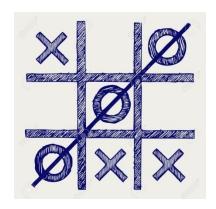
# CSC 422 Class #3

**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.

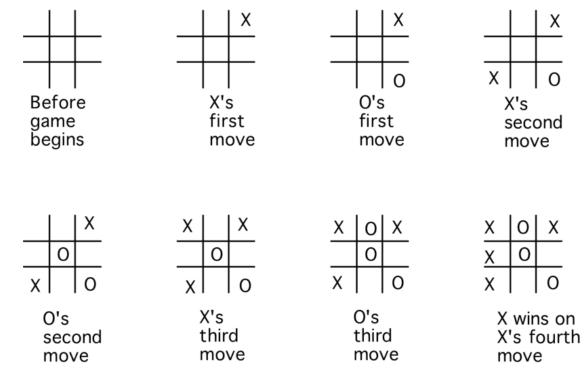


Fig: Some states of Tic-Tac-Toe board.

# Why searching in AI?

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

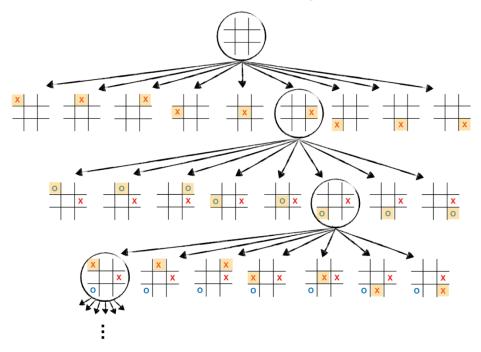
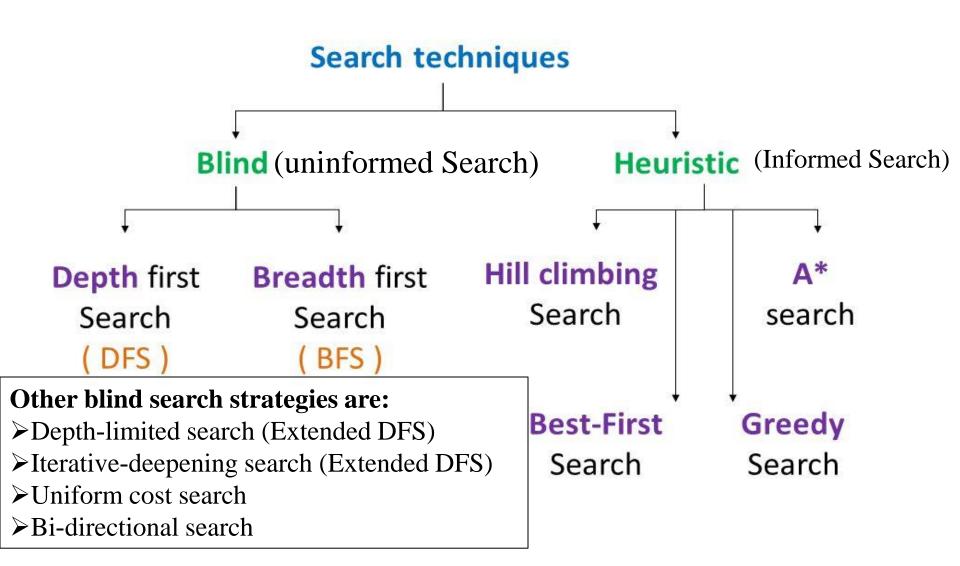


Fig: Game search tree for Tic-Tac-Toe board (partial view)

## **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:**

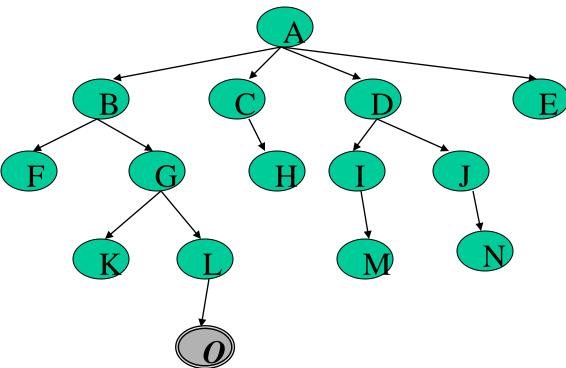
**optimization** in which the search is to find an optimal value of an objective function: hill climbing, simulated annealing, genetic algorithms, Ant Colony Optimization

