

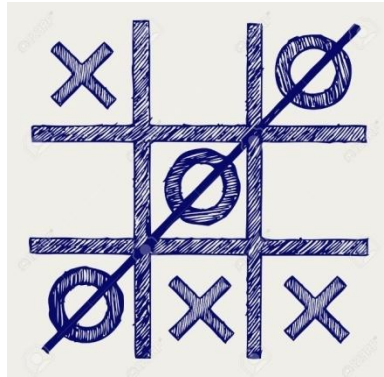
**CSE422: Artificial intelligence**

# **Problem Solving as Search**

**Uninformed Search Techniques**

**Asif Shahriar**  
**Lecturer, CSE, BRACU**

# Why searching in AI?

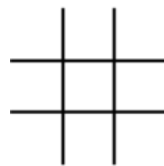


Maybe you all are familiar with the board game Tic-Tac-Toe.

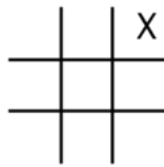
As an AI student, you want to design an intelligent agent, which can play this game with you as an opponent.

# Why searching in AI?

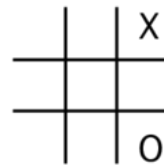
Intelligent agent should know the current state of the Tic-Tac-Toe board.



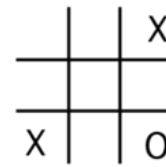
Before  
game  
begins



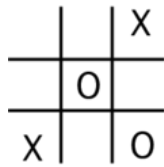
X's  
first  
move



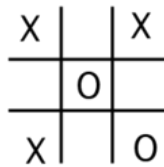
O's  
first  
move



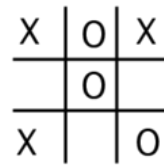
X's  
second  
move



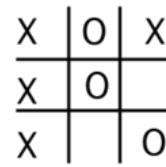
O's  
second  
move



X's  
third  
move



O's  
third  
move



X wins on  
X's fourth  
move

Fig: Some states of Tic-Tac-Toe board.

# Why searching in AI?

Intelligent agent should explore the next possible movement and search for the best move to win the game.

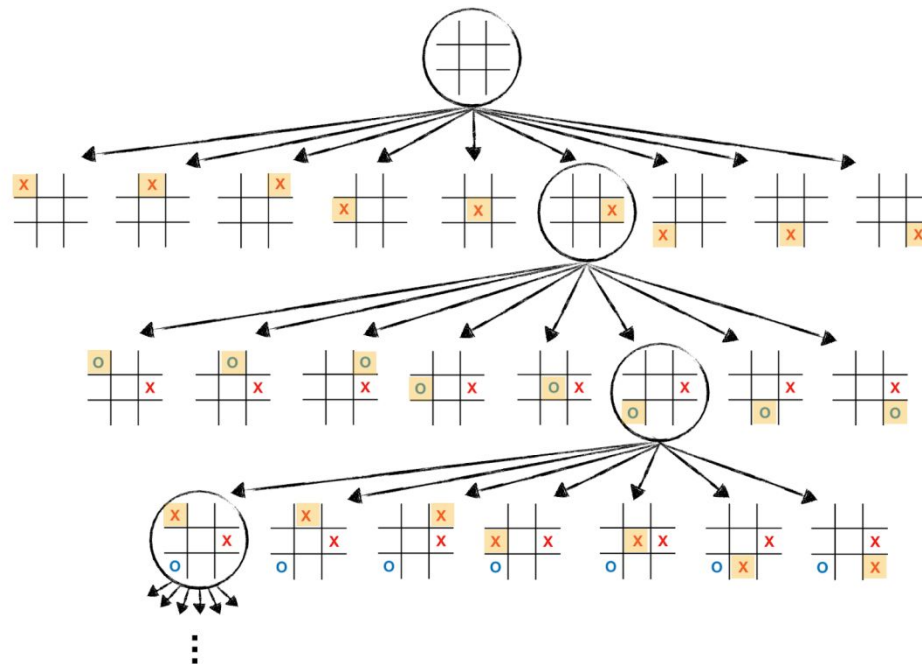
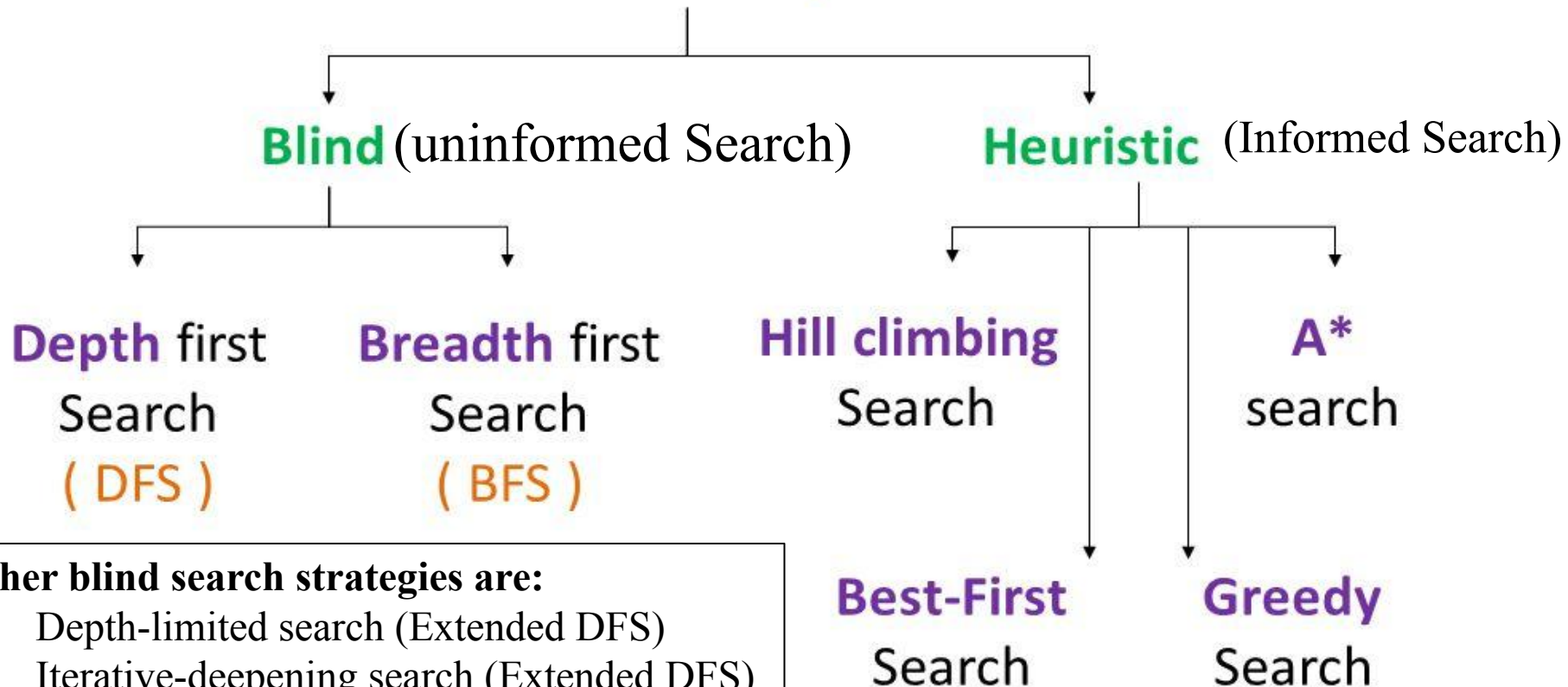


Fig: Game search tree for Tic-Tac-Toe board (partial view)

# SEARCH TECHNIQUES

## Search techniques



### Other blind search strategies are:

- Depth-limited search (Extended DFS)
- Iterative-deepening search (Extended DFS)
- Uniform cost search
- Bi-directional search

# Uninformed Vs Informed Search

**Uninformed search:** Use only the information available in the problem definition. Example: breadth-first, depth-first, depth limited, iterative deepening, uniform cost and bidirectional search

**Informed search:** Use domain knowledge or heuristic to choose the best move. Example. Greedy best-first, A\*, IDA\*, and beam search

## **Additional Note:**

**optimization** in which the search is to find an optimal value of an objective function: hill climbing, simulated annealing, genetic algorithms, Ant Colony Optimization

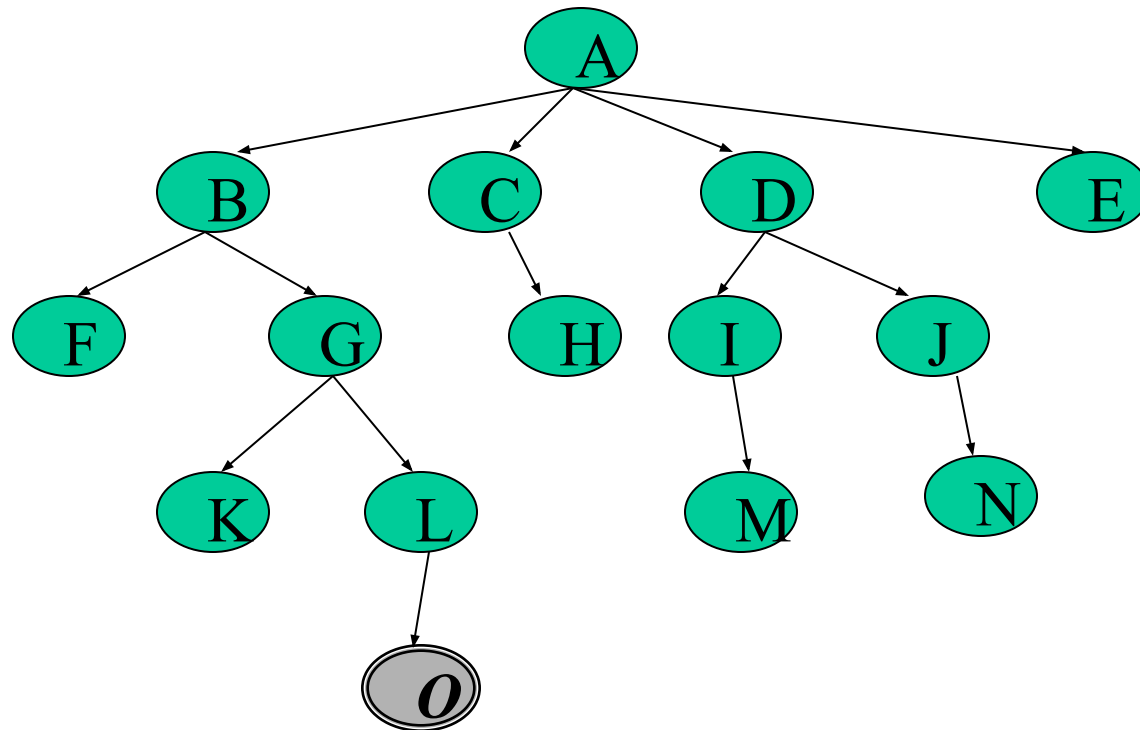
**Game playing**, an adversarial search: minimax algorithm, alpha-beta pruning

# Uninformed Search

# Breadth First Search

- Application 1:

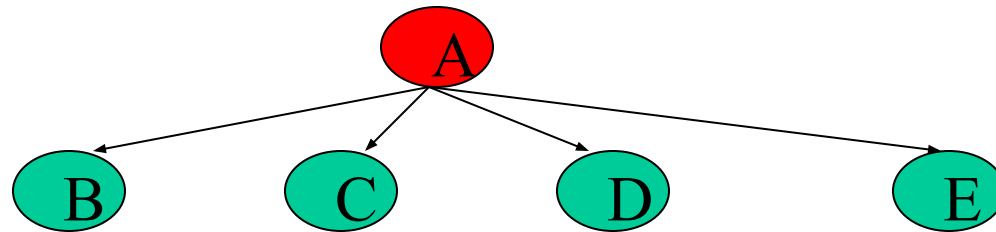
Given the following state space (tree search), give the sequence of visited nodes when using BFS (assume that the node **O** is the goal state):





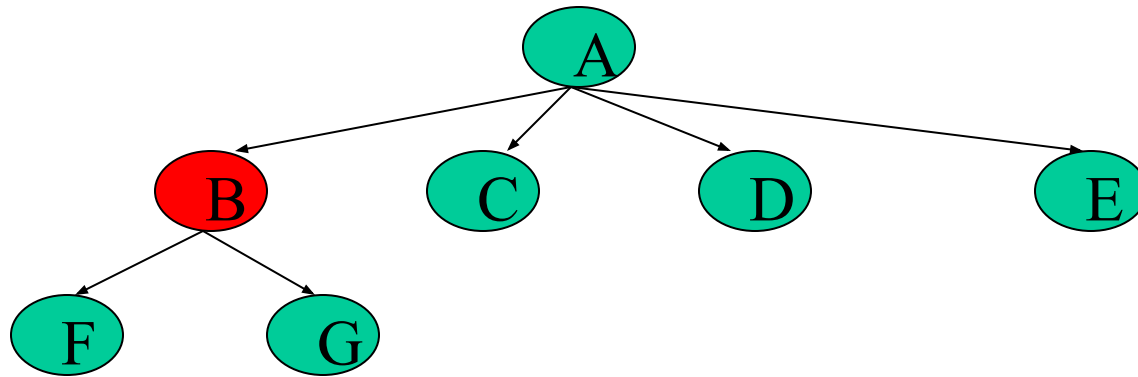
# Breadth First Search

- A,



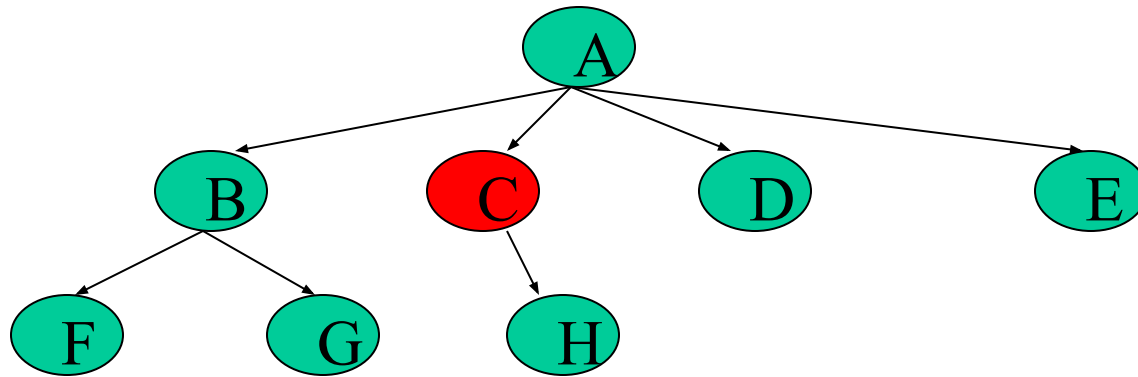
# Breadth First Search

- A,
- B,



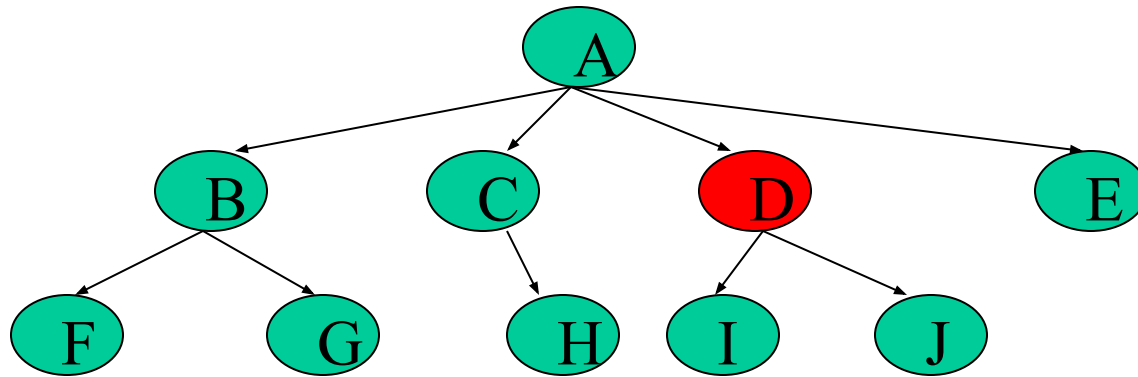
# Breadth First Search

- A,
- B,C



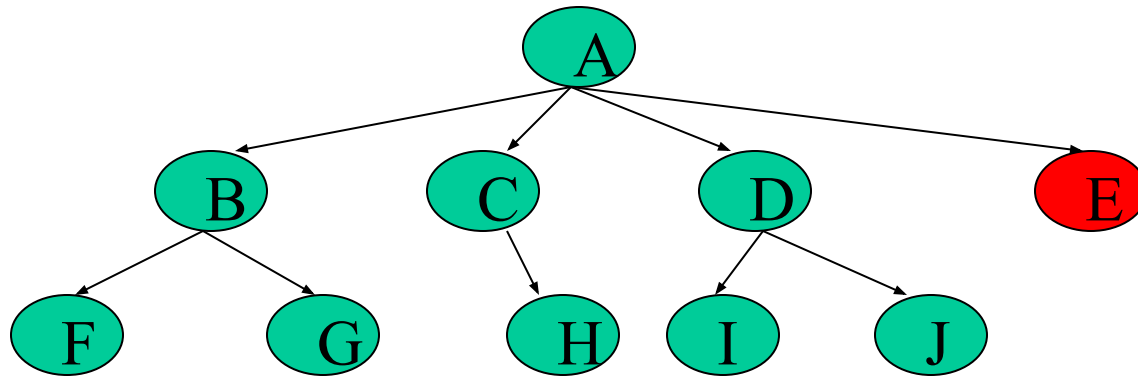
# Breadth First Search

- A,
- B,C,D



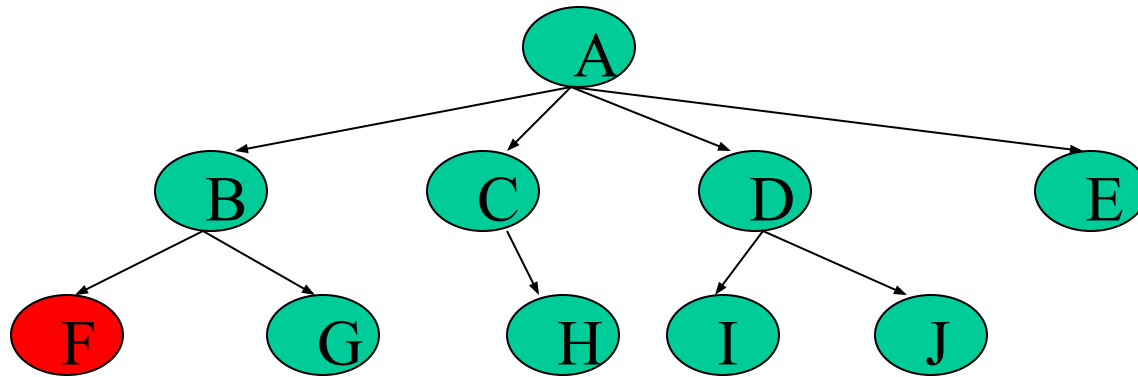
# Breadth First Search

- A,
- B,C,D,E



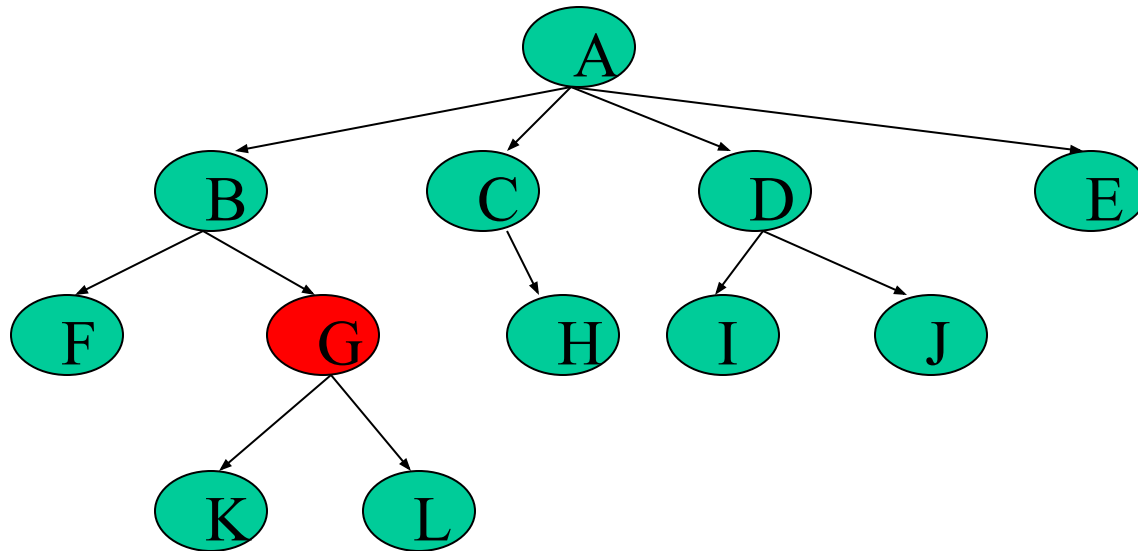
# Breadth First Search

- A,
- B,C,D,E,
- F,



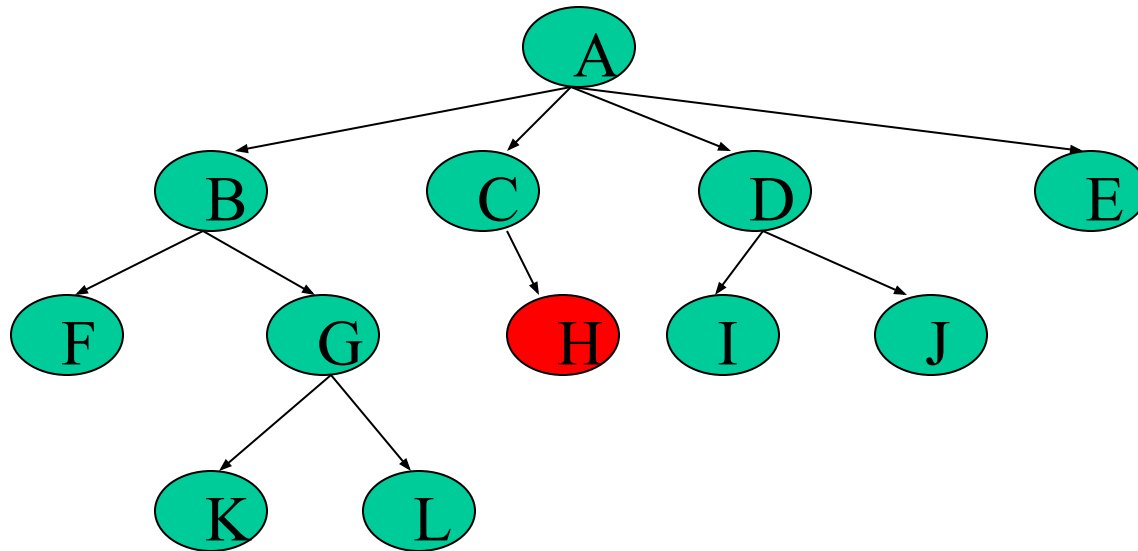
# Breadth First Search

- A,
- B,C,D,E,
- F,G



# Breadth First Search

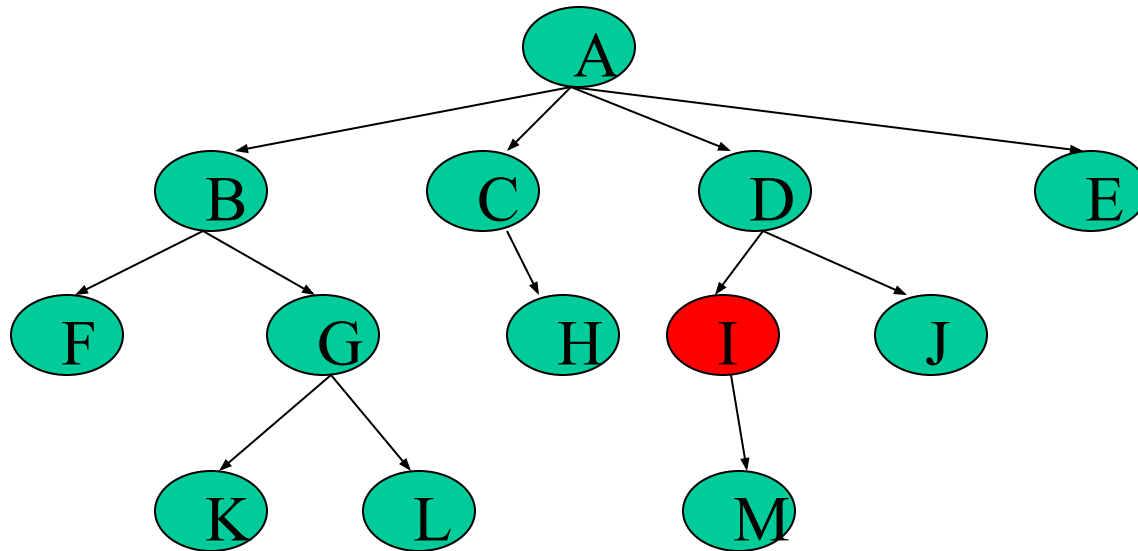
- A,
- B,C,D,E,
- F,G,H





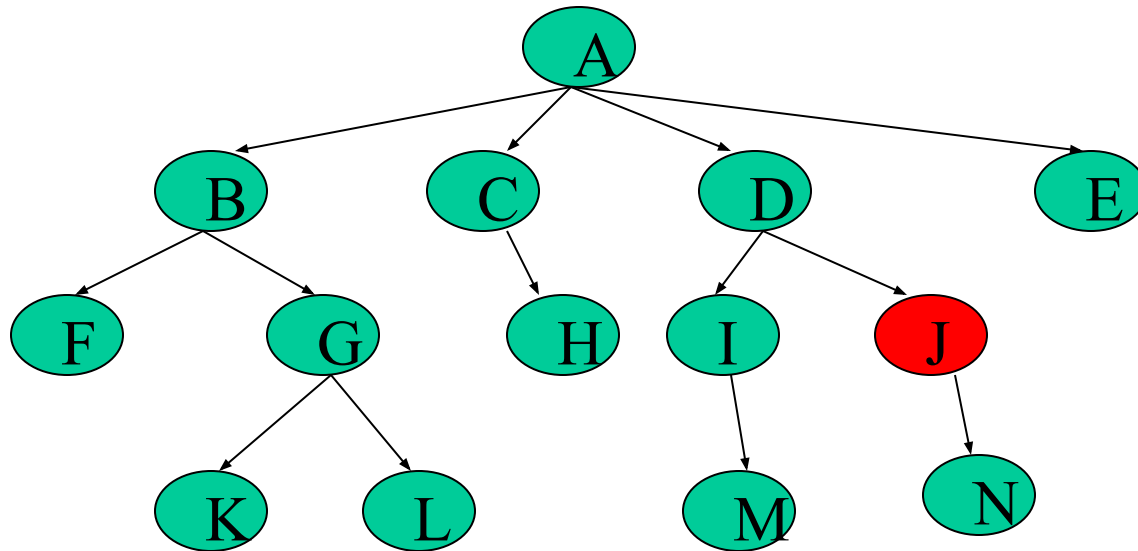
# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I



