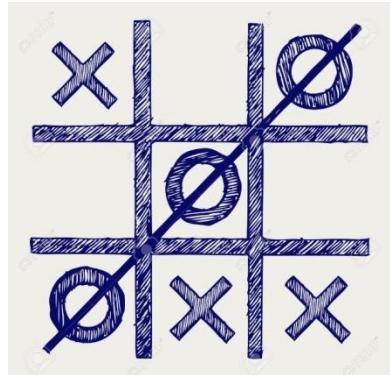


CSC 422

Class #3

Problem Solving as Search

Why searching in AI?

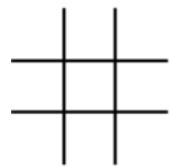


May be you all are familiar with the board game Tic-Tac-Toe.

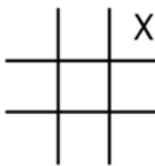
As an AI student, you want design an intelligent agent, which can play this game with you as an opponent.

Why searching in AI?

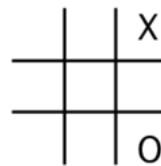
Intelligent agent should know the current state of the Tic-Tac-Toe board.



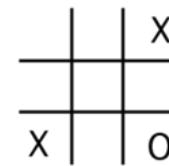
Before game begins



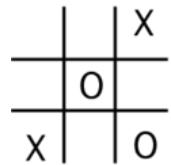
X's first move



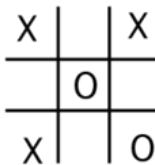
O's first move



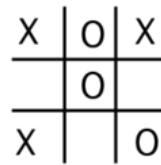
X's second move



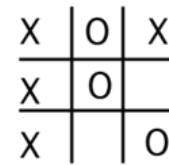
O's second move



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.

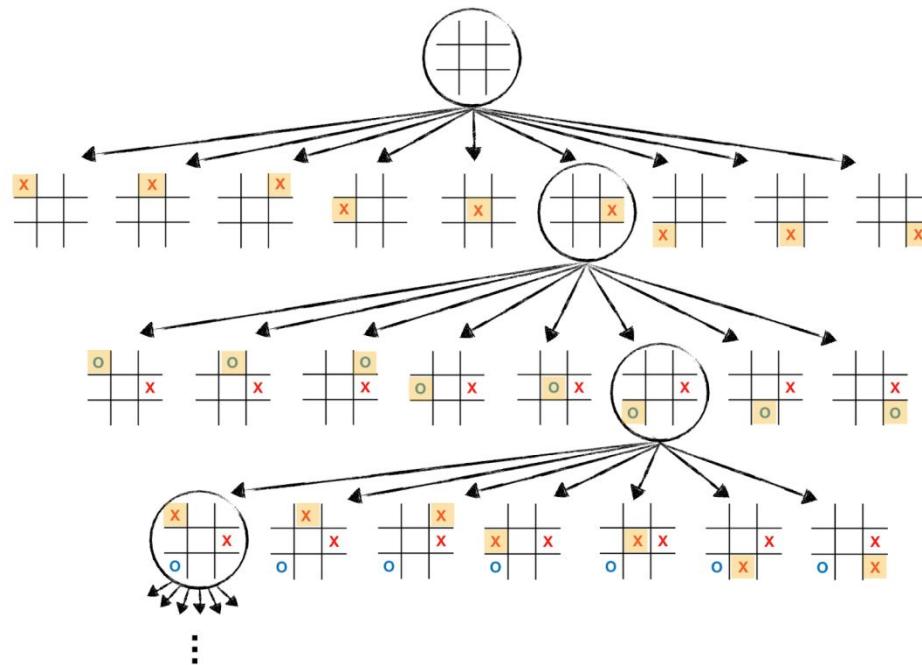
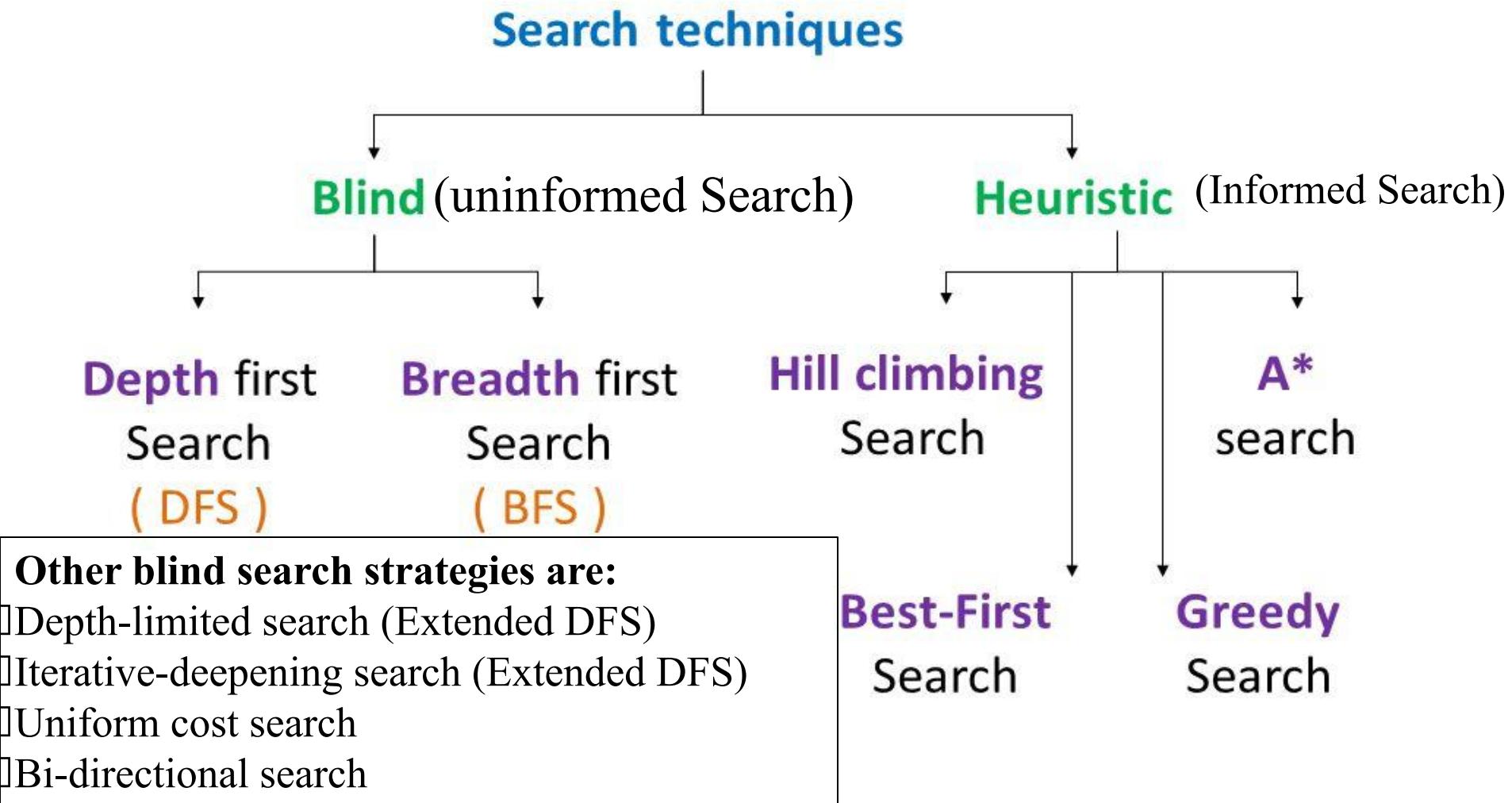


Fig: Game search tree for Tic-Tac-Toe board (partial view)

SEARCH TECHNIQUES



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:

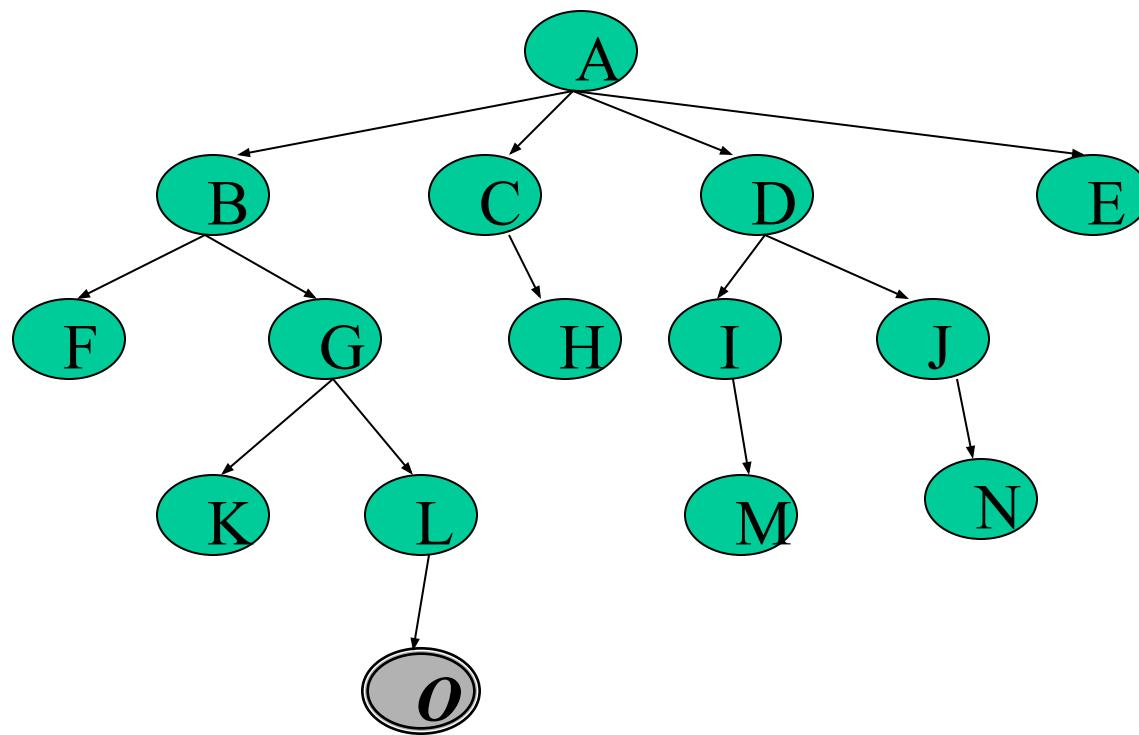
optimization in which the search is to find an optimal value of an objective function: hill climbing, simulated annealing, genetic algorithms, Ant Colony Optimization

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

Uninformed Search

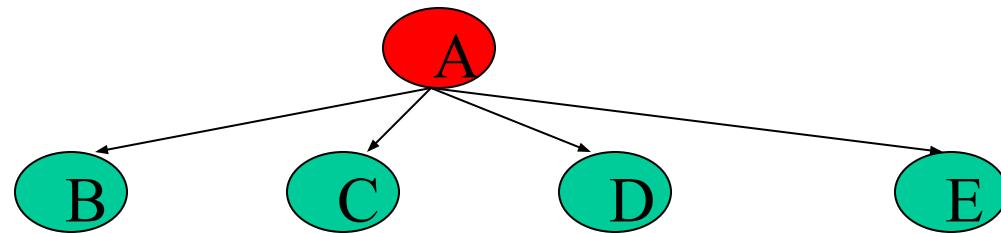
Breadth First Search

- Application1:
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):