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

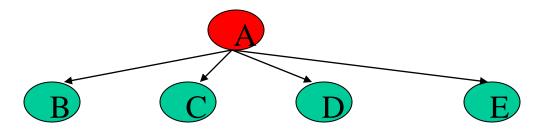
## **Uninformed Search**

#### Application1:

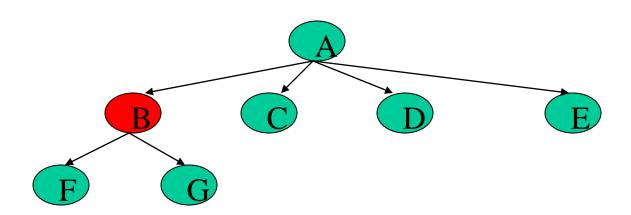
Given the following state space (tree search), give the sequence of visited nodes when using BFS (assume that the node *O* is the goal state):



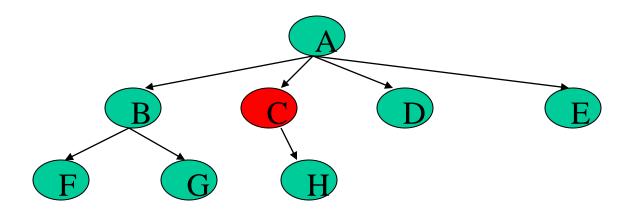
A,



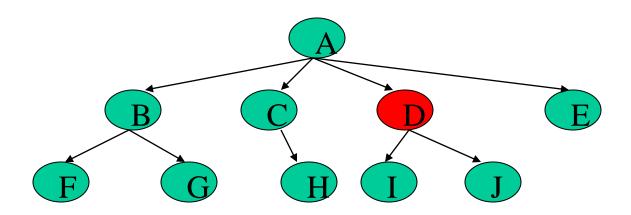
- A,
- **B**,



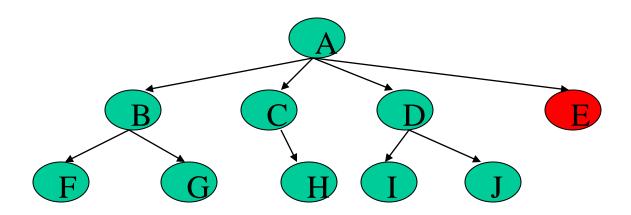
- A,
- **■** B,C



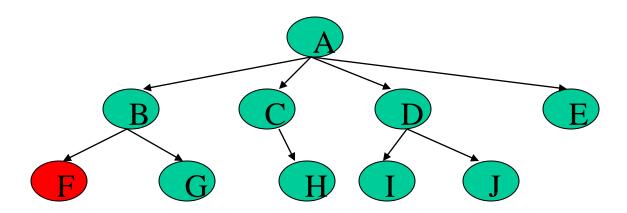
