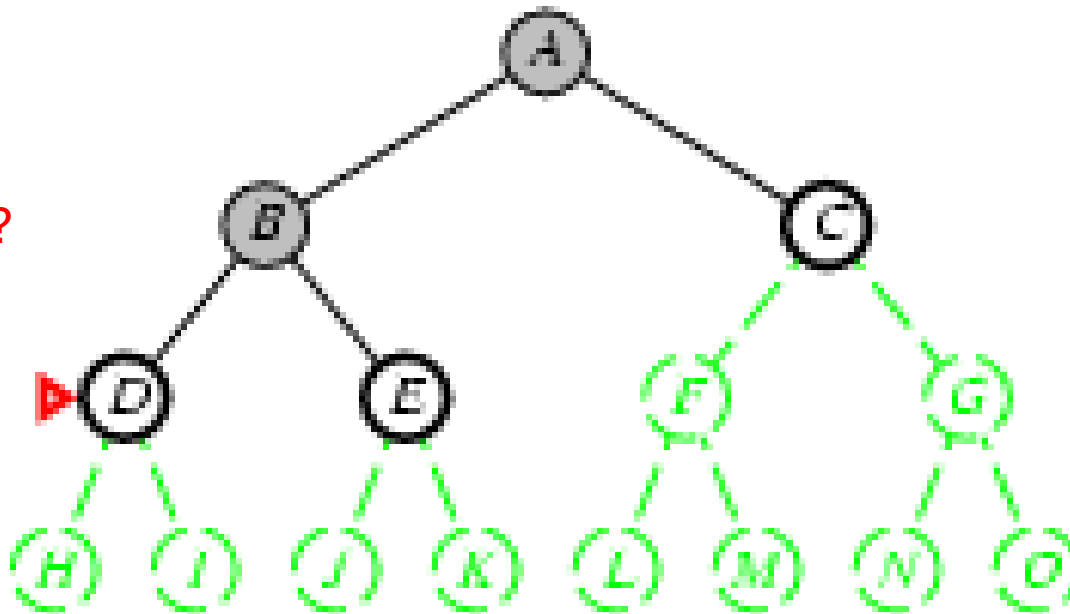


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[D,E,C]

Is D = goal state?



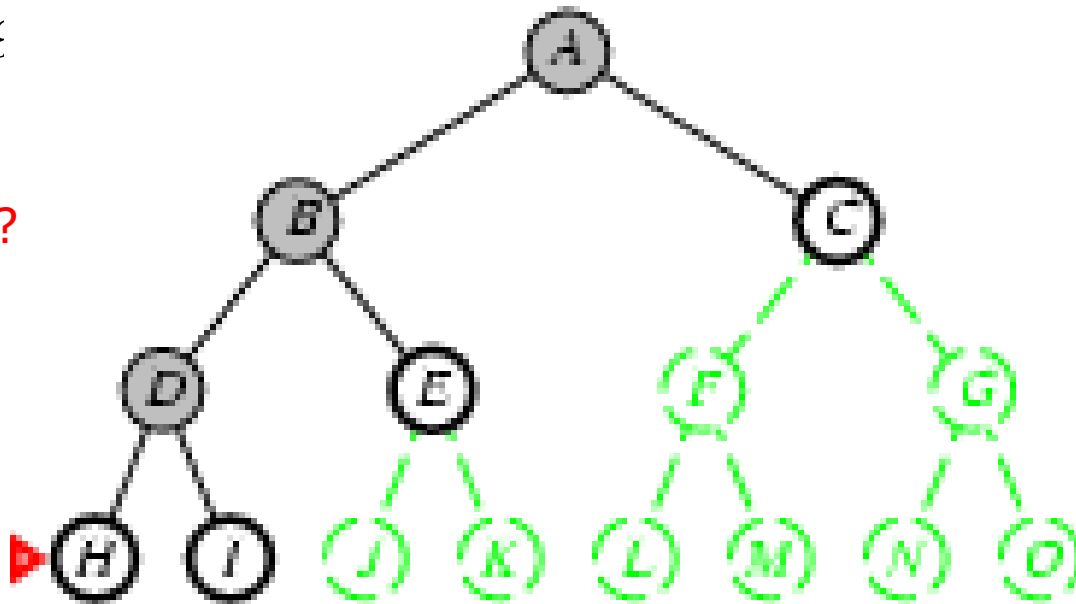


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[H,I,E,C]

Is H = goal state?

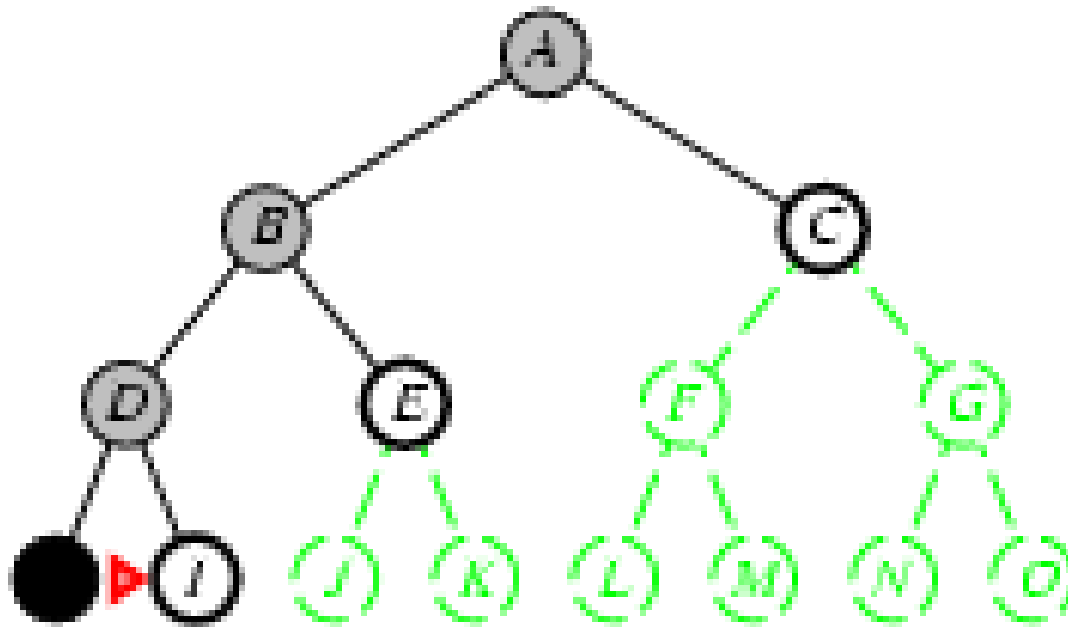


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[I,E,C]

Is I = goal state?