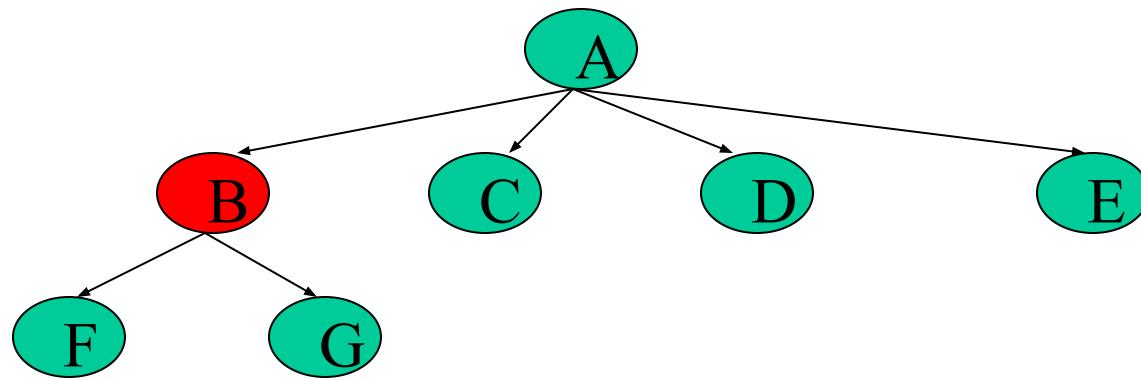
Breadth First Search

- A,



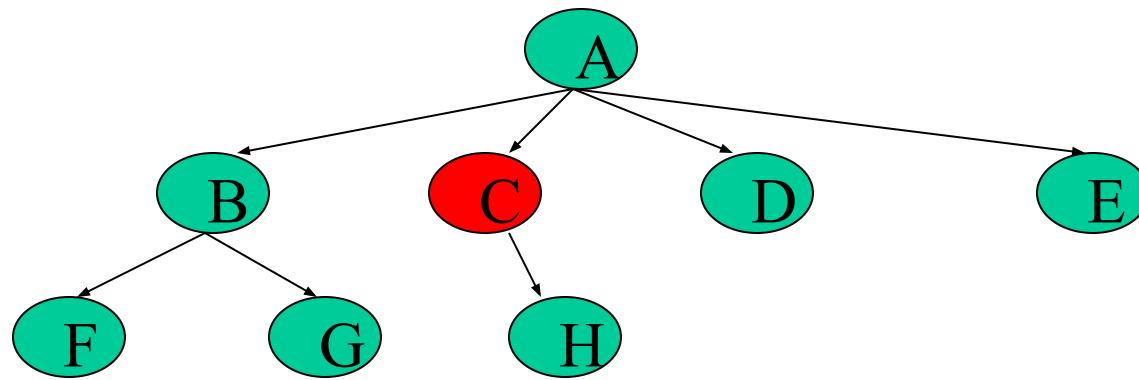
Breadth First Search

- A,
- B,



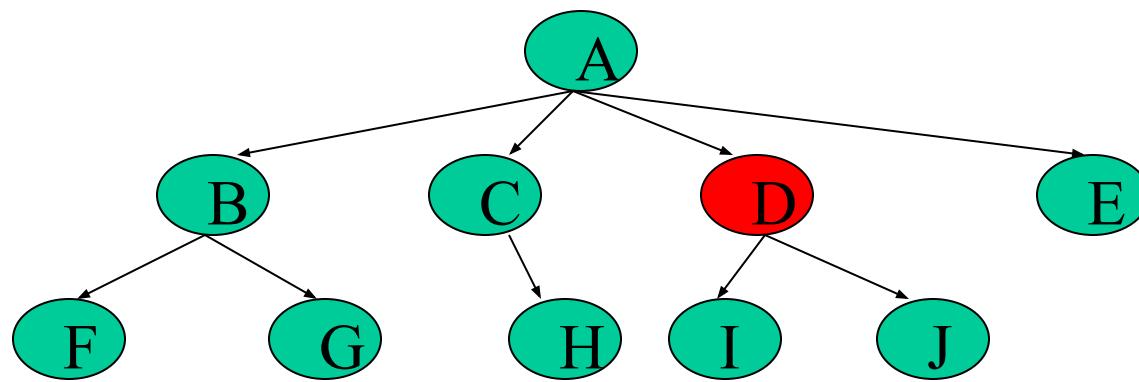
Breadth First Search

- A,
- B,C



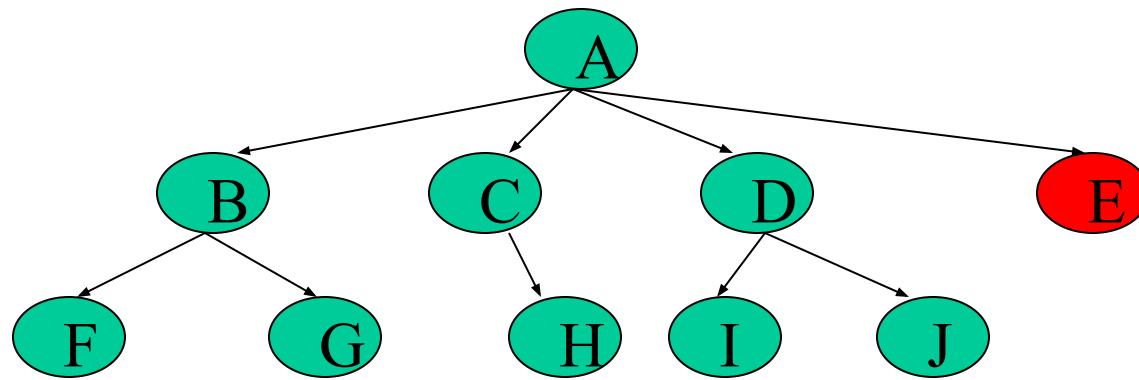
Breadth First Search

- A,
- B,C,D



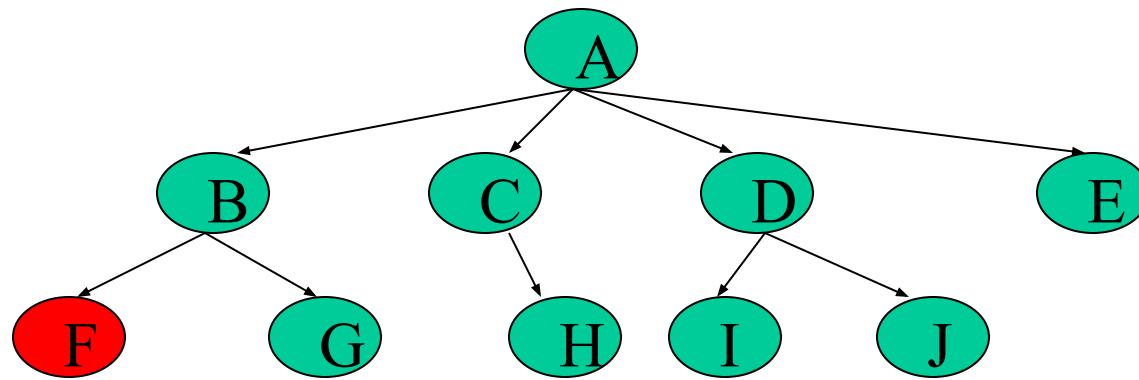
Breadth First Search

- A,
- B,C,D,E



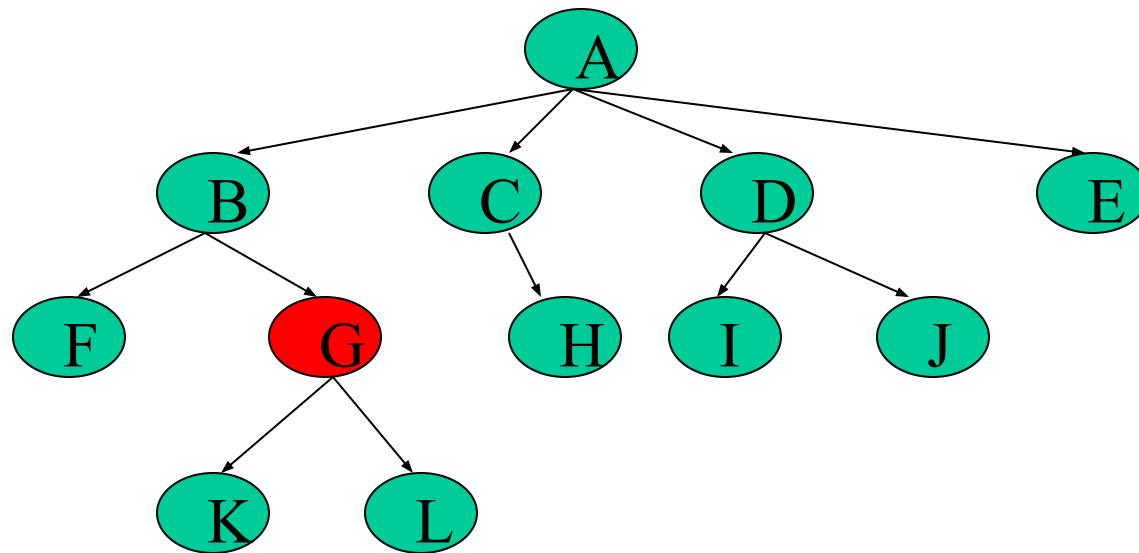
Breadth First Search

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



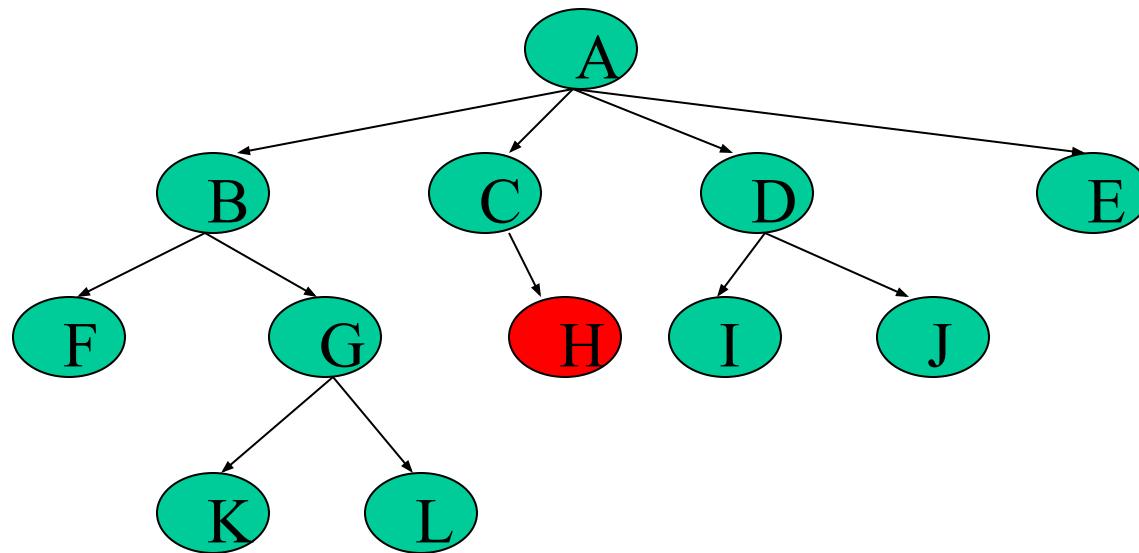
Breadth First Search

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



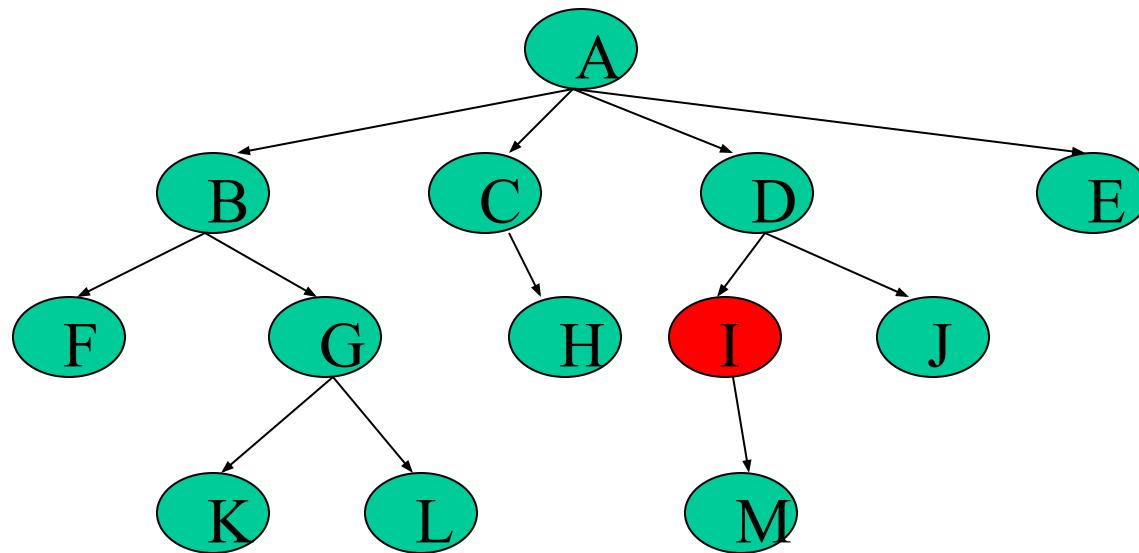
Breadth First Search

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



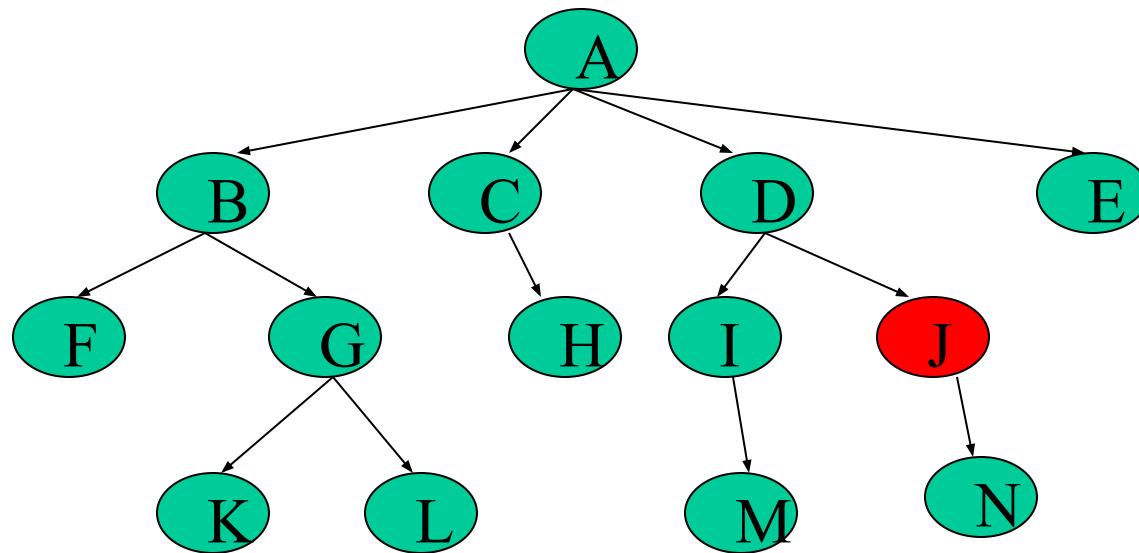
Breadth First Search

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



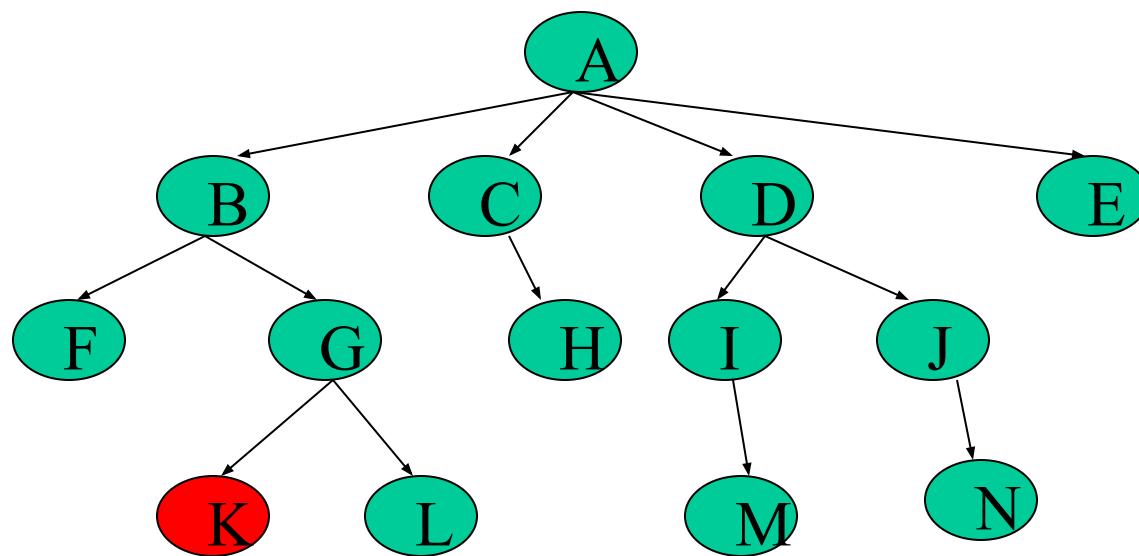
Breadth First Search

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



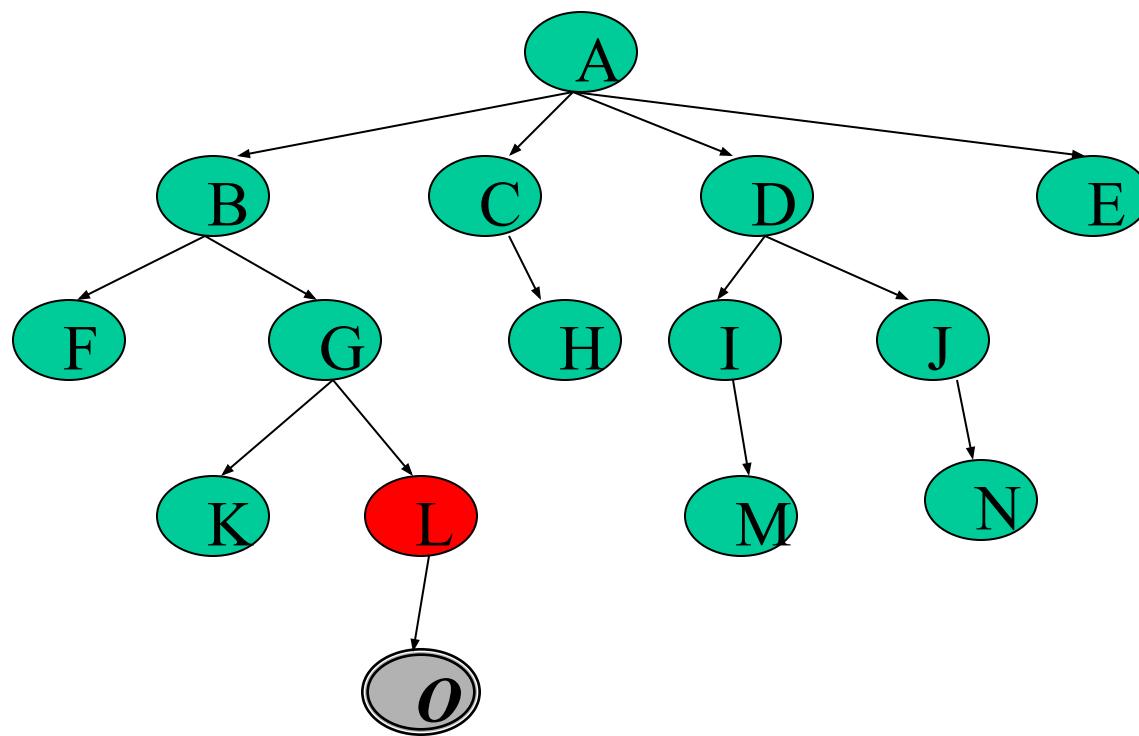
Breadth First Search

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



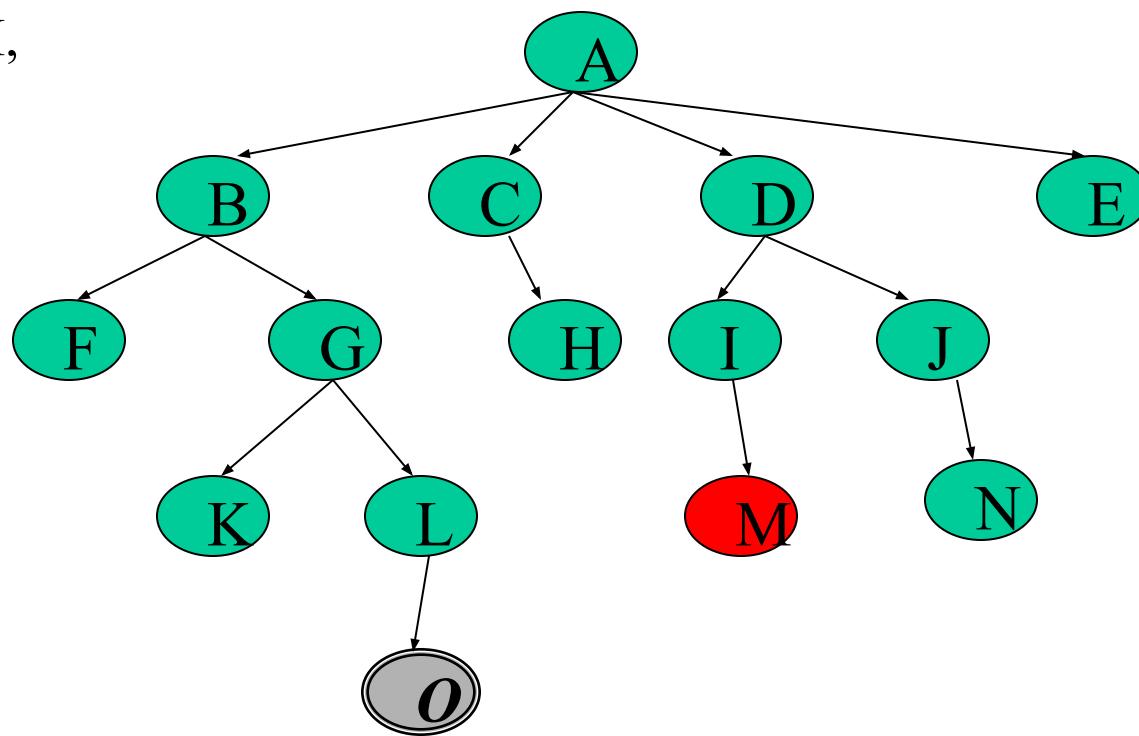
Breadth First Search

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



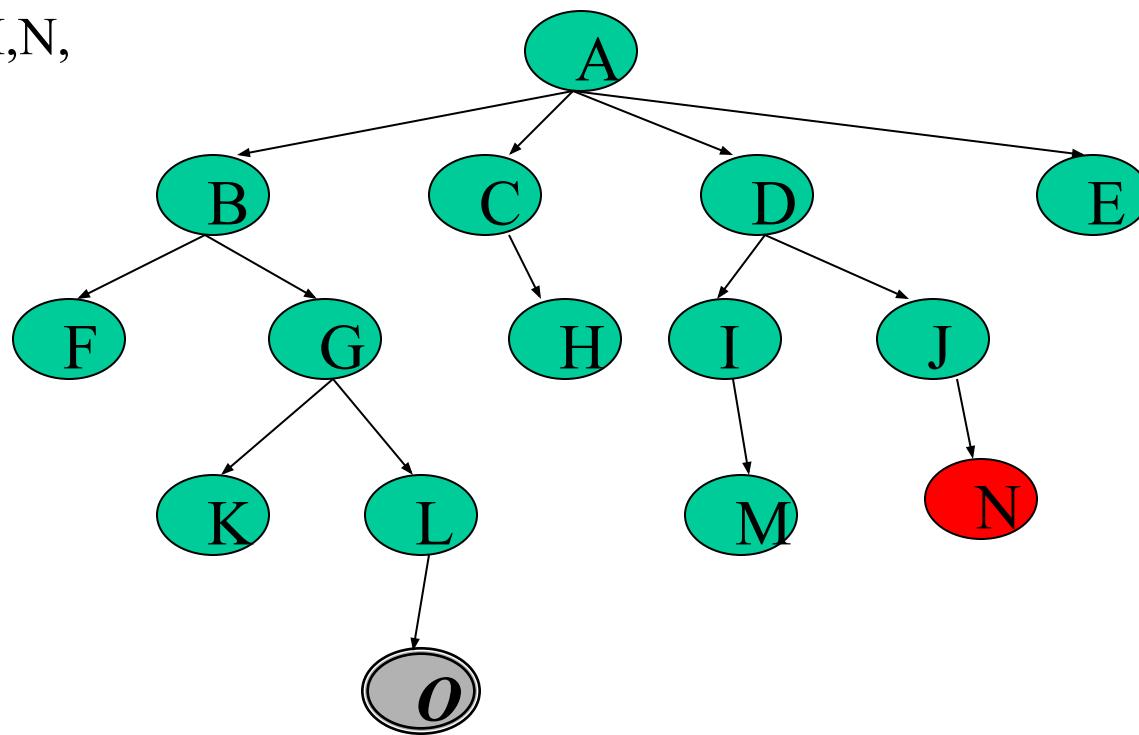
Breadth First Search

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



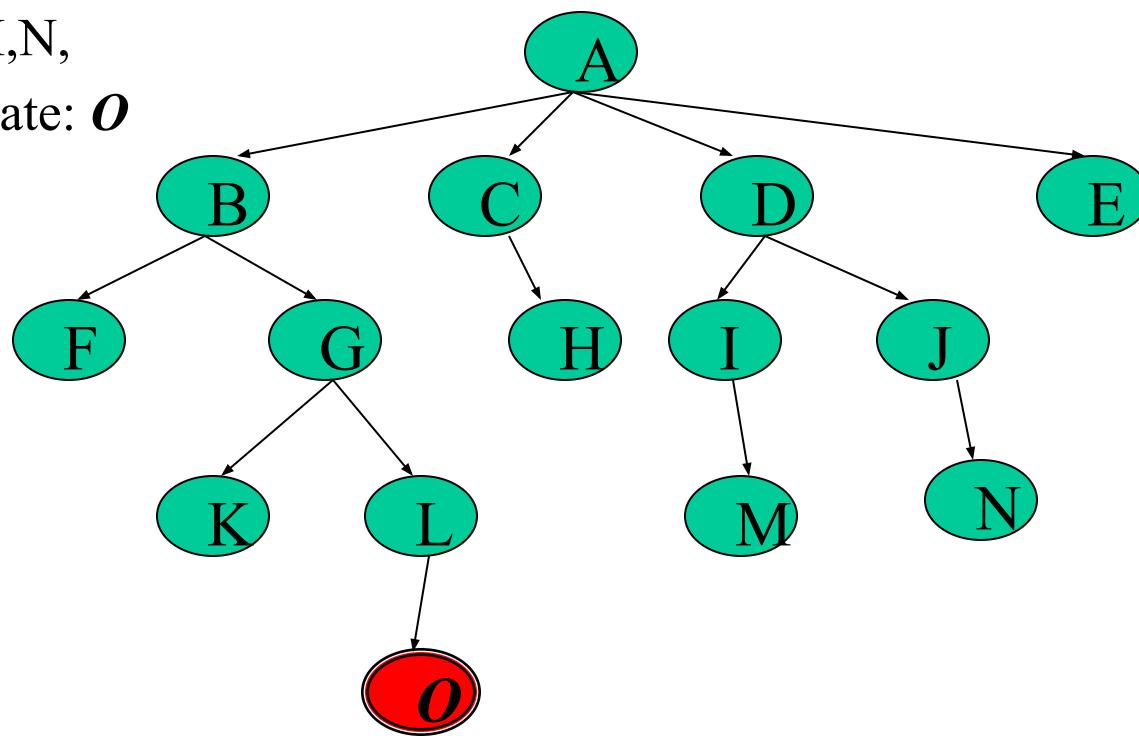
Breadth First Search

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



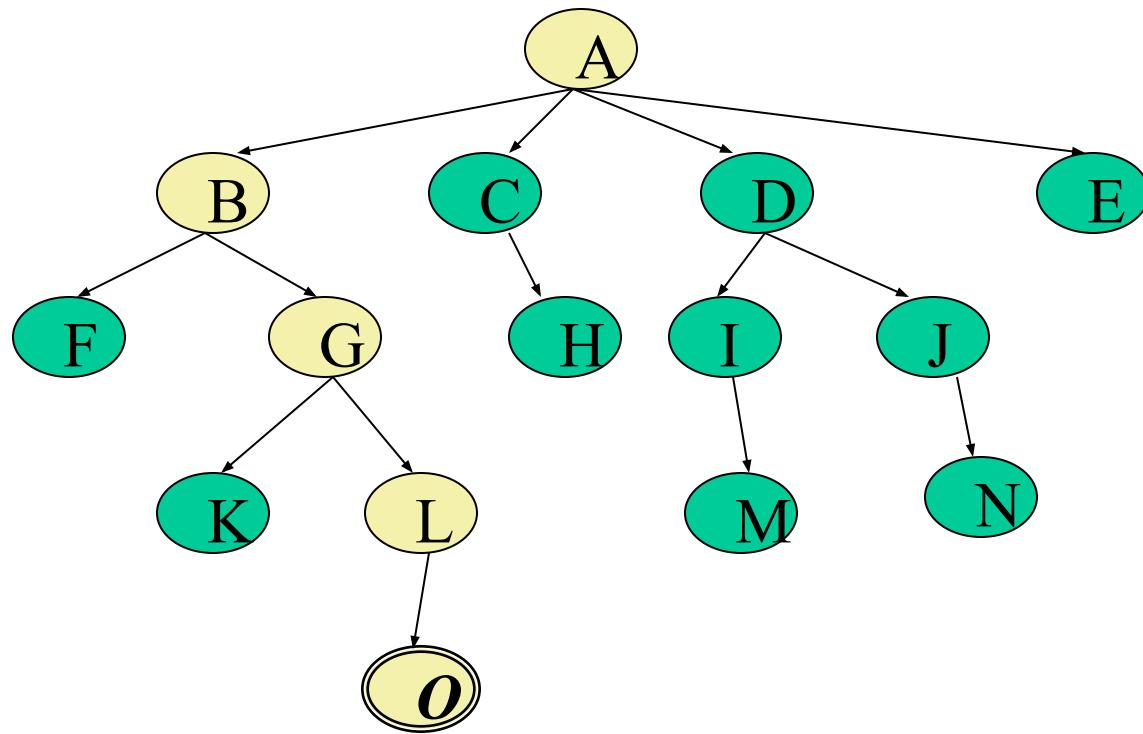
Breadth First Search

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



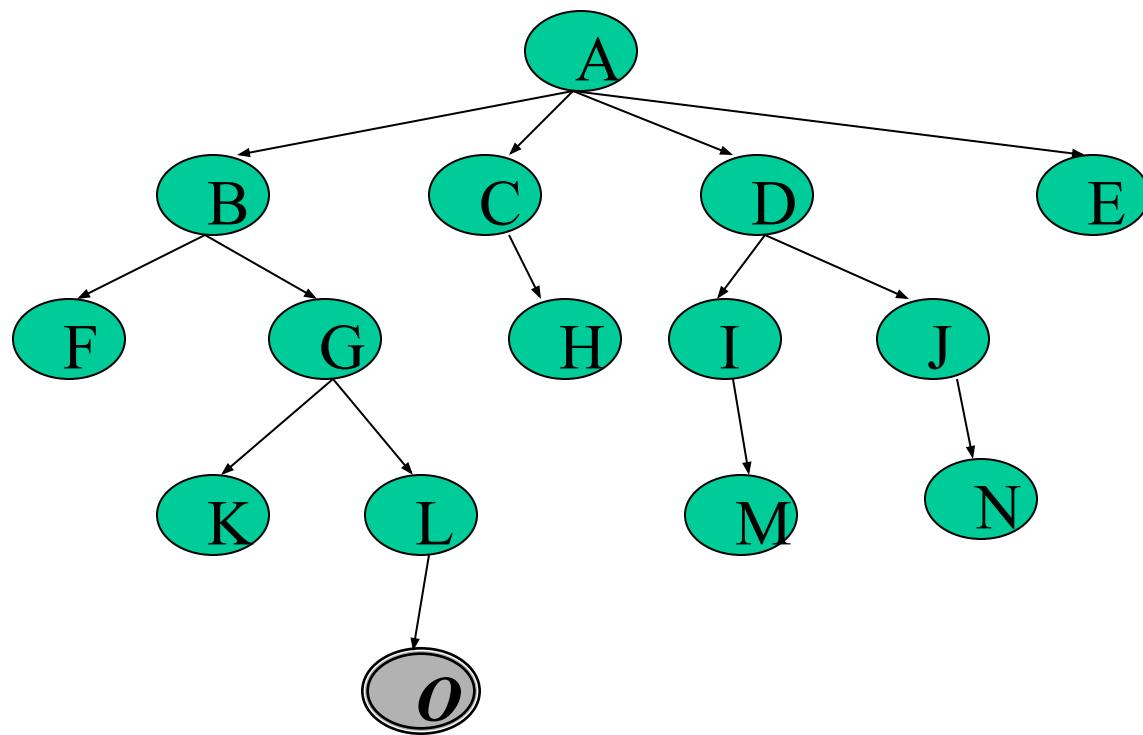
Breadth First Search

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



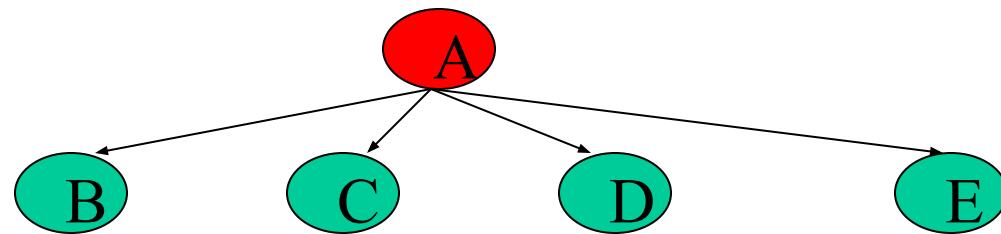
Depth First Search (DFS)

- 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):