- A,
- B,C,D



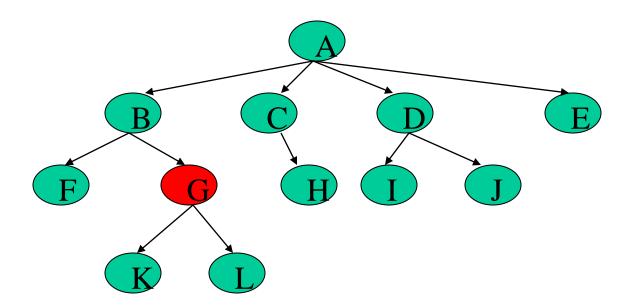
- A,
- B,C,D,E



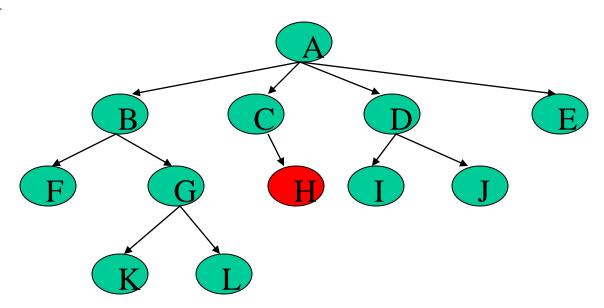
- A,
- B,C,D,E,
- **F**,



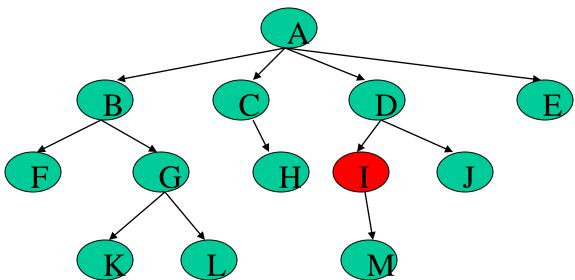
- A,
- **■** B,C,D,E,
- F,G



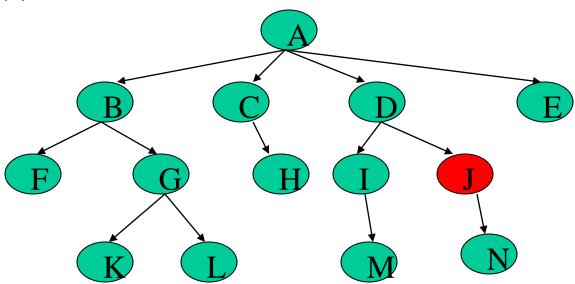
- A,
- **■** B,C,D,E,
- F,G,H



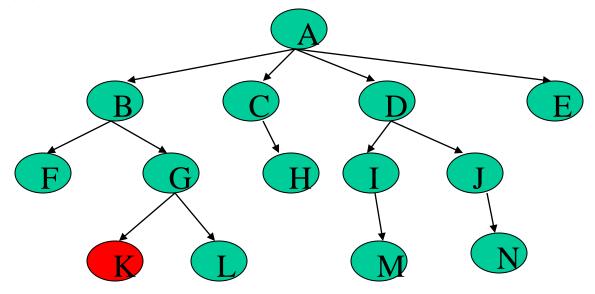
- A,
- **■** B,C,D,E,
- F,G,H,I



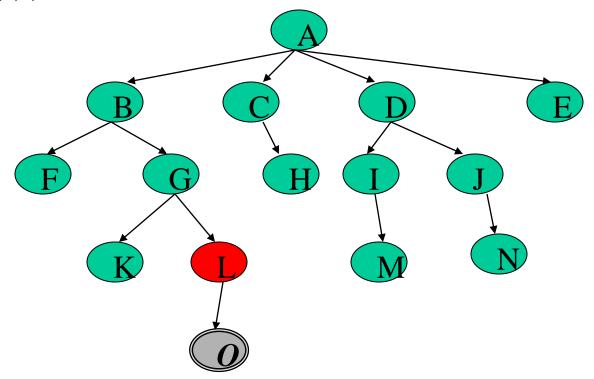
- A,
- **■** B,C,D,E,
- F,G,H,I,J,



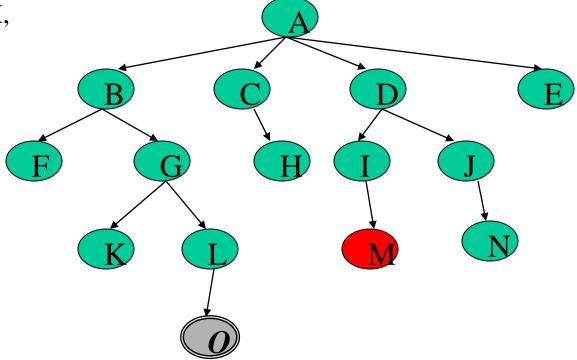
- A,
- **■** B,C,D,E,
- F,G,H,I,J,
- **K**,