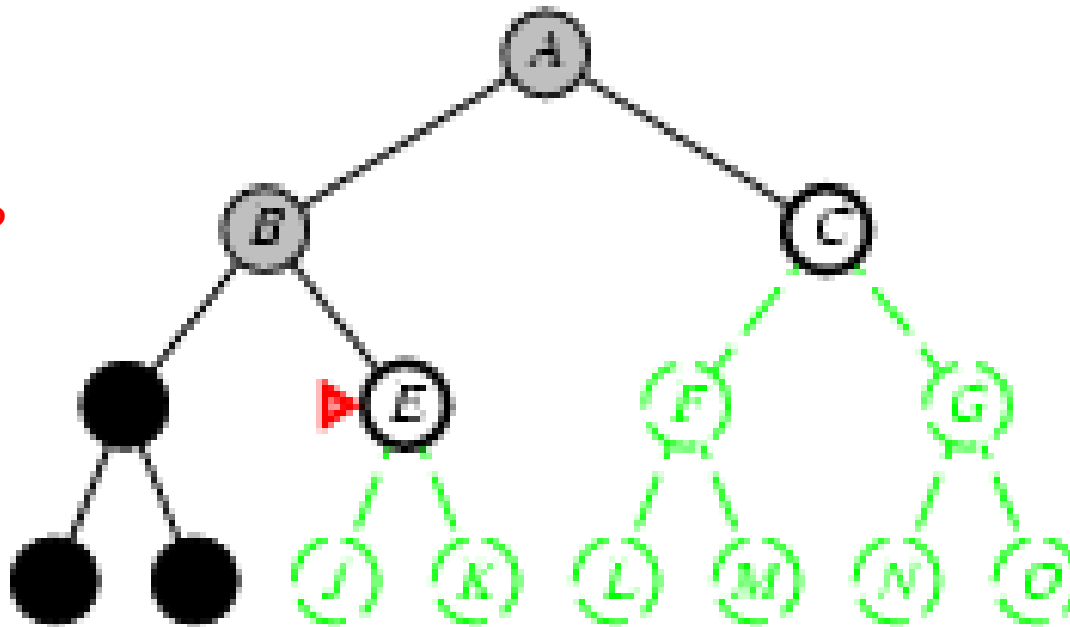


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[E,C]

Is E = goal state?

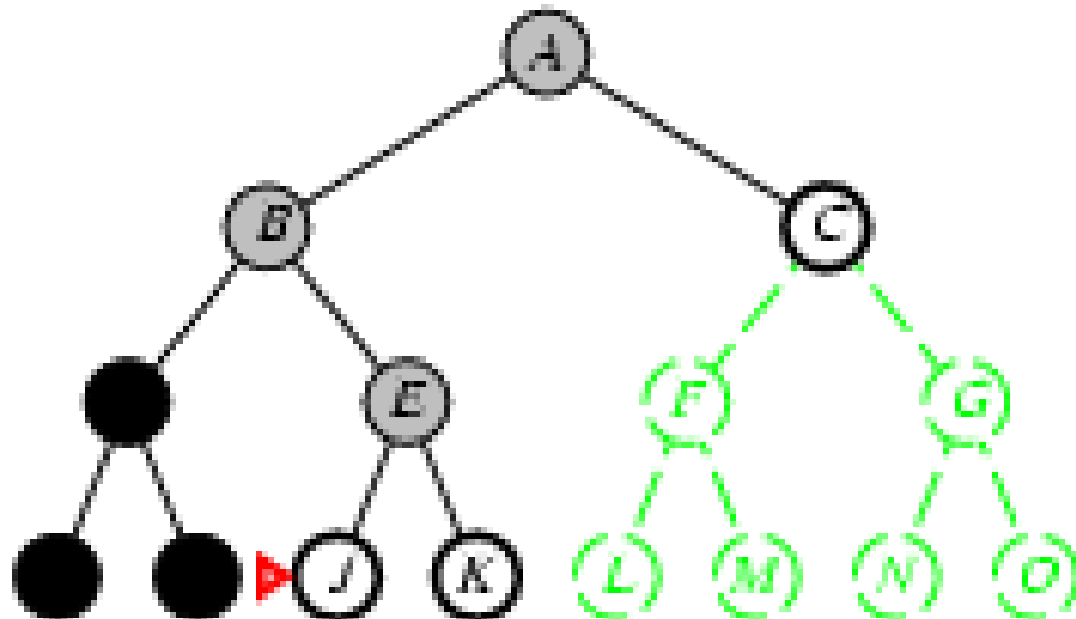


# Depth-first search

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

queue=[J,K,C]

Is J = goal state?