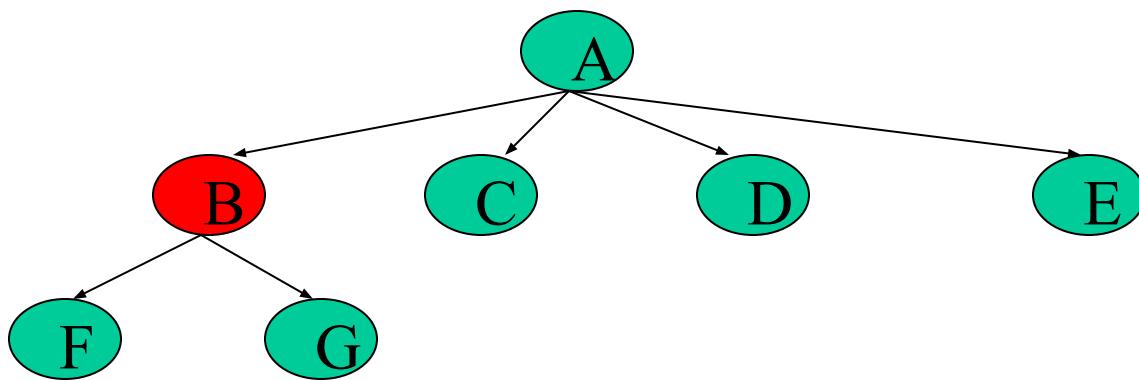
Depth First Search

- A,



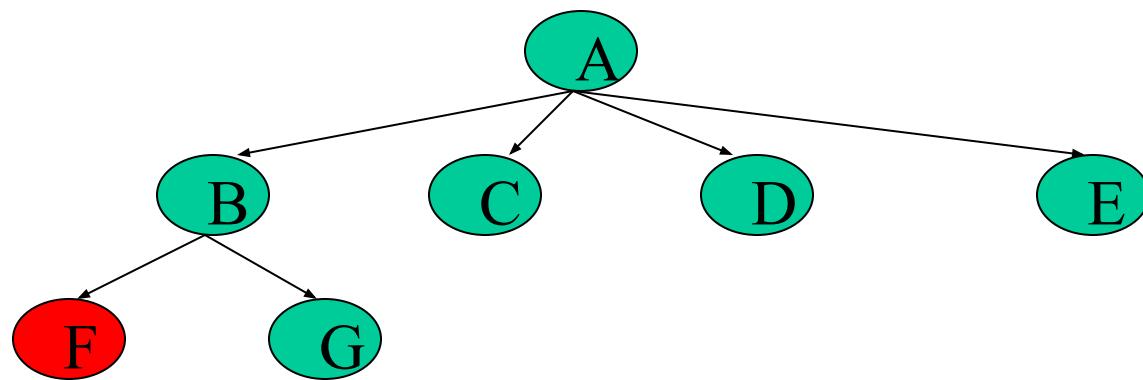
Depth First Search

- A,B,



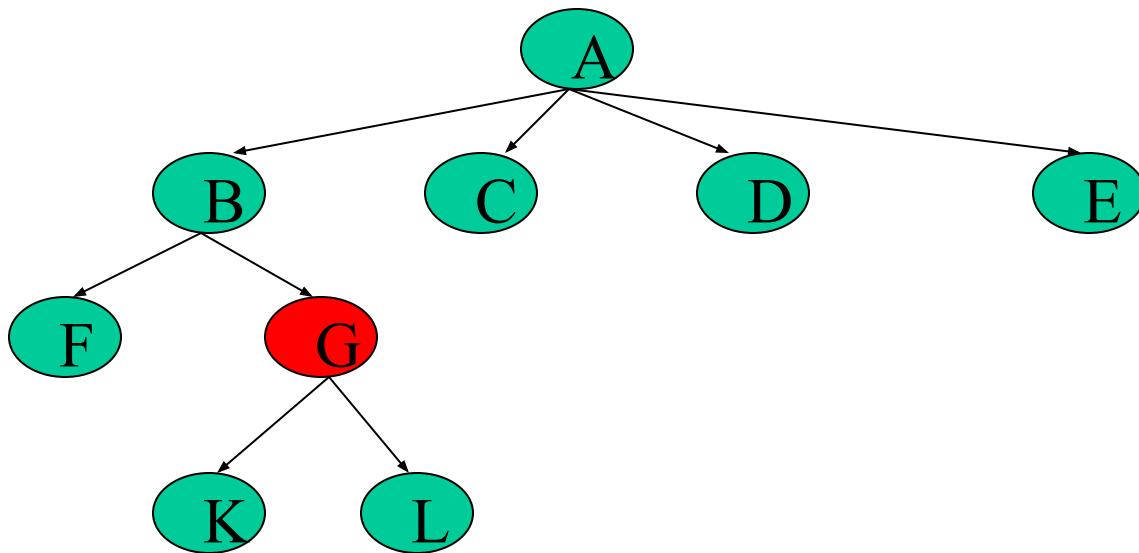
Depth First Search

- A,B,F,



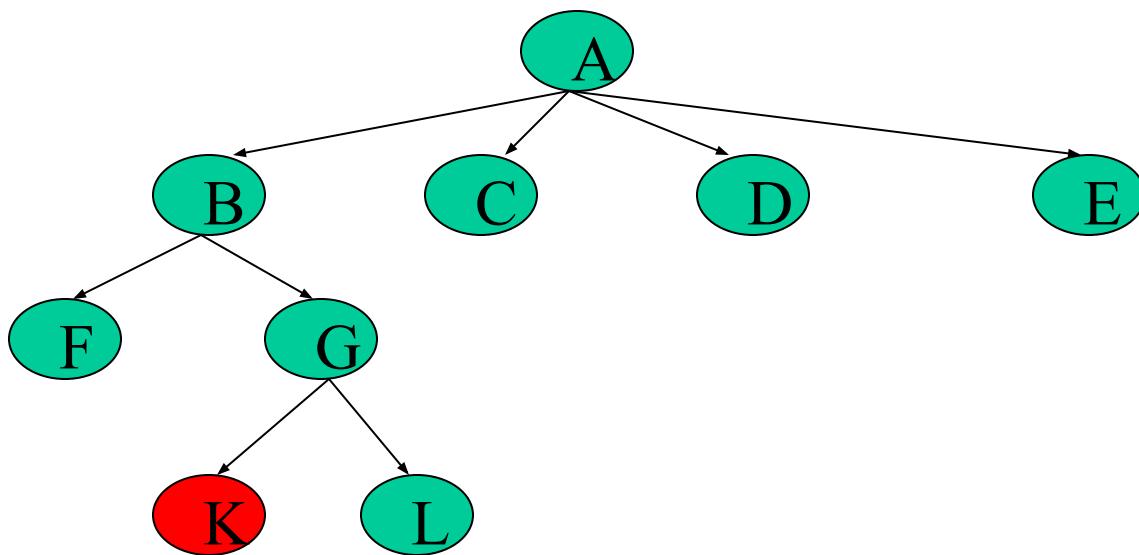
Depth First Search

- A,B,F,
- G,



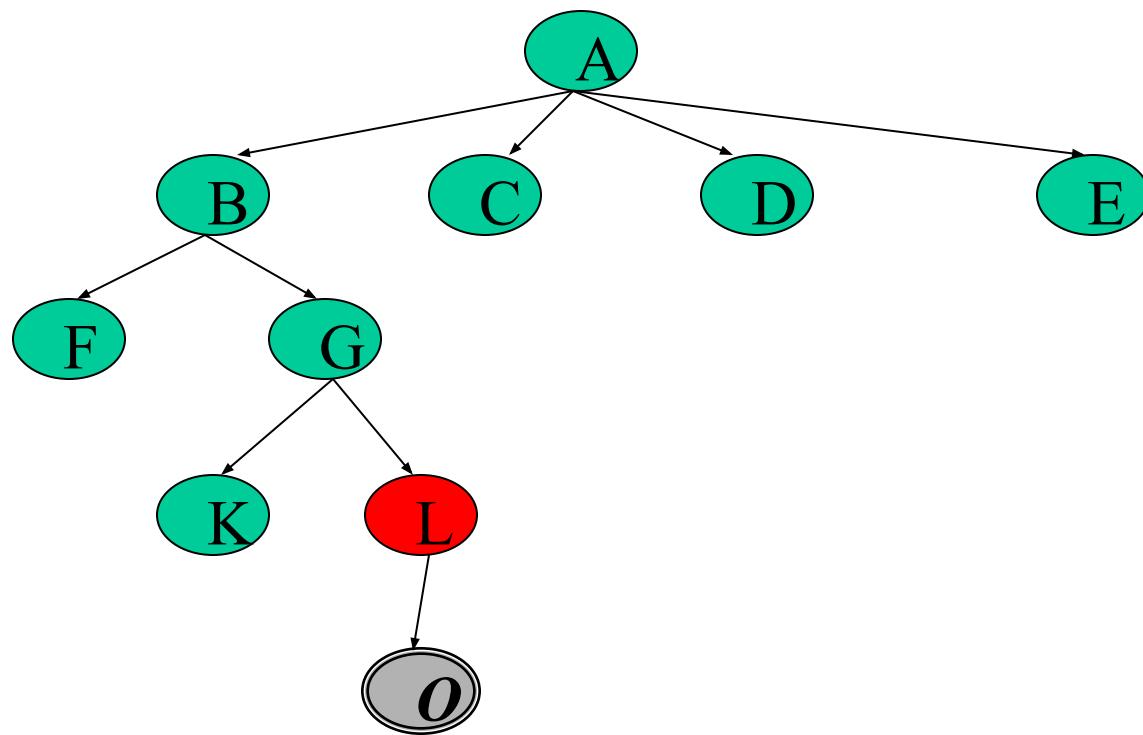
Depth First Search

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



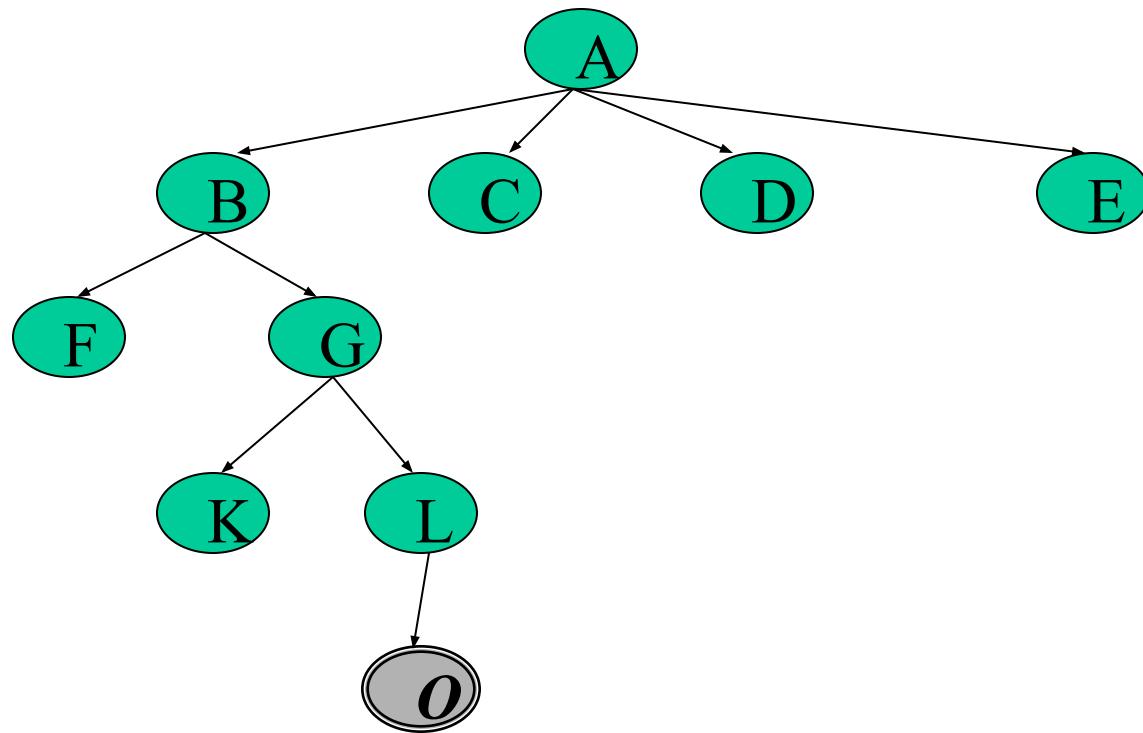
Depth First Search

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



Depth First Search

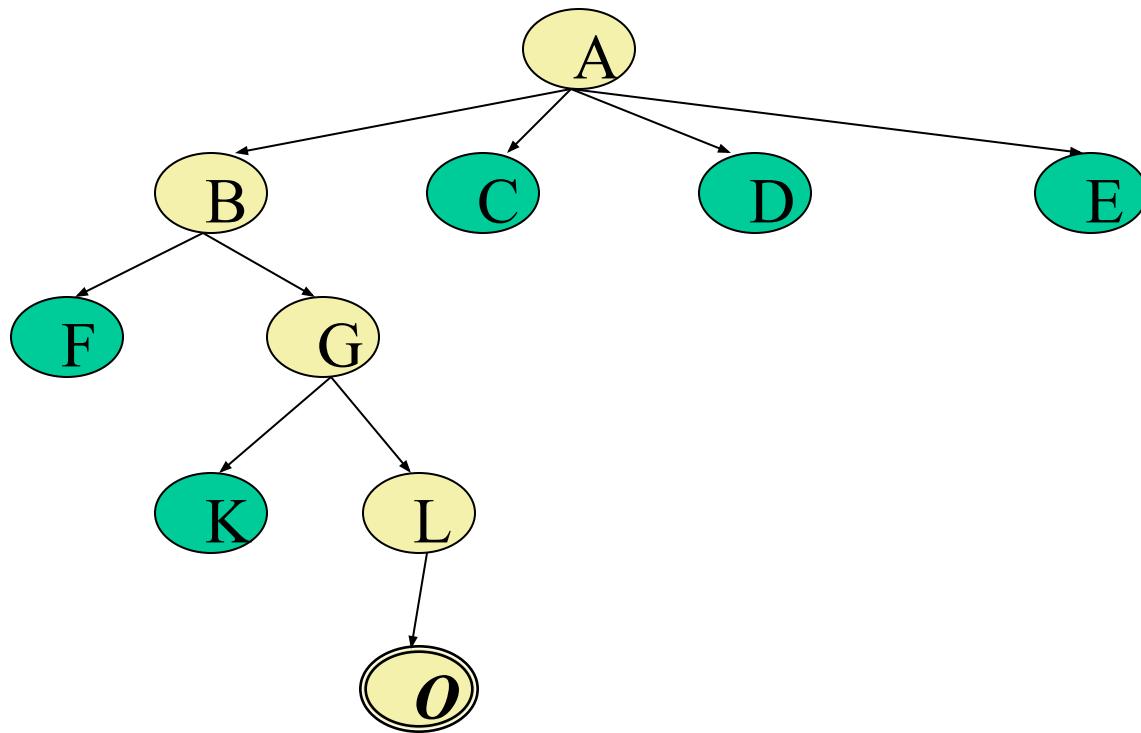
- A,B,F,
- G,K,
- L, O : Goal State



Depth First Search

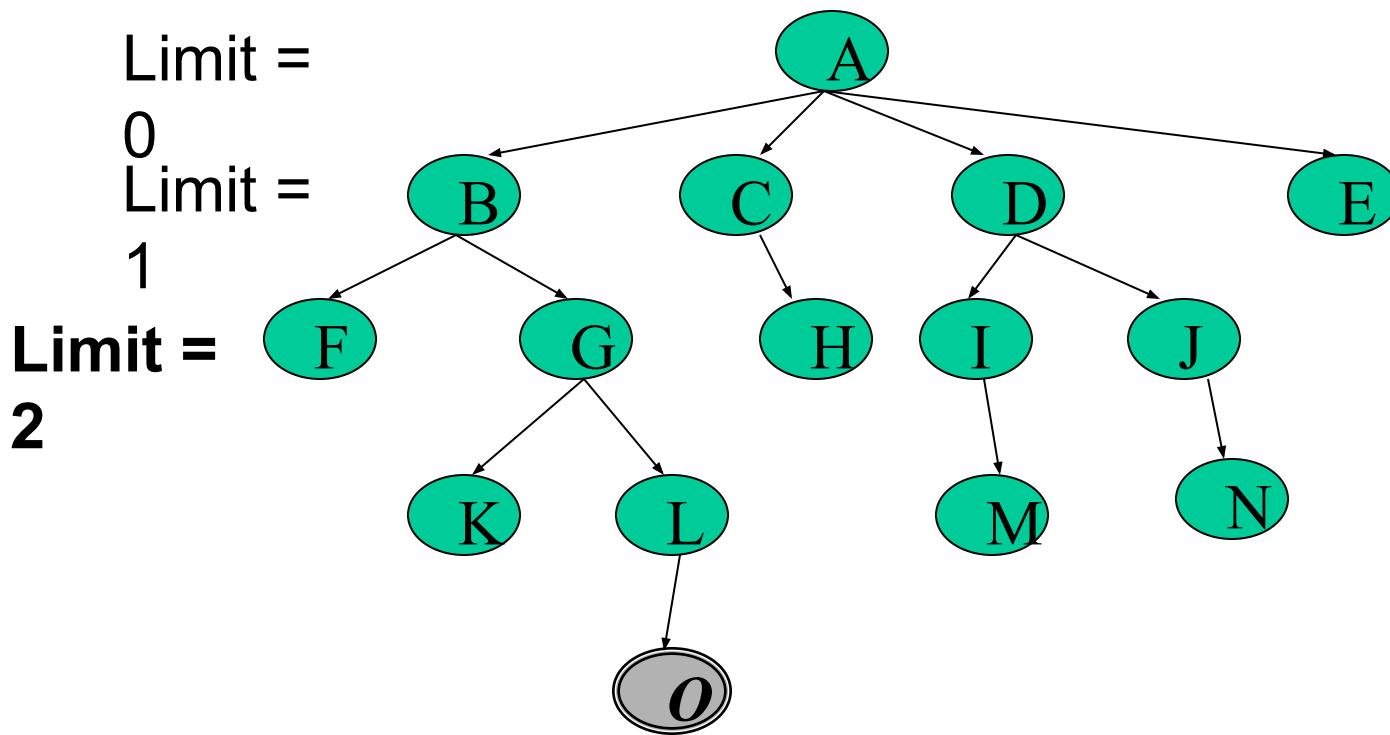
The returned solution is the sequence of operators in the path:

A, B, G, L, O



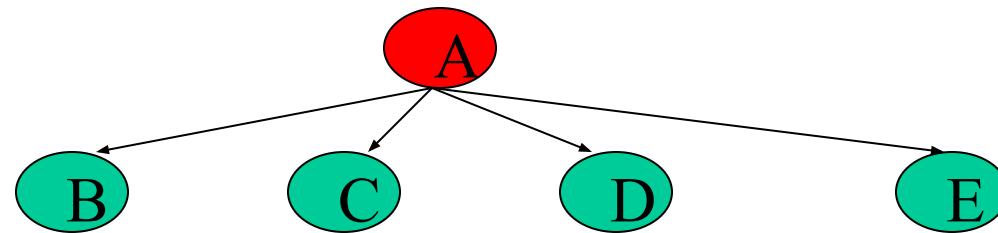
Depth-Limited Search (DLS)

- Application3:
Given the following state space (tree search), give the sequence of visited nodes when using DLS (Limit = 2):



Depth-Limited Search (DLS)

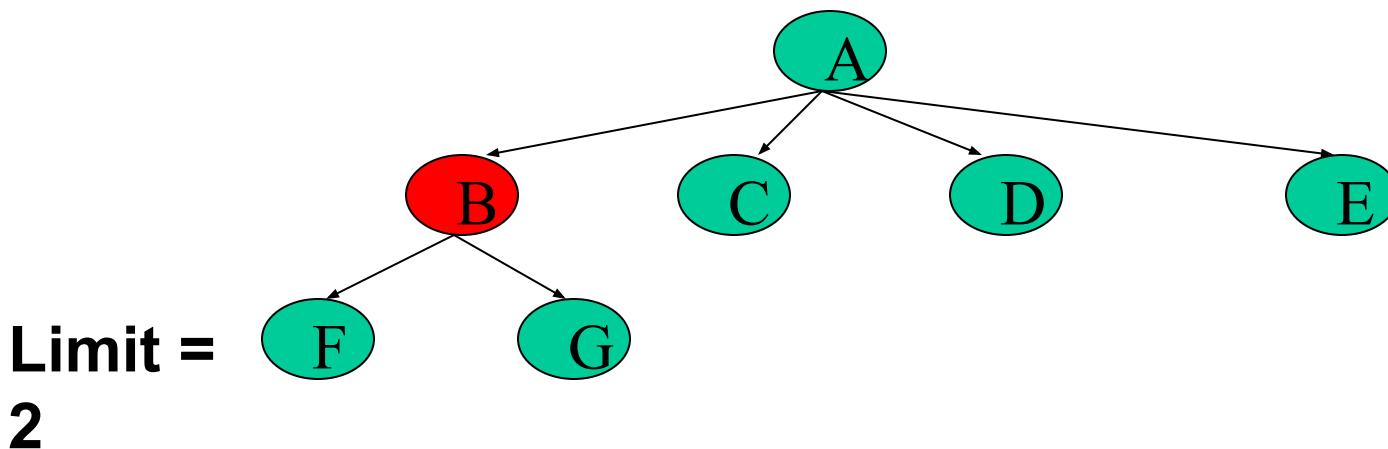
- A,



**Limit =
2**

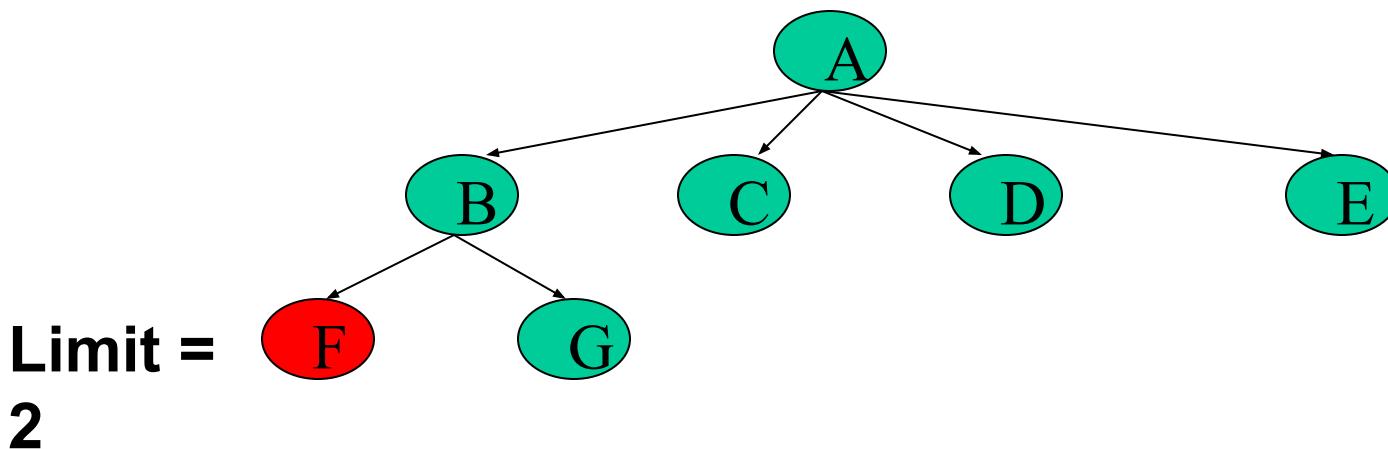
Depth-Limited Search (DLS)

- A,B,



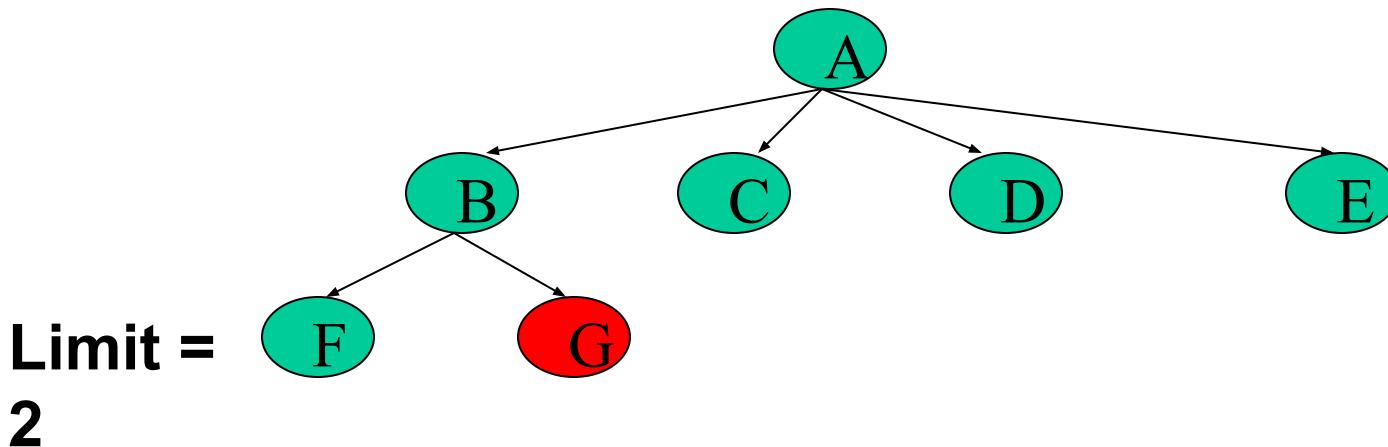
Depth-Limited Search (DLS)

- A,B,F,



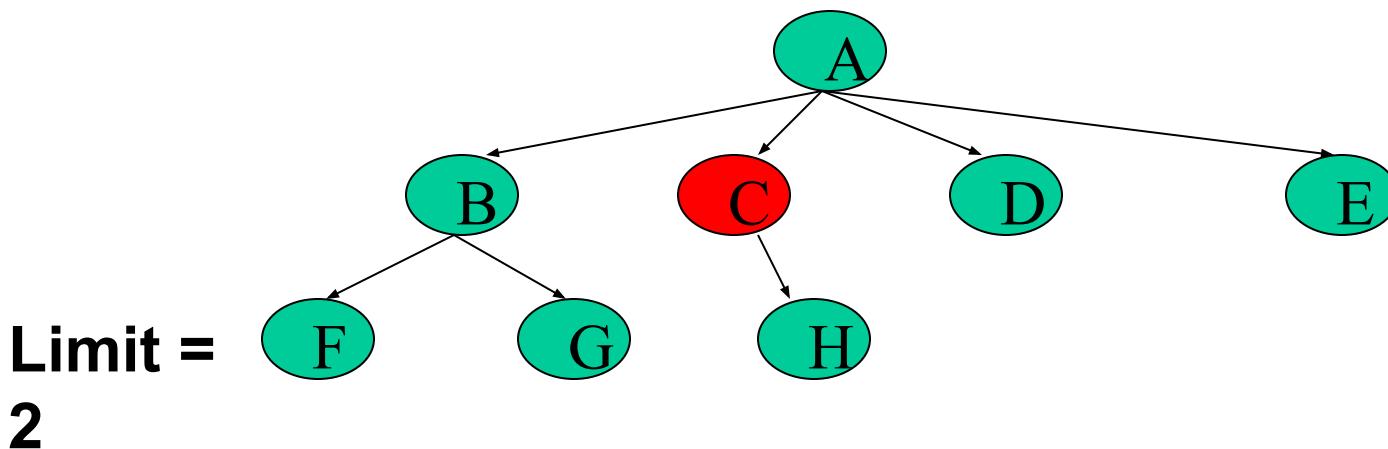
Depth-Limited Search (DLS)

- A,B,F,
- G,



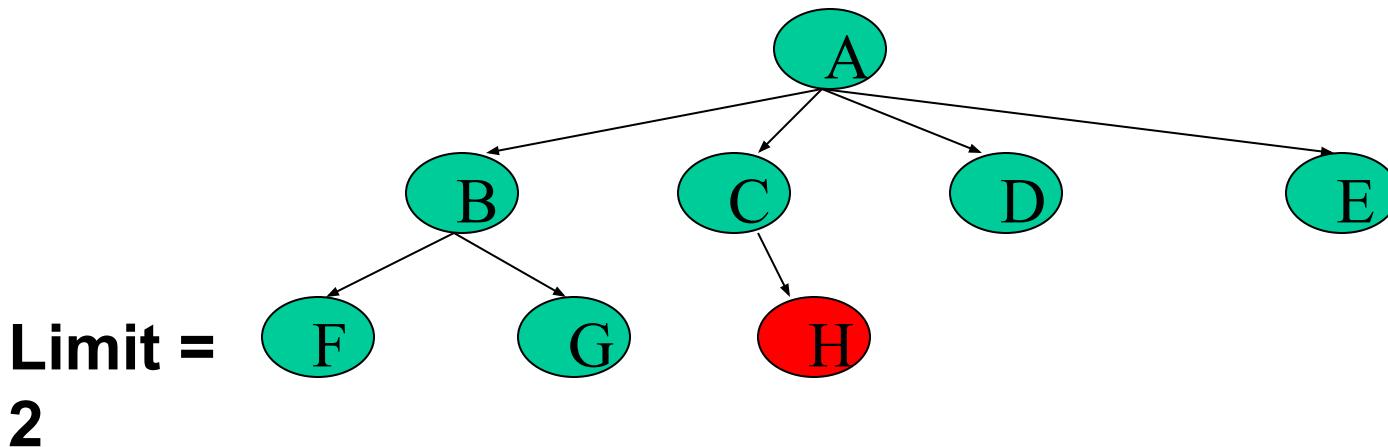
Depth-Limited Search (DLS)