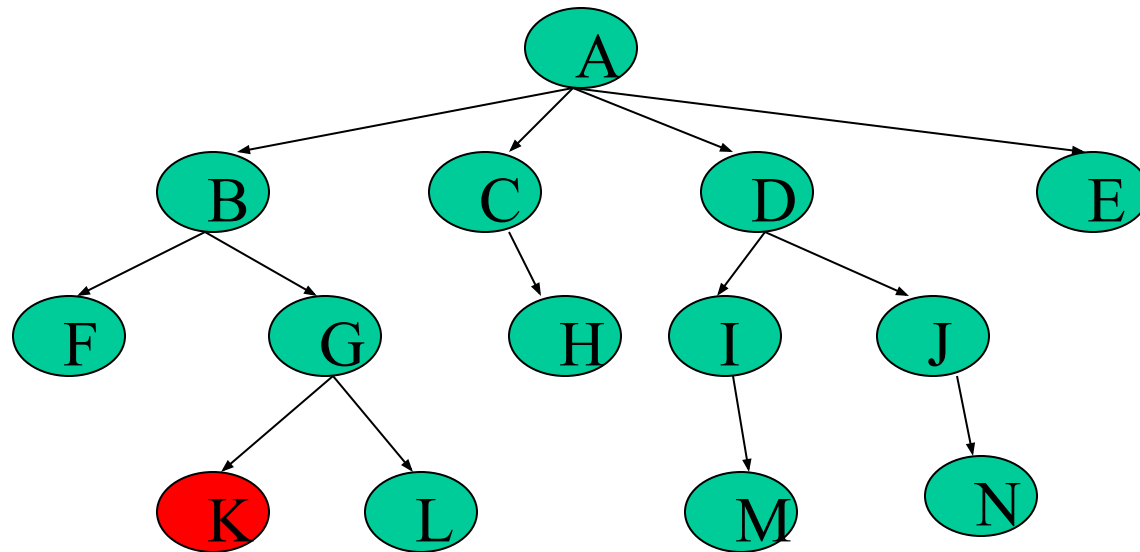
# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I,J,



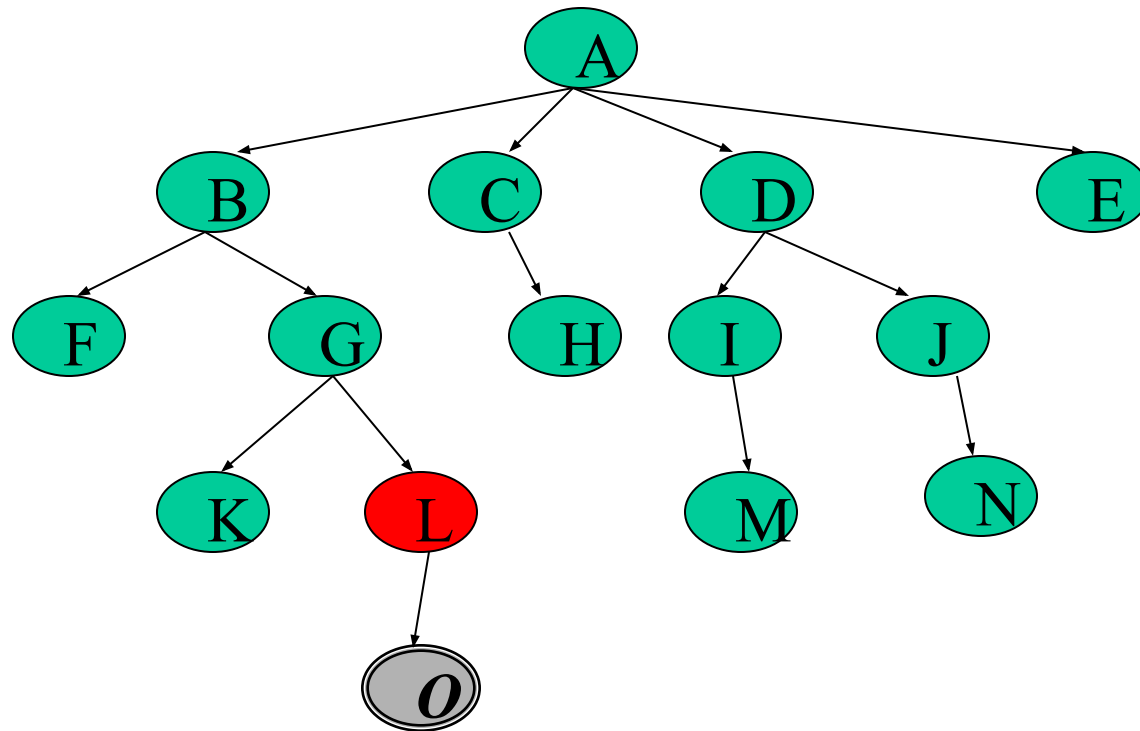
# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I,J,
- K,



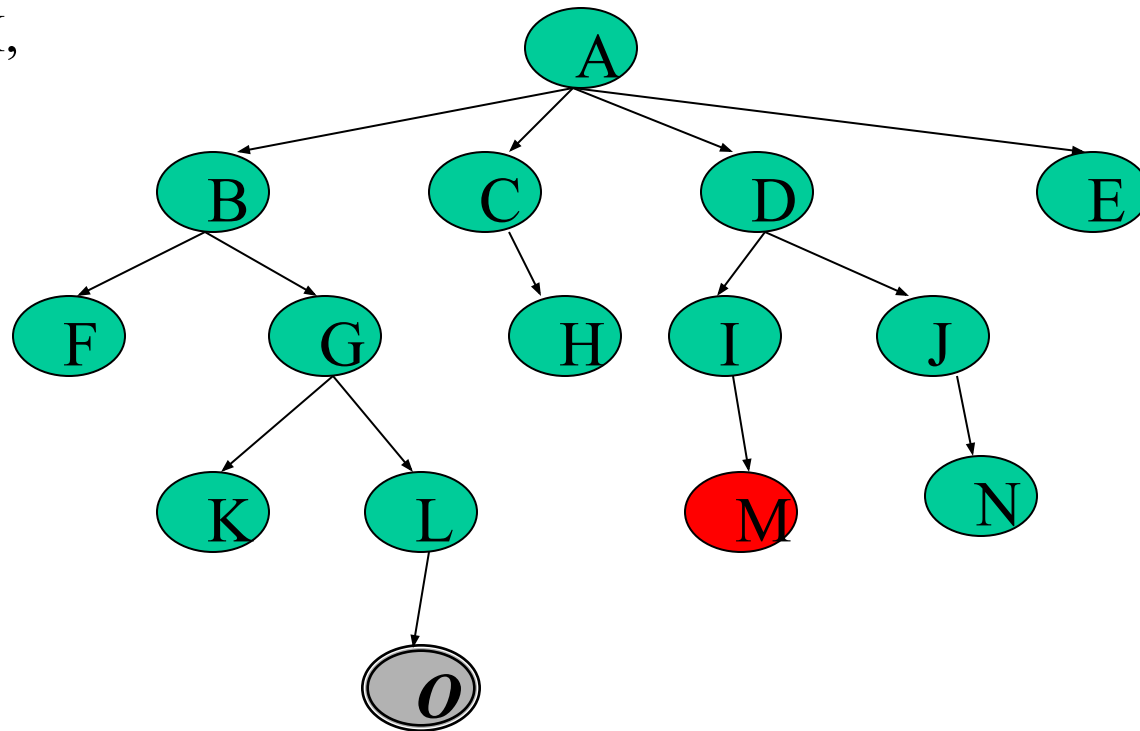
# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L

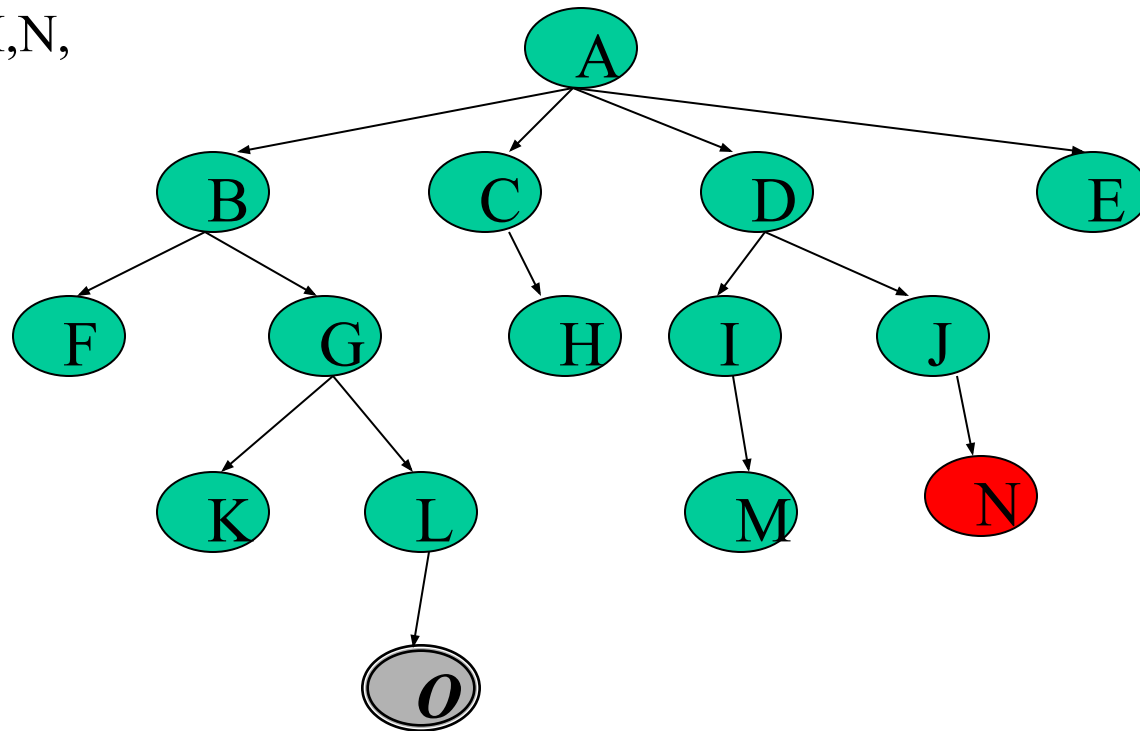


# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L, M,

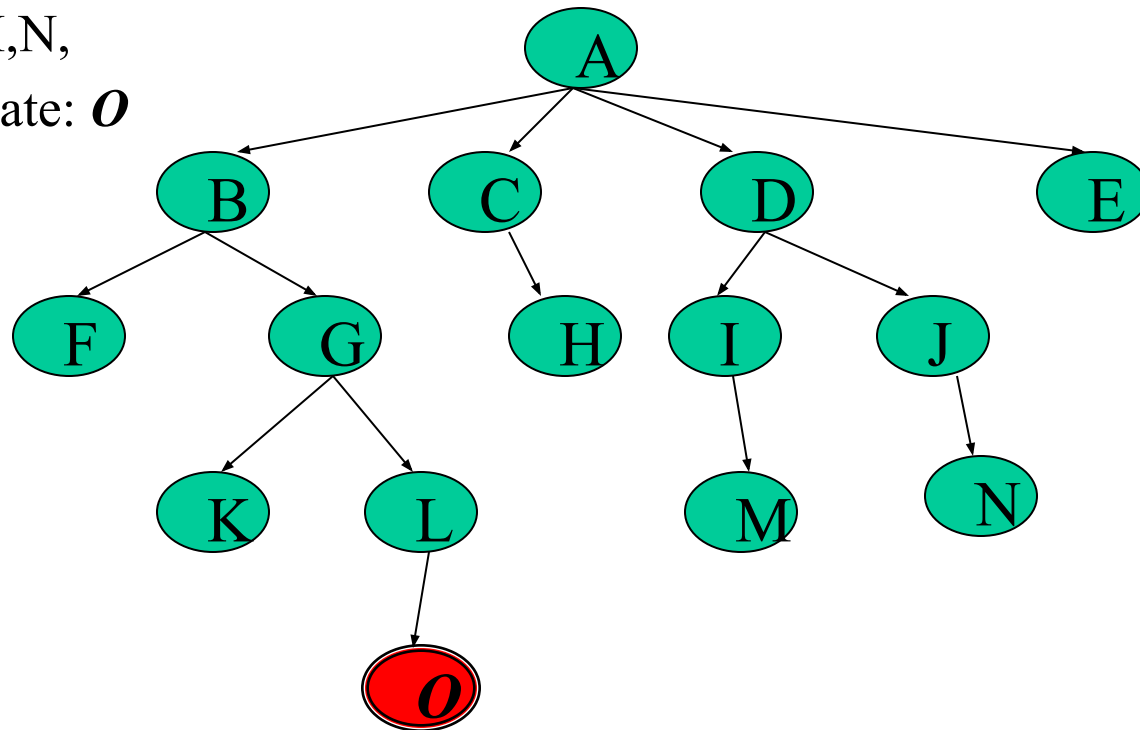


- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L, M,N,



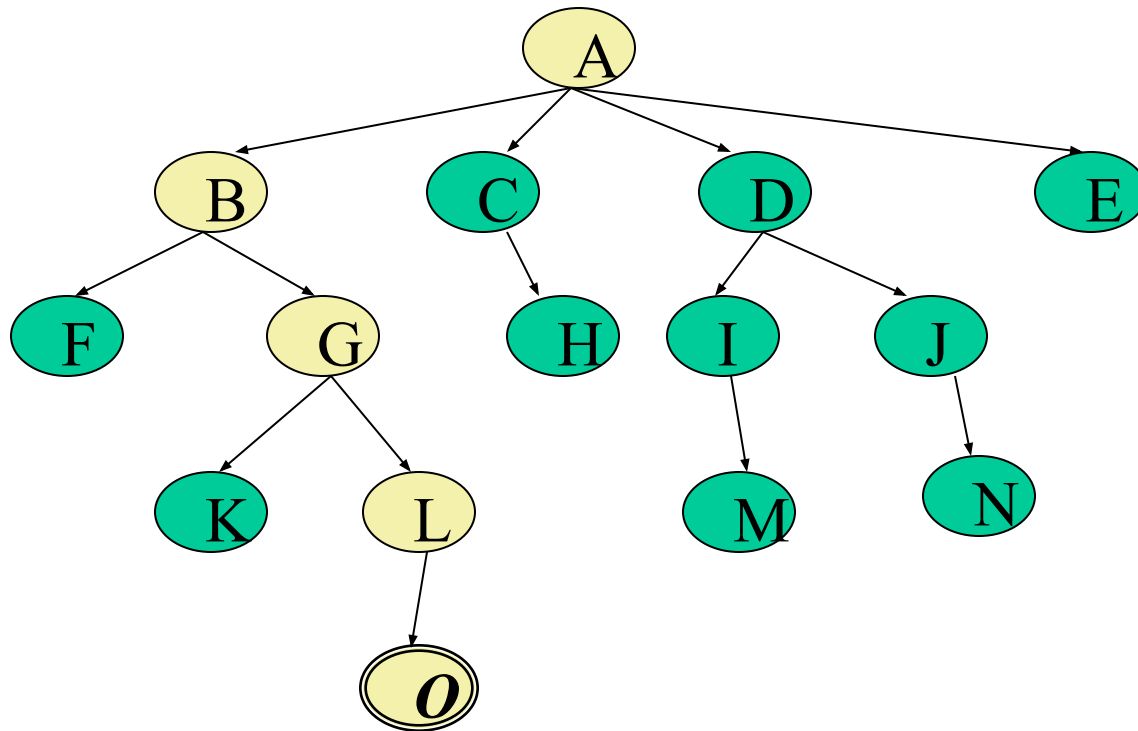
# Breadth First Search

- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L, M,N,
- Goal state: ***O***



# Breadth First Search

- The returned solution is the sequence of operators in the path:  
*A, B, G, L, O*

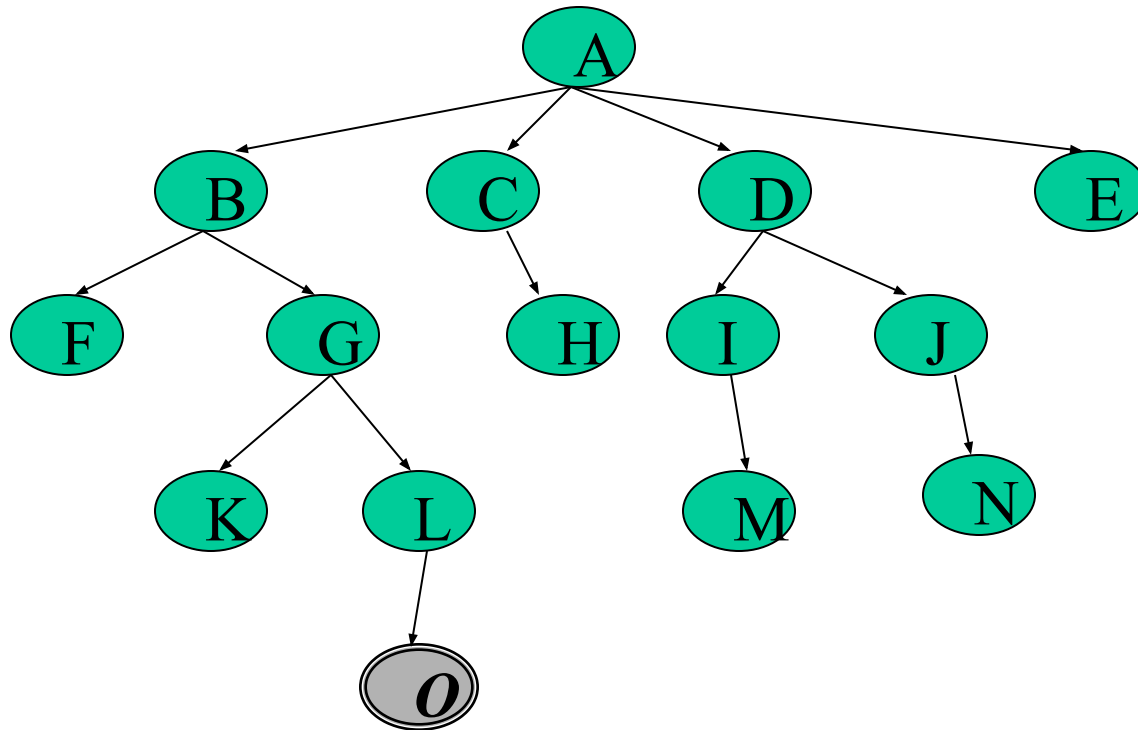




# Depth First Search (DFS)

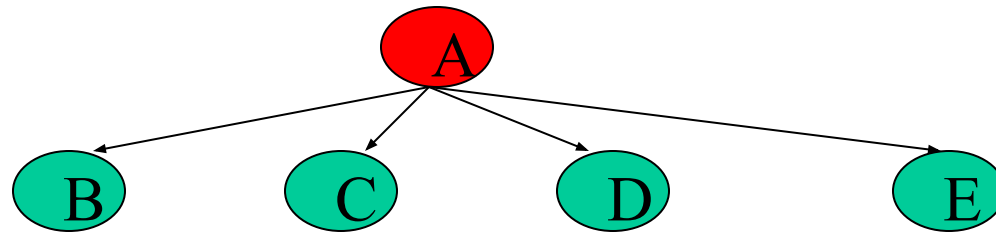
- Application2:

Given the following state space (tree search), give the sequence of visited nodes when using DFS (assume that the node **O** is the goal state):