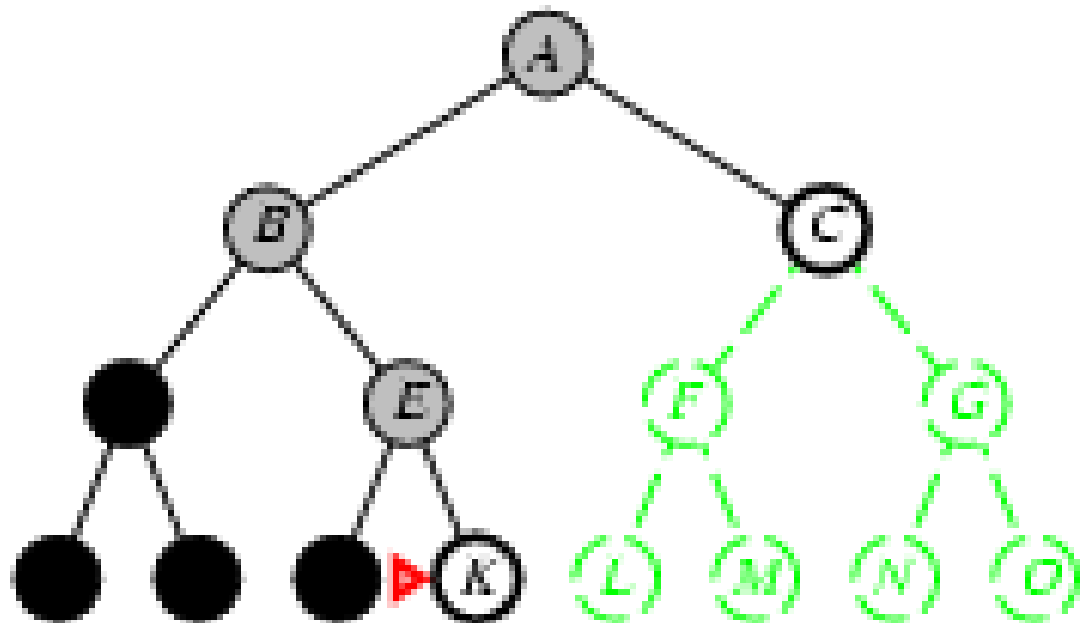


# Depth-first search

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

queue=[K,C]

Is K = goal state?

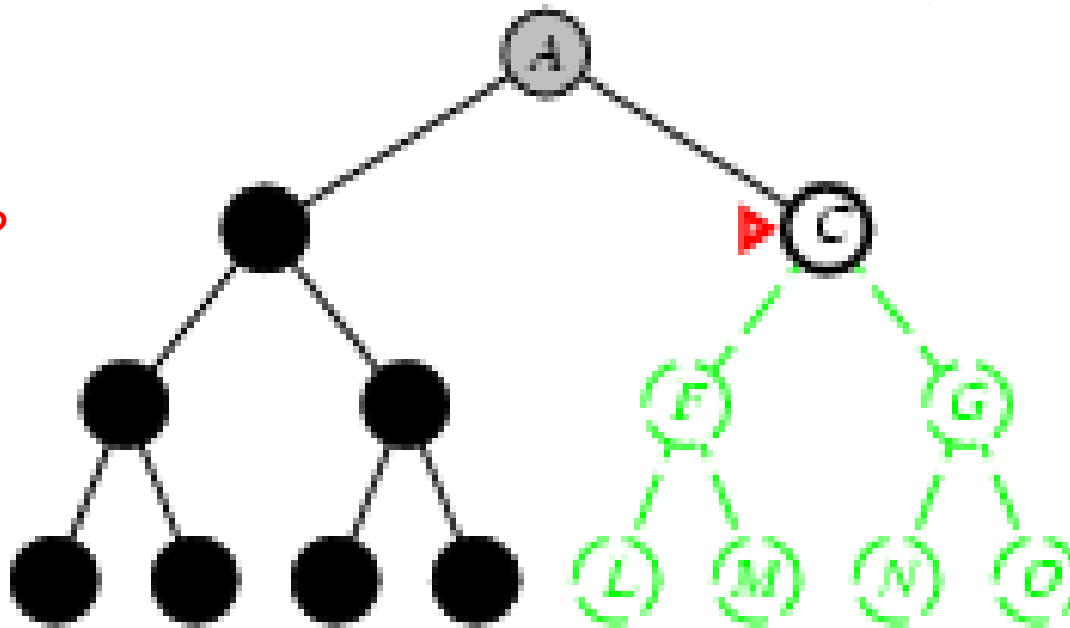


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[C]•

Is C = goal state?

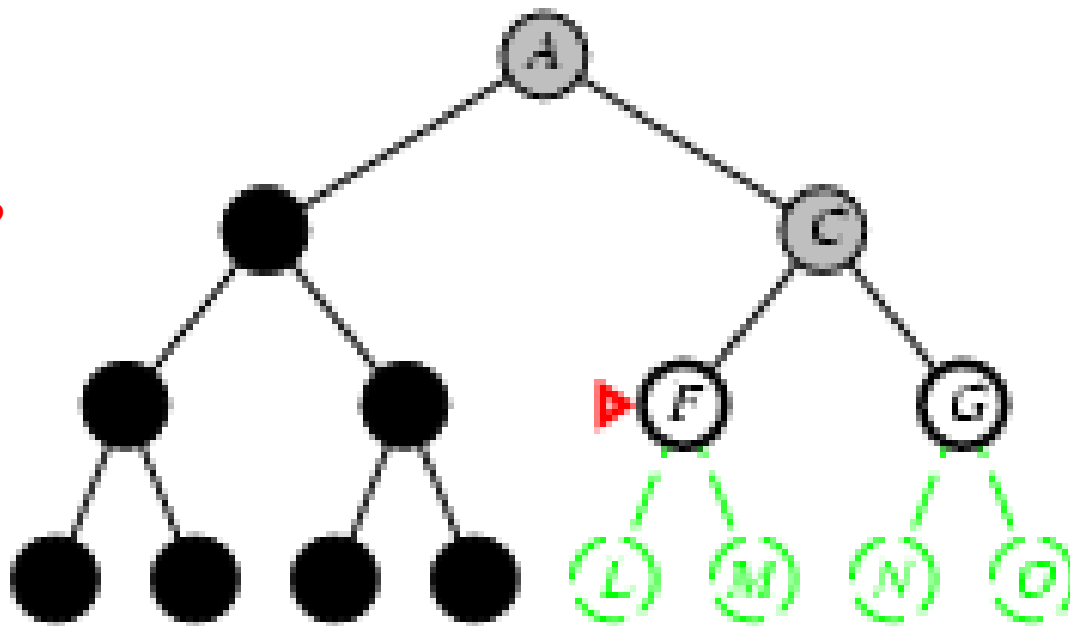


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[F,G]

Is F = goal state?