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



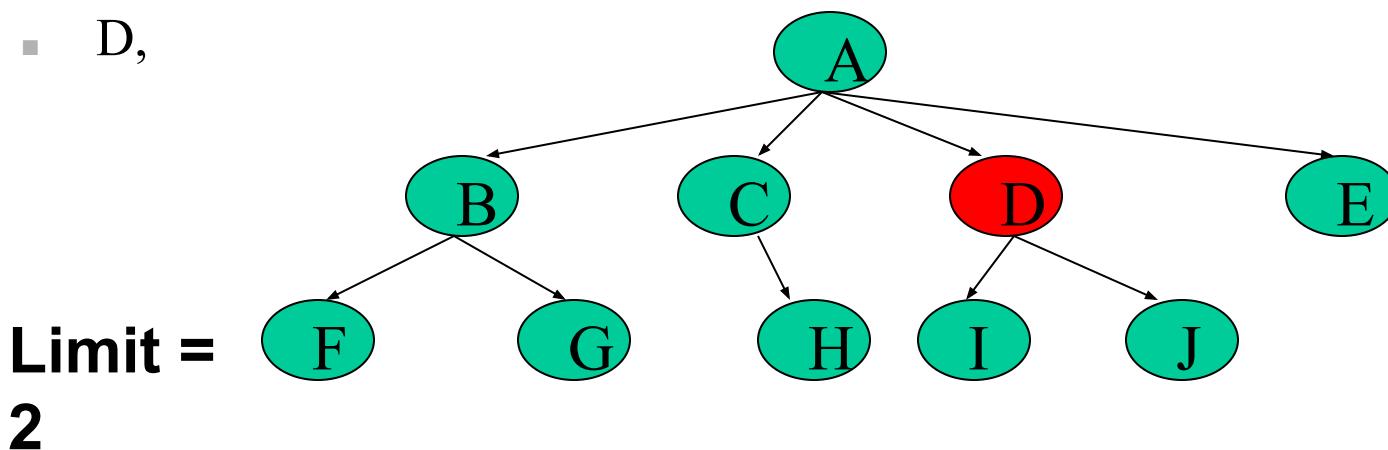
Depth-Limited Search (DLS)

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



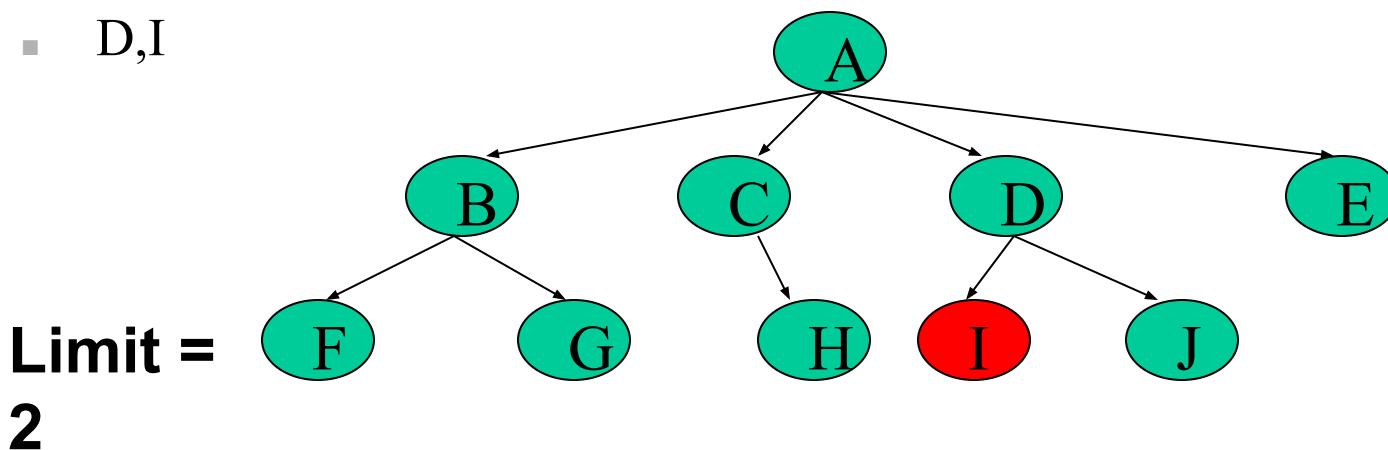
Depth-Limited Search (DLS)

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



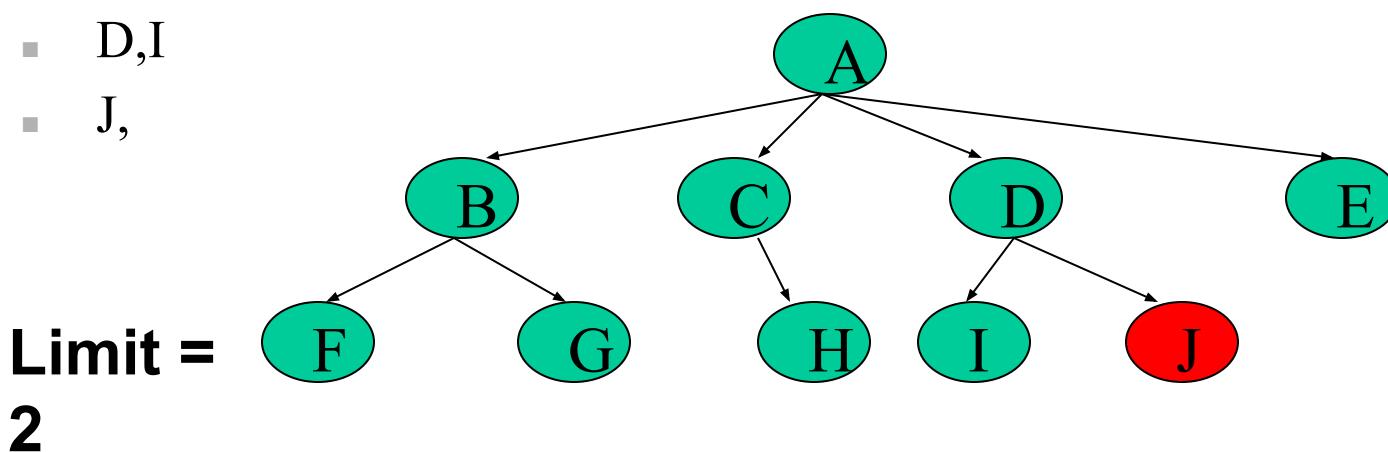
Depth-Limited Search (DLS)

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



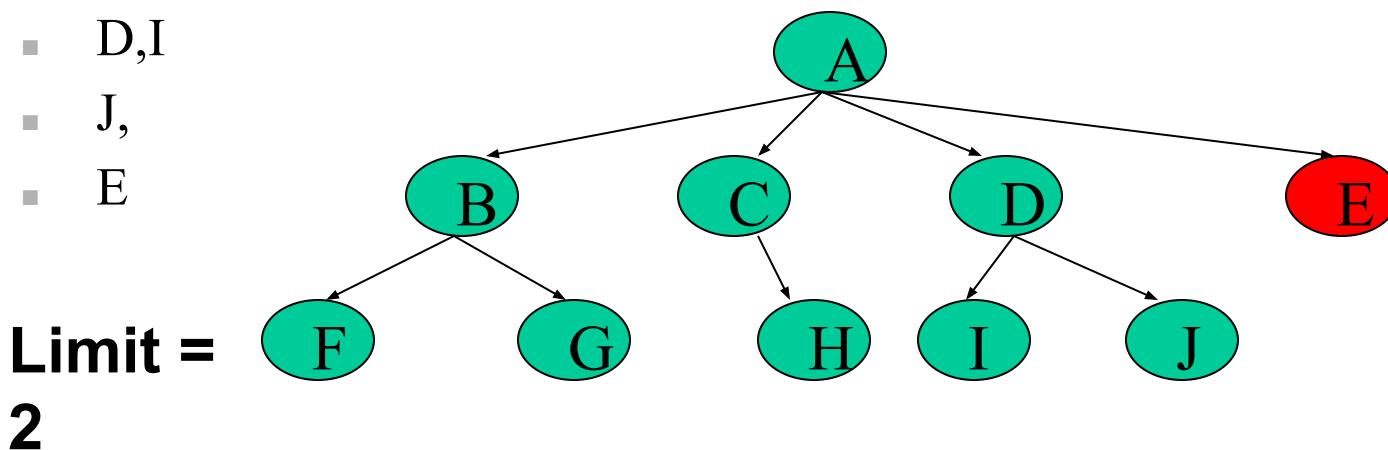
Depth-Limited Search (DLS)

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



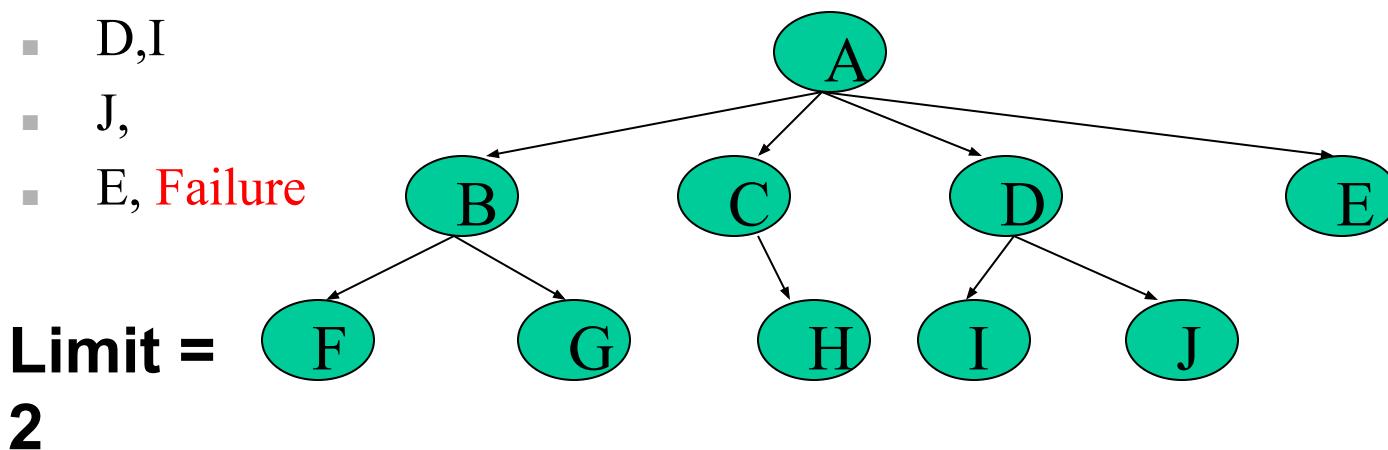
Depth-Limited Search (DLS)

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



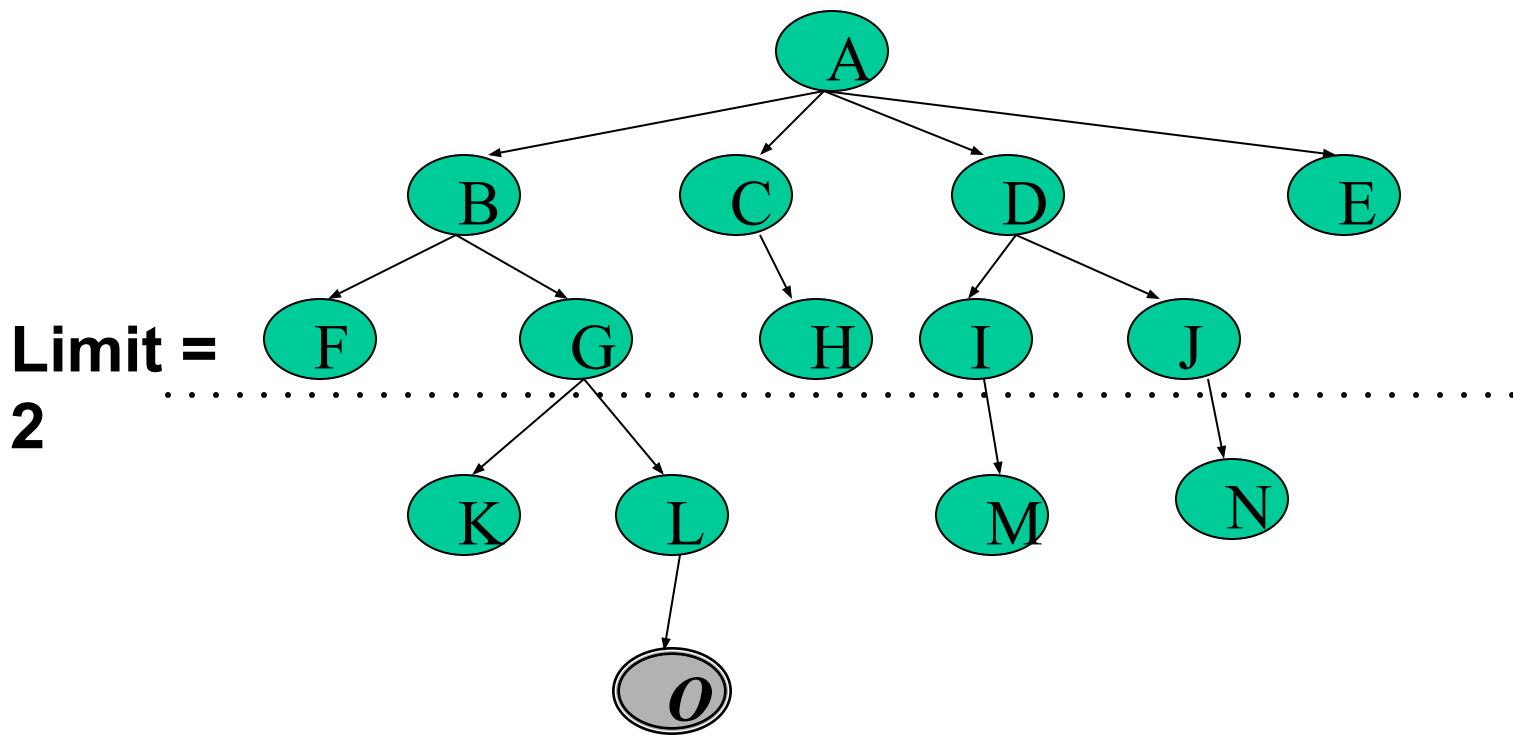
Depth-Limited Search (DLS)

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



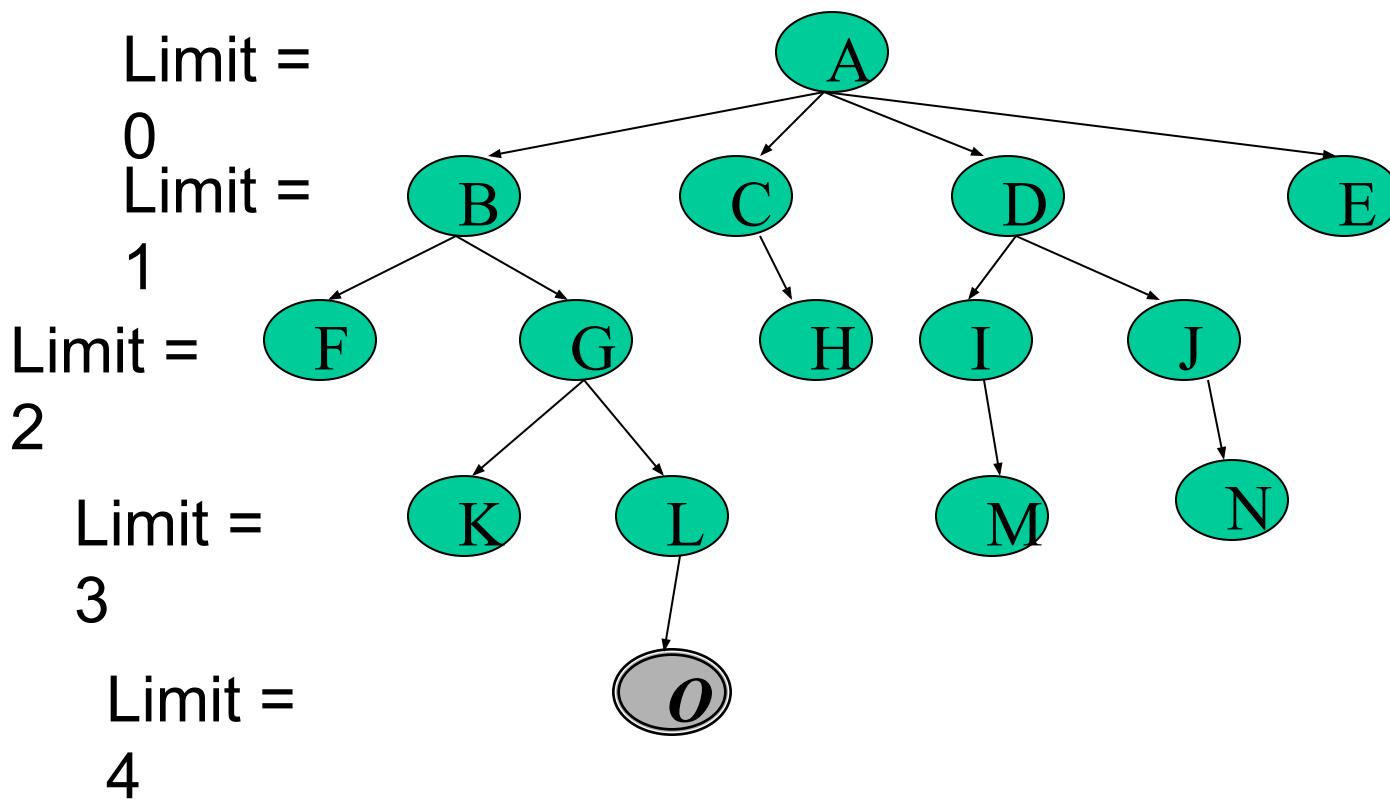
Depth-Limited Search (DLS)

- 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 (IDS)

DLS with bound = 0

Iterative Deepening Search (IDS)

- A,

Limit =
0



Iterative Deepening Search (IDS)

- A, Failure

**Limit =
0**

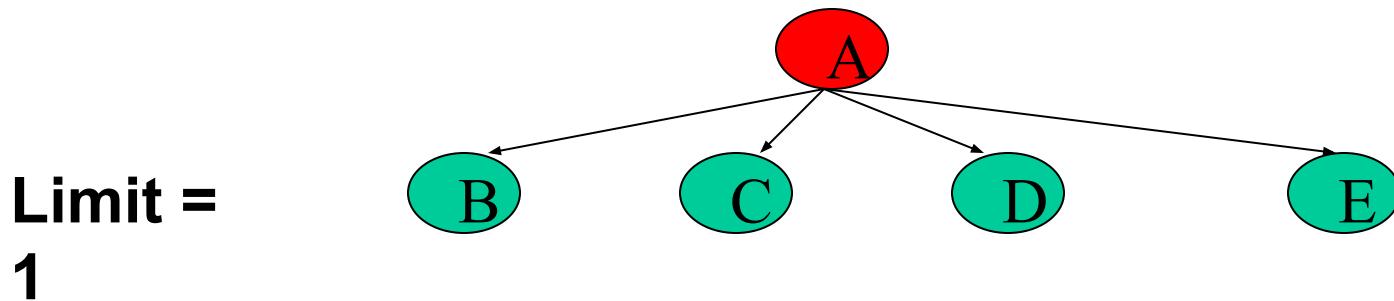


Iterative Deepening Search (IDS)

DLS with bound = 1

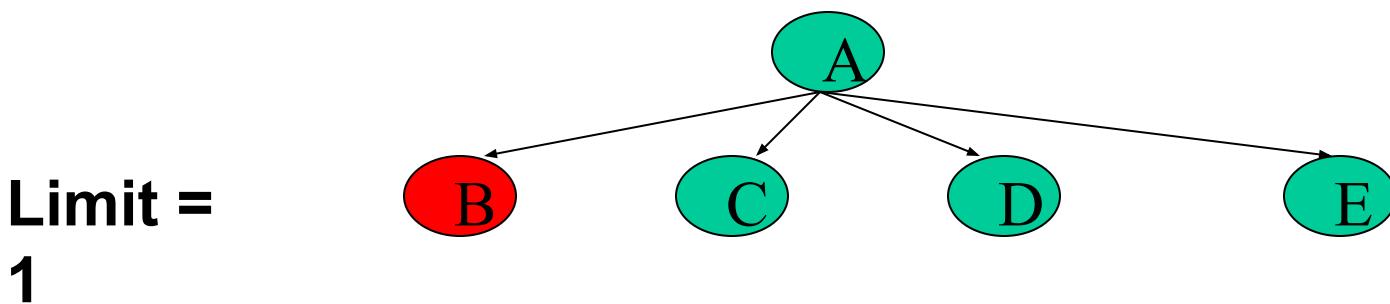
Iterative Deepening Search (IDS)

- A,



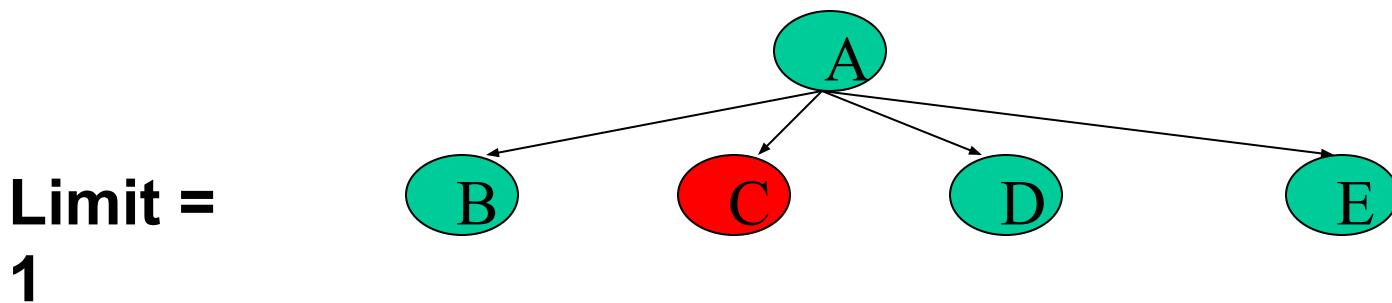
Iterative Deepening Search (IDS)

- A,B,



Iterative Deepening Search (IDS)

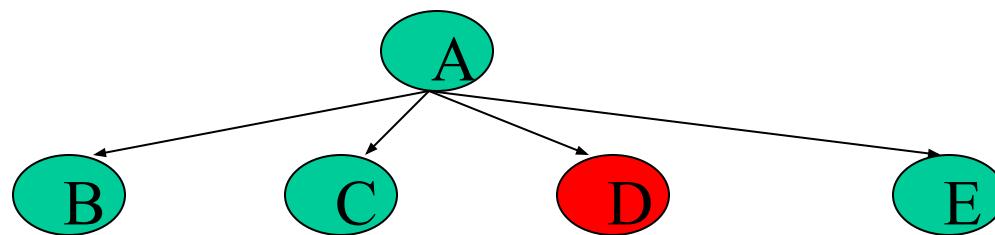
- A,B,
- C,



Iterative Deepening Search (IDS)

- A,B,
- C,
- D,

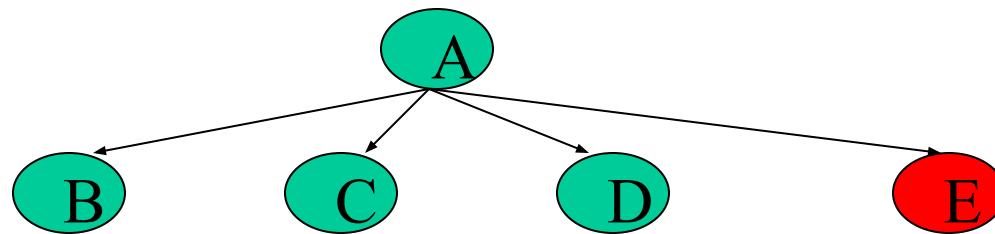
Limit =
1



Iterative Deepening Search (IDS)

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

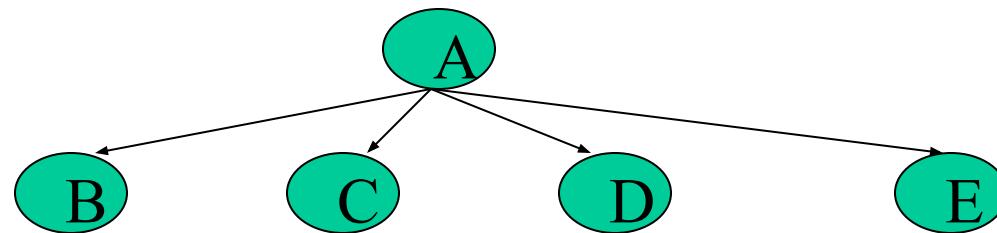
**Limit =
1**



Iterative Deepening Search (IDS)

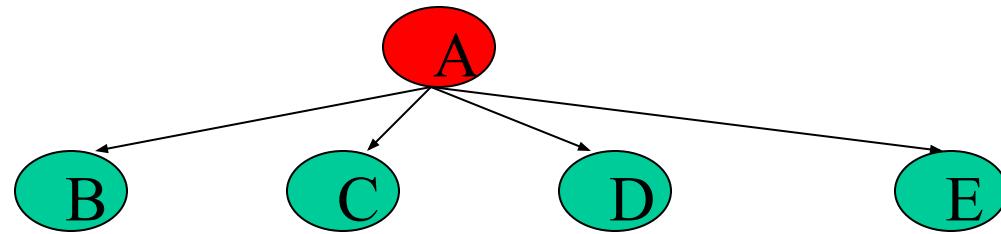
- A,B,
- C,
- D,
- E, **Failure**

**Limit =
1**



Iterative Deepening Search (IDS)