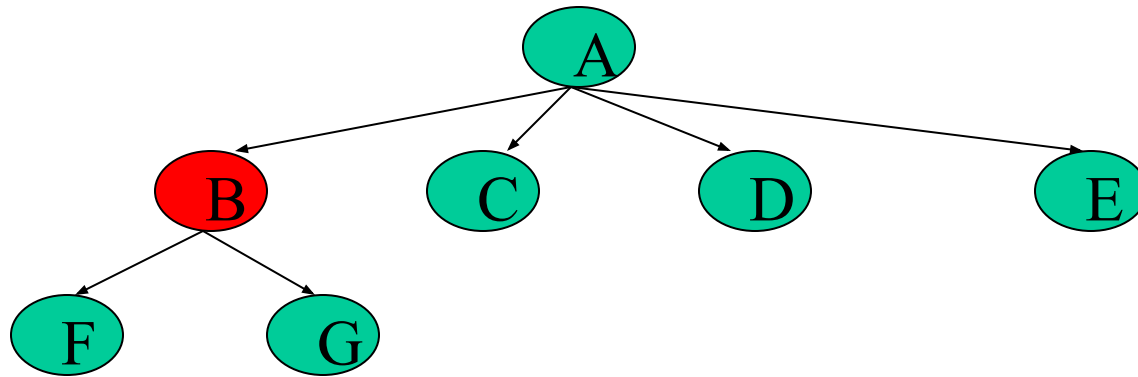
# Depth First Search

- A,



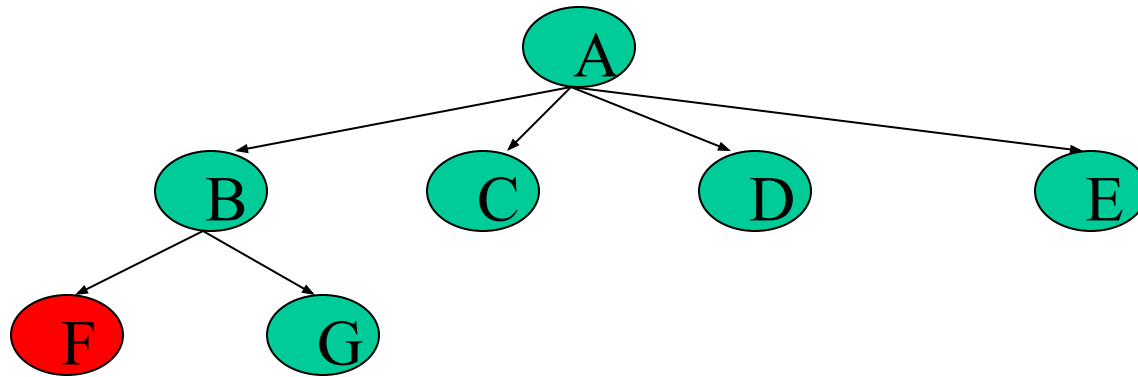
# Depth First Search

- A,B,



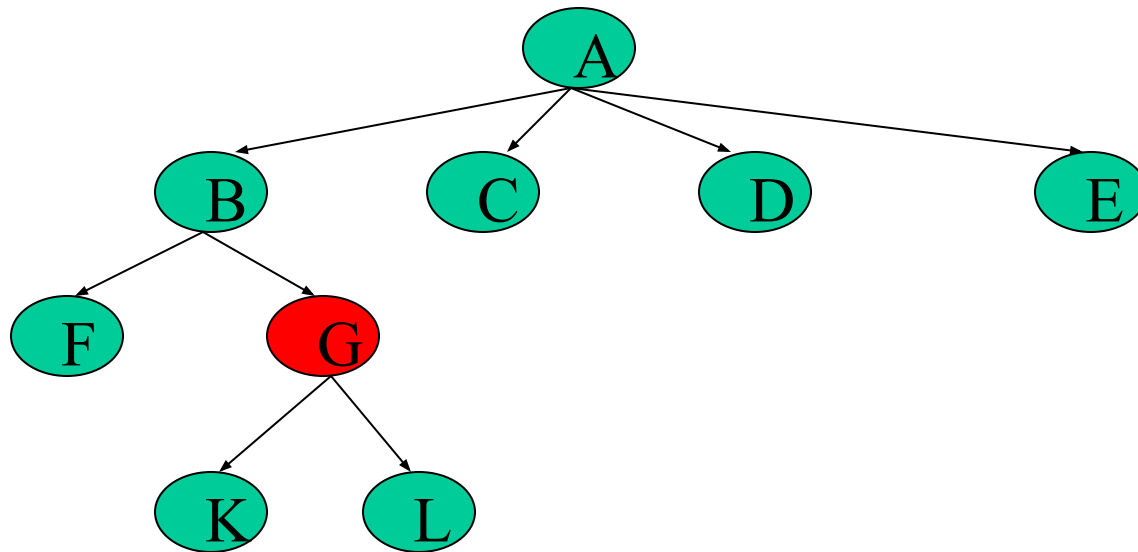
# Depth First Search

- A,B,F,



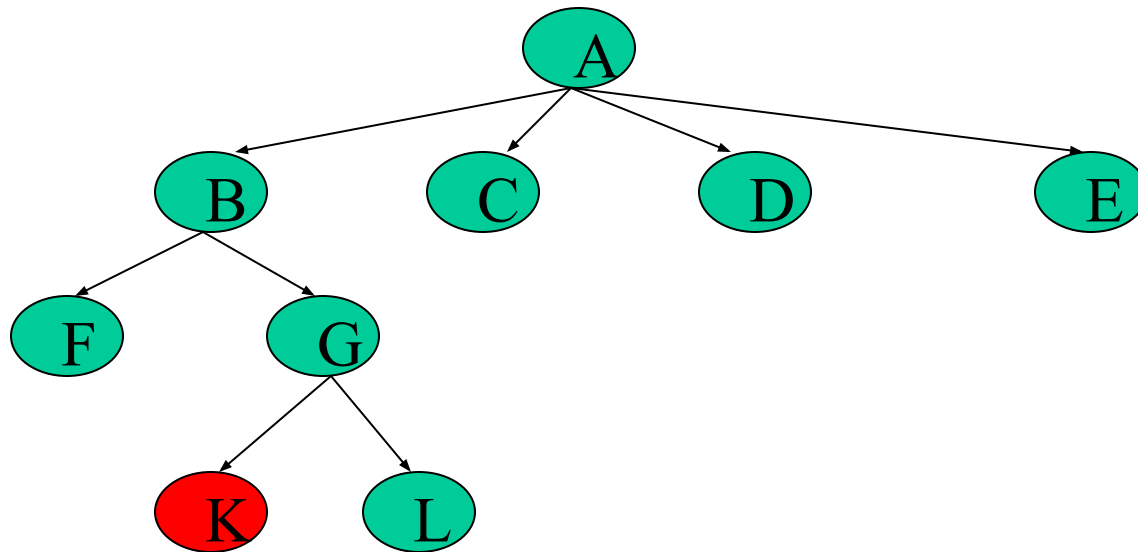
# Depth First Search

- A,B,F,
- G,



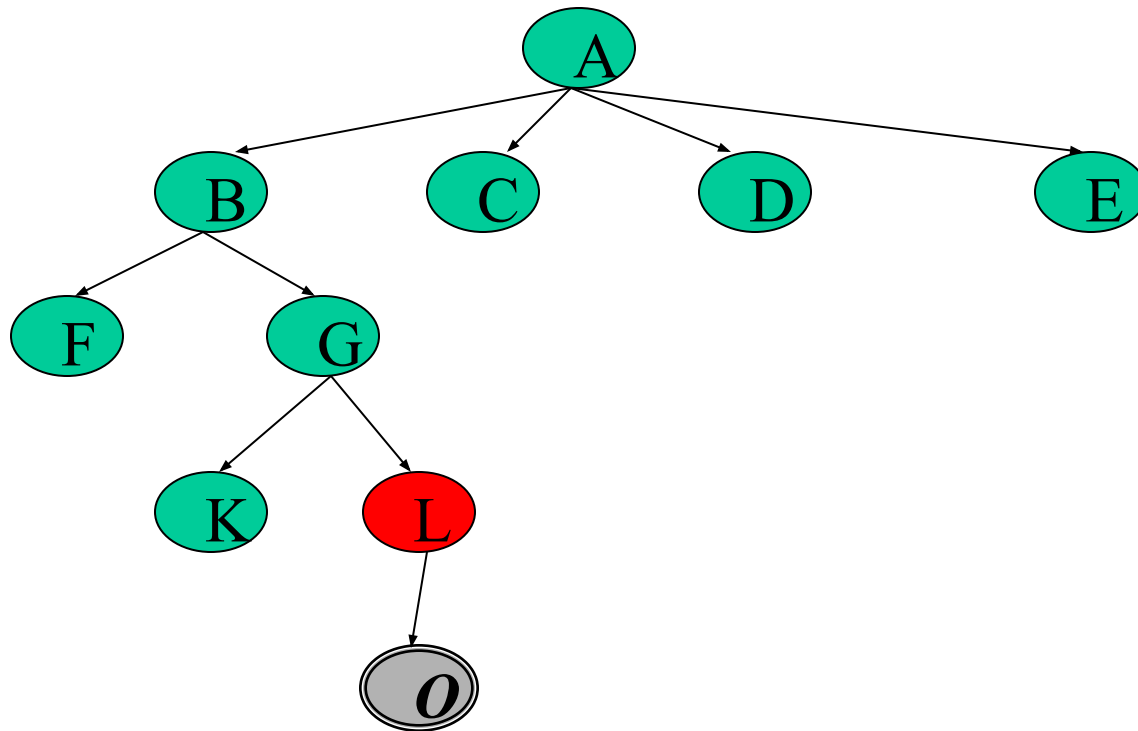
# Depth First Search

- A,B,F,
- G,K,



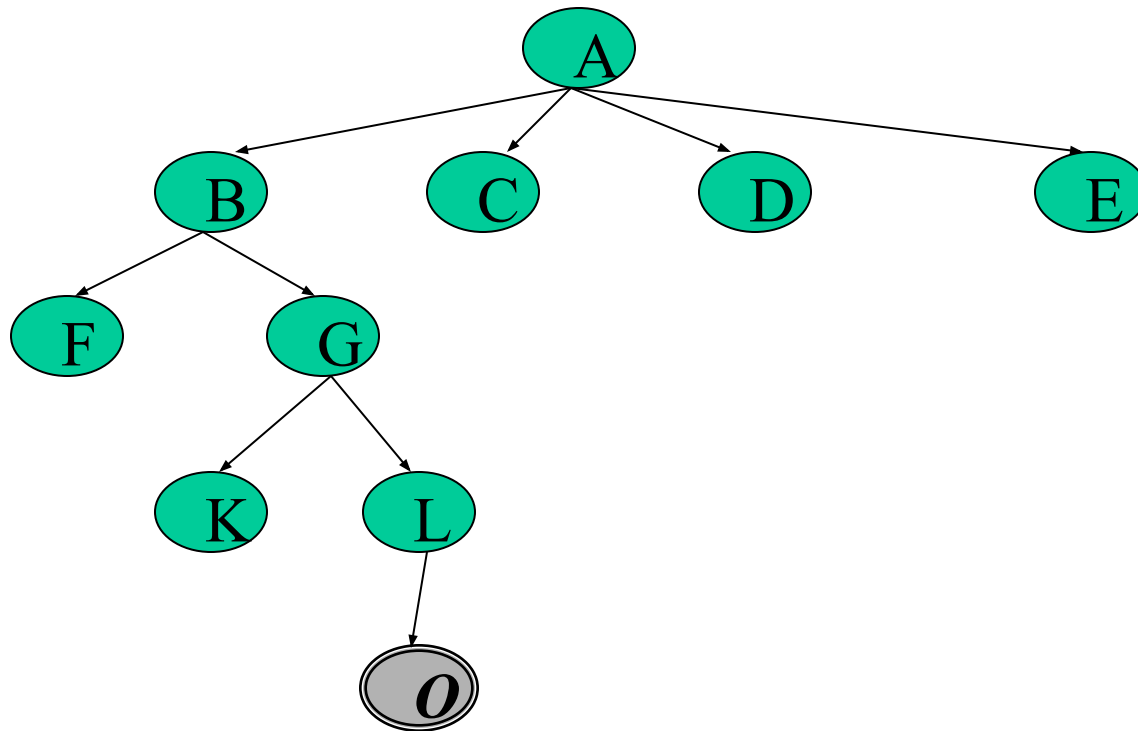
# Depth First Search

- A,B,F,
- G,K,
- L,



# Depth First Search

- A,B,F,
- G,K,
- L, *O*: Goal State

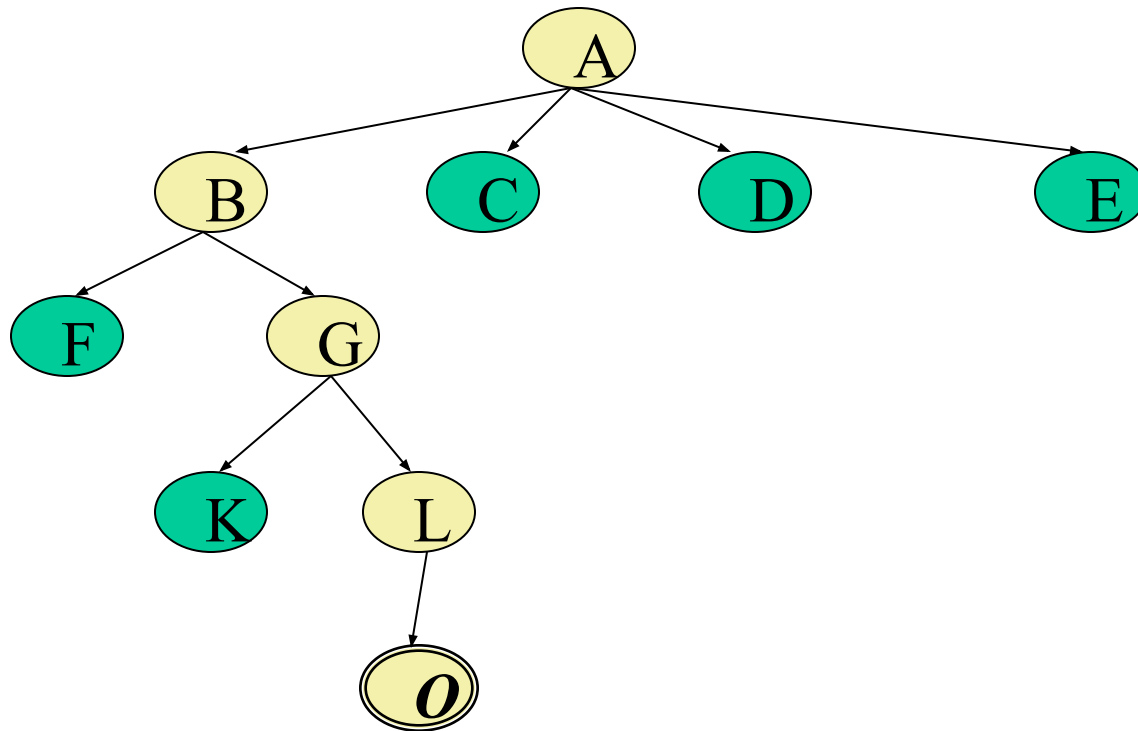




# Depth First Search

The returned solution is the sequence of operators in the path:

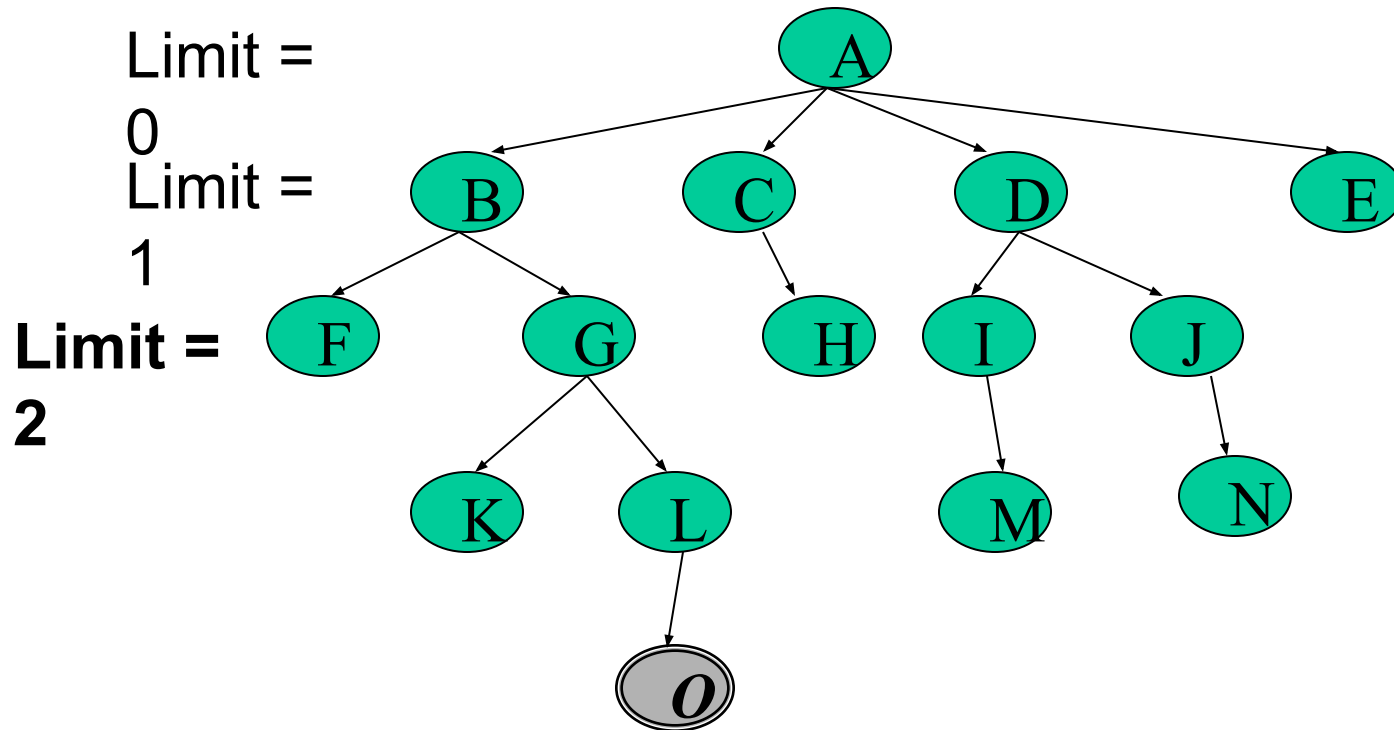
*A, B, G, L, O*



# Depth-Limited Search (DLS)

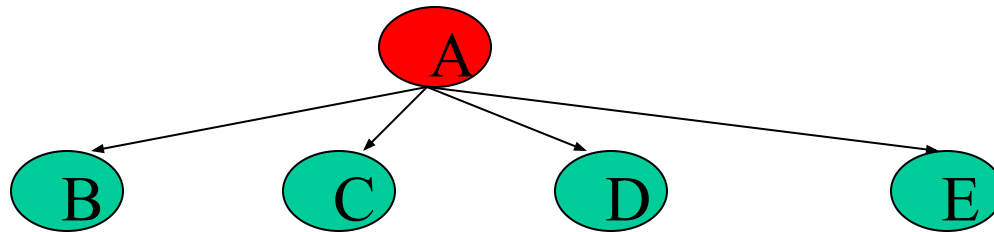
- Application3:

Given the following state space (tree search), give the sequence of visited nodes when using DLS (Limit = 2):