- Expand deepest unexpanded node

- Implementation:

queue=[L,M,G]

A search tree diagram illustrating a breadth-first search process. The root node is A (gray). Level 1 nodes are black circles. Level 2 nodes include L (white), M (white), F (gray), and G (white). Level 3 nodes include N (green dashed) and O (green dashed). A red arrow points from the fourth black circle at level 2 to node L.

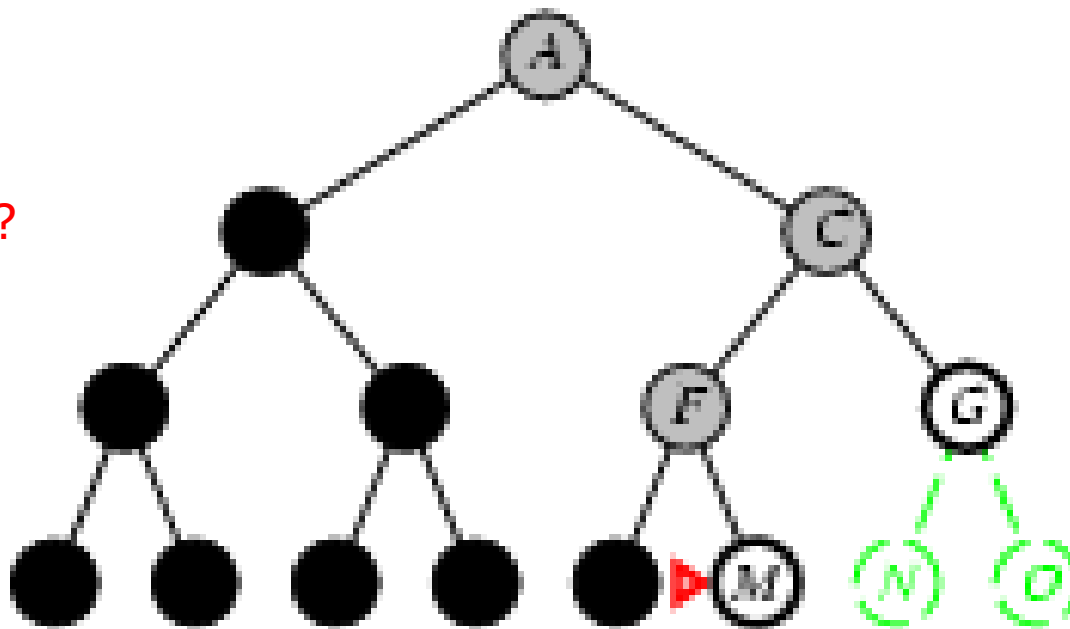


# Depth-first search

- Expand deepest unexpanded node
- 
- Implementation:
  - $fringe$

queue=[M,G]

Is M = goal state?



# Properties of depth-first search

- Complete? No: fails in infinite-depth spaces

Can modify to avoid repeated states along path

- Time?  $O(b^m)$  with  $m$  = maximum depth
- 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! (we only need to remember a single path + expanded unexplored nodes)
- Optimal? No (It may find a non-optimal goal first)

# 8 Puzzel Game

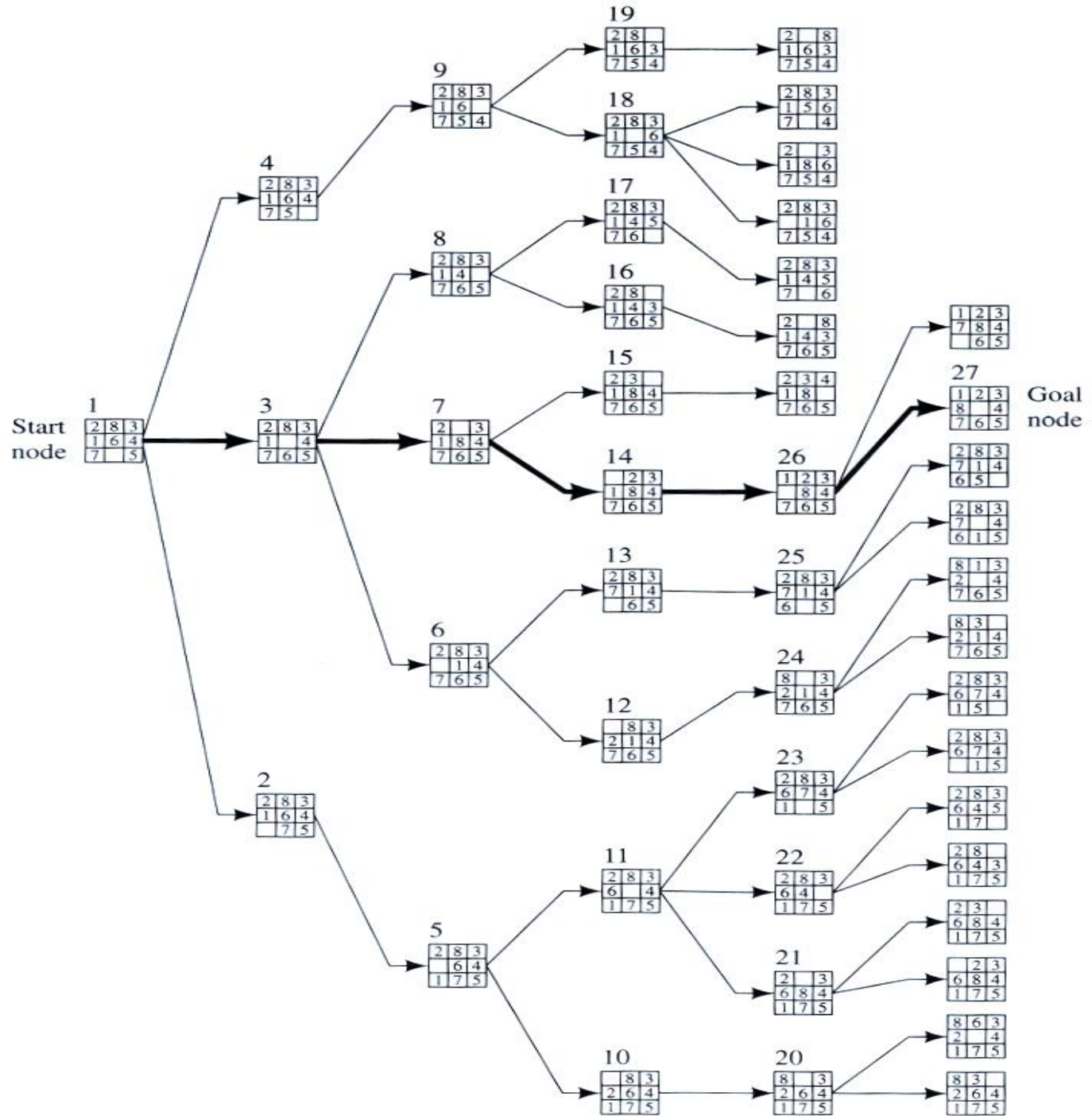
|   |   |   |
|---|---|---|
| 2 | 8 | 3 |
| 1 | 6 | 4 |
| 7 |   | 5 |