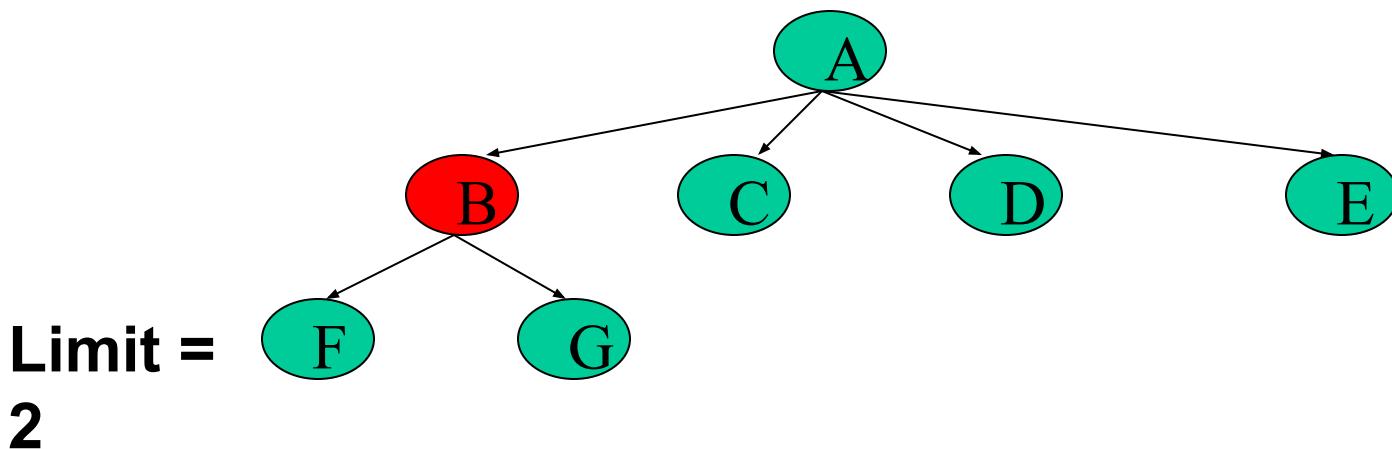
- A,



Limit =
2

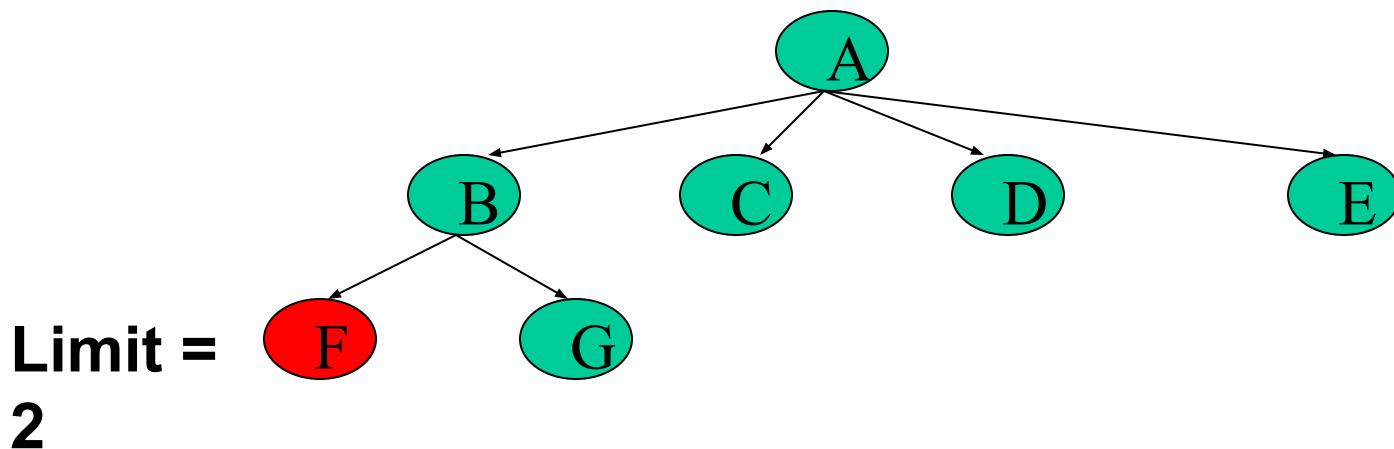
Iterative Deepening Search (IDS)

- A,B,



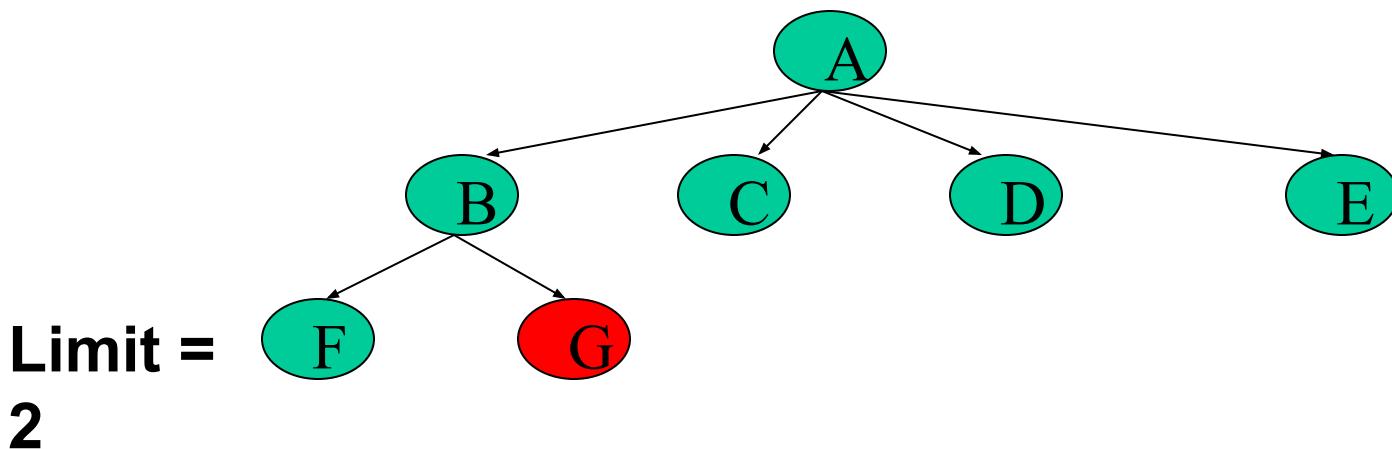
Iterative Deepening Search (IDS)

- A,B,F,



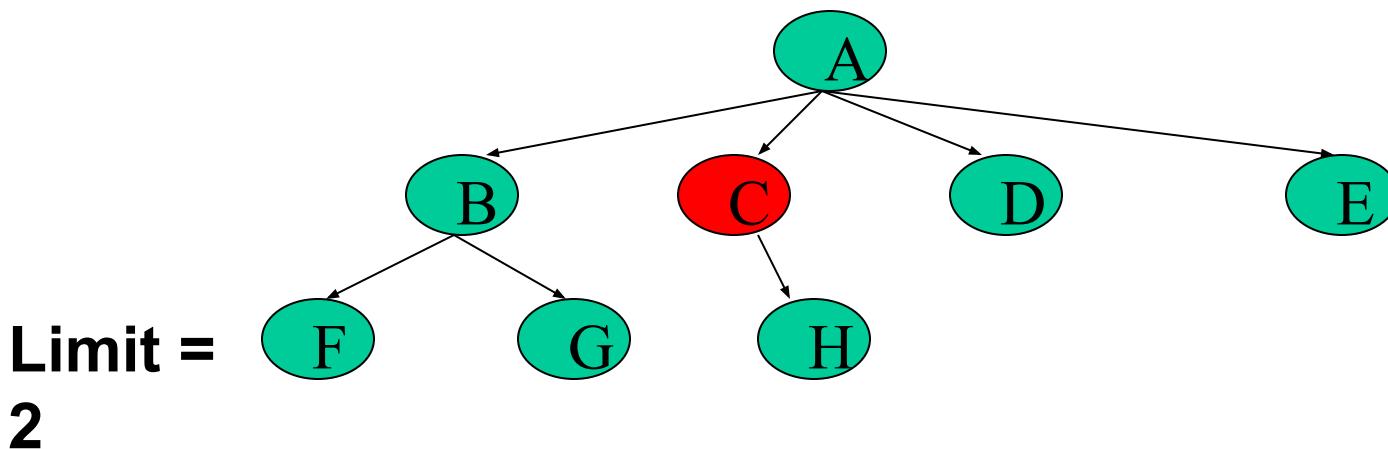
Iterative Deepening Search (IDS)

- A,B,F,
- G,



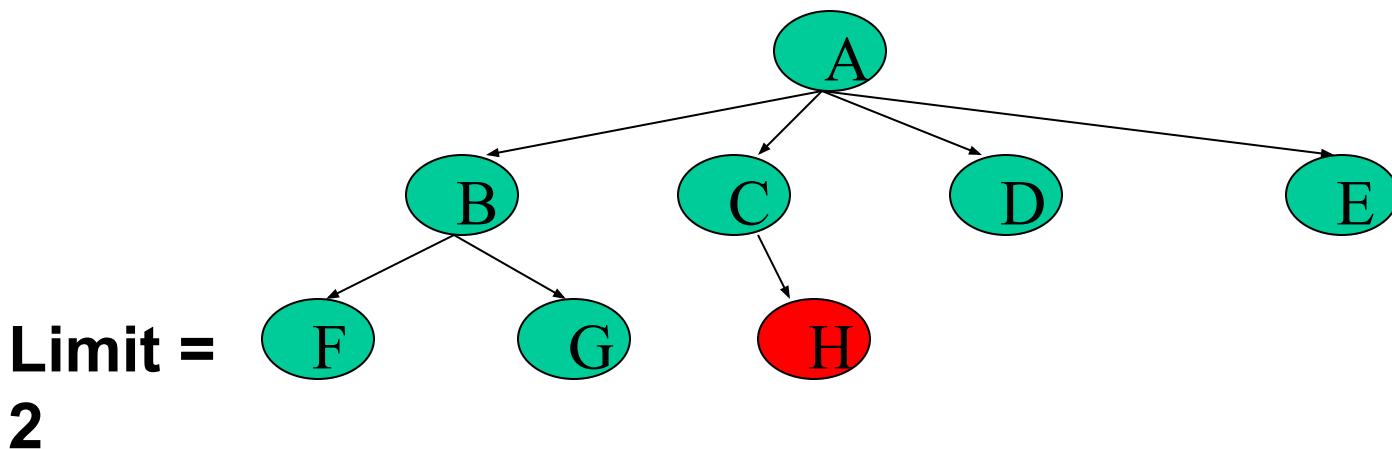
Iterative Deepening Search (IDS)

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



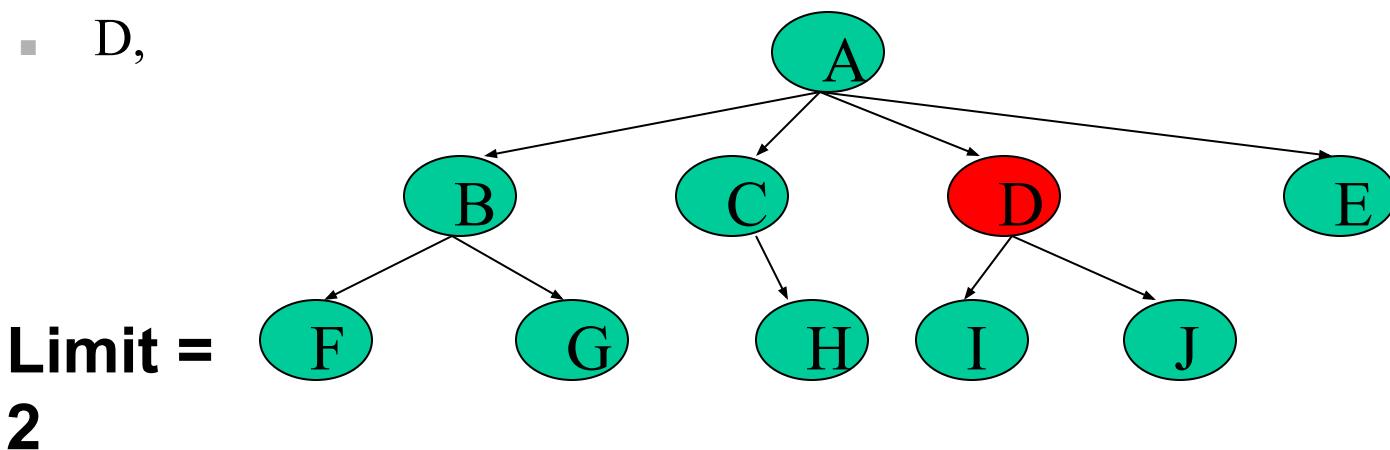
Iterative Deepening Search (IDS)

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



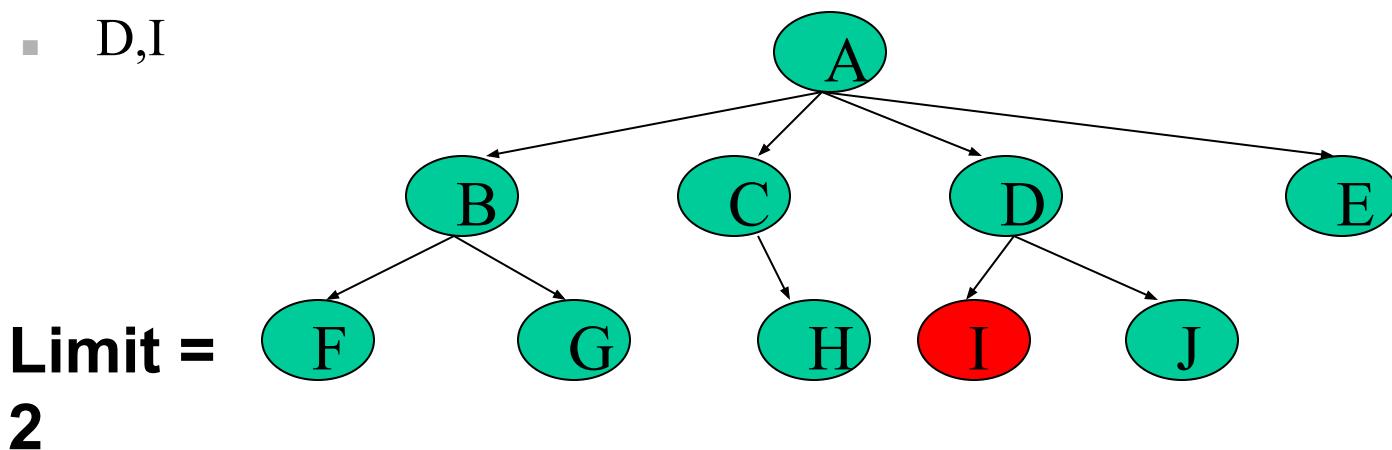
Iterative Deepening Search (IDS)

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



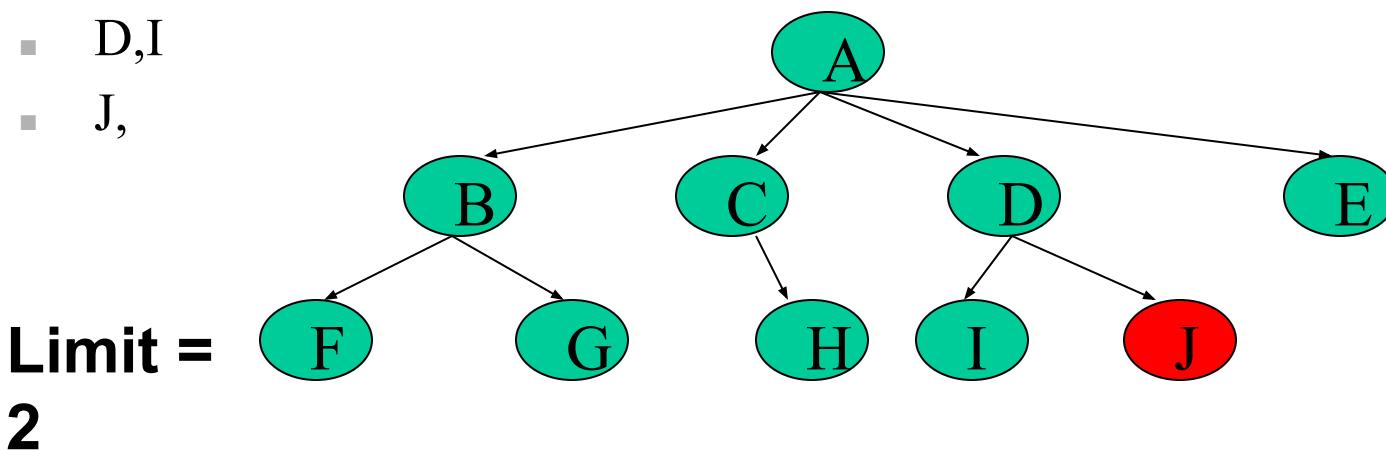
Iterative Deepening Search (IDS)

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



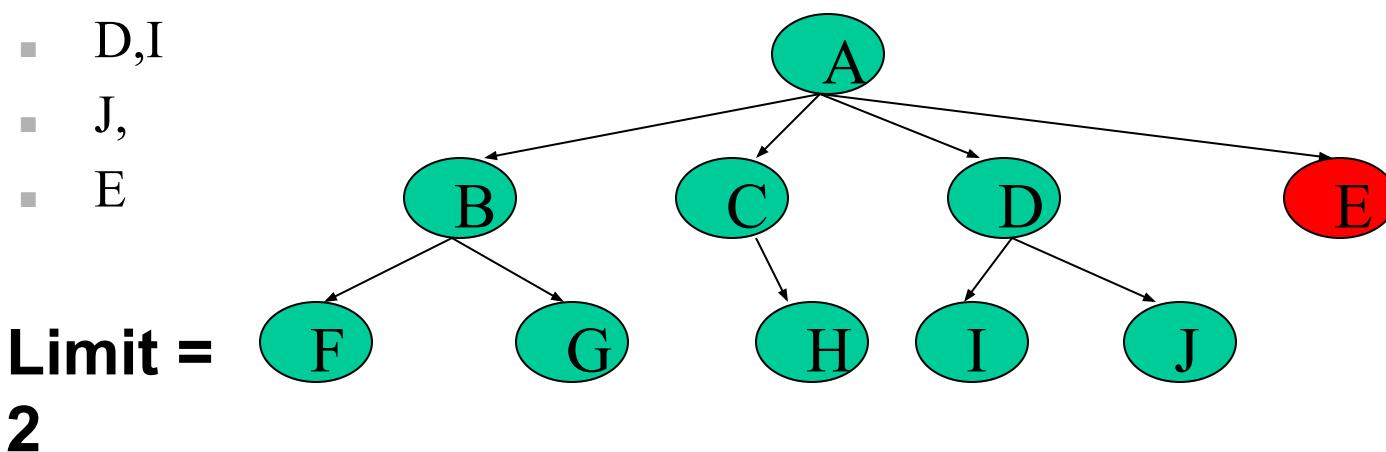
Iterative Deepening Search (IDS)

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



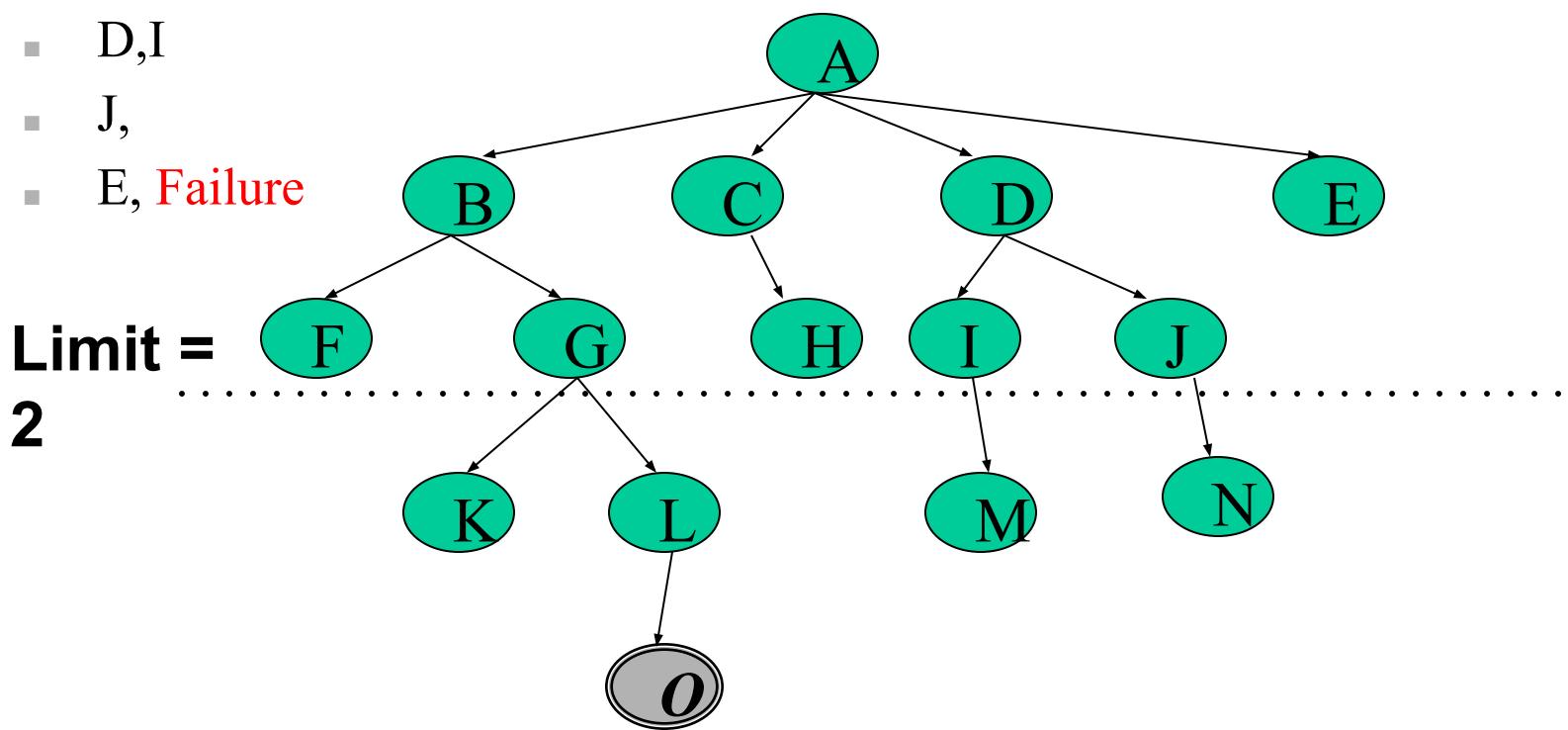
Iterative Deepening Search (IDS)

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



Iterative Deepening Search (IDS)

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

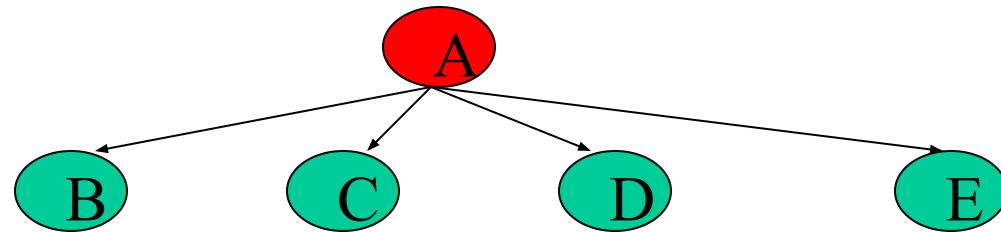


Iterative Deepening Search (IDS)

DLS with bound = 3

Iterative Deepening Search (IDS)

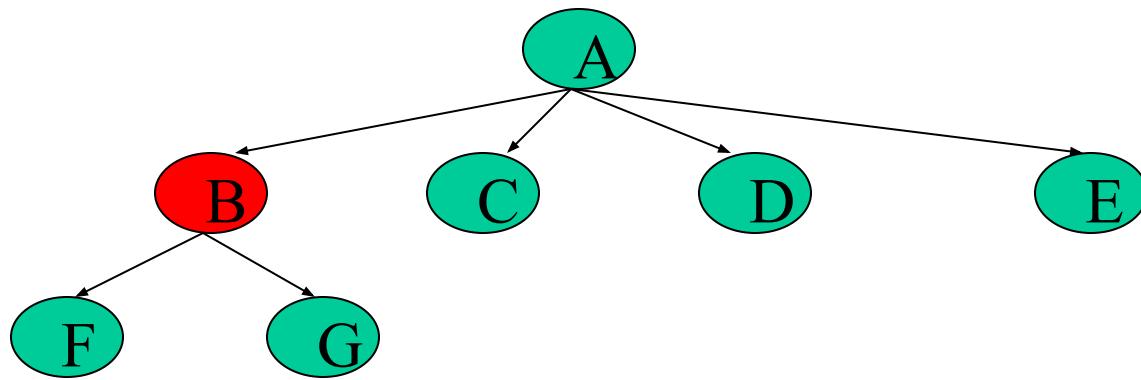
- A,



Limit =
3

Iterative Deepening Search (IDS)