# Depth Limited Search

- A,

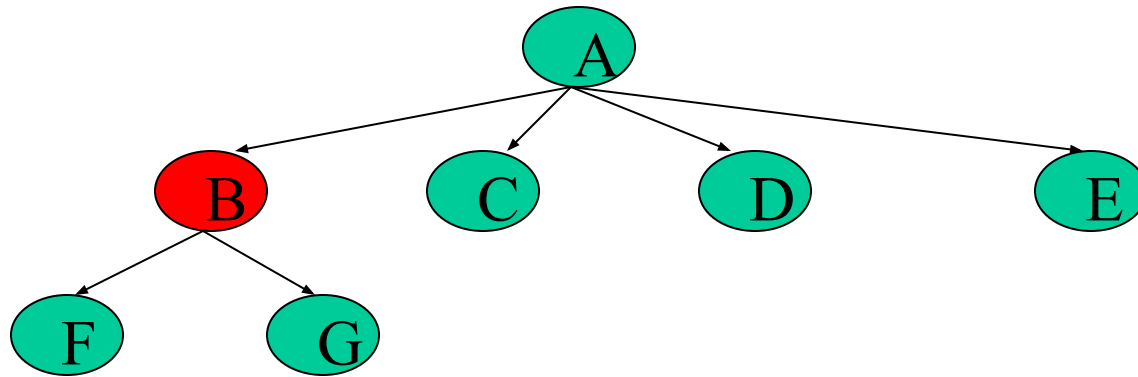


**Limit =**  
**2**

# Depth Limited Search

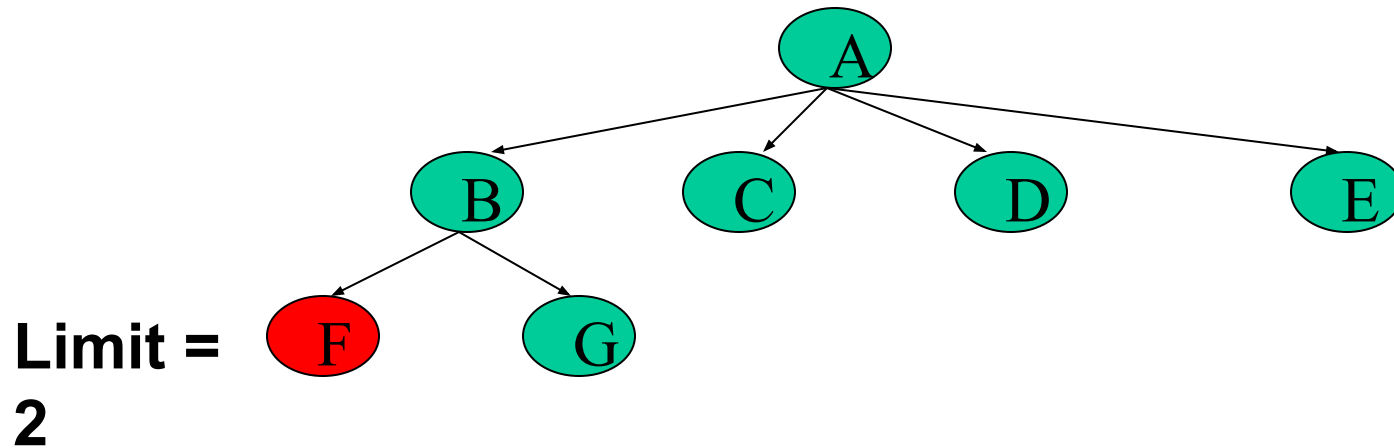
- A,B,

**Limit =  
2**



# Depth Limited Search

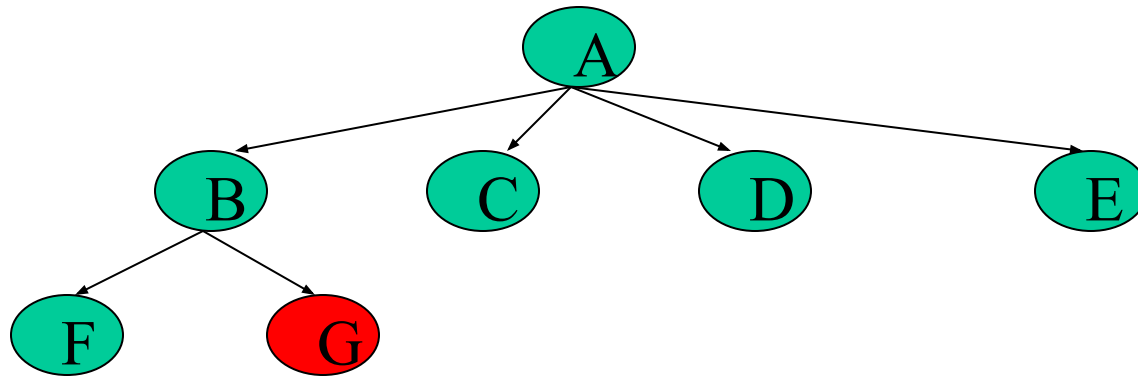
- A,B,F,



# Depth Limited Search

- A,B,F,
- G,

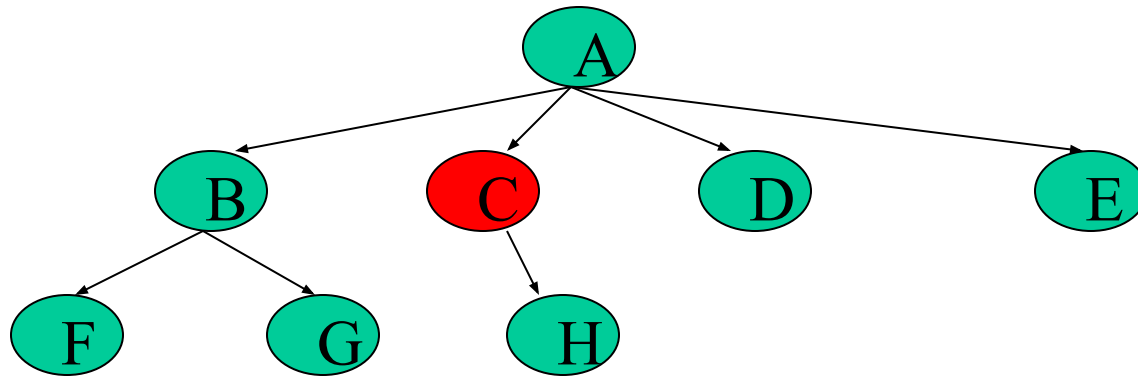
**Limit =**  
**2**



# Depth Limited Search

- A,B,F,
- G,
- C,

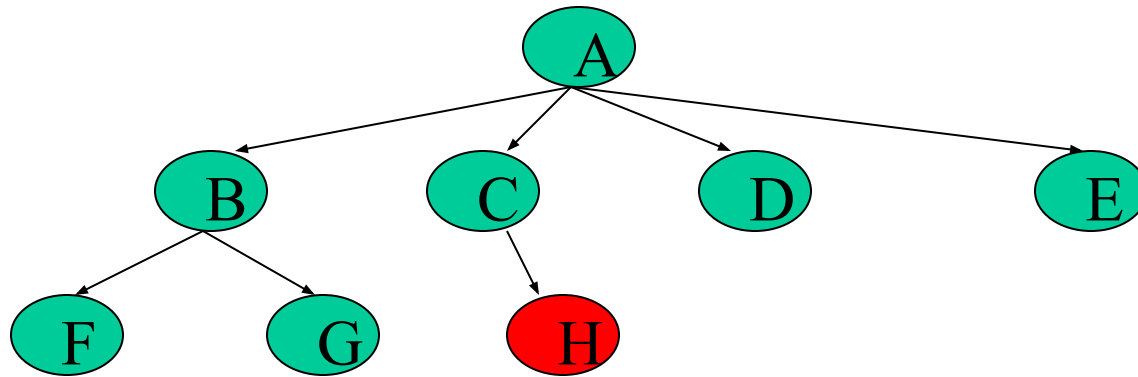
**Limit =  
2**



# Depth Limited Search

- A,B,F,
- G,
- C,H,

**Limit =  
2**

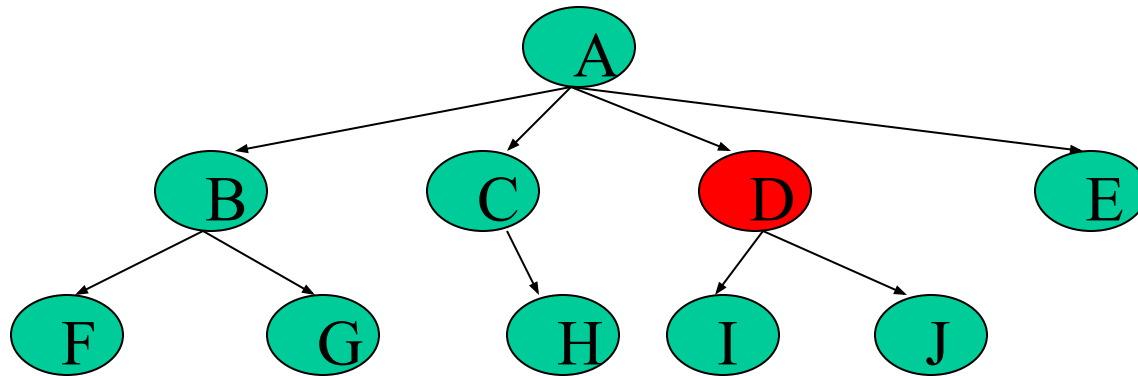




# Depth Limited Search

- A,B,F,
- G,
- C,H,
- D,

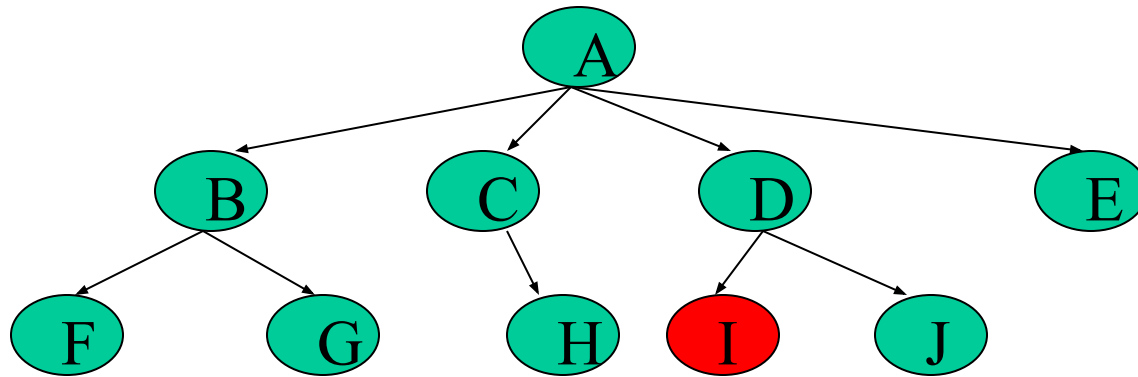
**Limit =  
2**



# Depth Limited Search

- A,B,F,
- G,
- C,H,
- D,I

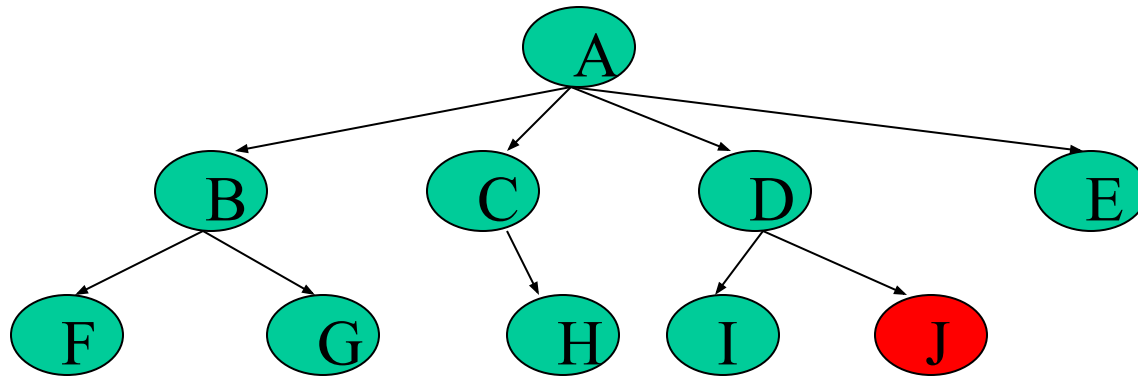
**Limit =  
2**



# Depth Limited Search

- A,B,F,
- G,
- C,H,
- D,I
- J,

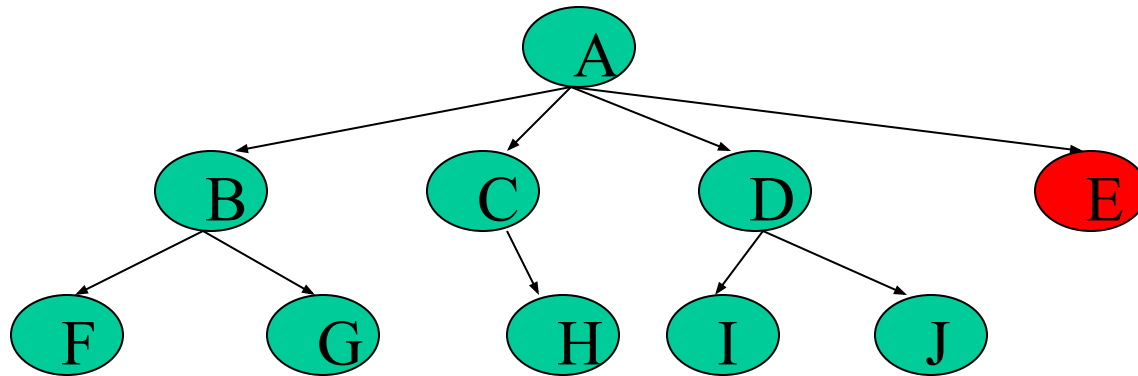
**Limit =  
2**



# Depth Limited Search

- A,B,F,
- G,
- C,H,
- D,I
- J,
- E

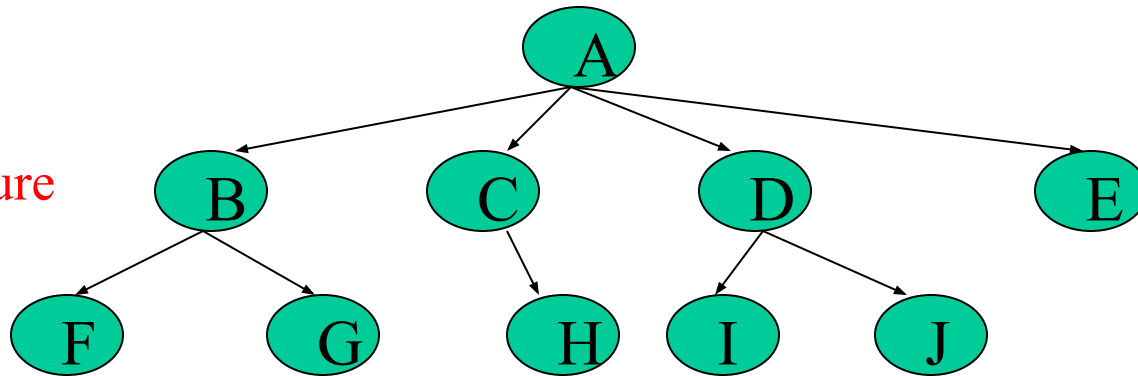
**Limit =  
2**



# Depth Limited Search

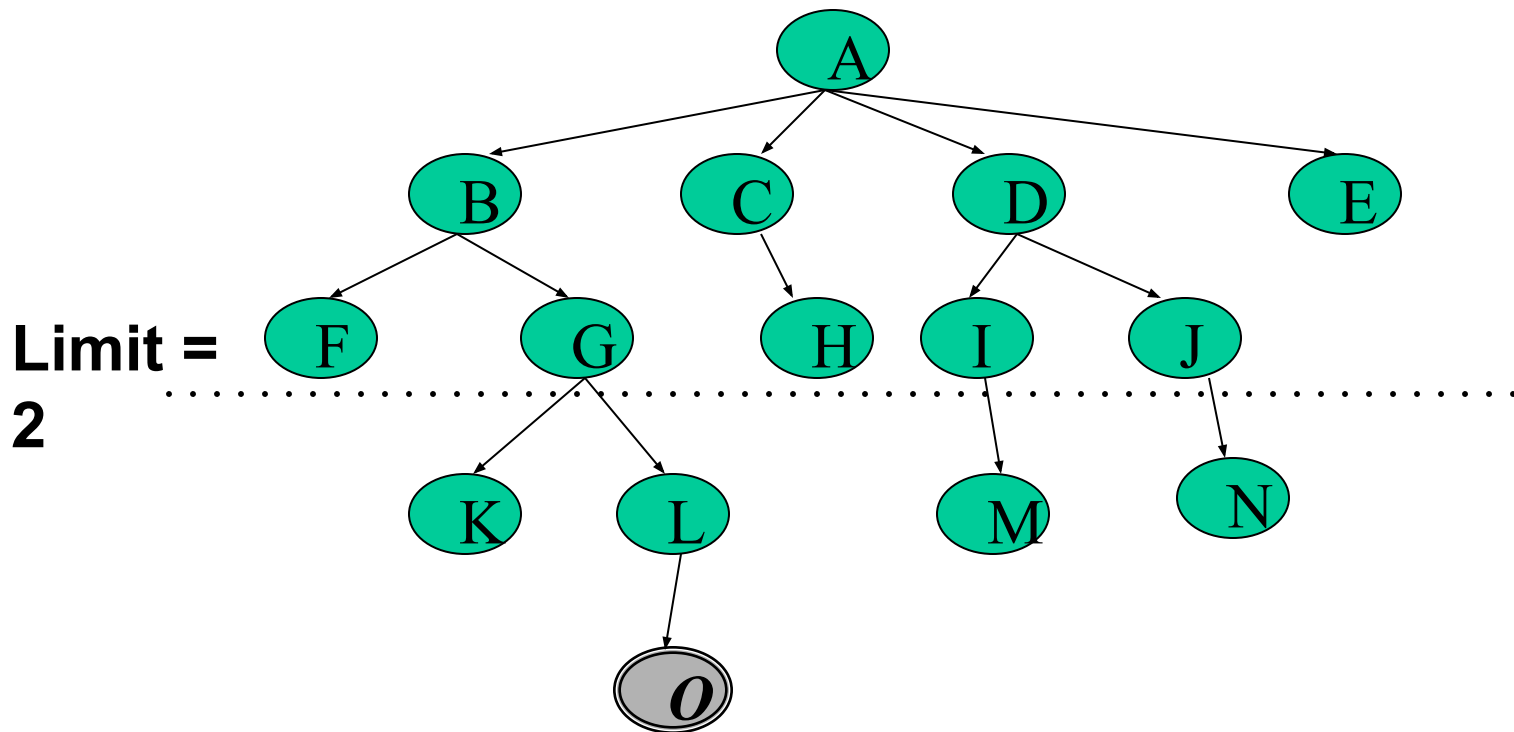
- A,B,F,
- G,
- C,H,
- D,I
- J,
- E, Failure

**Limit =**  
**2**



# Depth Limited Search

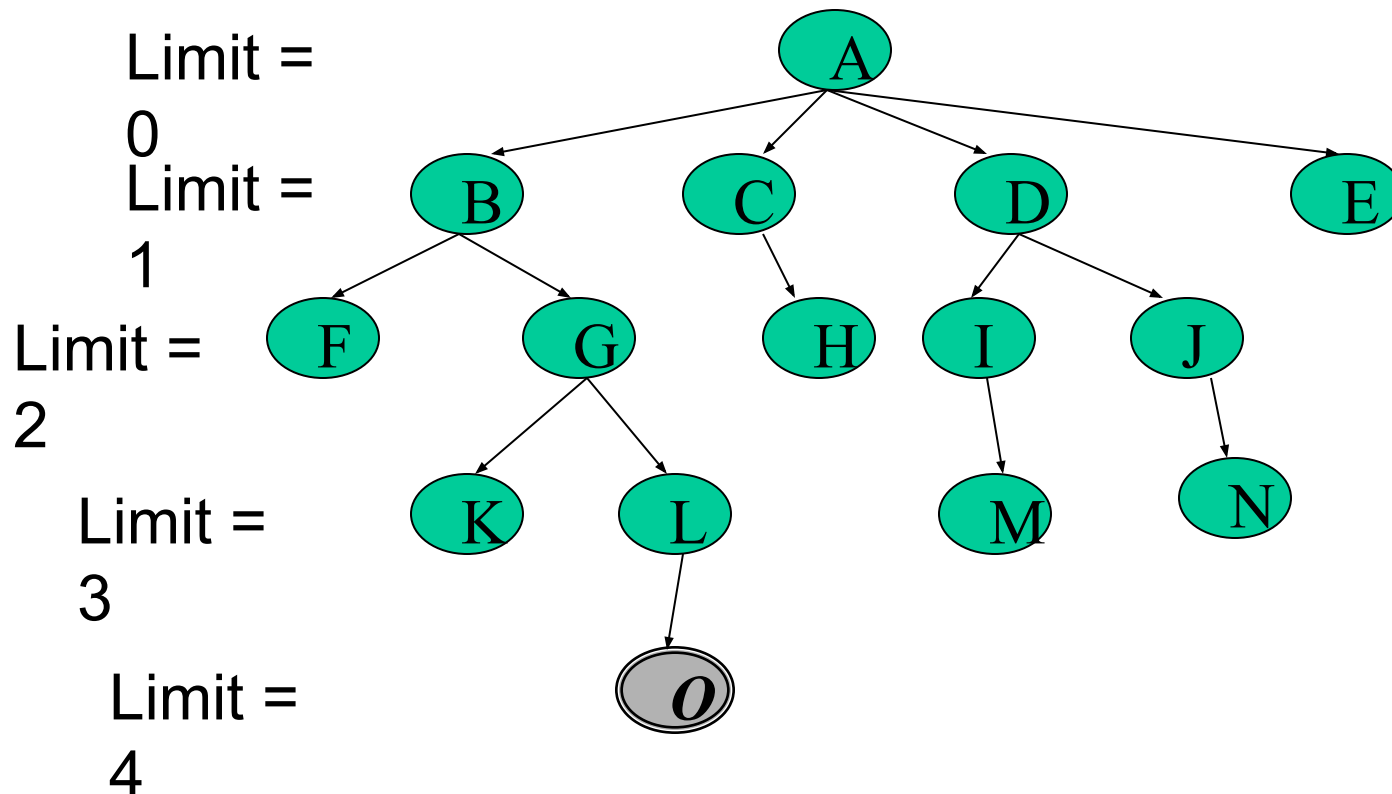
- DLS algorithm returns **Failure (no solution)**
- The reason is that the goal is beyond the limit (Limit = 2): the goal depth is (d=4)



# Iterative Deepening Search (IDS)

- Application4:

Given the following state space (tree search), give the sequence of visited nodes when using IDS:



# Iterative Deepening Search

DLS with bound = 0



# Iterative Deepening Search

- A,

**Limit =**  
**0**



# Iterative Deepening Search

- A, Failure

**Limit =**  
**0**



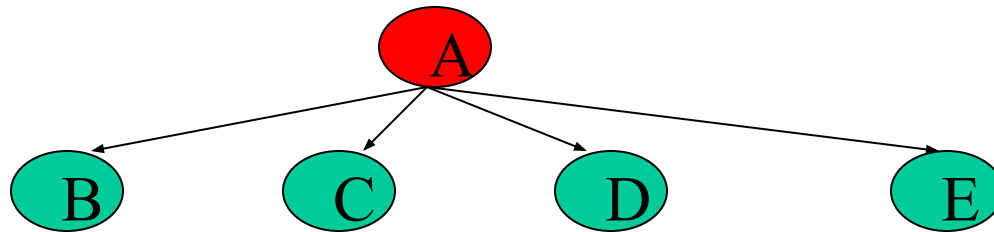
# Iterative Deepening Search (IDS)