|   |   |   |
|---|---|---|
| 8 |   | 3 |
| 2 | 6 | 4 |
| 1 | 7 | 5 |

Initial State

Goal State

# Example BFS



# Iterative deepening search

- To avoid the infinite depth problem of DFS, we can decide to only search until depth  $L$ , i.e. we don't expand beyond depth  $L$ .  
→ Depth-Limited Search
- What of solution is deeper than  $L$ ? → Increase  $L$  iteratively.  
→ Iterative Deepening Search
- As we shall see: this inherits the memory advantage of Depth-First search.

# Iterative deepening search $L=0$

Limit = 0



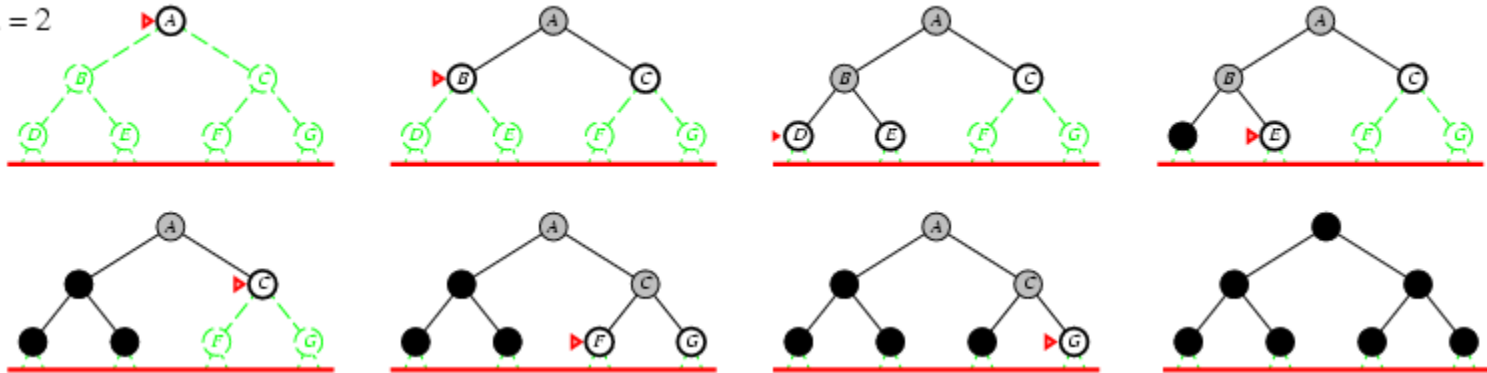
# Iterative deepening search $L=1$

Limit = 1



# Iterative deepening search $L=2$

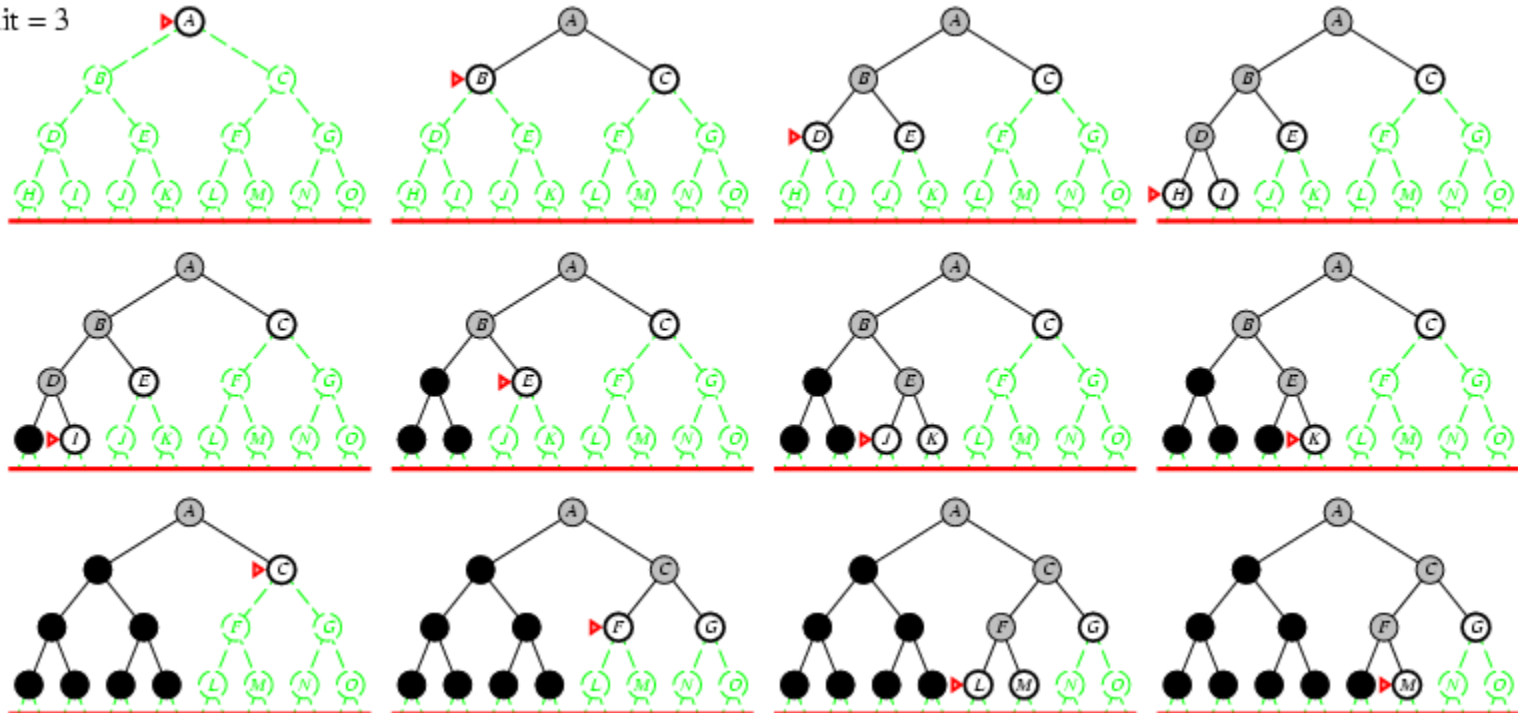