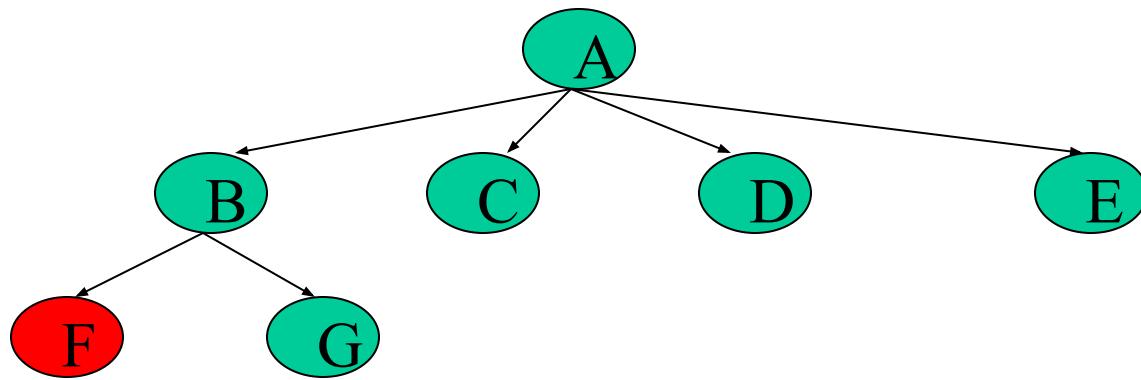
- A,B,



Limit =
3

Iterative Deepening Search (IDS)

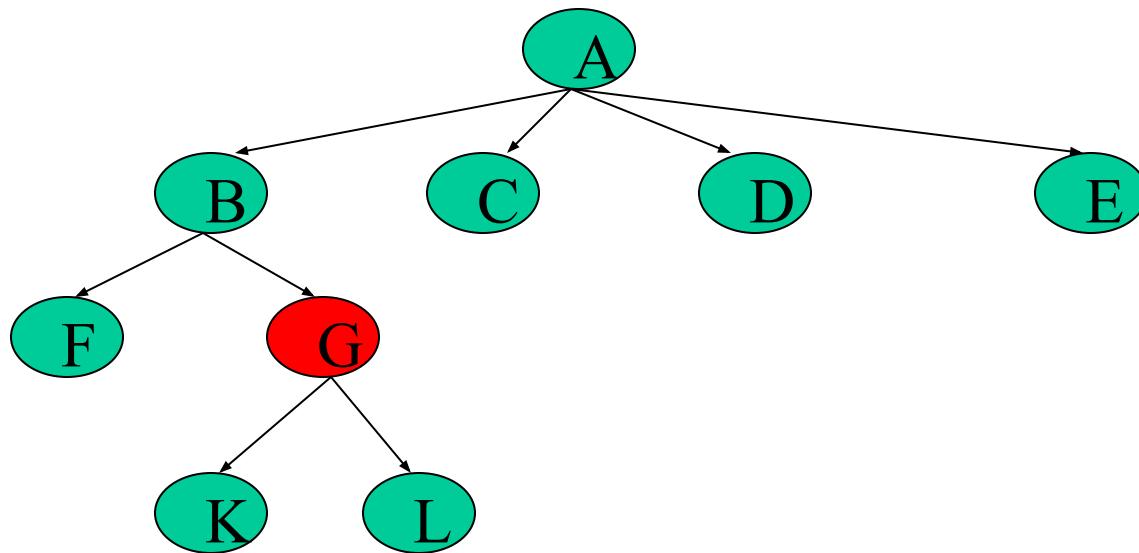
- A,B,F,



Limit =
3

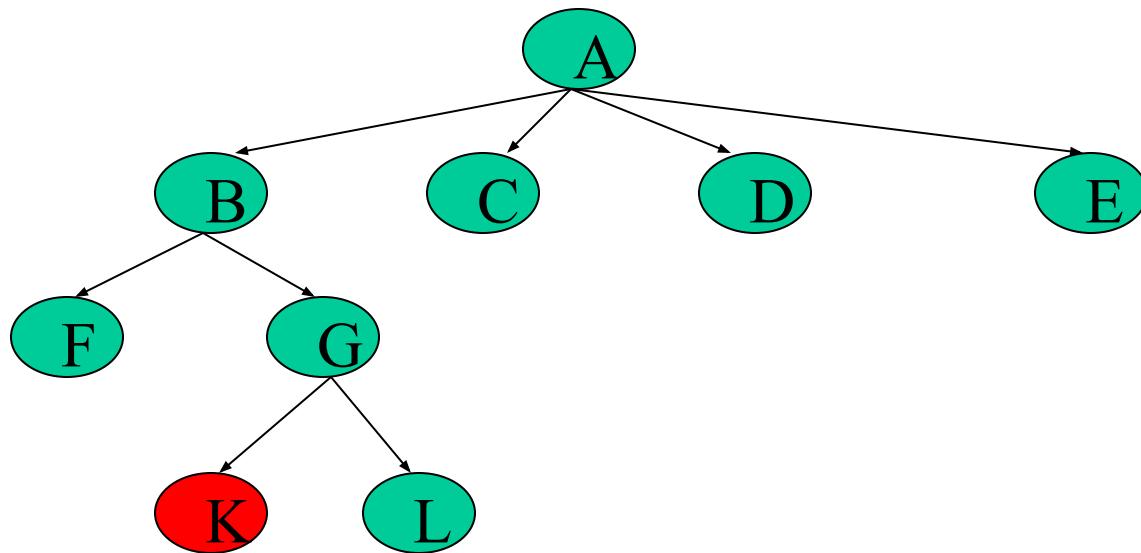
Iterative Deepening Search (IDS)

- A,B,F,
- G,



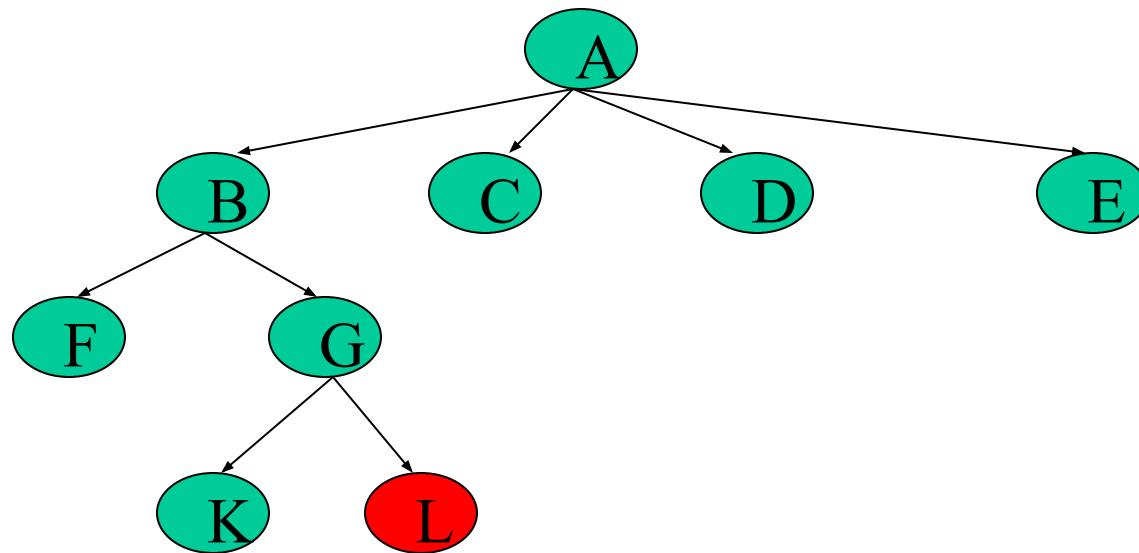
Iterative Deepening Search (IDS)

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



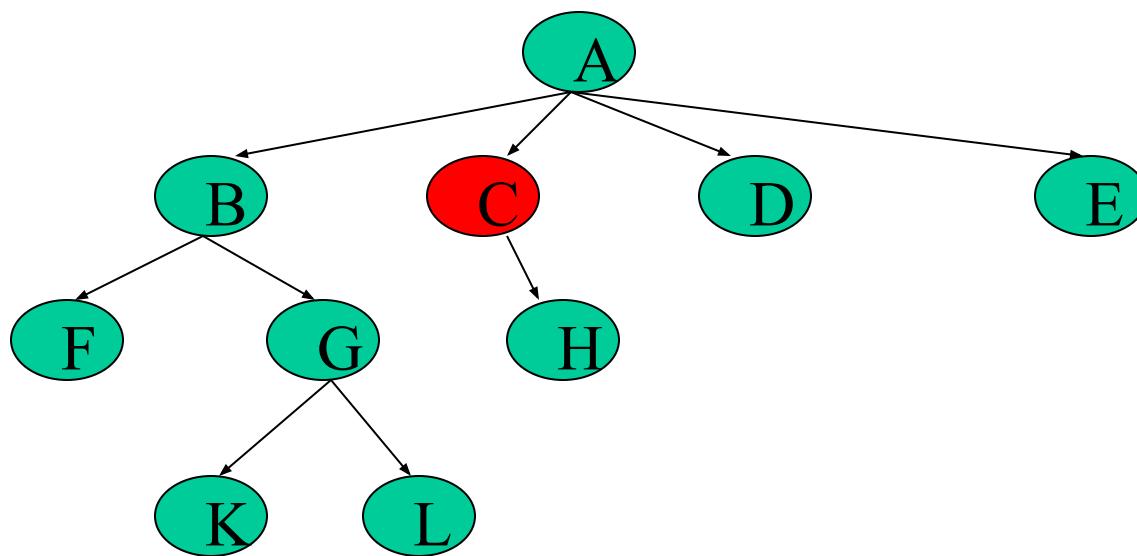
Iterative Deepening Search (IDS)

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



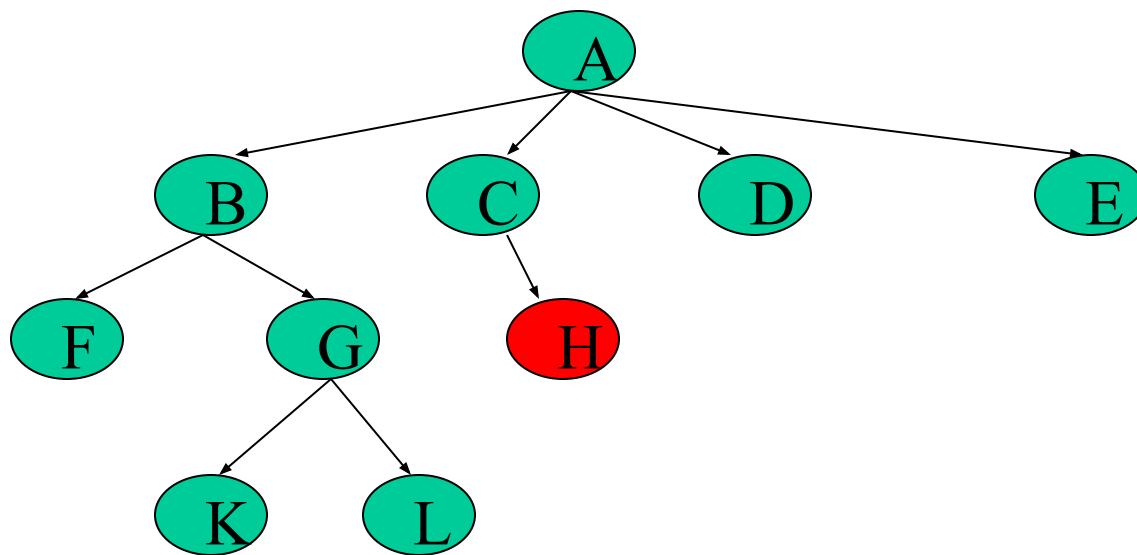
Iterative Deepening Search (IDS)

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



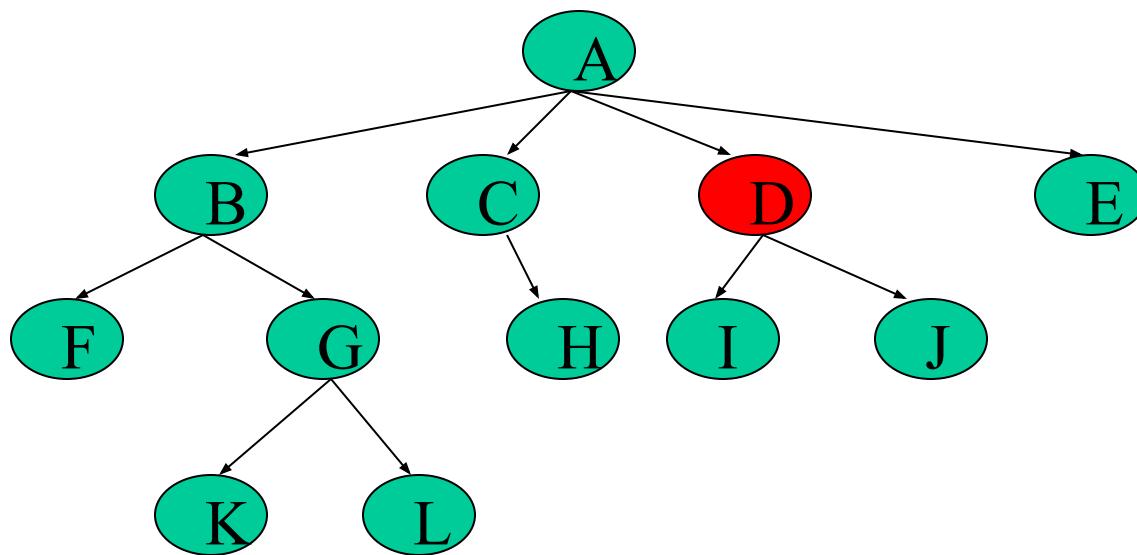
Iterative Deepening Search (IDS)

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



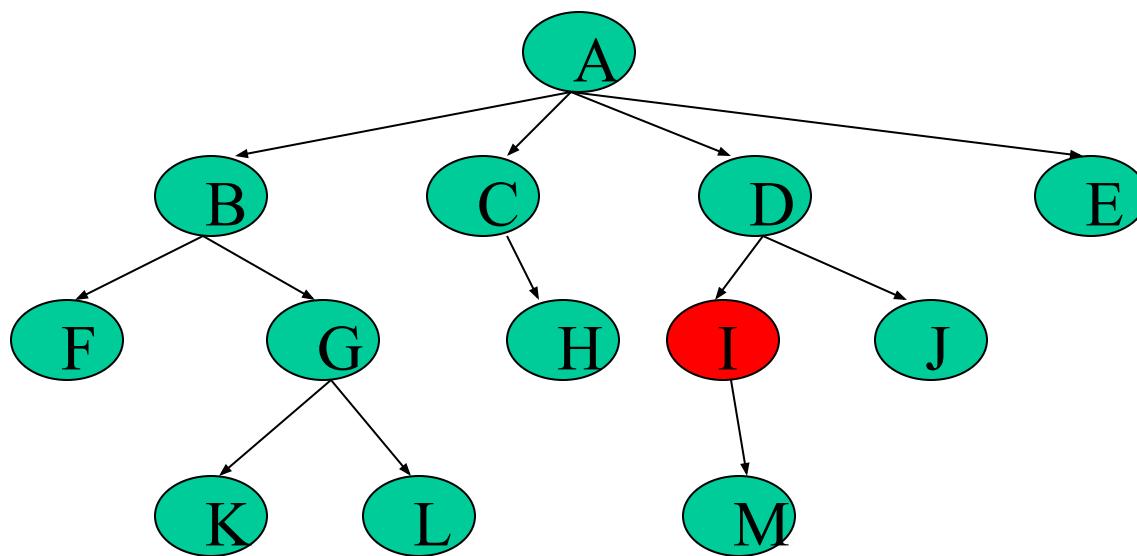
Iterative Deepening Search (IDS)

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



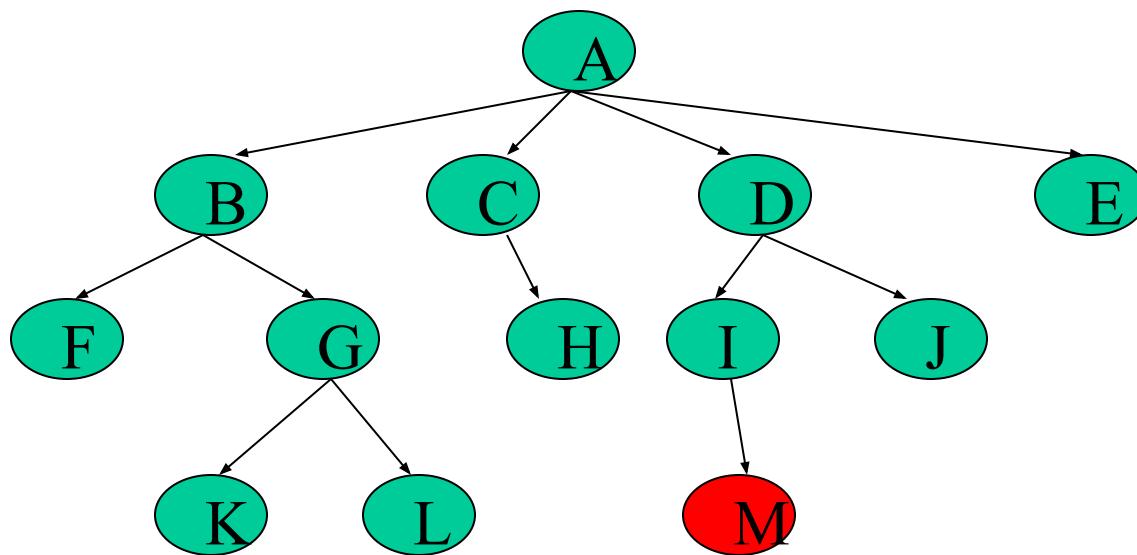
Iterative Deepening Search (IDS)

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



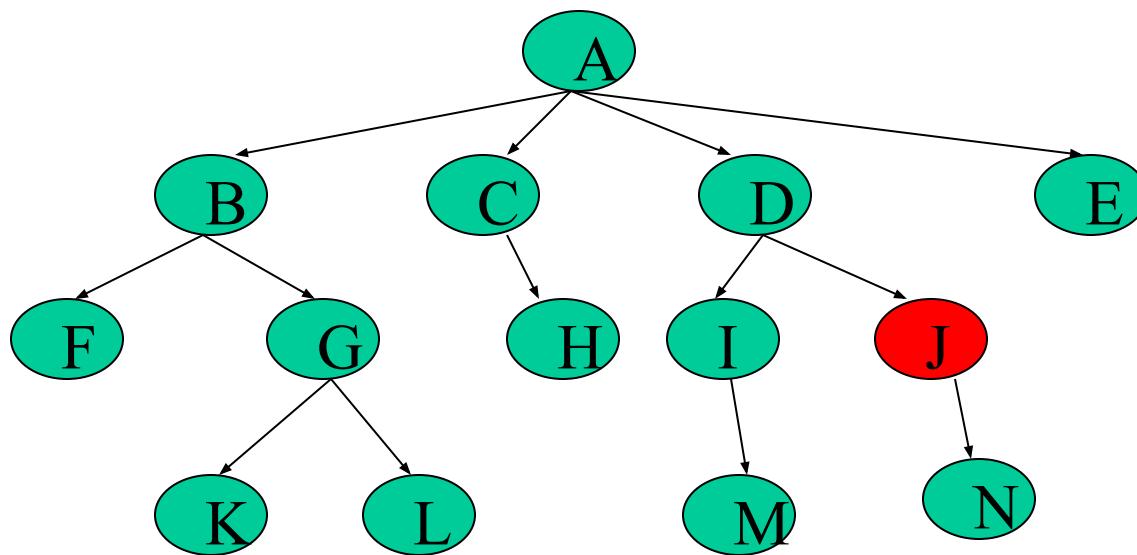
Iterative Deepening Search (IDS)

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



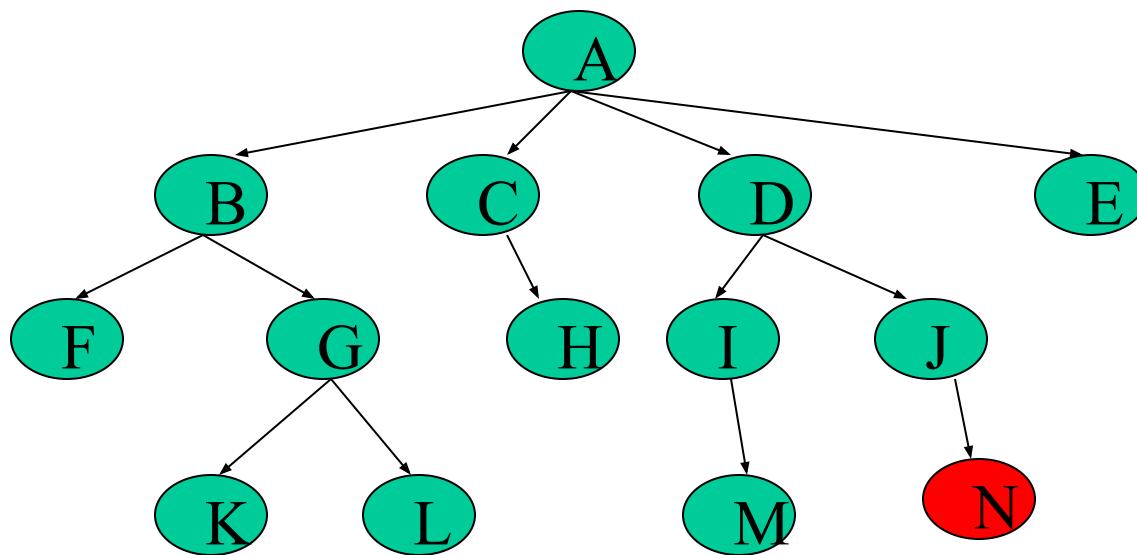
Iterative Deepening Search (IDS)

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



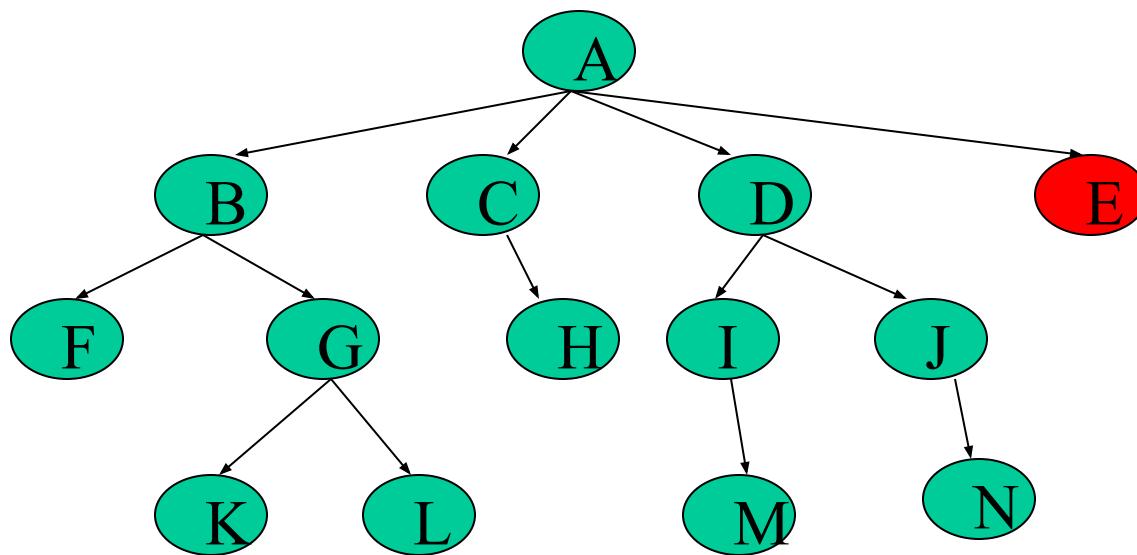
Iterative Deepening Search (IDS)