DLS with bound = 1

# Iterative Deepening Search

- A,

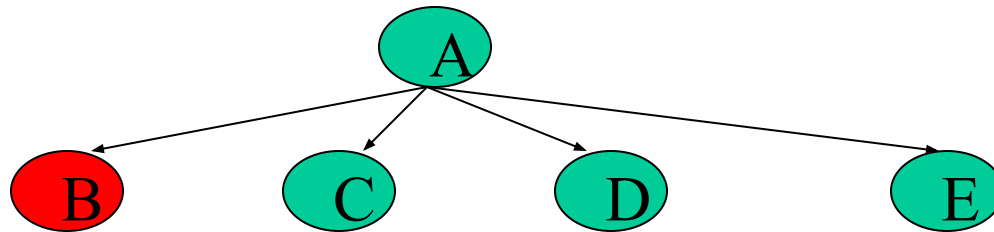
**Limit =  
1**



# Iterative Deepening Search

- A,B,

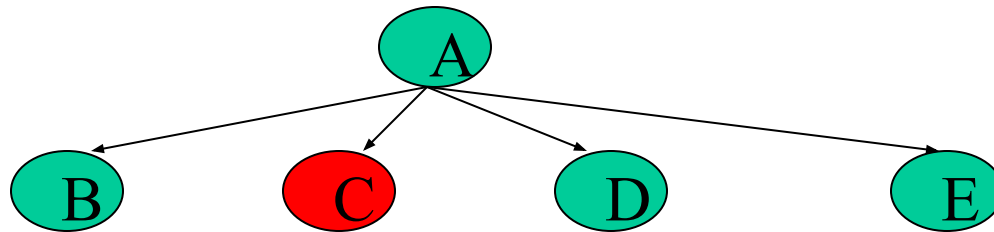
**Limit =  
1**



# Iterative Deepening Search

- A,B,
- C,

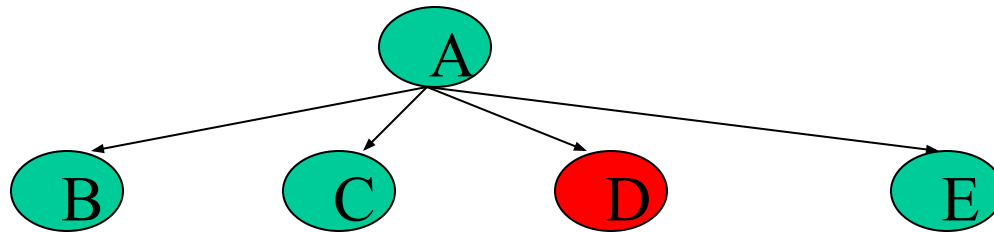
**Limit =  
1**



# Iterative Deepening Search

- A,B,
- C,
- D,

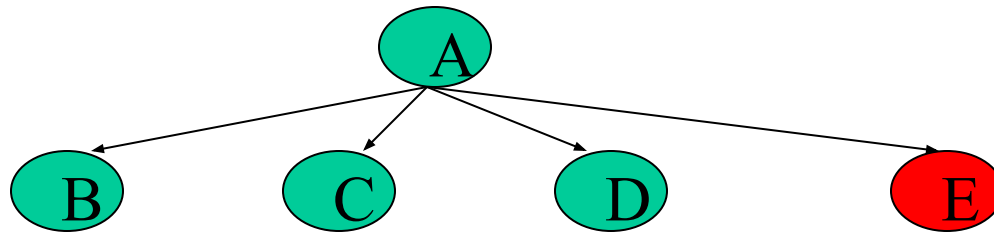
**Limit =**  
**1**



# Iterative Deepening Search

- A,B
- C,
- D,
- E,

**Limit =**  
**1**

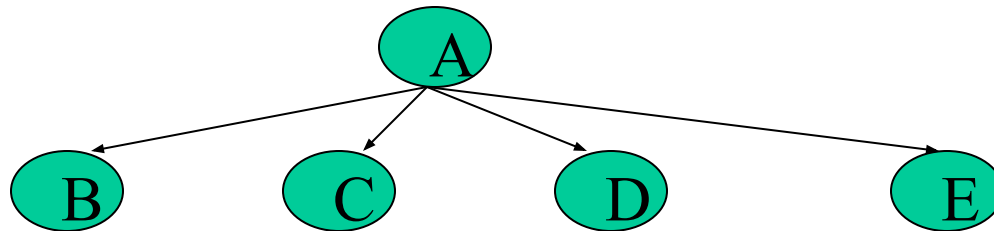




# Iterative Deepening Search

- A,B,
- C,
- D,
- E, Failure

**Limit =**  
**1**

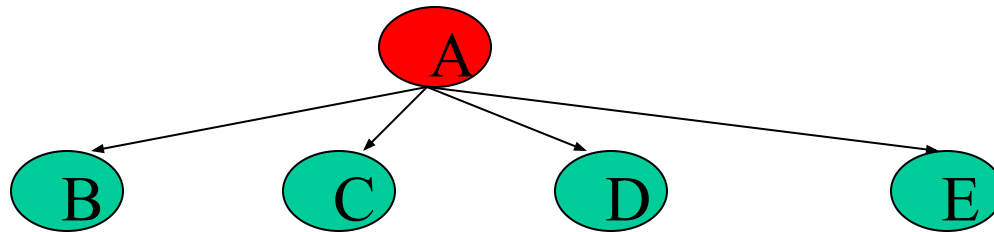


# Iterative Deepening Search (IDS)

DLS with bound = 2

# Iterative Deepening Search

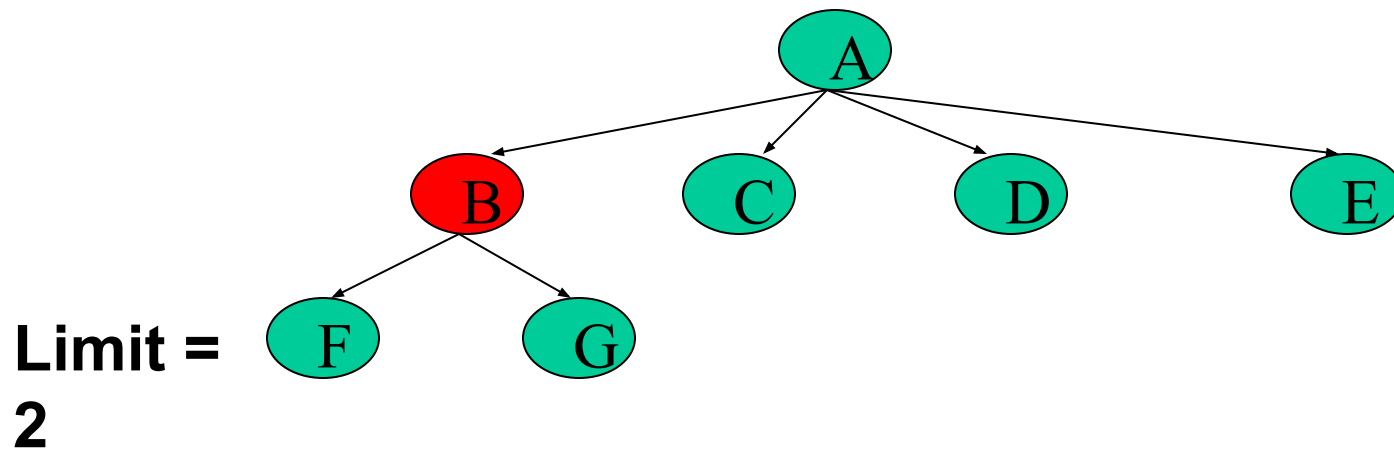
- A,



**Limit =**  
**2**

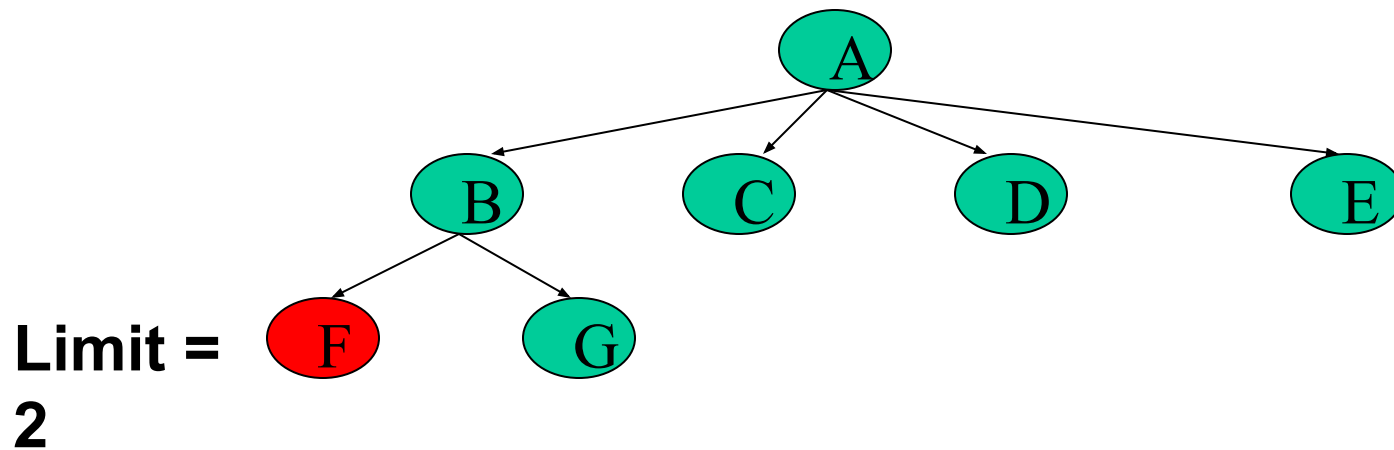
# Iterative Deepening Search

- A,B,



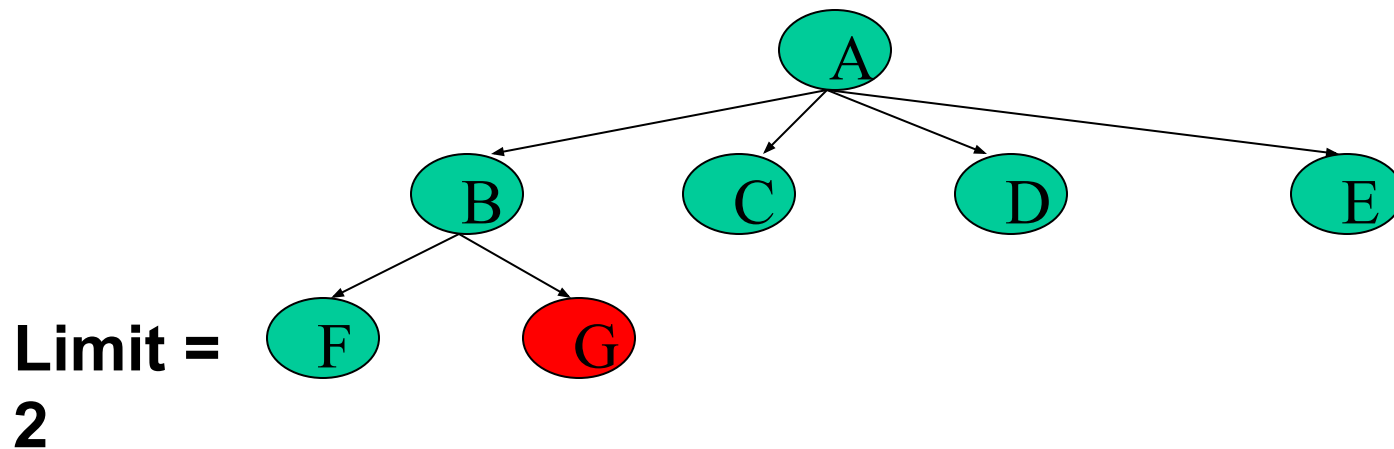
# Iterative Deepening Search

- A,B,F,



# Iterative Deepening Search

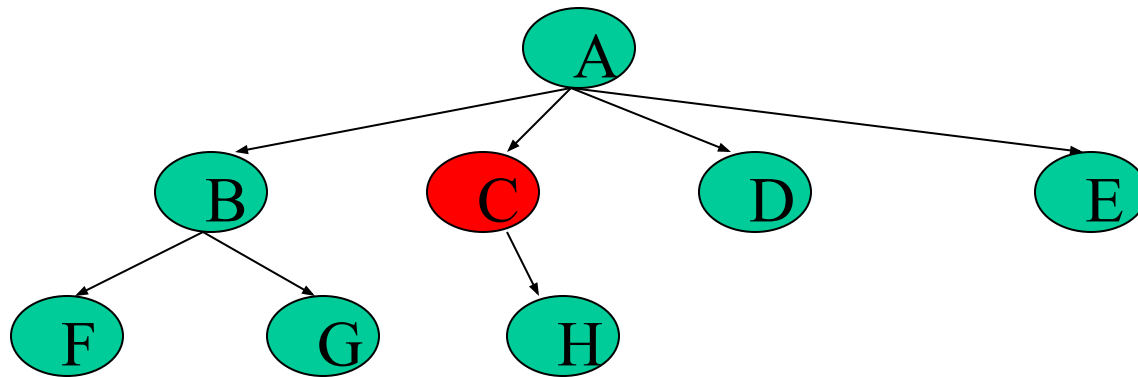
- A,B,F,
- G,



# Iterative Deepening Search

- A,B,F,
- G,
- C,

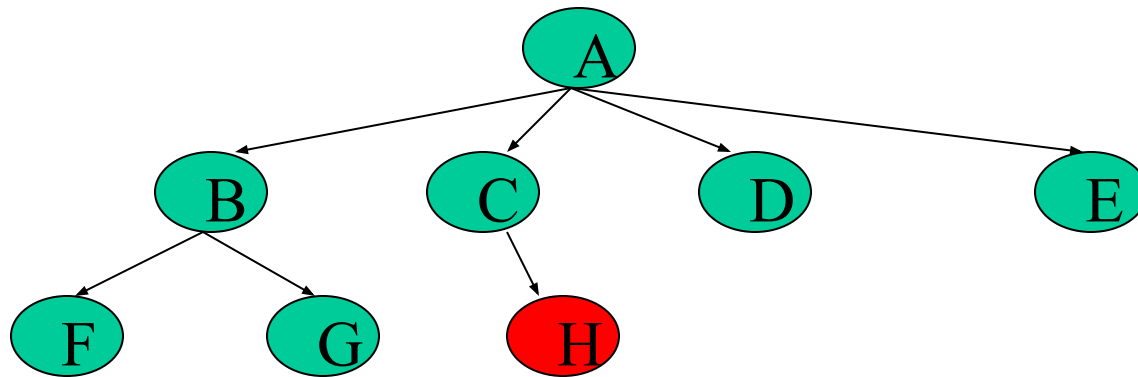
**Limit =**  
**2**



# Iterative Deepening Search

- A,B,F,
- G,
- C,H,

**Limit =**  
**2**

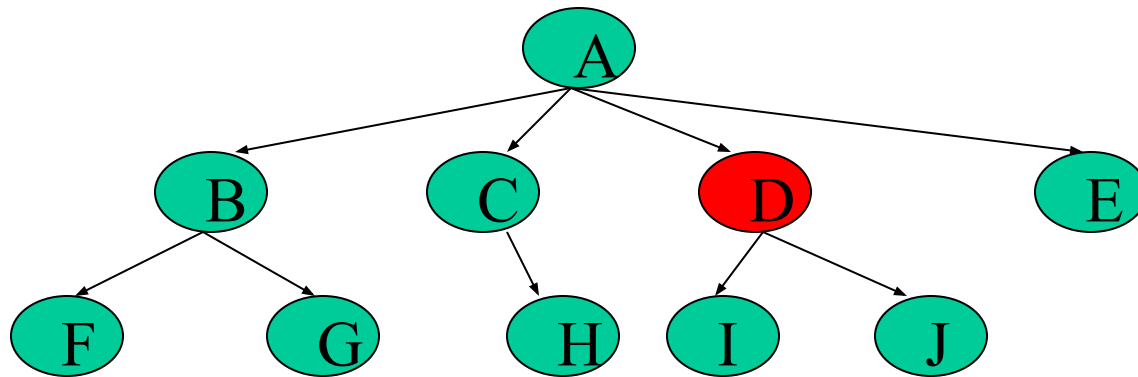




# Iterative Deepening Search

- A,B,F,
- G,
- C,H,
- D,

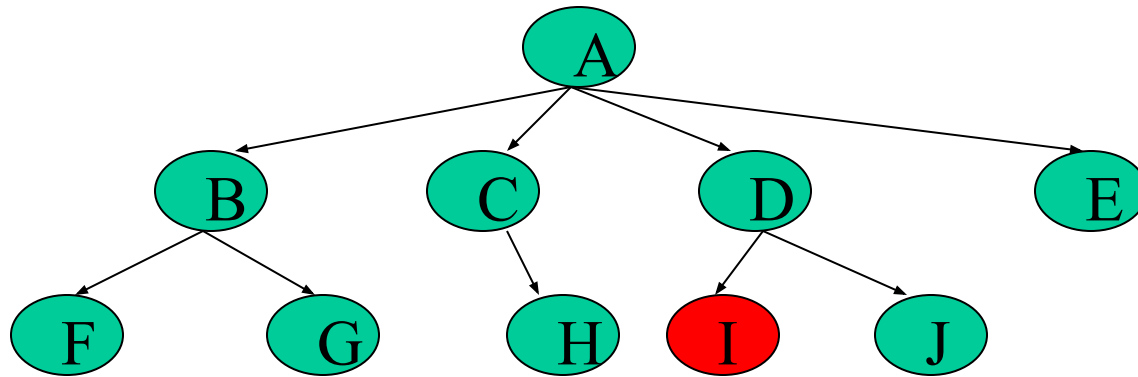
**Limit =**  
**2**



# Iterative Deepening Search

- A,B,F,
- G,
- C,H,
- D,I

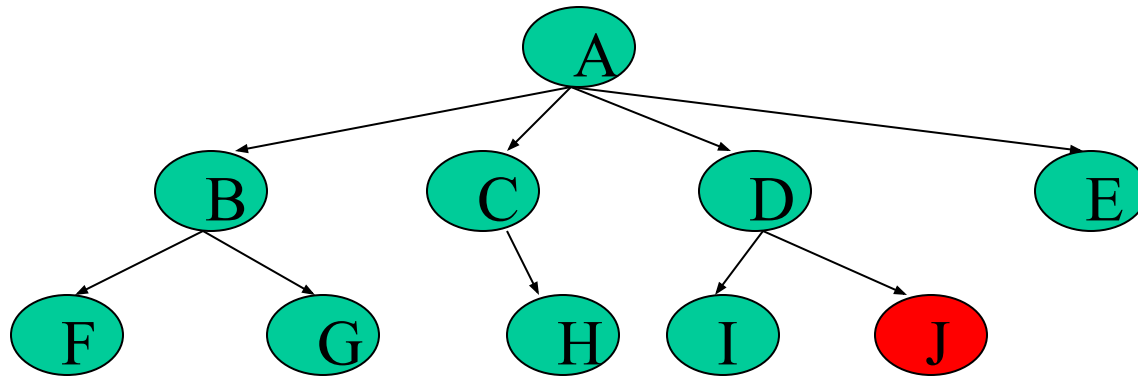
**Limit =  
2**



# Iterative Deepening Search

- A,B,F,
- G,
- C,H,
- D,I
- J,

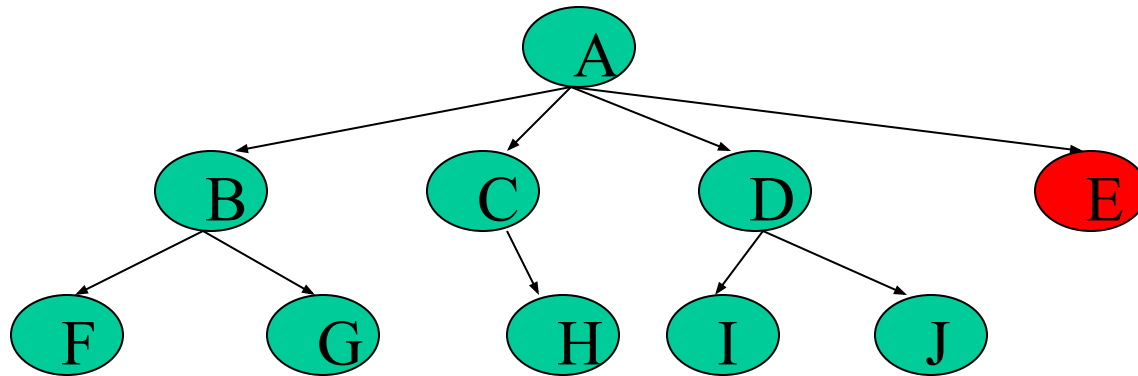
**Limit =**  
**2**



# Iterative Deepening Search

- A,B,F,
- G,
- C,H,
- D,I
- J,
- E

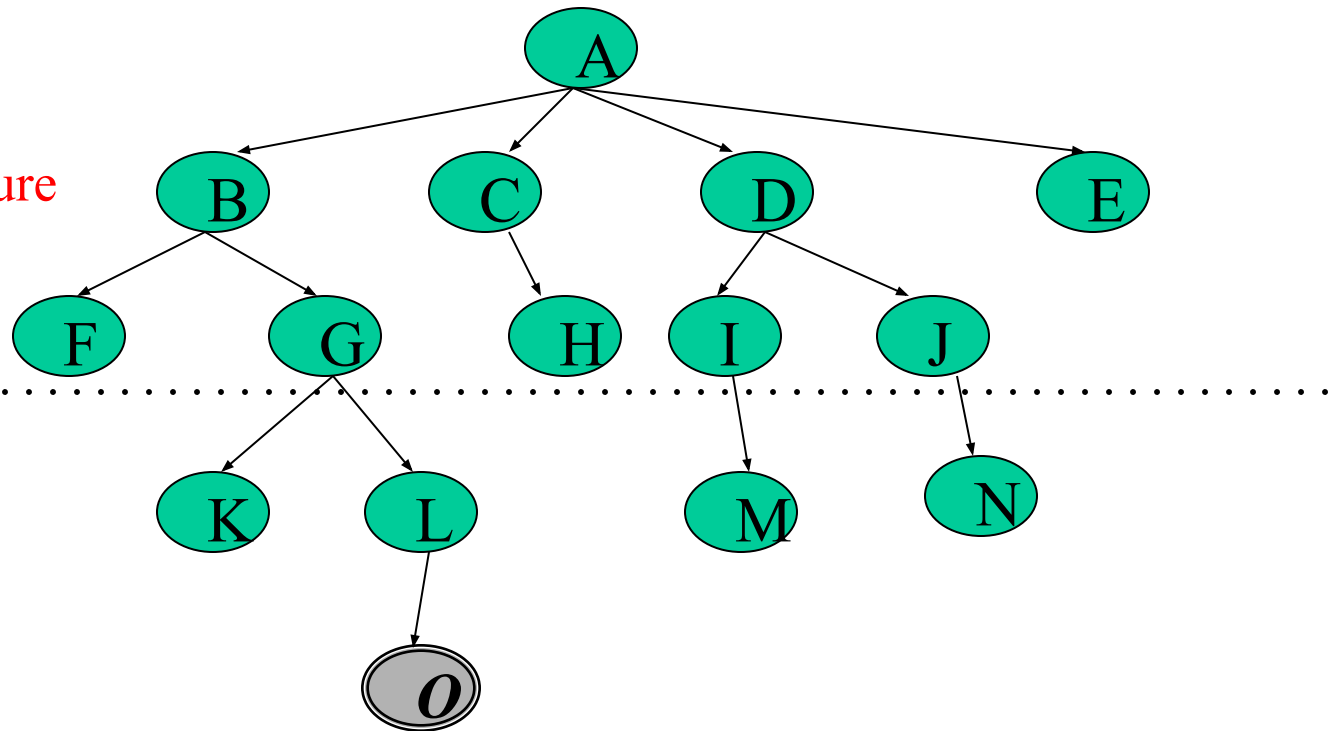
**Limit =**  
**2**



# Iterative Deepening Search

- A,B,F,