Limit = 2





# Iterative deepening search $L=3$

Limit = 3



# Iterative deepening search

- Number of nodes generated in a depth-limited search to depth  $d$  with branching factor  $b$ :

$$N_{DLS} = b^0 + b^1 + b^2 + \dots + b^{d-2} + b^{d-1} + b^d$$

- Number of nodes generated in an iterative deepening search to depth  $d$  with branching factor  $b$ :

$$N_{IDS} = (d+1)b^0 + d b^1 + (d-1)b^2 + \dots + 3b^{d-2} + 2b^{d-1} + 1b^d =$$

$$O(b^d) \neq O(b^{d+1})$$

BFS

- For  $b = 10, d = 5$ ,

- $N_{DLS} = 1 + 10 + 100 + 1,000 + 10,000 + 100,000 = 111,111$

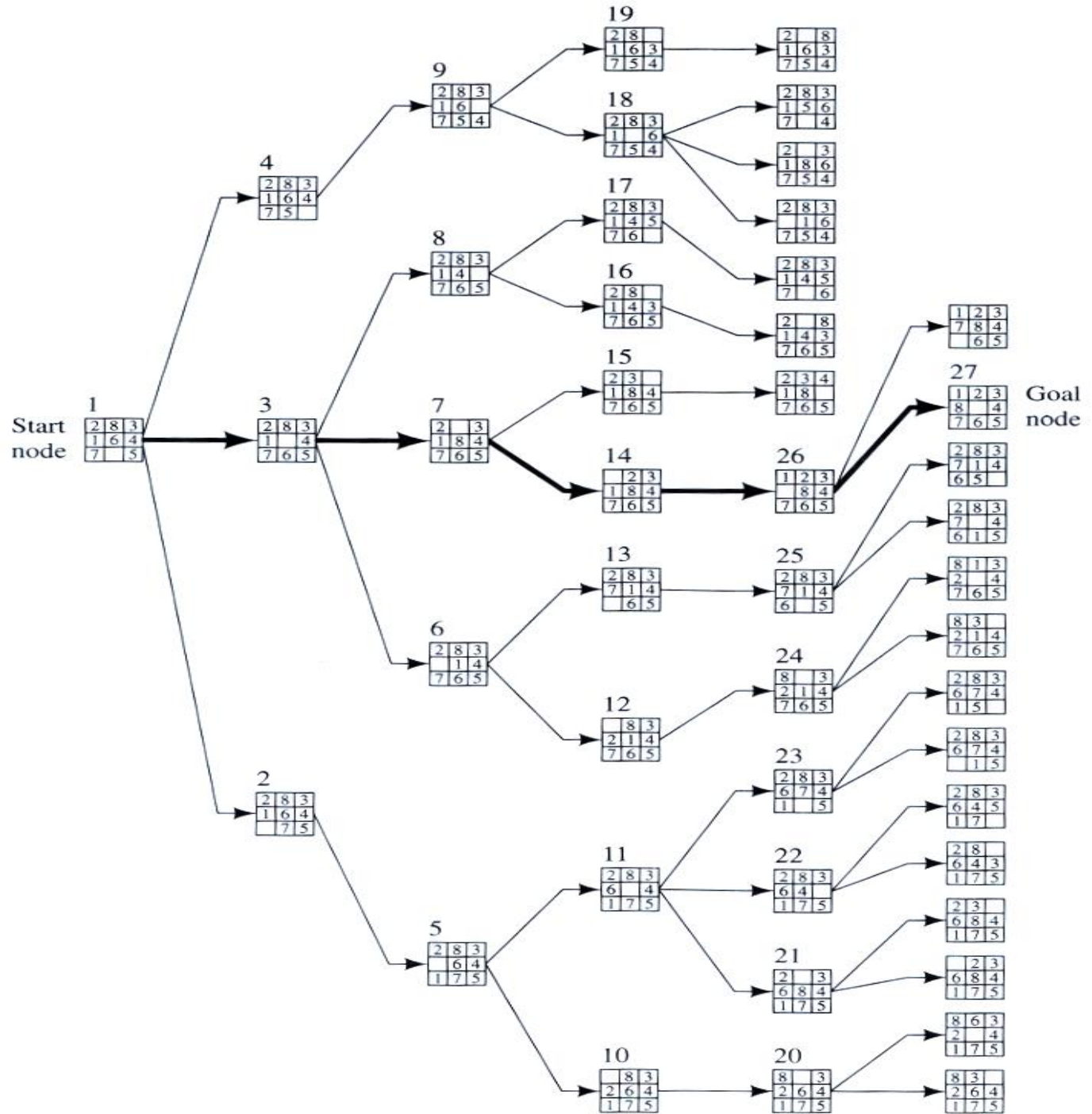
- $N_{IDS} = 6 + 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450$