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



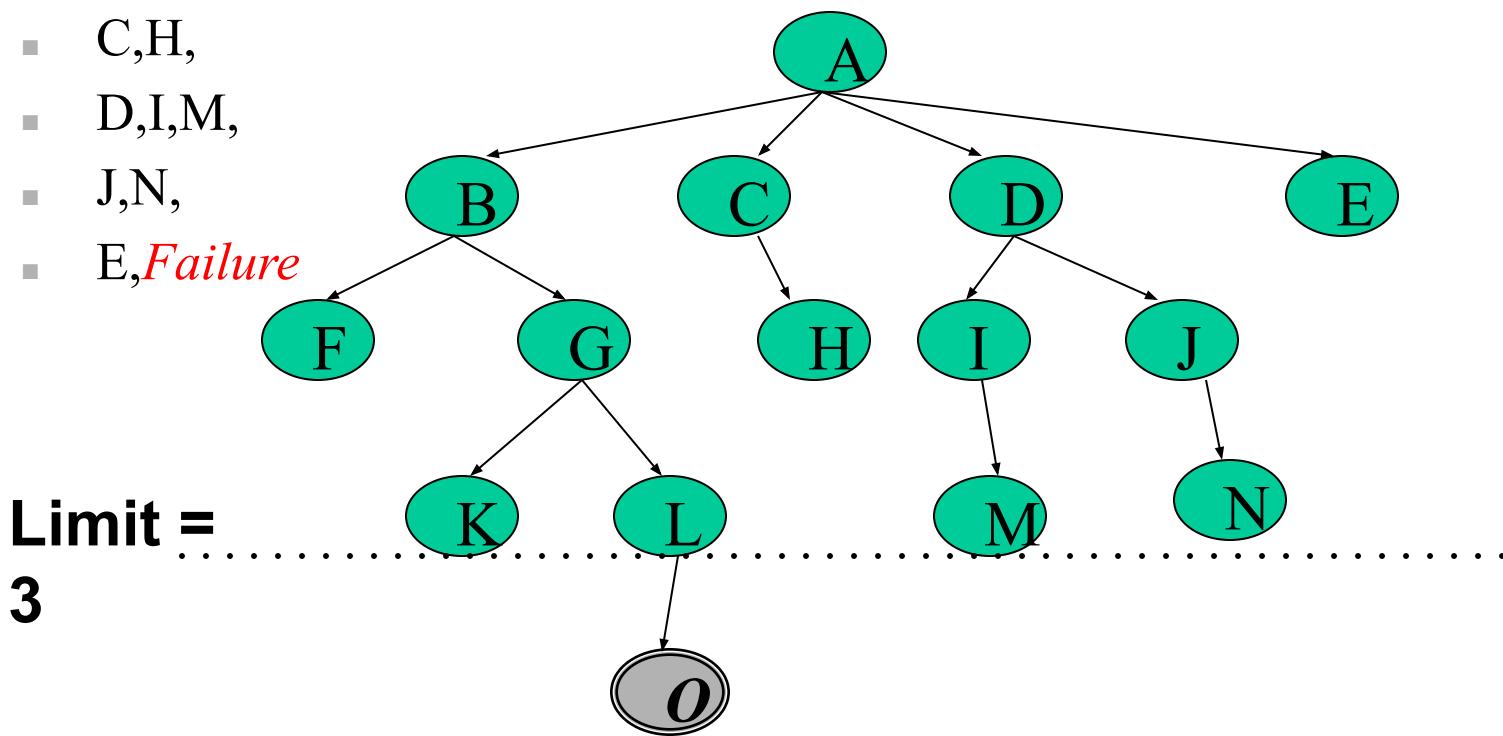
Iterative Deepening Search (IDS)

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



Iterative Deepening Search (IDS)

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

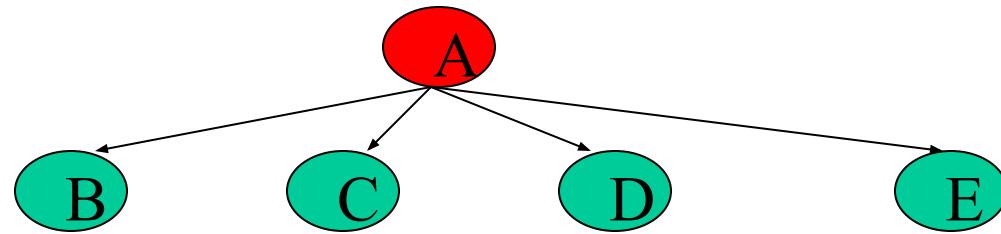


Iterative Deepening Search (IDS)

DLS with bound = 4

Iterative Deepening Search (IDS)

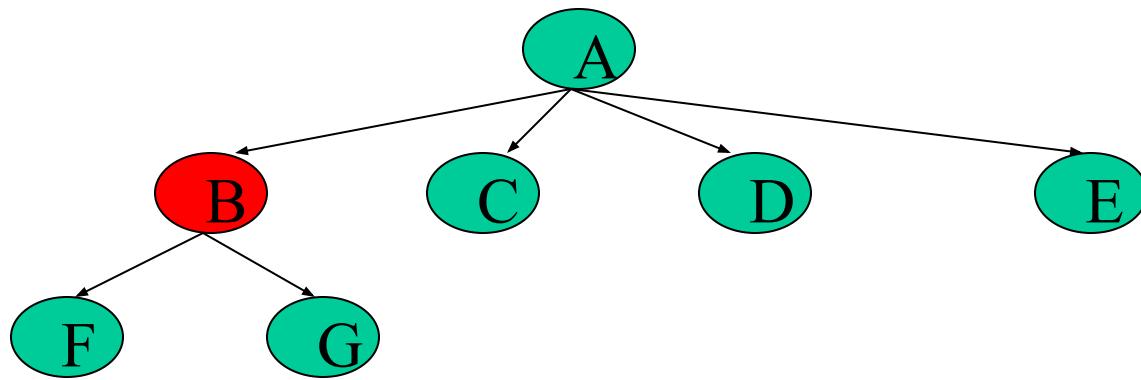
- A,



Limit =
4

Iterative Deepening Search (IDS)

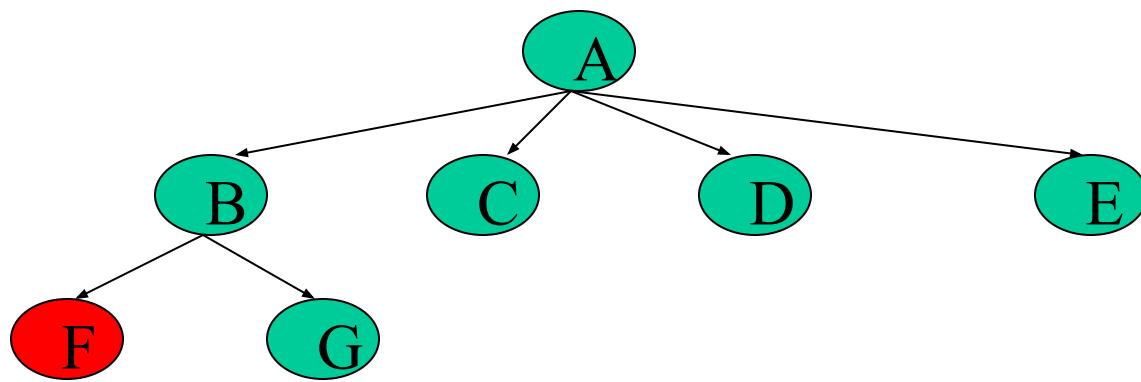
- A,B,



Limit =
4

Iterative Deepening Search (IDS)

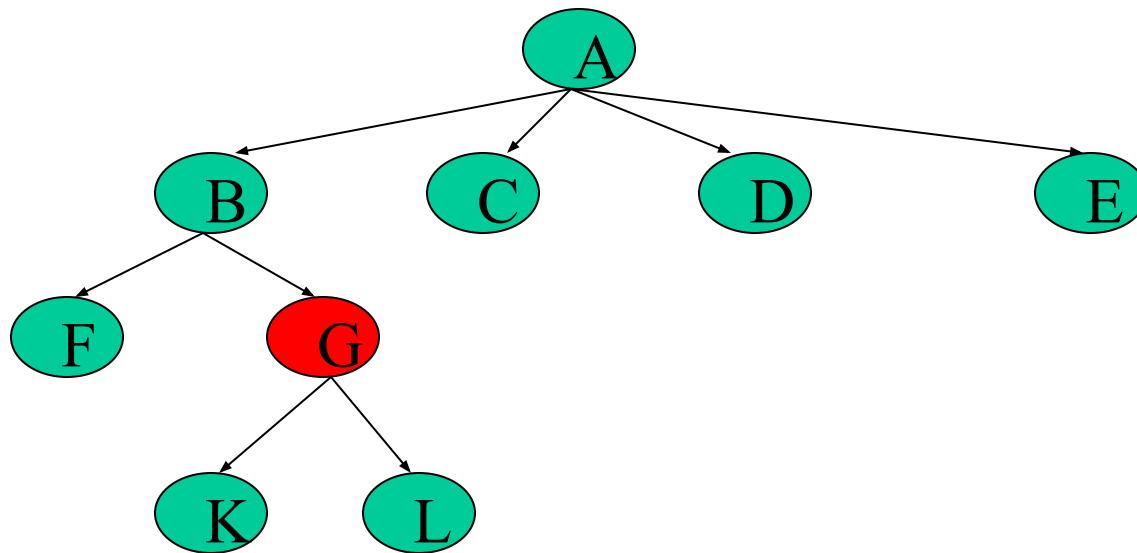
- A,B,F,



Limit =
4

Iterative Deepening Search (IDS)

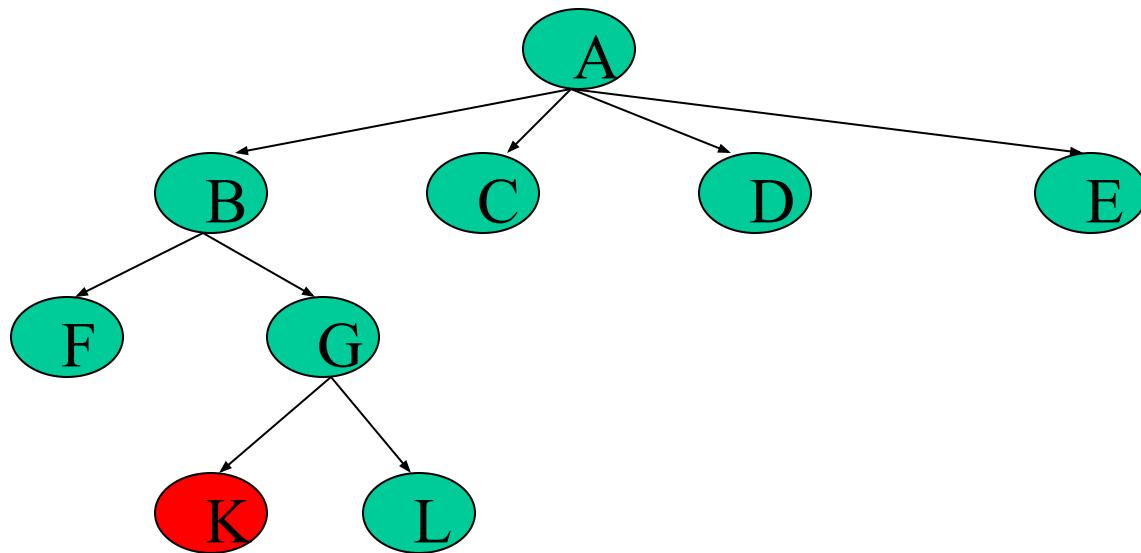
- A,B,F,
- G,



Limit =
4

Iterative Deepening Search (IDS)

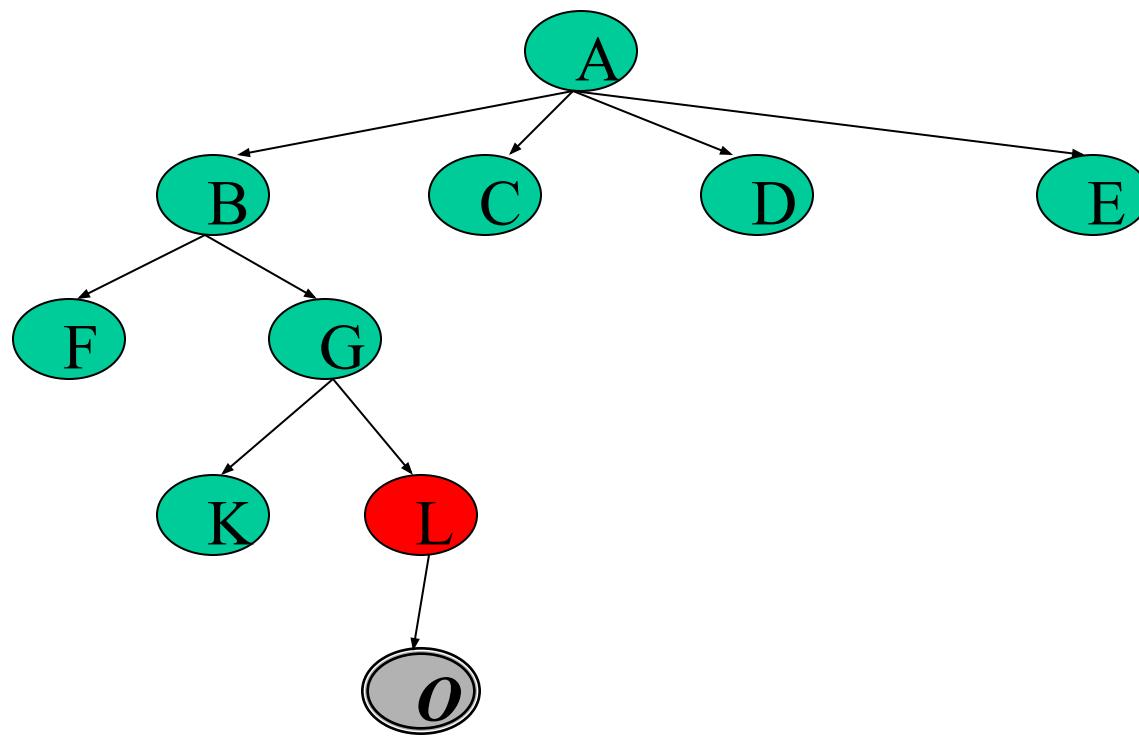
- A,B,F,
- G,K,



Limit =
4

Iterative Deepening Search (IDS)

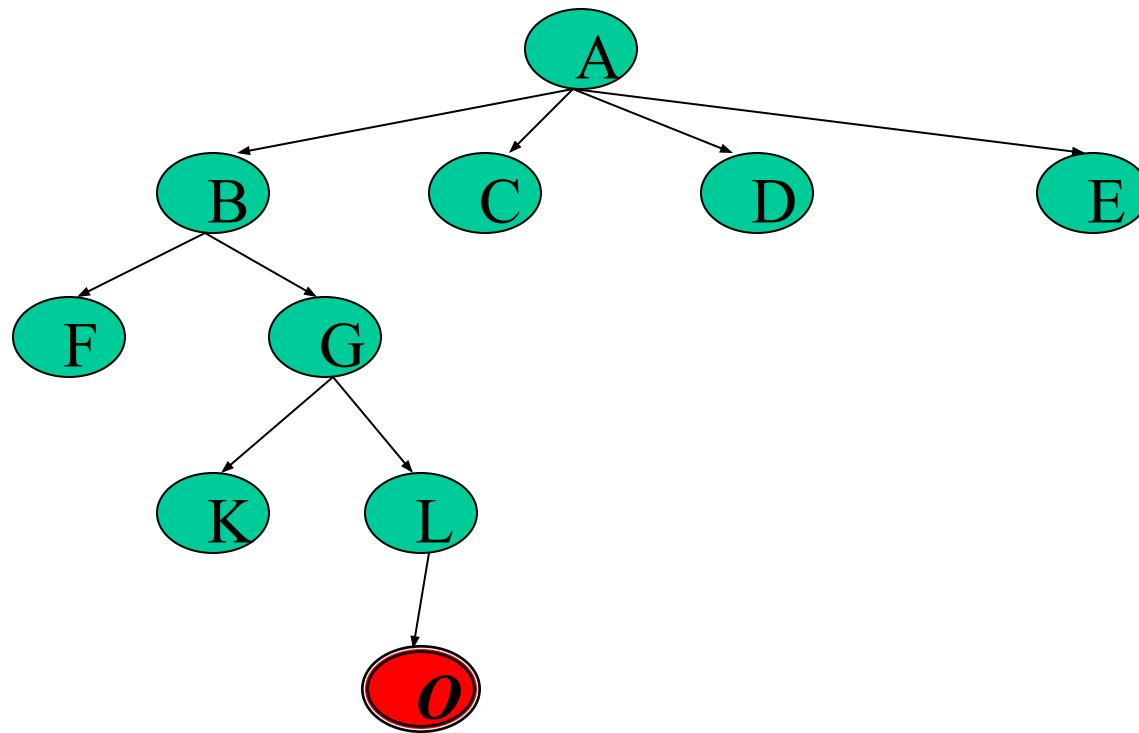
- A,B,F,
- G,K,
- L,



Limit =
4

Iterative Deepening Search (IDS)

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

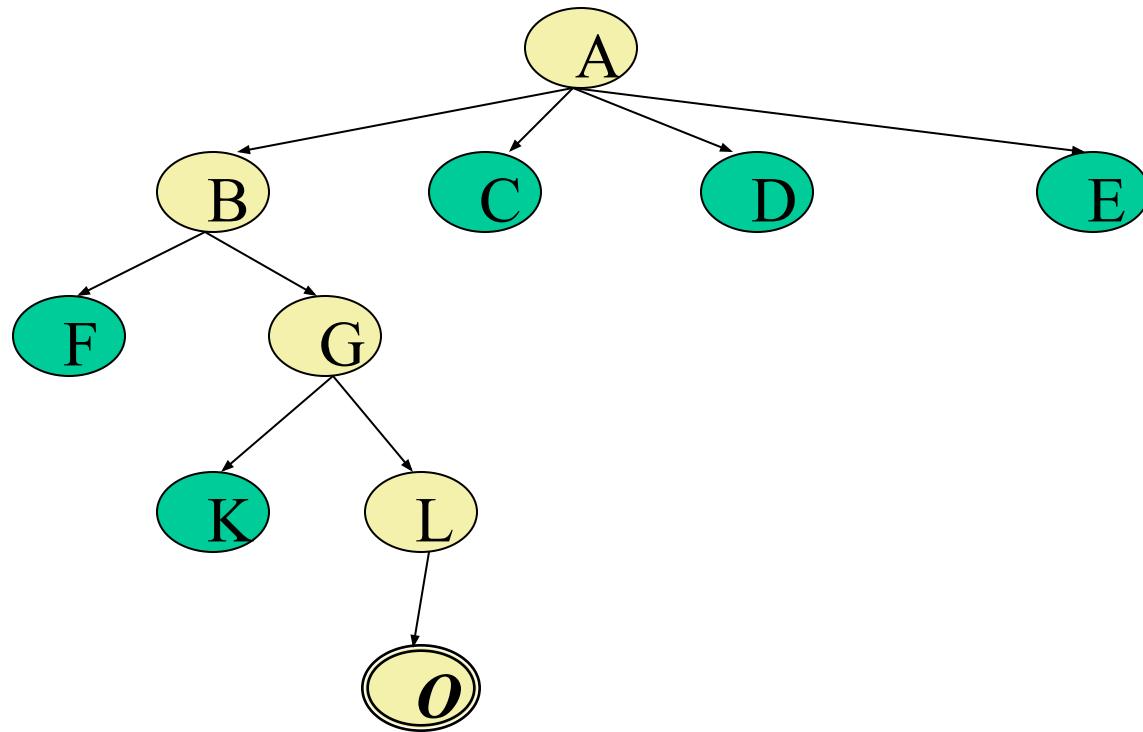


Limit =
4

Iterative Deepening Search (IDS)

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:**

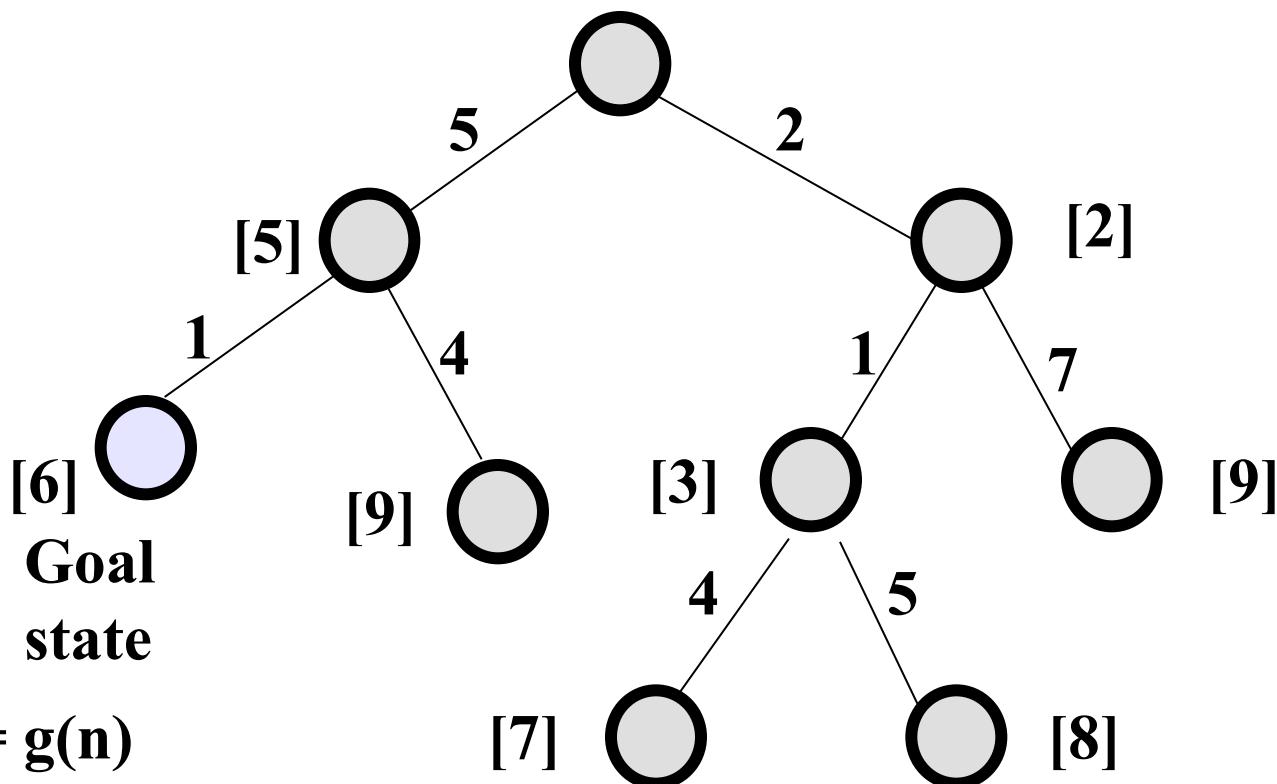
Enqueue nodes in order of cost $g(n)$.

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.

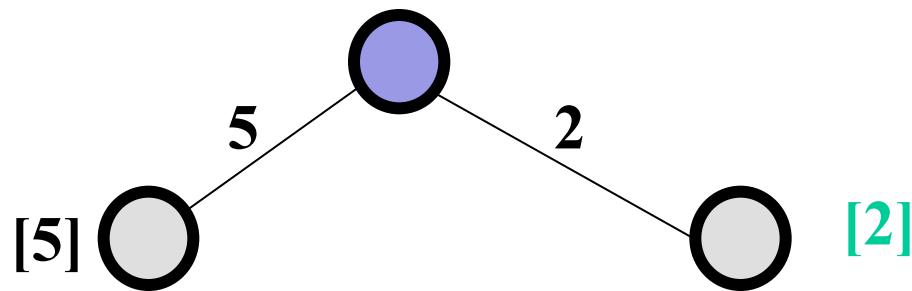
- Complete? Yes.
- Optimal? Yes, if path cost is nondecreasing function of depth
- Time Complexity: $O(b^d)$
- Space Complexity: $O(b^d)$, note that every node in the fringe keep in the queue.