- G,
- C,H,
- D,I
- J,
- E, **Failure**

**Limit =**  
**2**

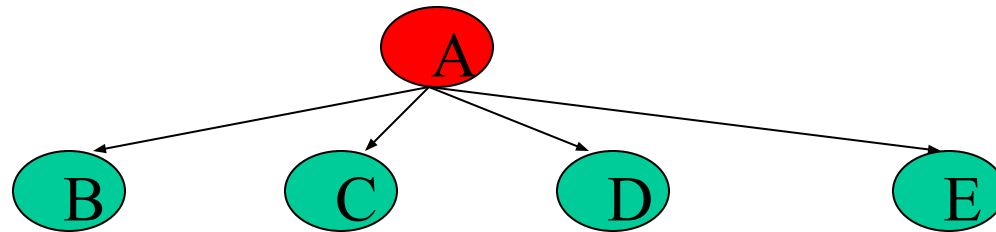


# Iterative Deepening Search (IDS)

DLS with bound = 3

# Iterative Deepening Search

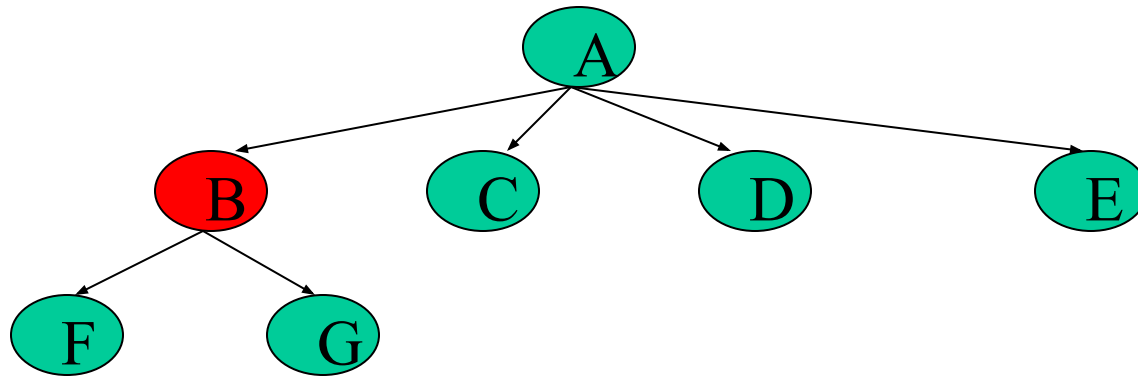
- A,



**Limit =**  
**3**

# Iterative Deepening Search

- A,B,

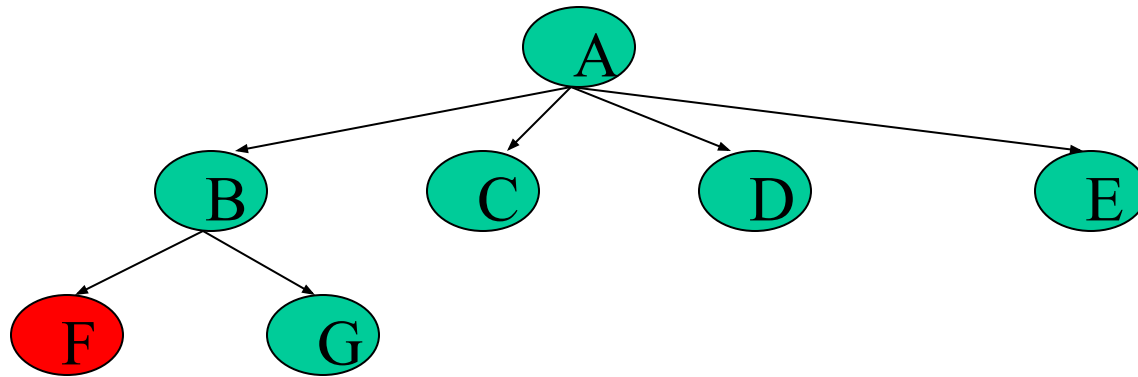


**Limit =**  
**3**



# Iterative Deepening Search

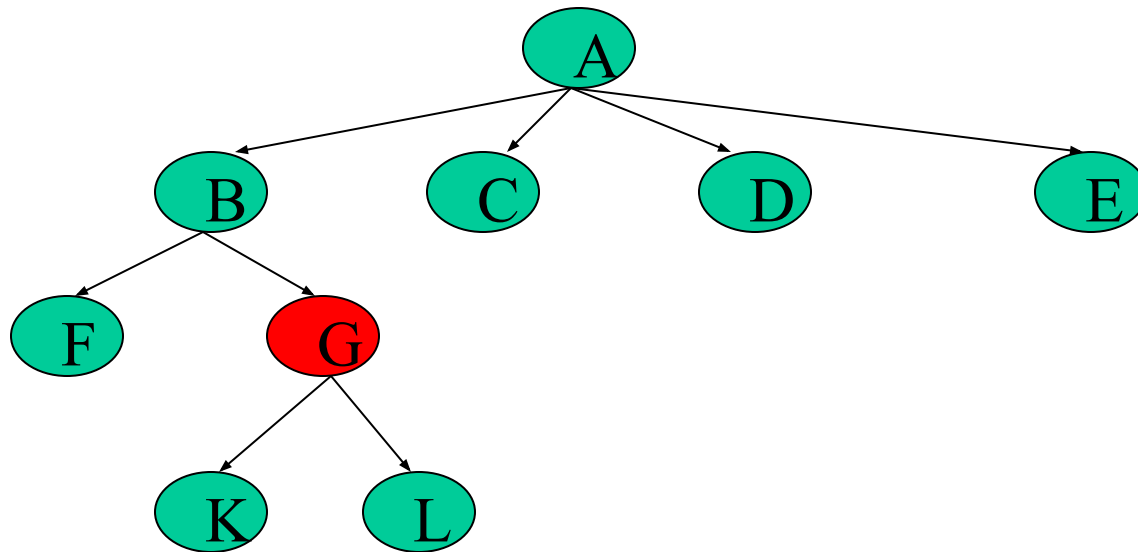
- A,B,F,



**Limit =**  
**3**

# Iterative Deepening Search

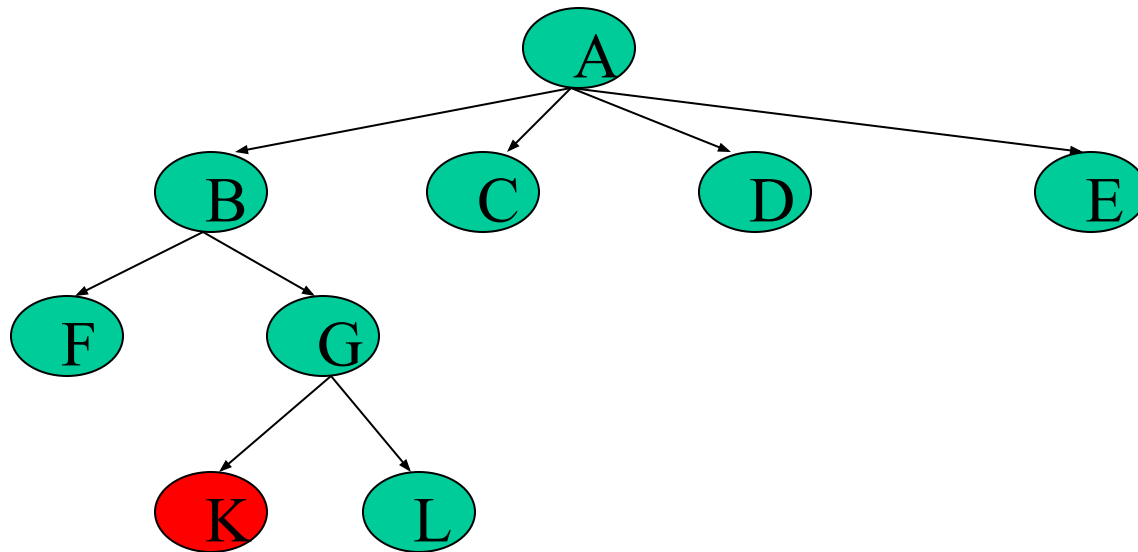
- A,B,F,
- G,



**Limit =**  
**3**

# Iterative Deepening Search

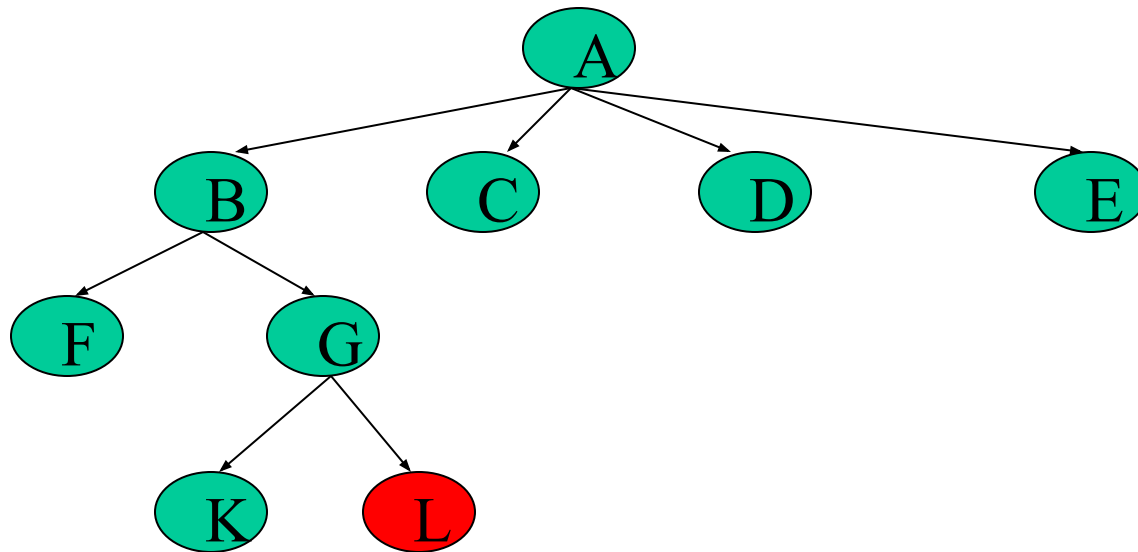
- A,B,F,
- G,K,



**Limit =  
3**

# Iterative Deepening Search

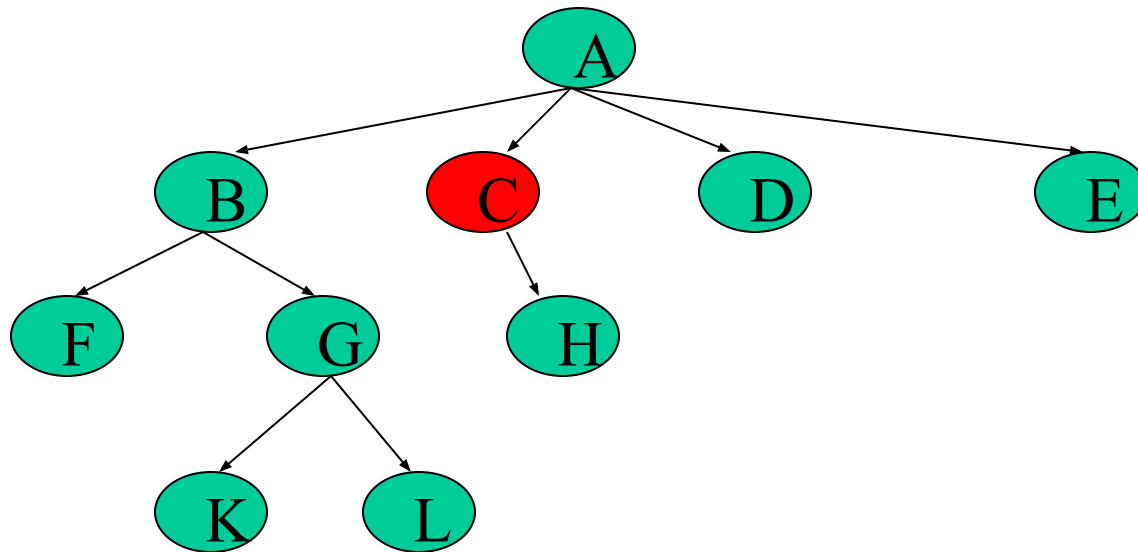
- A,B,F,
- G,K,
- L,



**Limit =**  
**3**

# Iterative Deepening Search

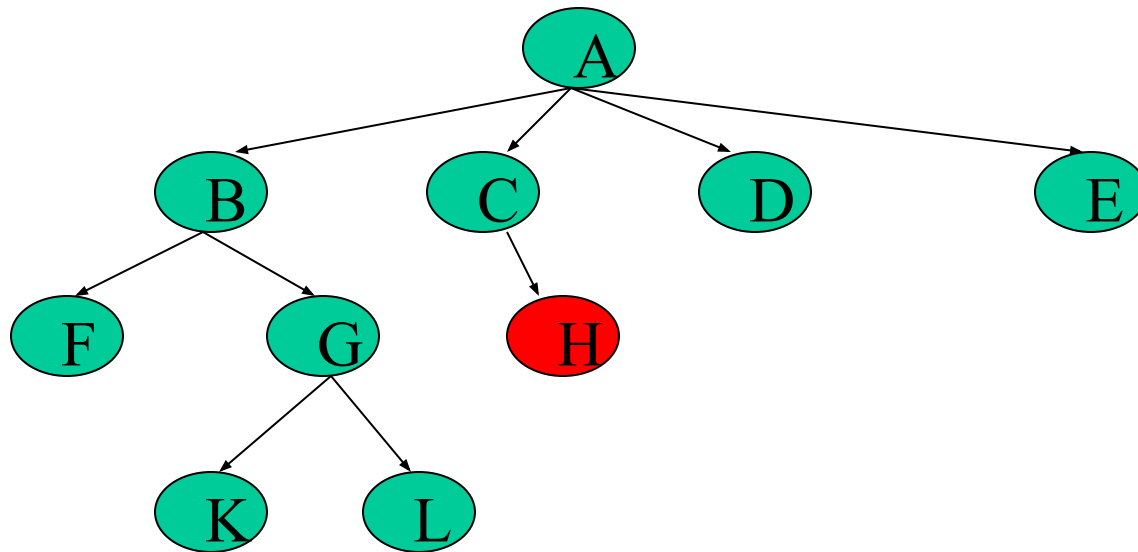
- A,B,F,
- G,K,
- L,
- C,



**Limit =  
3**

# Iterative Deepening Search

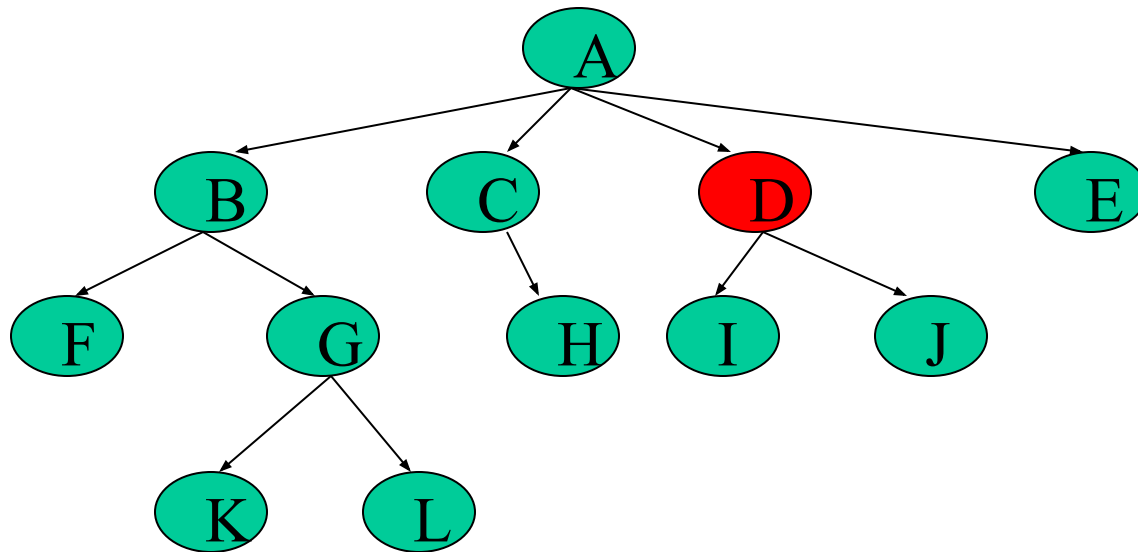
- A,B,F,
- G,K,
- L,
- C,H,



**Limit =  
3**

# Iterative Deepening Search

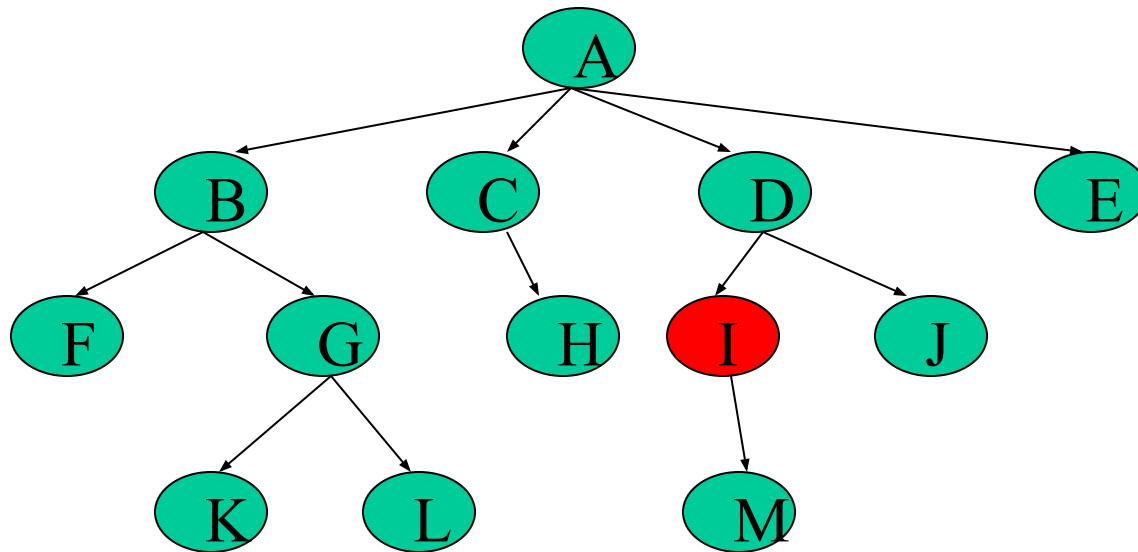
- A,B,F,
- G,K,
- L,
- C,H,
- D,



**Limit =  
3**

# Iterative Deepening Search

- A,B,F,
- G,K,
- L,
- C,H,
- D,I,

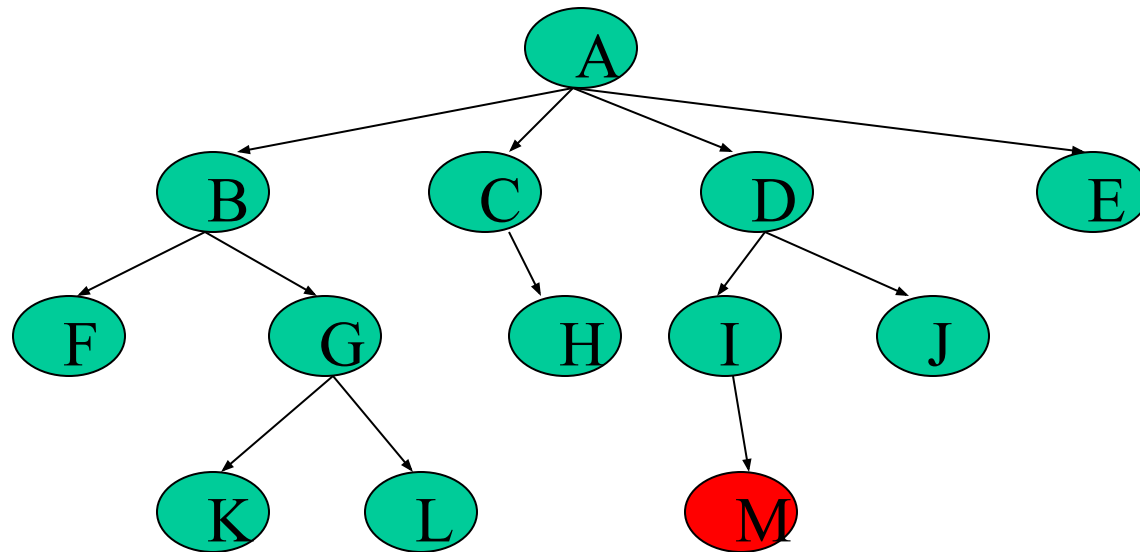


**Limit =  
3**



# Iterative Deepening Search

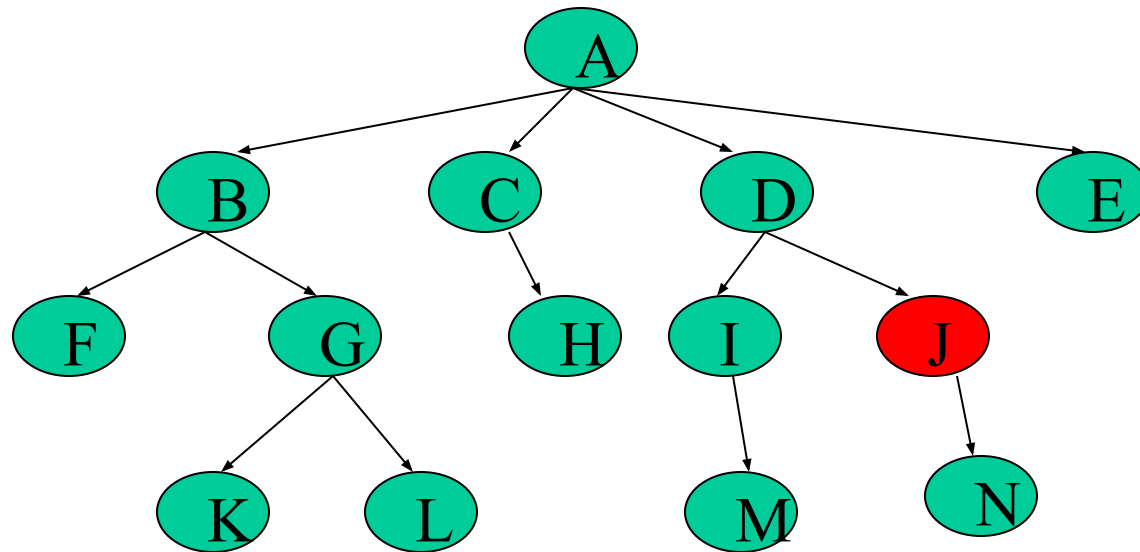
- A,B,F,
- G,K,
- L,
- C,H,
- D,I,M,



**Limit =**  
**3**

# Iterative Deepening Search

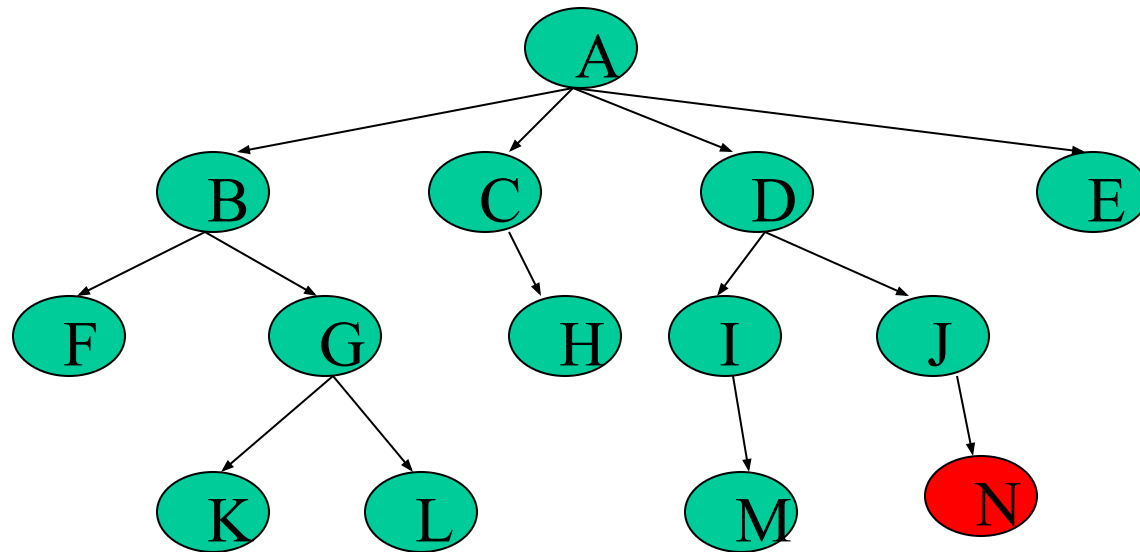
- A,B,F,
- G,K,
- L,
- C,H,
- D,I,M,
- J,



**Limit =**  
**3**

# Iterative Deepening Search

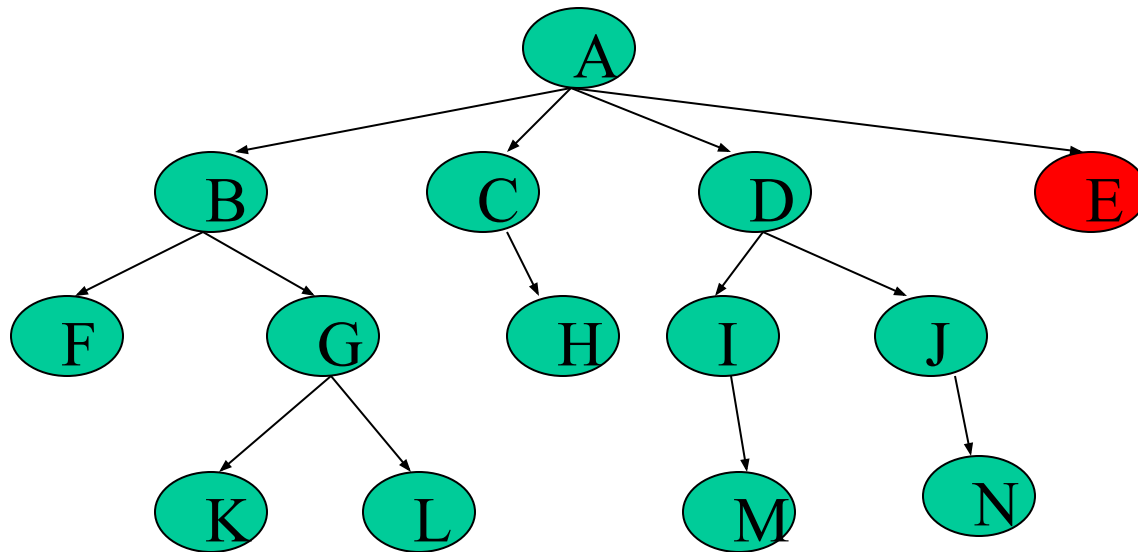
- A,B,F,
- G,K,
- L,
- C,H,
- D,I,M,
- J,N,



**Limit =**  
**3**

# Iterative Deepening Search

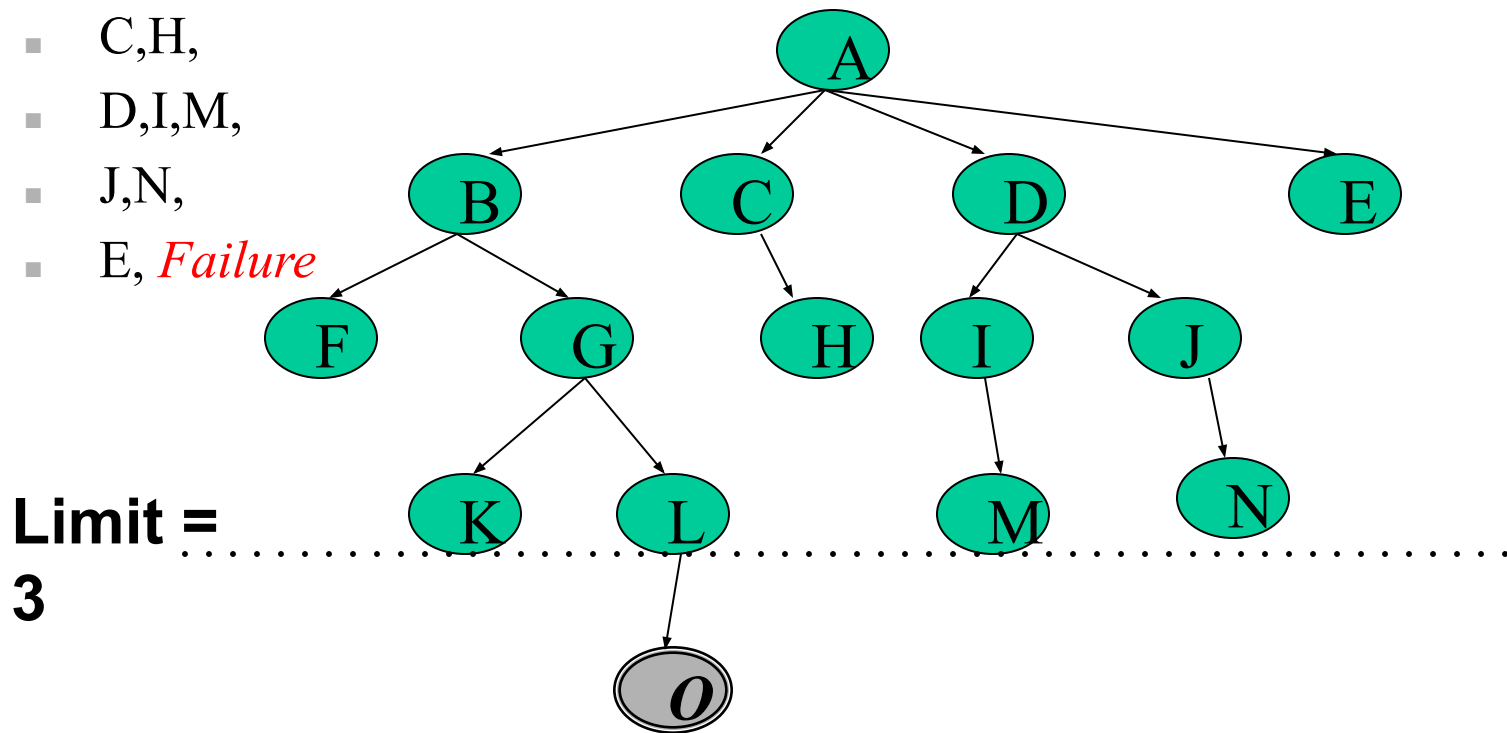
- A,B,F,
- G,K,
- L,
- C,H,
- D,I,M,
- J,N,
- E,