- $N_{BFS} = \dots = 1,111,100$

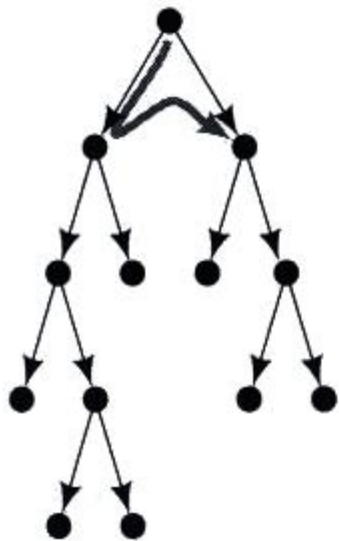
# Properties of iterative deepening search

- 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 or increasing function of depth.

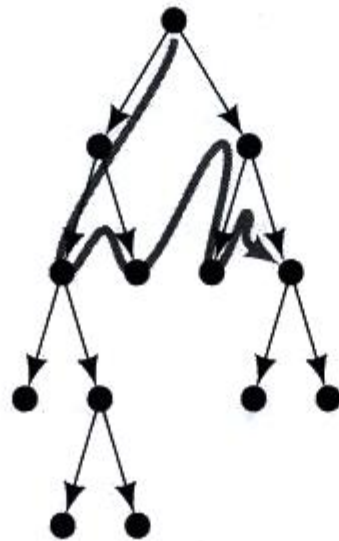
# Example BFS



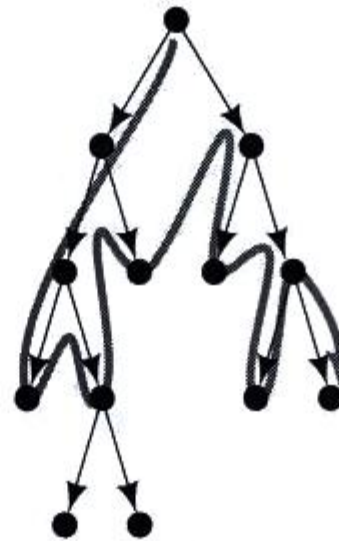
# Example IDS



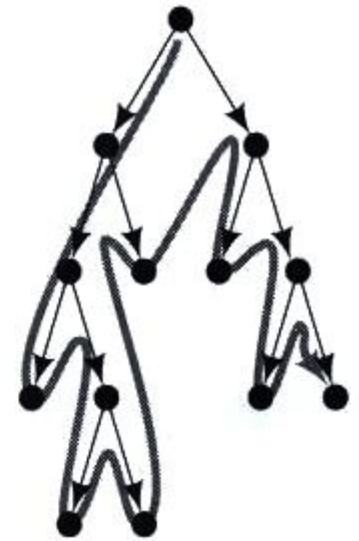
Depth bound = 1



Depth bound = 2



Depth bound = 3



Depth bound = 4

---

Stages in Iterative-Deepening Search

# Bidirectional Search

- Idea
  - simultaneously search forward from S and backwards from G
  - stop when both “meet in the middle”
  - need to keep track of the intersection of 2 open sets of nodes
- What does searching backwards from G mean
  - need a way to specify the predecessors of G
    - this can be difficult,
    - e.g., predecessors of checkmate in chess?
  - what if there are multiple goal states?
  - what if there is only a goal test, no explicit list?

# Bi-Directional Search

Complexity: time and space complexity are:  $O(b^{d/2})$

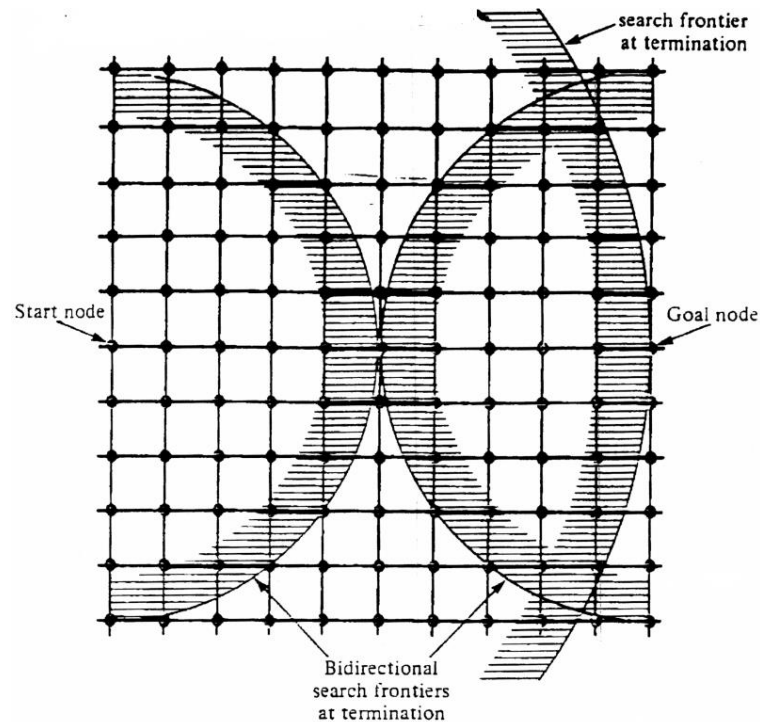


Fig. 2.10 Bidirectional and unidirectional breadth-first searches.