Uniform Cost Search (UCS)

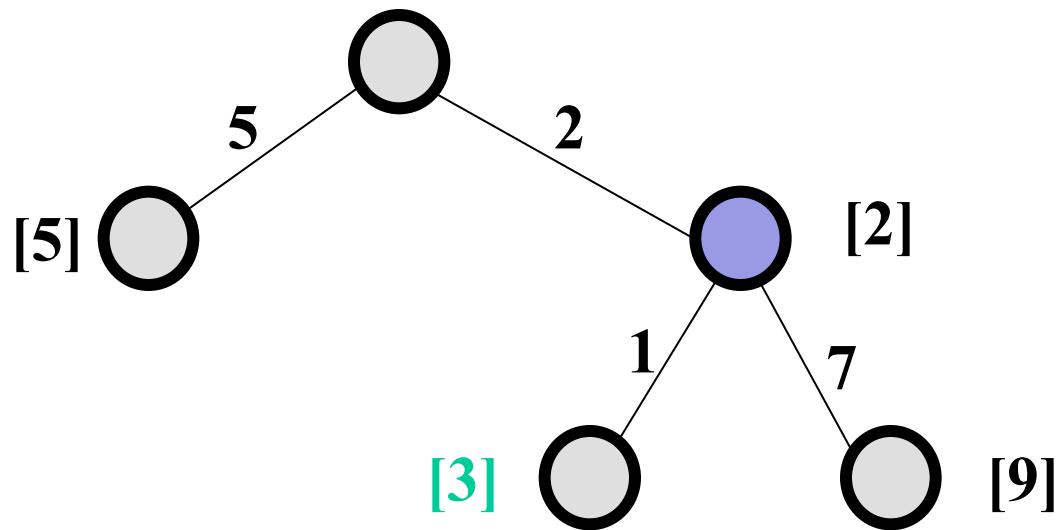


**path cost of node
 n**

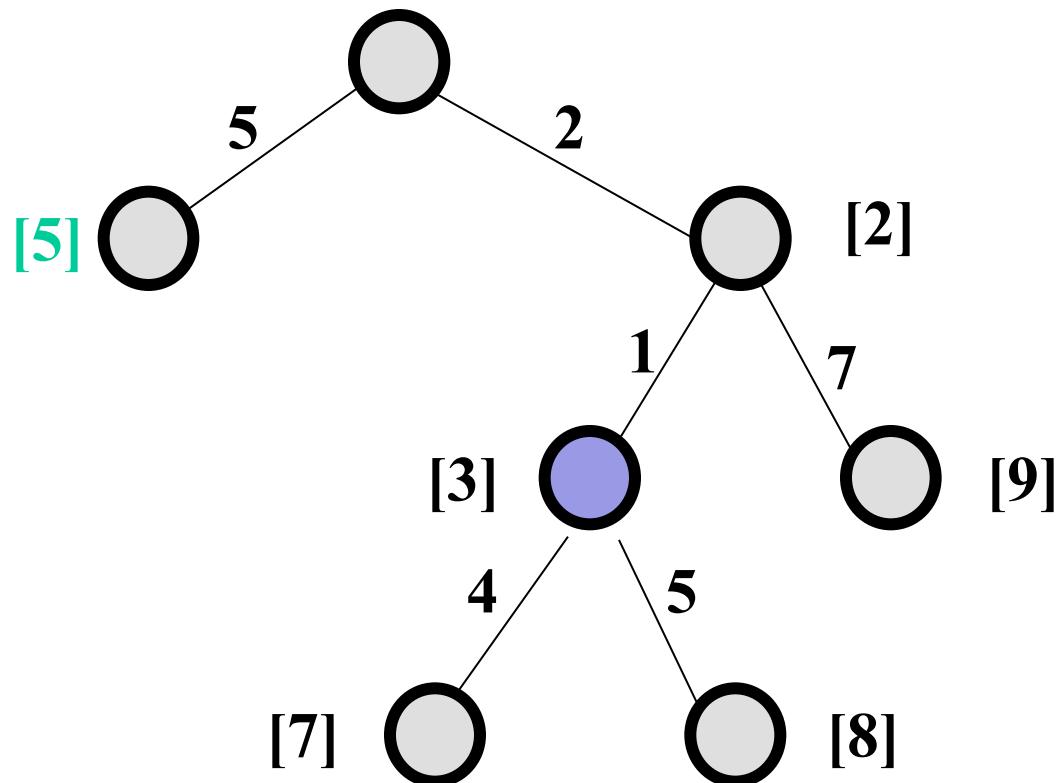
Uniform Cost Search (UCS)



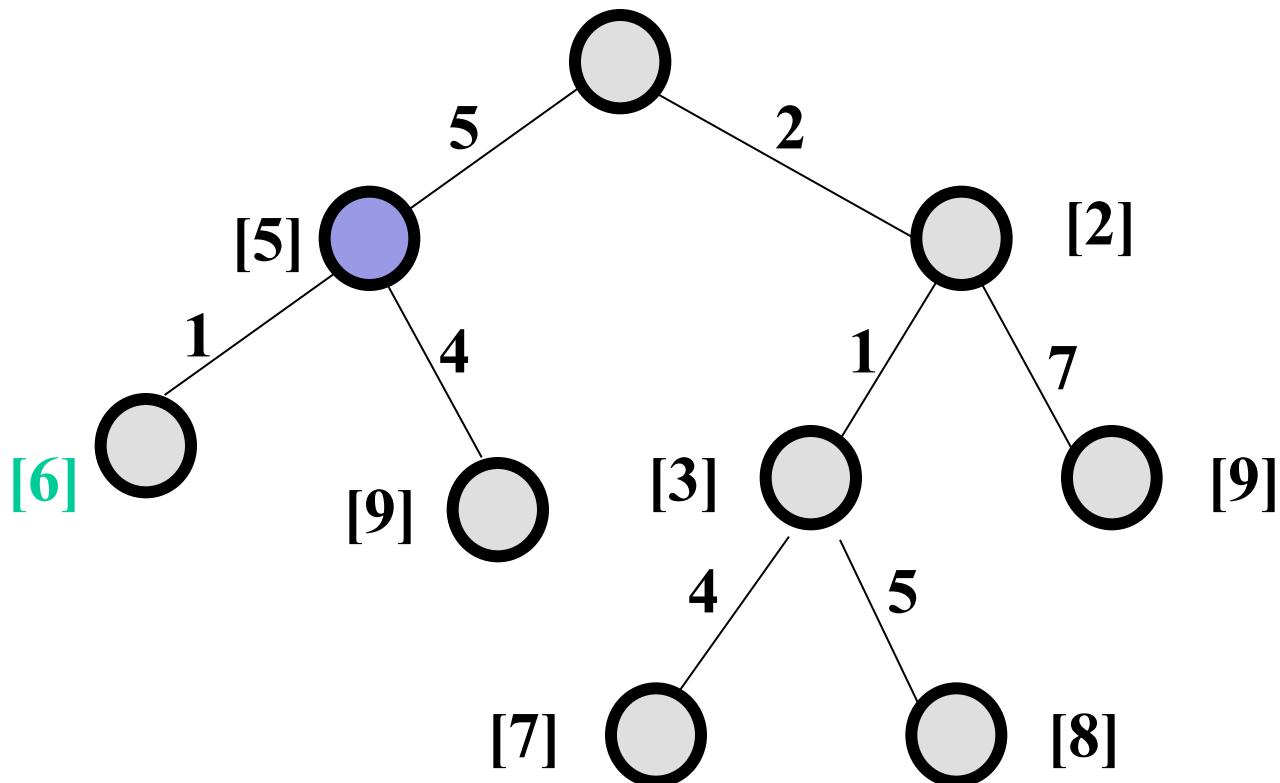
Uniform Cost Search (UCS)



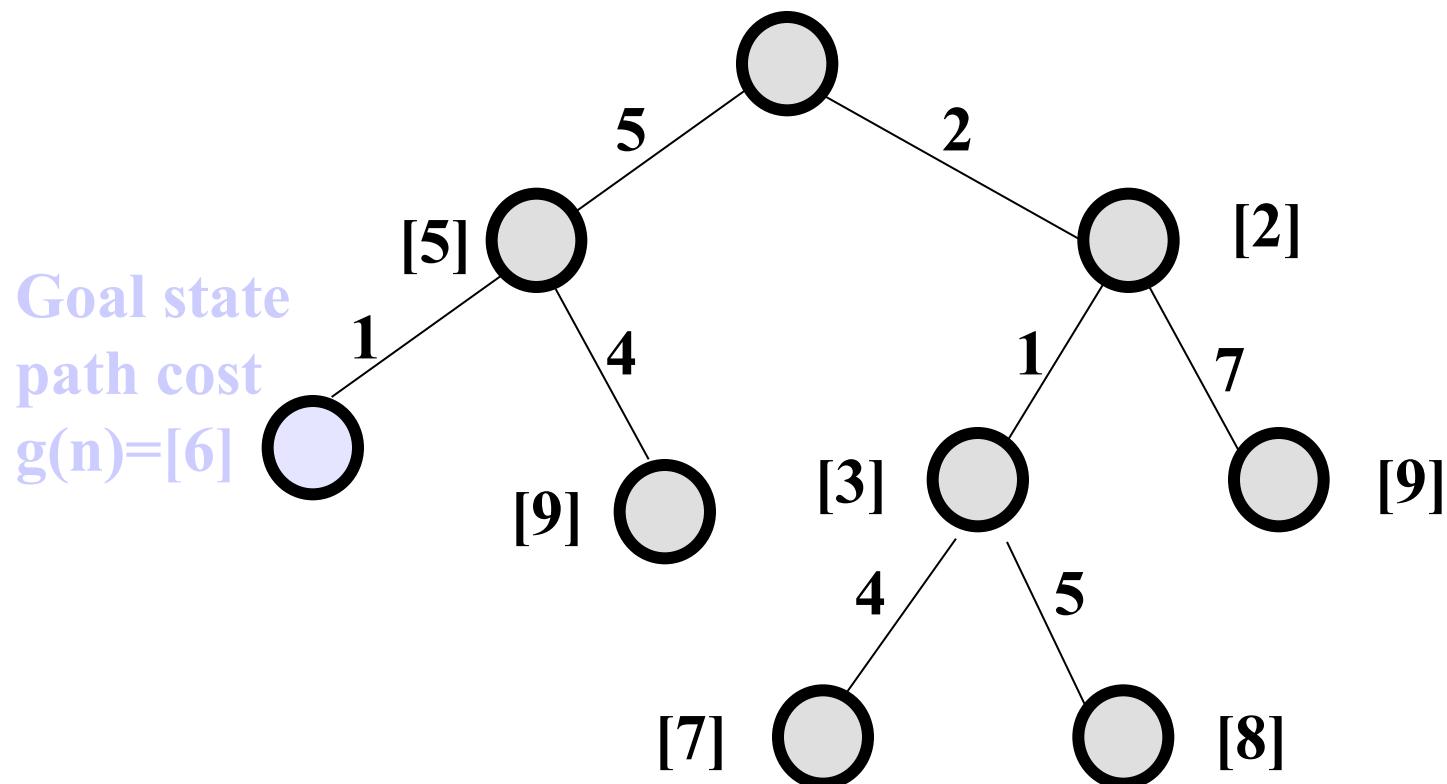
Uniform Cost Search (UCS)



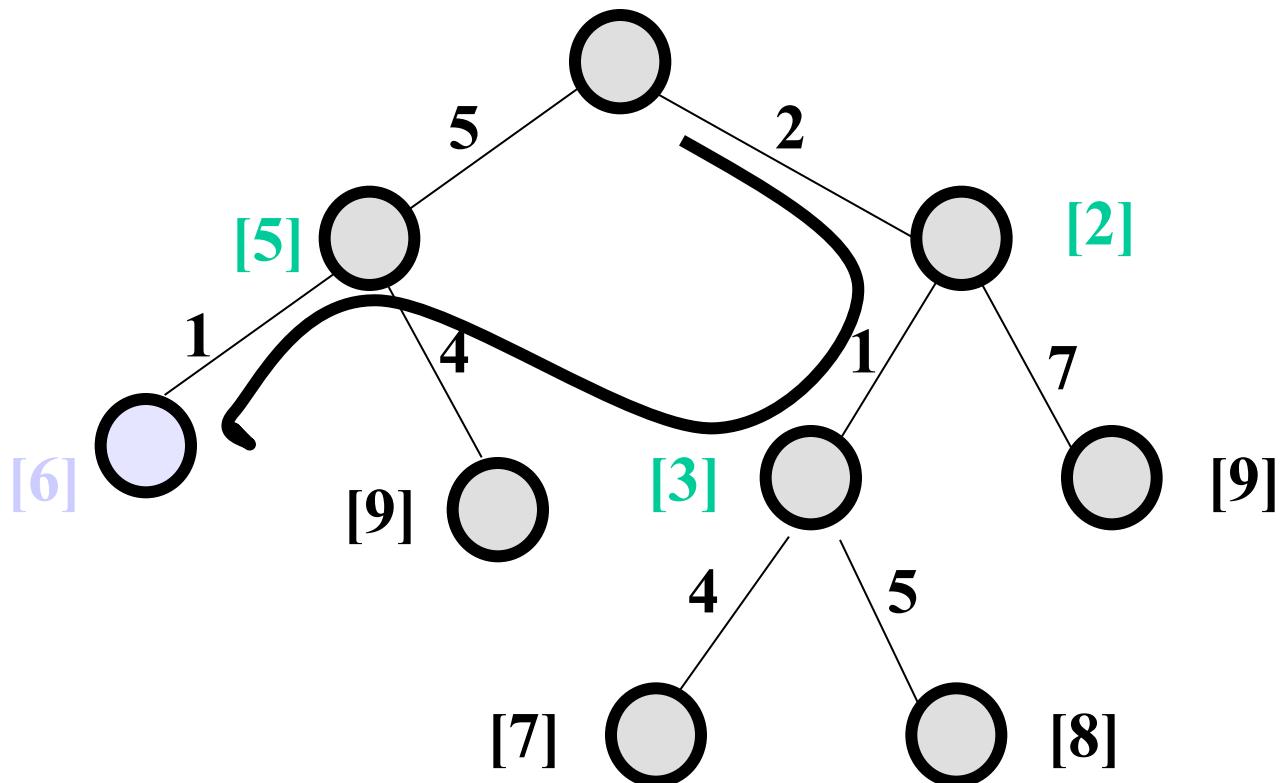
Uniform Cost Search (UCS)



Uniform Cost Search (UCS)



Uniform Cost Search (UCS)



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?

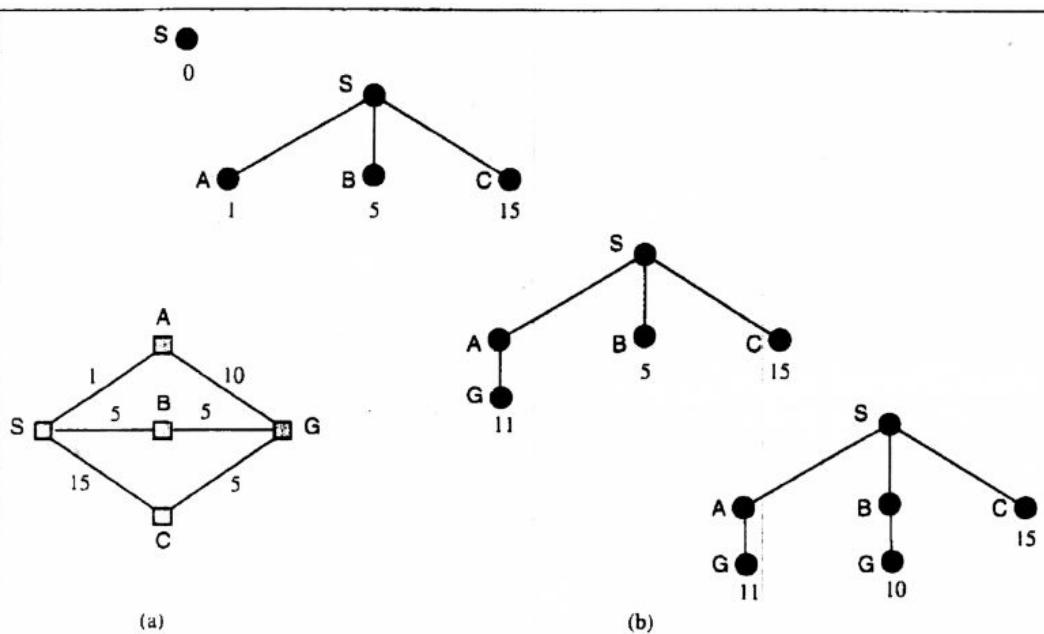
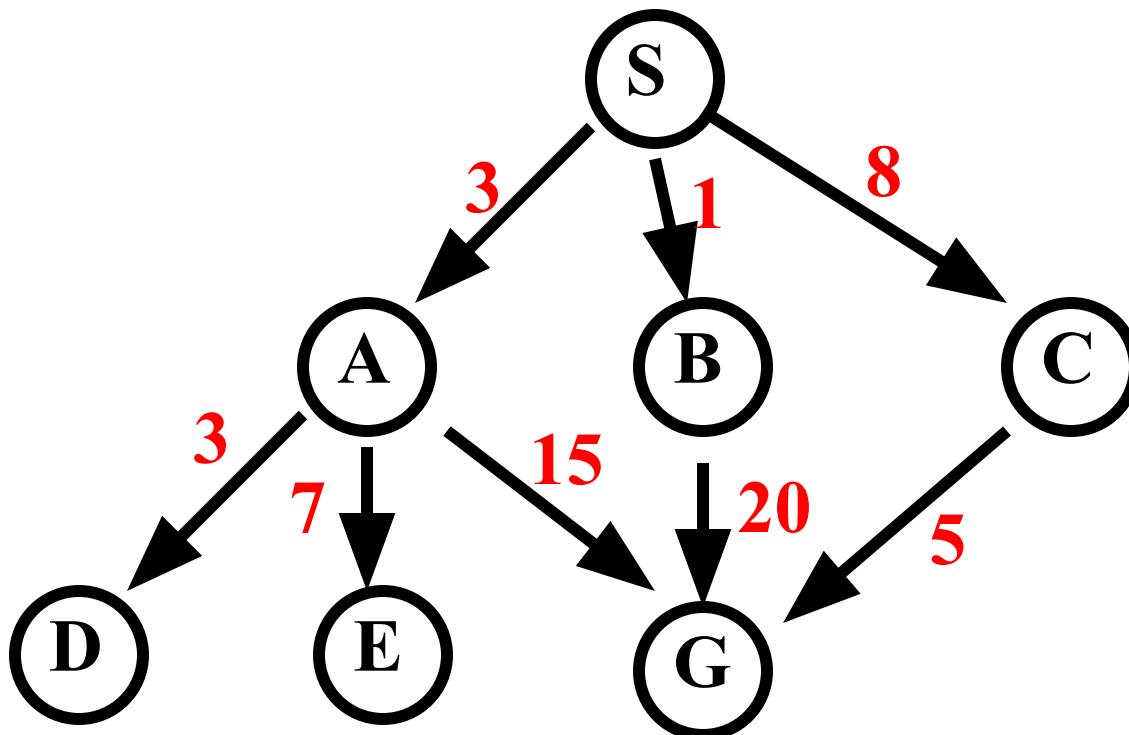


Figure 3.13 A route-finding problem. (a) The state space, showing the cost for each operator. (b) Progression of the search. Each node is labelled with $g(n)$. At the next step, the goal node with $g = 10$ will be selected.

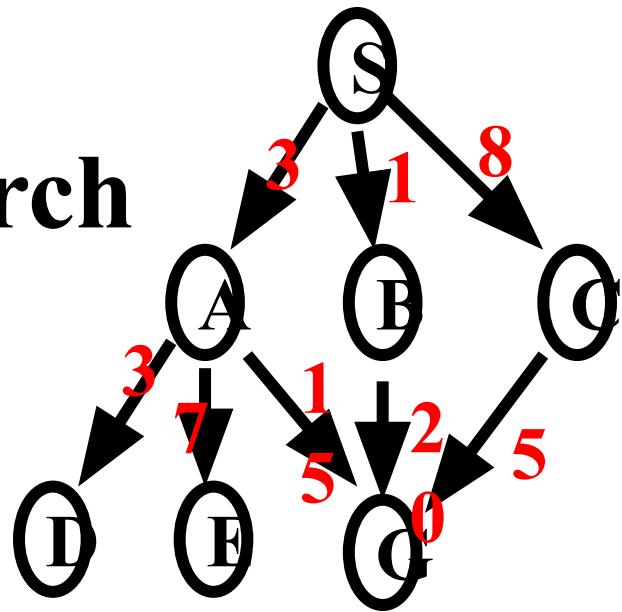
Example for Illustrating Search Strategies



Depth-First Search

Expanded node Nodes list

- { S⁰ }
- S⁰ { A³ B¹ C⁸ }
- A³ { D⁶ E¹⁰ G¹⁸ B¹ C⁸ }
- D⁶ { E¹⁰ G¹⁸ B¹ C⁸ }
- E¹⁰ { G¹⁸ B¹ C⁸ }
- G¹⁸ { B¹ C⁸ }



Solution path found is S A G, cost 18

Number of nodes expanded (including goal node) = 5

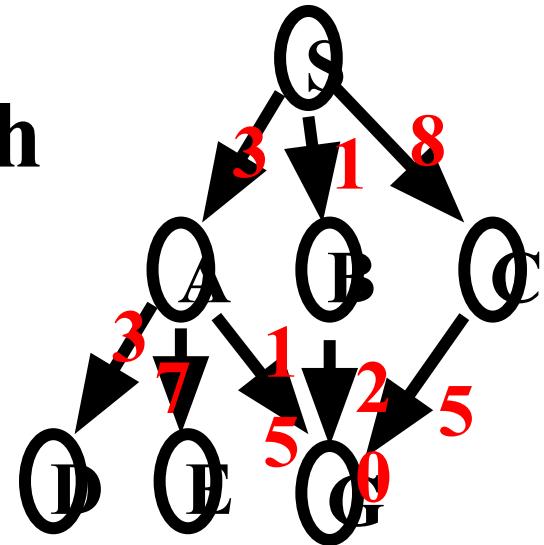
Breadth-First Search

Expanded node Nodes list

- { S⁰ }
- S⁰ { A³ B¹ C⁸ }
- A³ { B¹ C⁸ D⁶ E¹⁰ G¹⁸ }
- B¹ { C⁸ D⁶ E¹⁰ G¹⁸ G²¹ }
- C⁸ { D⁶ E¹⁰ G¹⁸ G²¹ G¹³ }
- D⁶ { E¹⁰ G¹⁸ G²¹ G¹³ }
- E¹⁰ { G¹⁸ G²¹ G¹³ }
- G¹⁸ { G²¹ G¹³ }

Solution path found is S A G , cost 18

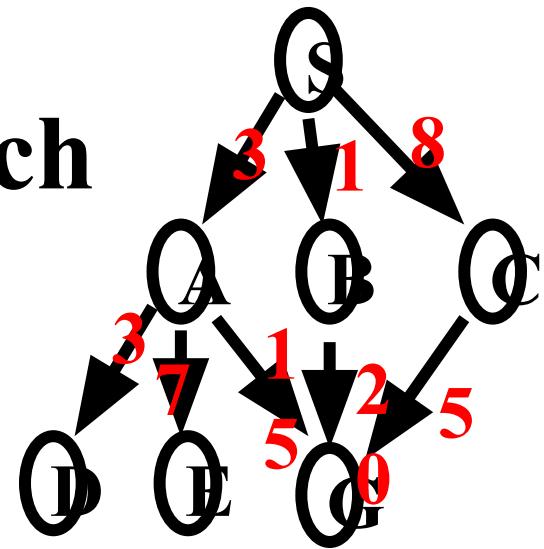
Number of nodes expanded (including goal node) = 7



Uniform-Cost Search

Expanded node Nodes list

- { S⁰ }
- S⁰ { B¹ A³ C⁸ }
- B¹ { A³ C⁸ G²¹ }
- A³ { D⁶ C⁸ E¹⁰ G¹⁸ }
- D⁶ { C⁸ E¹⁰ G¹⁸ }
- C⁸ { E¹⁰ G¹³ }
- E¹⁰ { G¹³ }
- G¹³ {}



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.

Uniform-cost search

Implementation: $\text{fringe} = \text{queue ordered by path cost}$
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 $\text{path cost} \leq \text{cost of optimal solution.}$

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

Optimal? Yes, for any step cost.

Bi-Directional Search

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

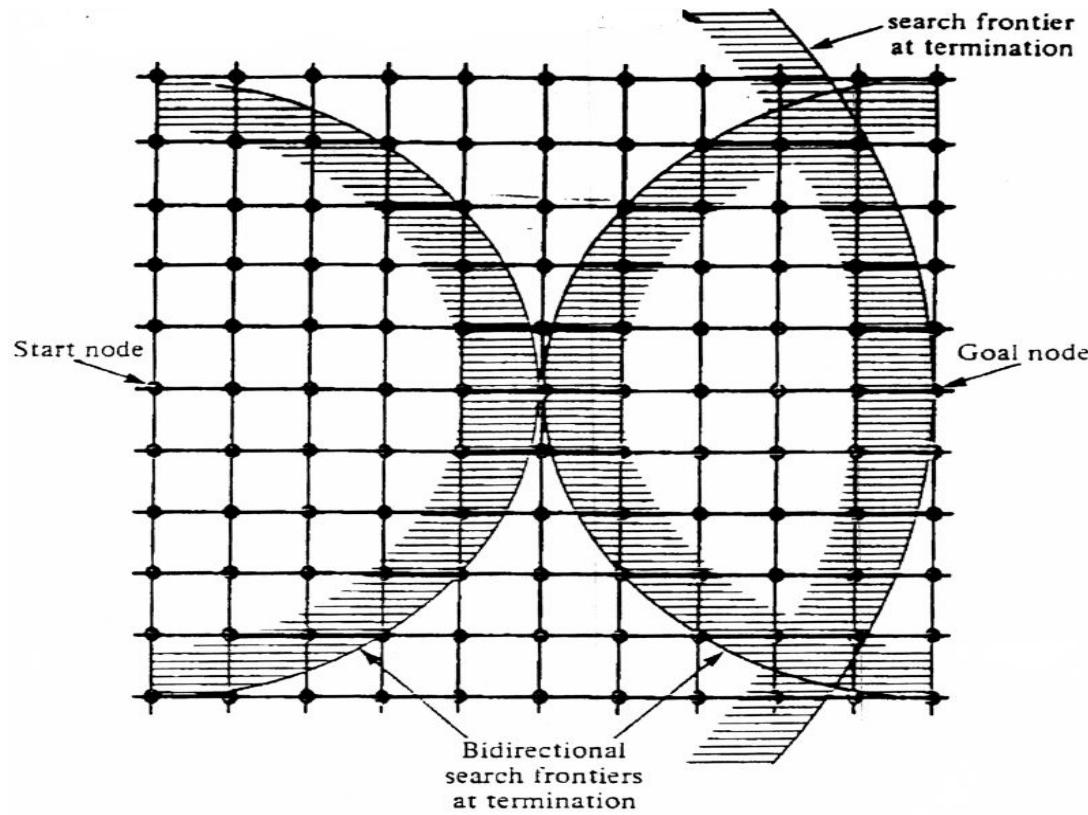


Fig. 2.10 Bidirectional and unidirectional breadth-first searches.

Summary of 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