**Limit =**  
**3**

# Iterative Deepening Search

- A,B,F,
- G,K,
- L,
- C,H,
- D,I,M,
- J,N,
- E, *Failure*

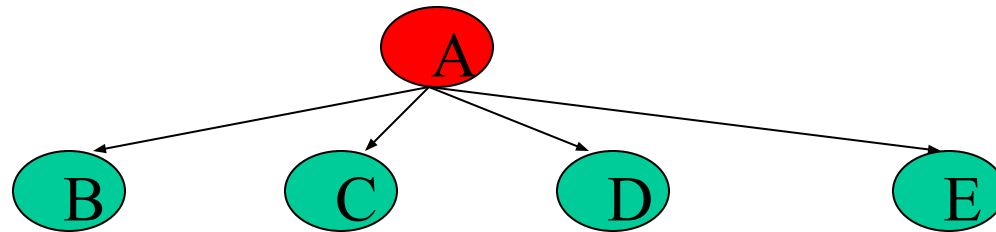


# Iterative Deepening Search (IDS)

DLS with bound = 4

# Iterative Deepening Search

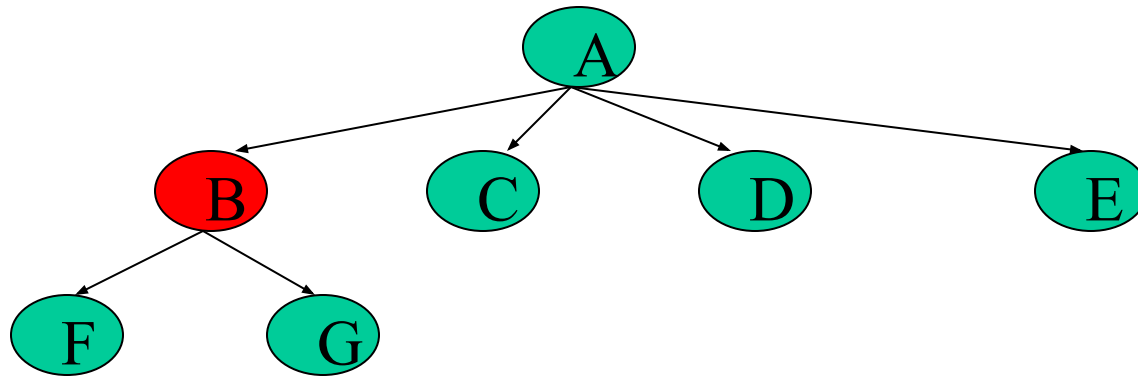
- A,



**Limit =**  
**4**

# Iterative Deepening Search

- A,B,

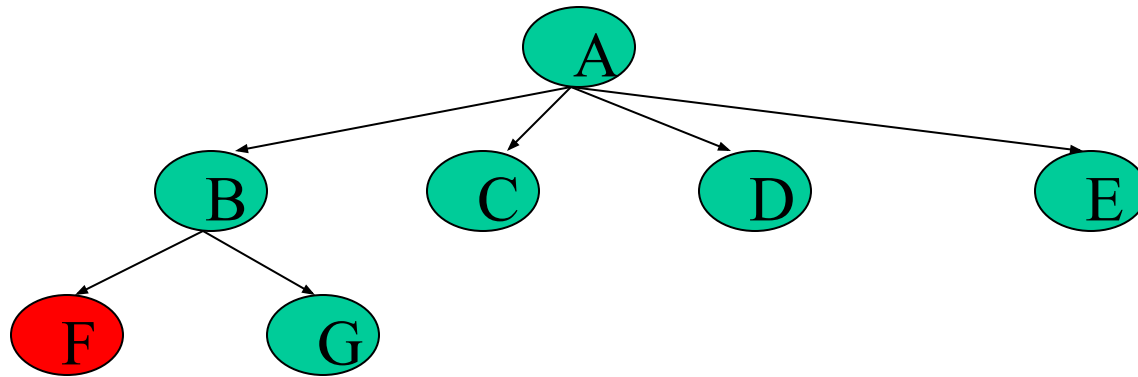


**Limit =**  
**4**



# Iterative Deepening Search

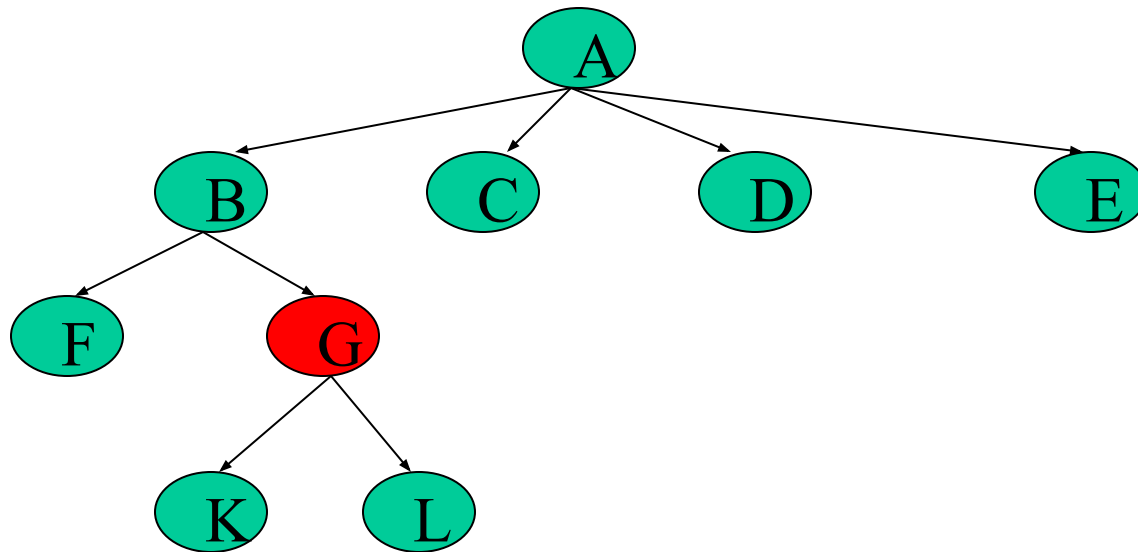
- A,B,F,



**Limit =**  
**4**

# Iterative Deepening Search

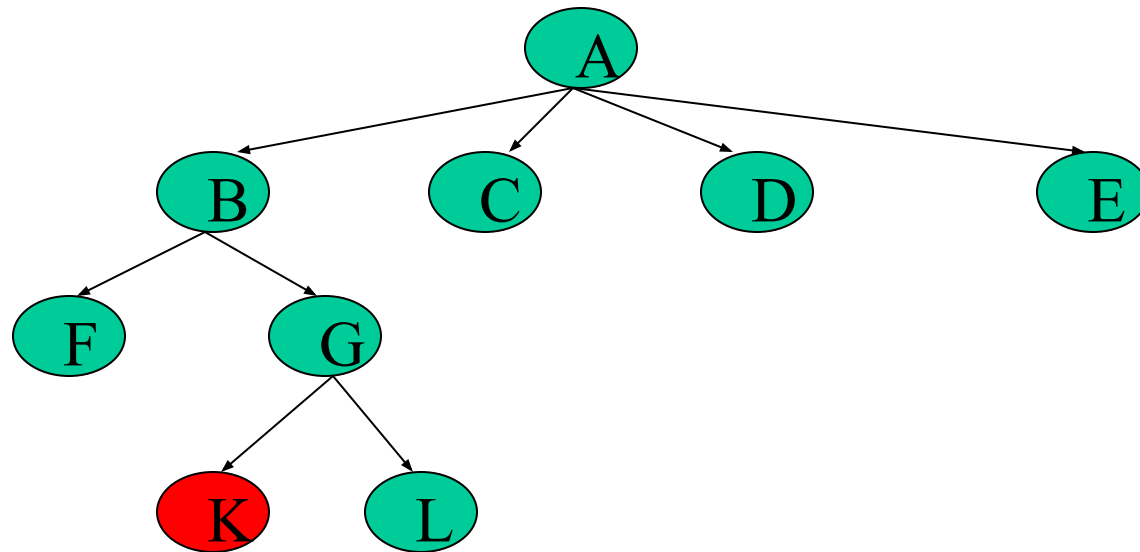
- A,B,F,
- G,



**Limit =**  
**4**

# Iterative Deepening Search

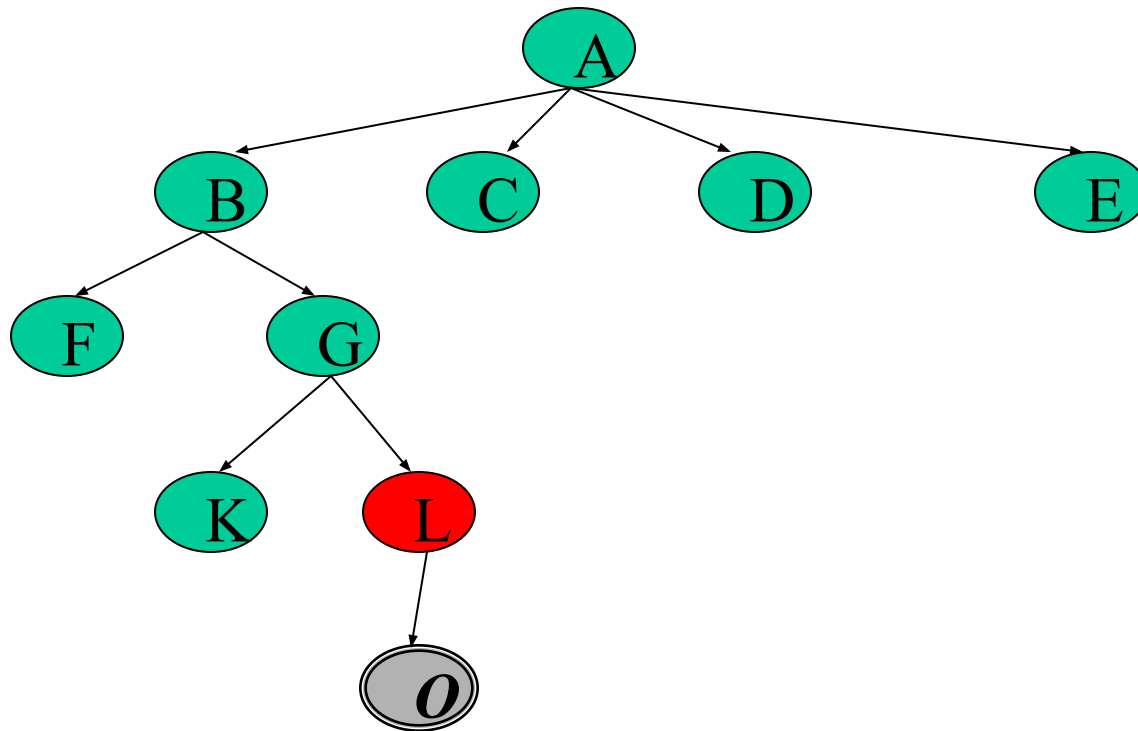
- A,B,F,
- G,K,



**Limit =**  
**4**

# Iterative Deepening Search

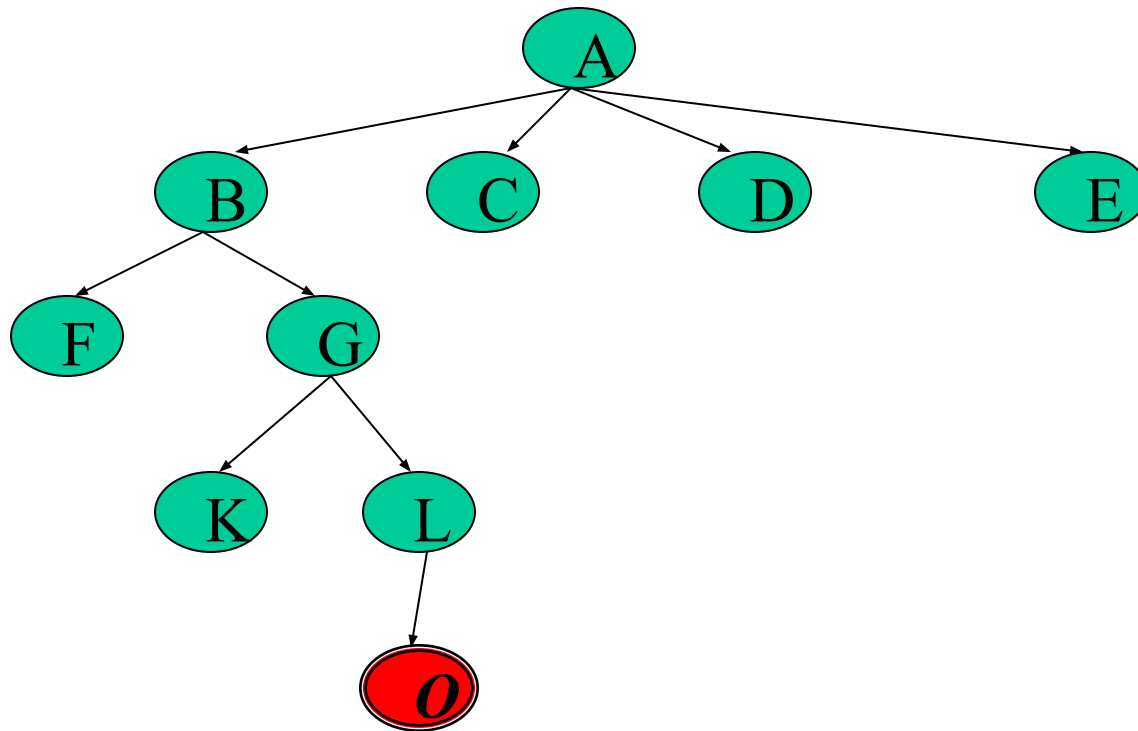
- A,B,F,
- G,K,
- L,



**Limit =**  
**4**

# Iterative Deepening Search

- A,B,F,
- G,K,
- L, O: *Goal State*

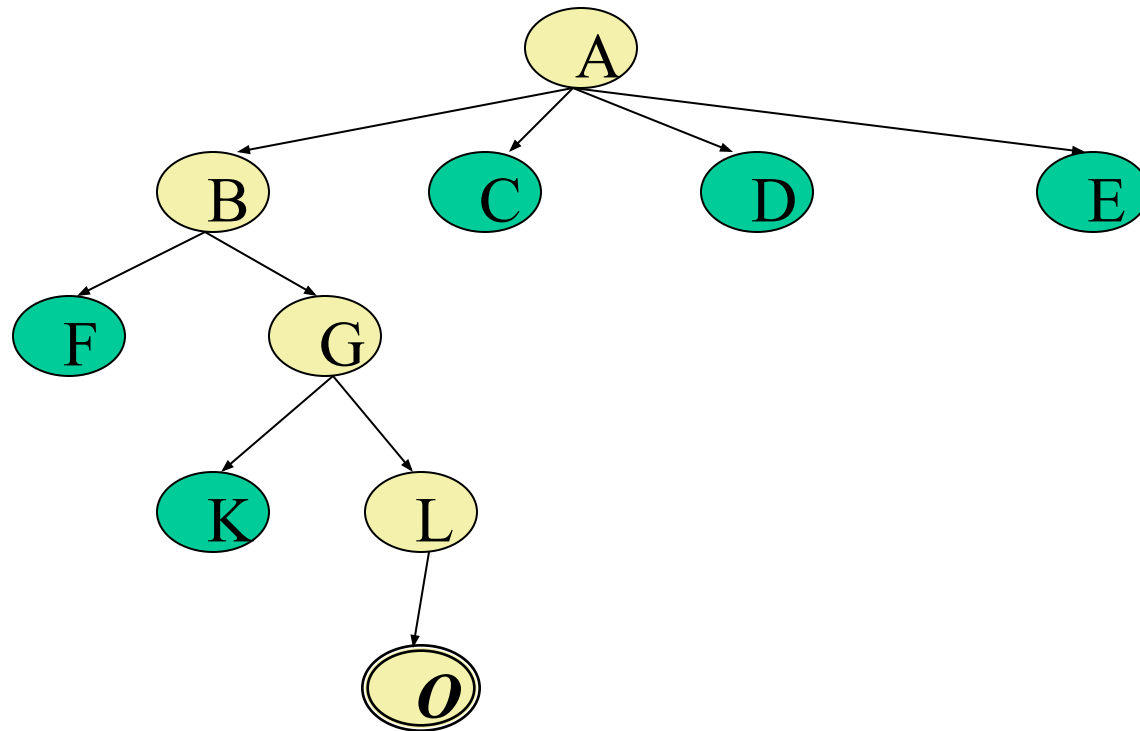


Limit =  
4

# Iterative Deepening Search

The returned solution is the sequence of operators in the path:

*A, B, G, L, O*



# Uniform Cost Search (UCS)

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

- **Implementation:**

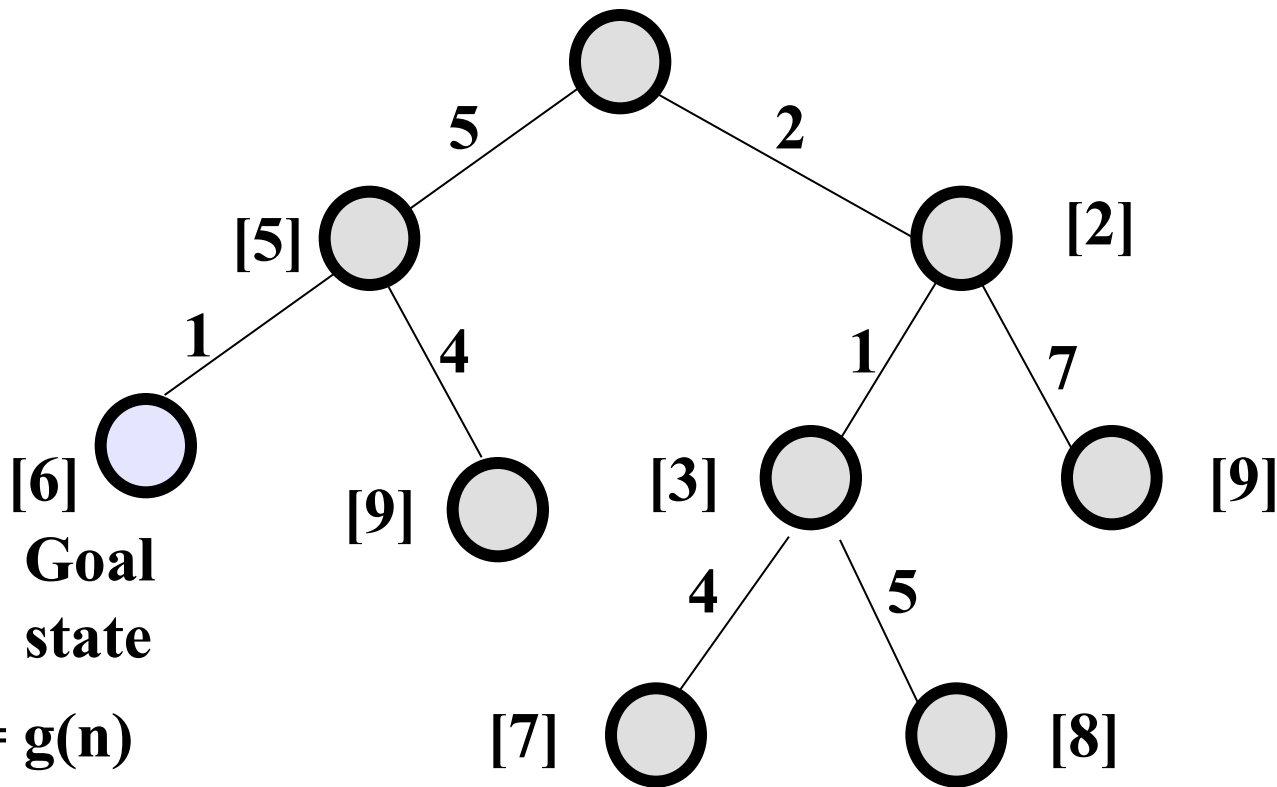
*Enqueue nodes in order of cost  $g(n)$ .*

QUEUING-FN:- insert in order of increasing path cost.

*Enqueue new node at the appropriate position in the queue so that we dequeue the cheapest node.*

- Complete? Yes.
- Optimal? Yes, if path cost is nondecreasing function of depth
- Time Complexity:  $O(b^d)$
- Space Complexity:  $O(b^d)$ , note that every node in the fringe keep in the queue.