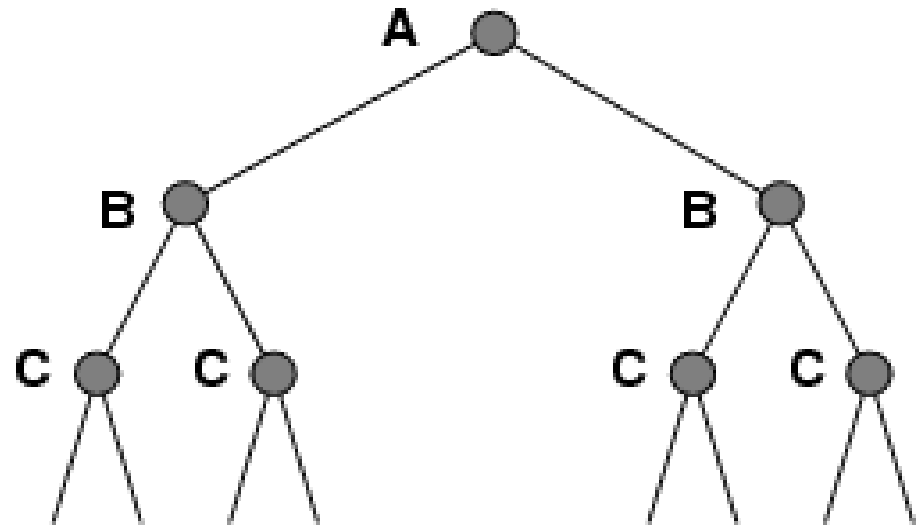
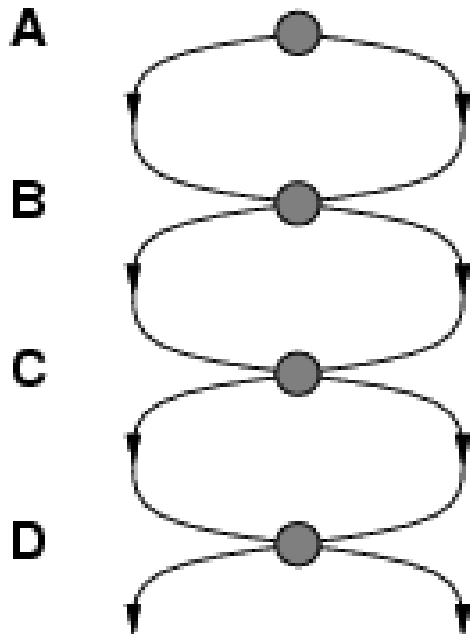
# Summary of algorithms

| Criterion | Breadth-First | Uniform-Cost                        | Depth-First | Depth-Limited | Iterative Deepening |
|-----------|---------------|-------------------------------------|-------------|---------------|---------------------|
| Complete? | Yes           | Yes                                 | No          | No            | Yes                 |
| Time      | $O(b^{d+1})$  | $O(b^{\lceil C^*/\epsilon \rceil})$ | $O(b^m)$    | $O(b^l)$      | $O(b^d)$            |
| Space     | $O(b^{d+1})$  | $O(b^{\lceil C^*/\epsilon \rceil})$ | $O(bm)$     | $O(bl)$       | $O(bd)$             |
| Optimal?  | Yes           | Yes                                 | No          | No            | Yes                 |

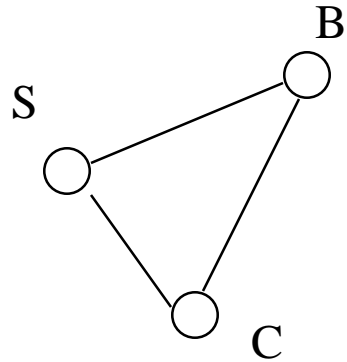


# Repeated states

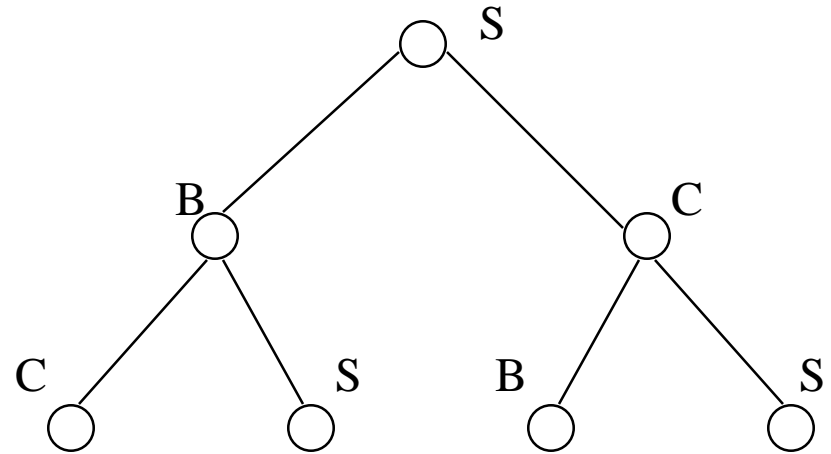
- Failure to detect repeated states can turn a linear problem into an exponential one!



# Solutions to Repeated States



State Space



Example of a Search Tree

- Method 1 ← suboptimal but practical
  - do not create paths containing cycles (loops)
- Method 2 ← optimal but memory inefficient
  - never generate a state generated before
    - must keep track of all possible states (uses a lot of memory)
    - e.g., 8-puzzle problem, we have  $9! = 362,880$  states