# Uniform Cost Search

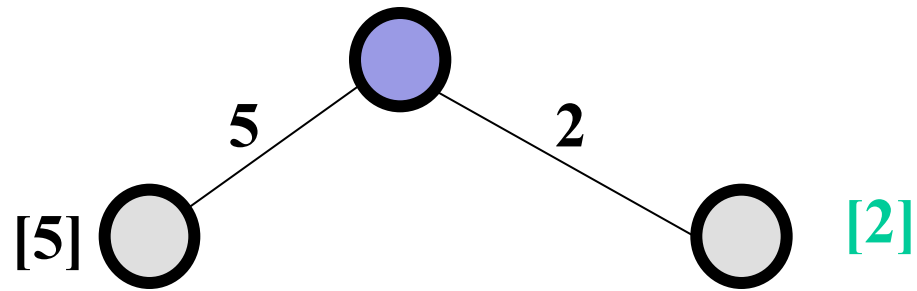


$[x] = g(n)$

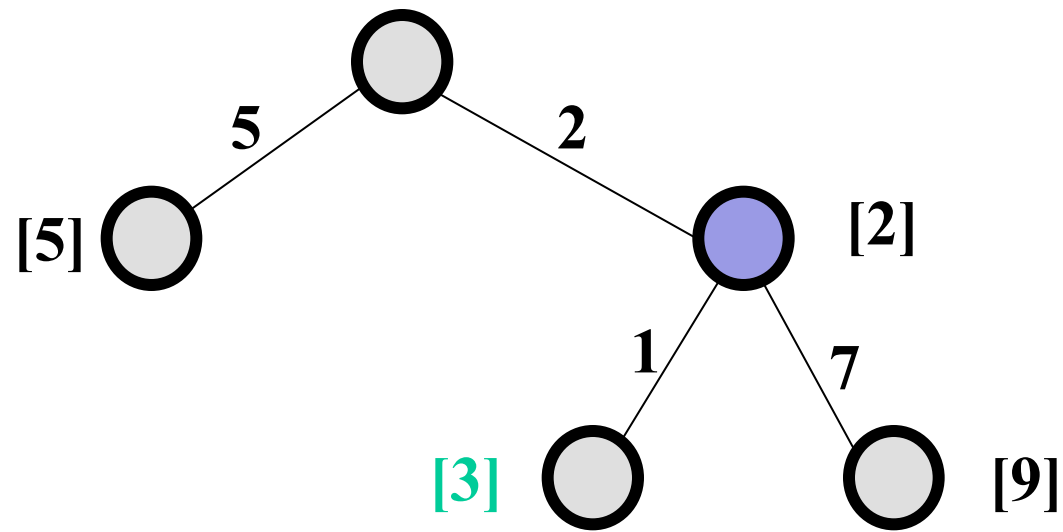
path cost of node  
 $n$



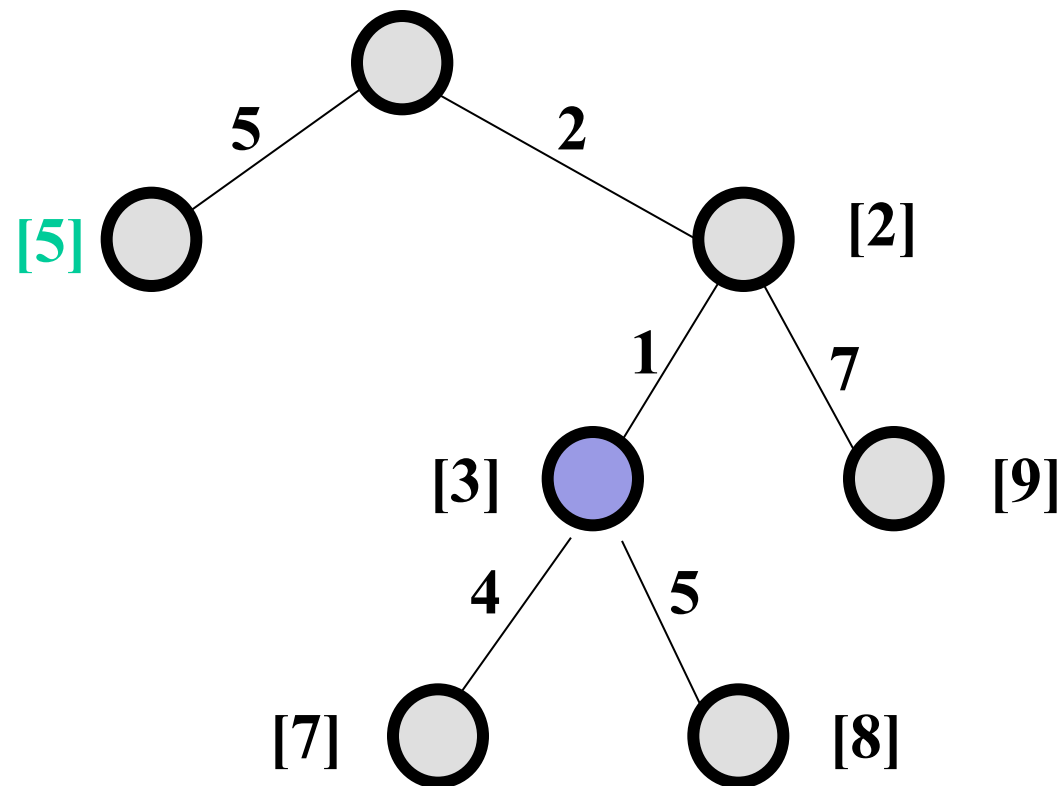
# Uniform Cost Search



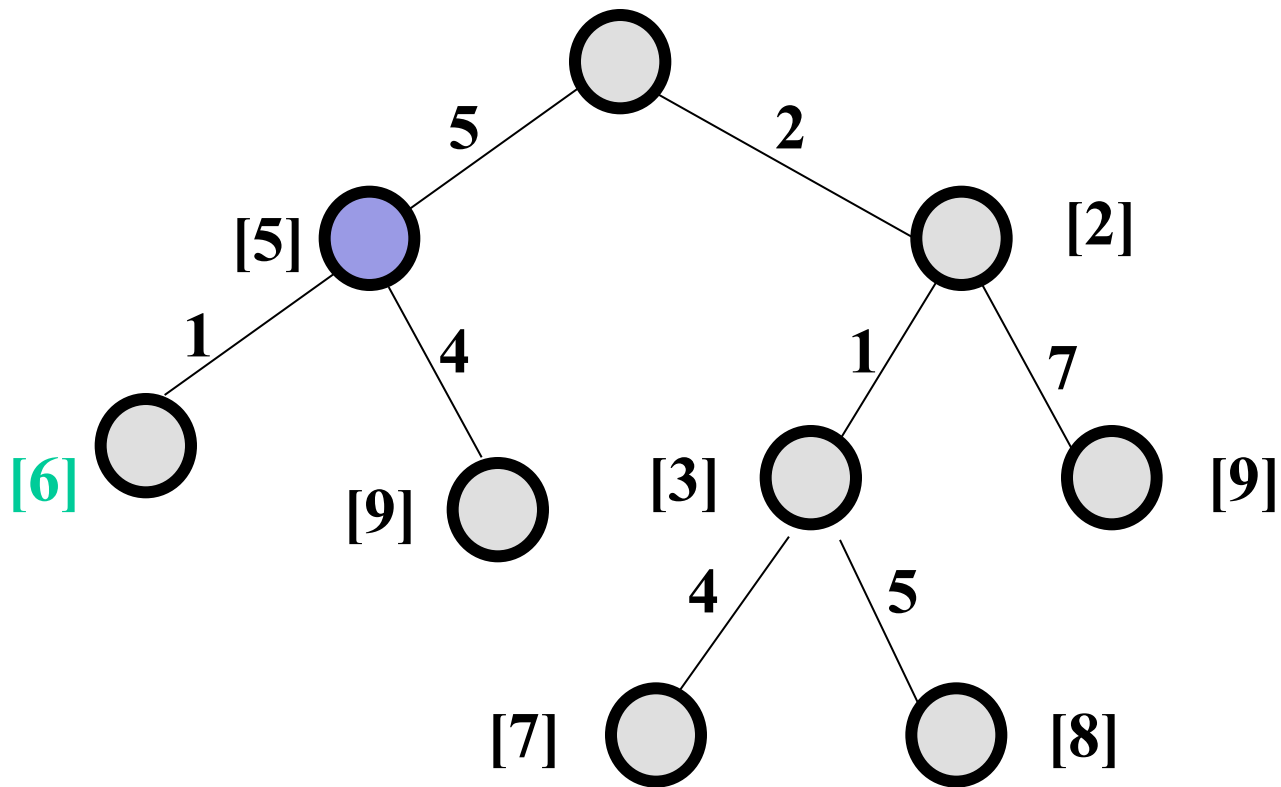
# Uniform Cost Search



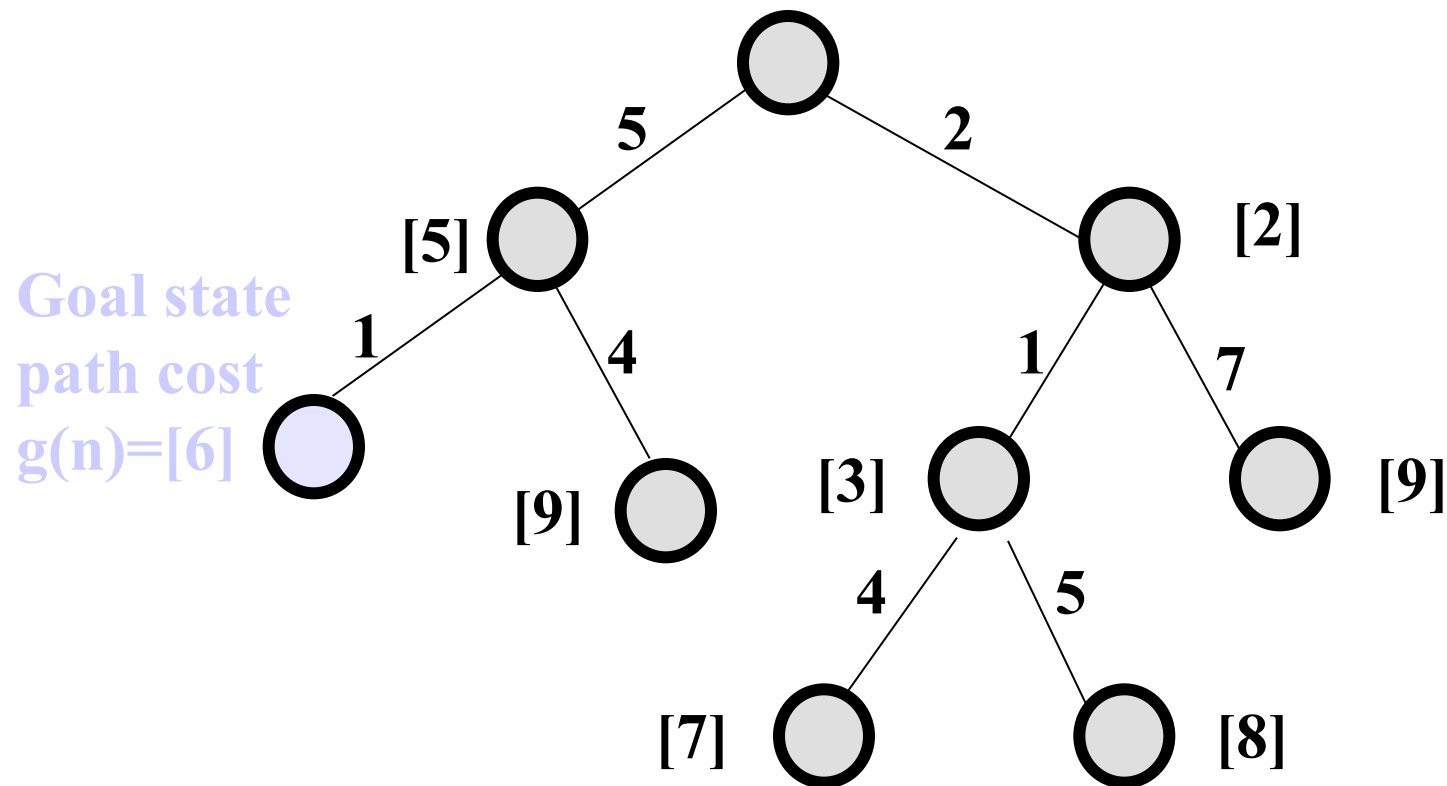
# Uniform Cost Search



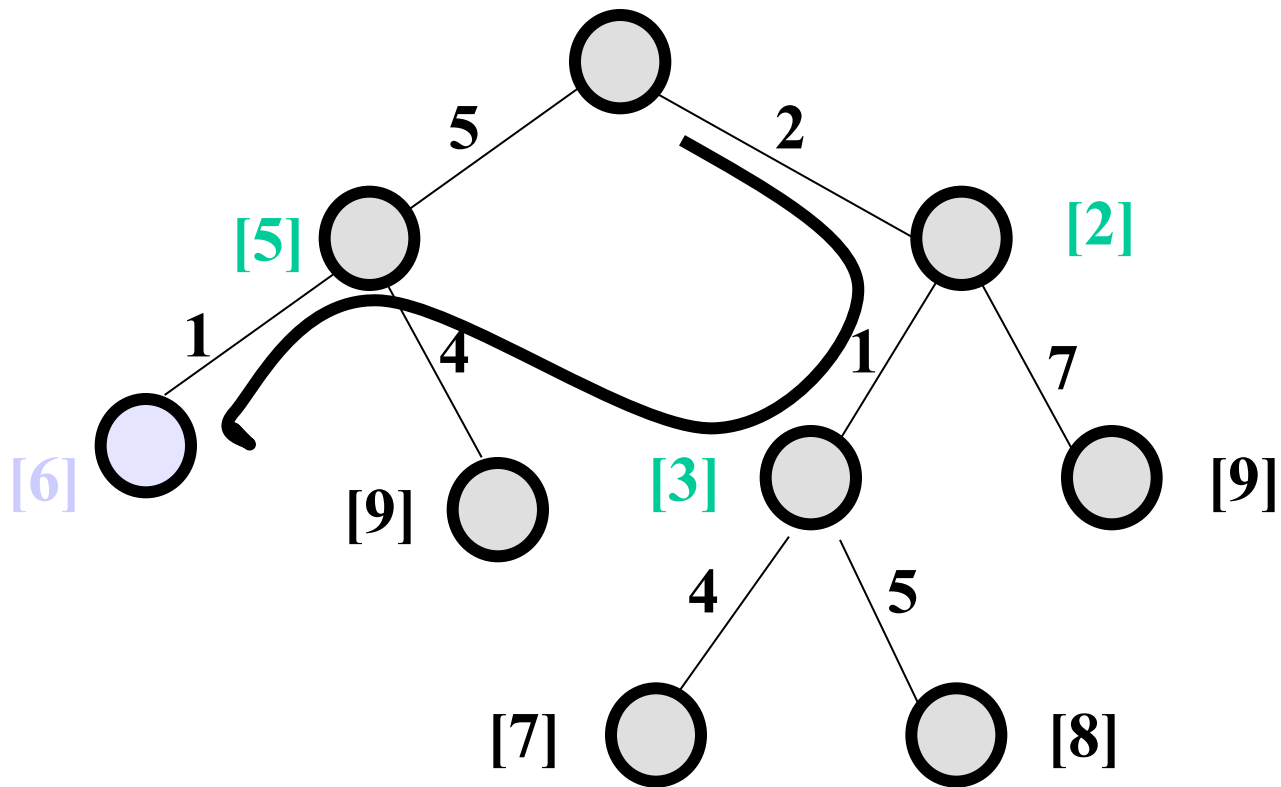
# Uniform Cost Search



# Uniform Cost Search

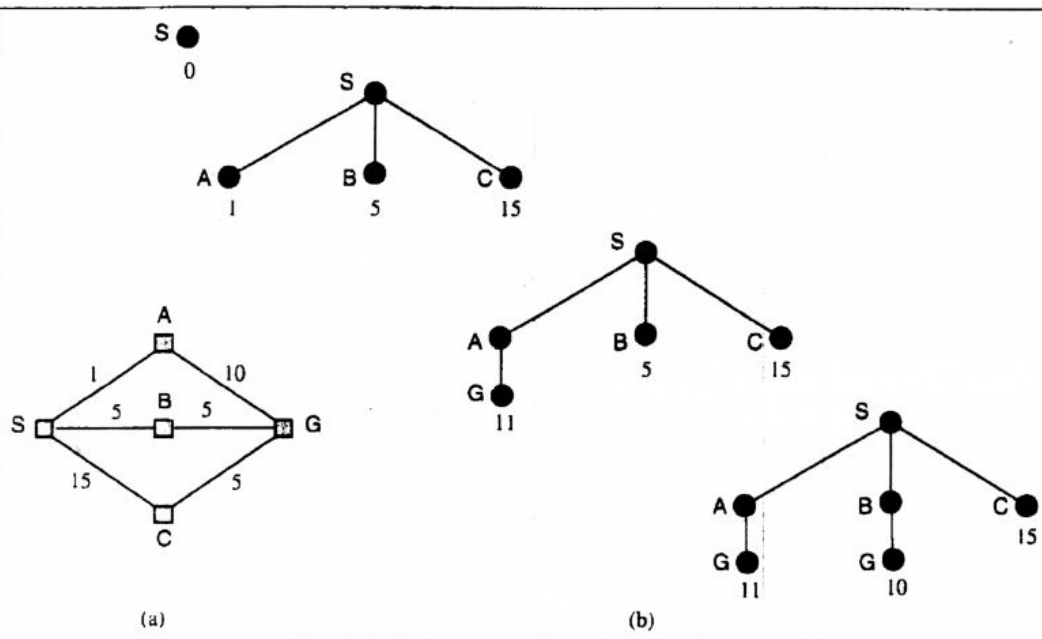


# Uniform Cost Search



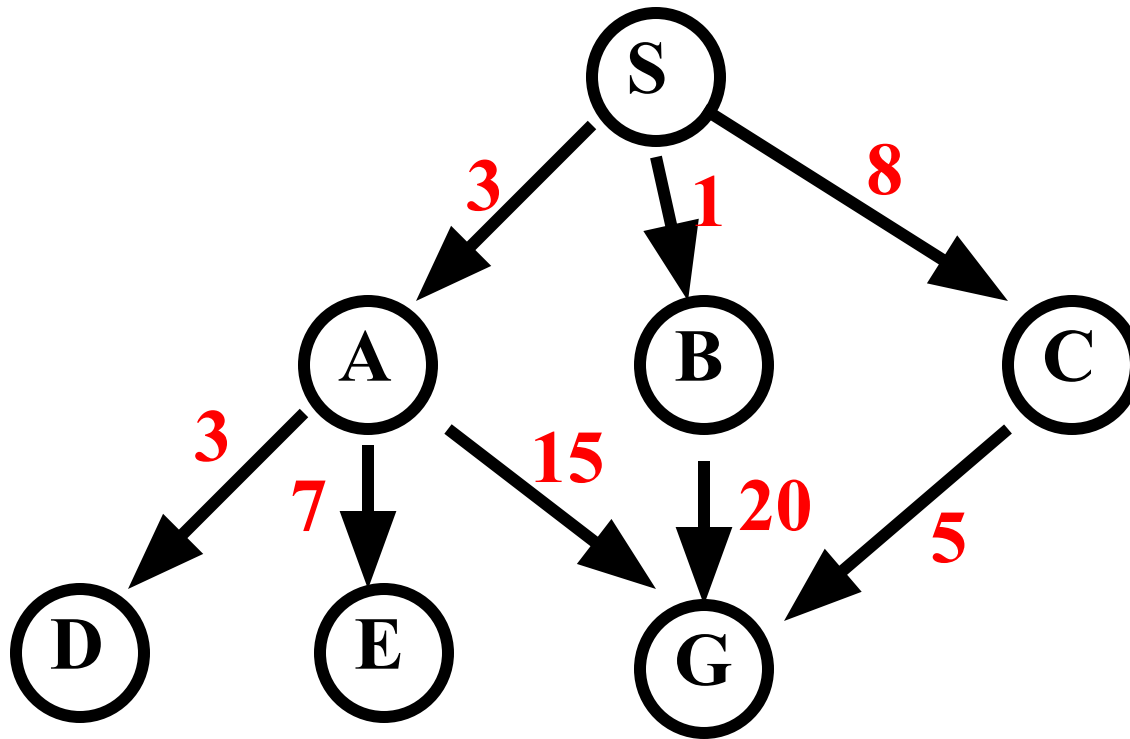
# Uniform Cost Search

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



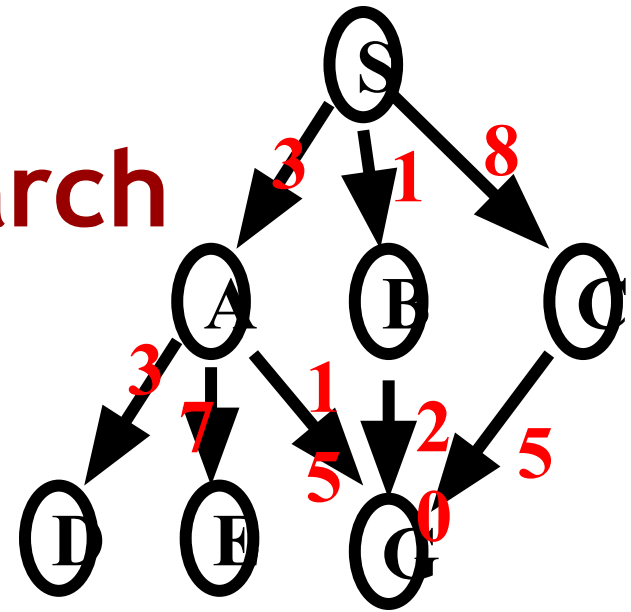
**Figure 3.13** A route-finding problem. (a) The state space, showing the cost for each operator. (b) Progression of the search. Each node is labelled with  $g(n)$ . At the next step, the goal node with  $g = 10$  will be selected.

## Example for Illustrating Search Strategies





# Depth-First Search

[illegible]
$$\{S^0\}$$
$$S^0 \quad \{ A^3 \ B^1 \ C^8 \}$$
$$\mathbf{A}^3 \quad \{ \mathbf{D}^6 \mathbf{E}^{10} \mathbf{G}^{18} \mathbf{B}^1 \mathbf{C}^8 \}$$
$$\mathbf{D}^6 \quad \{ \mathbf{E}^{10} \mathbf{G}^{18} \mathbf{B}^1 \mathbf{C}^8 \}$$
$$\mathbf{E}^{10} \{ \mathbf{G}^{18} \mathbf{B}^1 \mathbf{C}^8 \}$$
$$\mathbf{G}^{18} \{ \mathbf{B}^1 \mathbf{C}^8 \}$$

Solution path found is S A G, cost 18

Number of nodes expanded (including goal node) = 5

# Breadth-First Search

**Expanded node    Nodes list**

$\{ S^0 \}$

$S^0 \quad \{ A^3 \ B^1 \ C^8 \}$

$A^3 \quad \{ B^1 \ C^8 \ D^6 \ E^{10} \ G^{18} \}$

$B^1 \quad \{ C^8 \ D^6 \ E^{10} \ G^{18} \ G^{21} \}$

$C^8 \quad \{ D^6 \ E^{10} \ G^{18} \ G^{21} \ G^{13} \}$

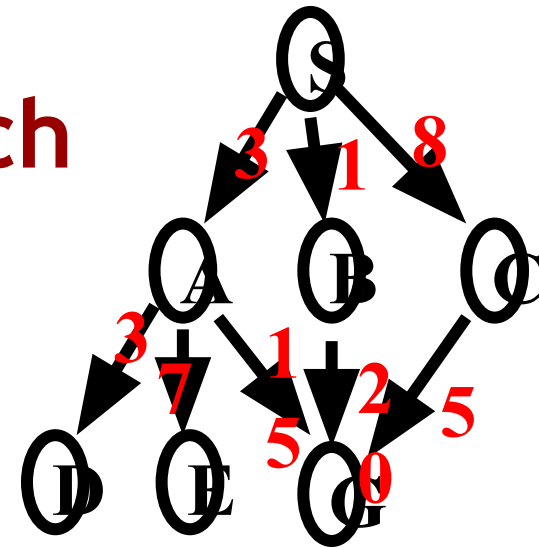
$D^6 \quad \{ E^{10} \ G^{18} \ G^{21} \ G^{13} \}$

$E^{10} \quad \{ G^{18} \ G^{21} \ G^{13} \}$

$G^{18} \quad \{ G^{21} \ G^{13} \}$

Solution path found is S A G , cost 18

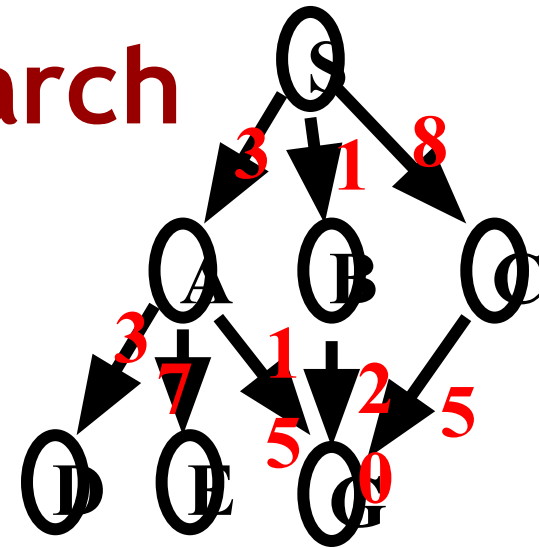
Number of nodes expanded (including goal node) = 7



# Uniform Cost Search

**Expanded node**   **Nodes list**

	$\{ S^0 \}$
$S^0$	$\{ B^1 A^3 C^8 \}$
$B^1$	$\{ A^3 C^8 G^{21} \}$
$A^3$	$\{ D^6 C^8 E^{10} G^{18} \}$
$D^6$	$\{ C^8 E^{10} G^{18} \}$
$C^8$	$\{ E^{10} G^{13} \}$
$E^{10}$	$\{ G^{13} \}$
$G^{13}$	$\{ \}$



Solution path found is S C G, cost 13

Number of nodes expanded (including goal node) = 7

# 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?

# What Criteria are used to Compare different search techniques ?

As we are going to consider different techniques to search the problem space, we need to consider what criteria we will use to compare them.

- **Completeness:** Is the technique guaranteed to find an answer (if there is one).
- **Optimality/Admissibility :** does it always find a least-cost solution?
  - an admissible algorithm will find a solution with minimum cost
- **Time Complexity:** How long does it take to find a solution.
- **Space Complexity:** How much memory does it take to find a solution.

# Time and Space Complexity

Time and space complexity are measured in terms of:

- The average number of new nodes we create when expanding a new node is the (effective) branching factor **b**.
- The (maximum) branching factor **b** is defined as the maximum nodes created when a new node is expanded.
- The length of a path to a goal is the depth **d**.
- The maximum length of any path in the state space **m**.

# 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)

# 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)



# 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.

# Properties of 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)

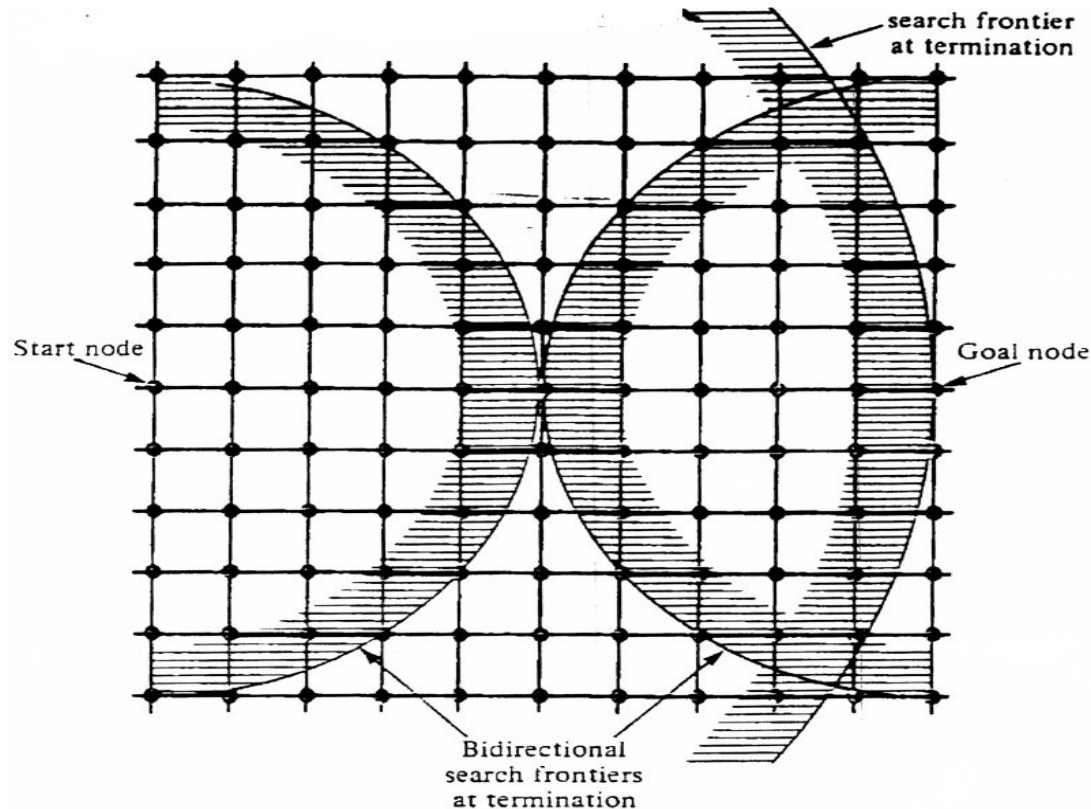
**Time?** # of nodes with *path cost*  $\leq$  cost of optimal solution.

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

**Optimal?** Yes, for any step cost.

# Bi-Directional Search

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



*Fig. 2.10 Bidirectional and unidirectional breadth-first searches.*

# Summary of search 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