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



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

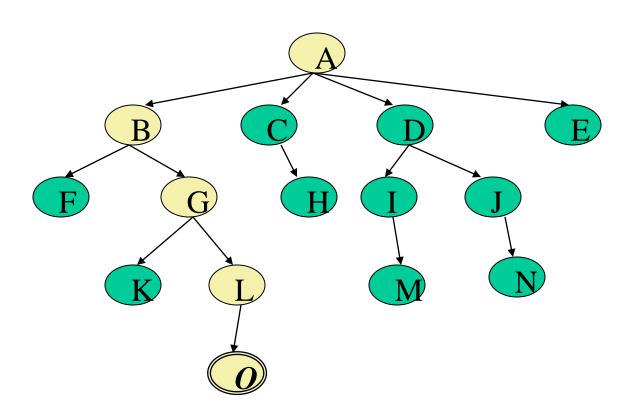


- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L, M,N,

  B
  C
  D
  E

- A,
- B,C,D,E,
- F,G,H,I,J,
- K,L, M,N,
  Goal state: 0
  F
  G
  H
  I
  N

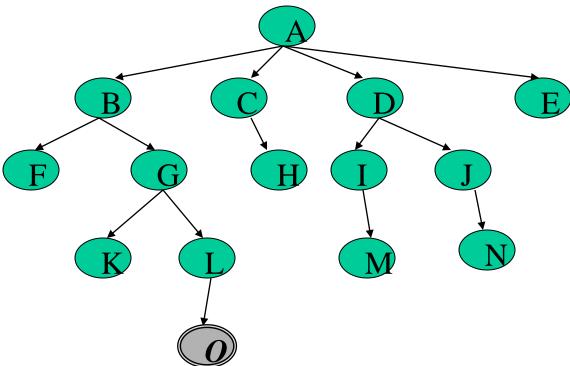
■ The returned solution is the sequence of operators in the path:



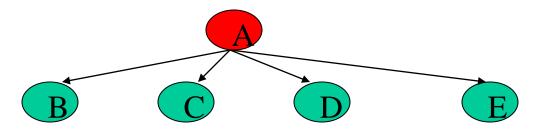
# **Depth First Search (DFS)**

#### Application2:

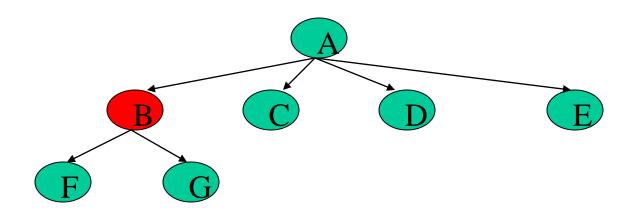
Given the following state space (tree search), give the sequence of visited nodes when using DFS (assume that the node *O* is the goal state):



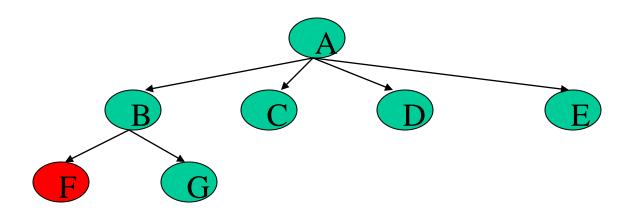
A,



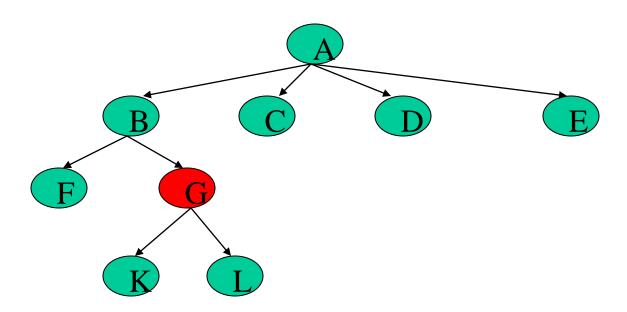
**A,B**,



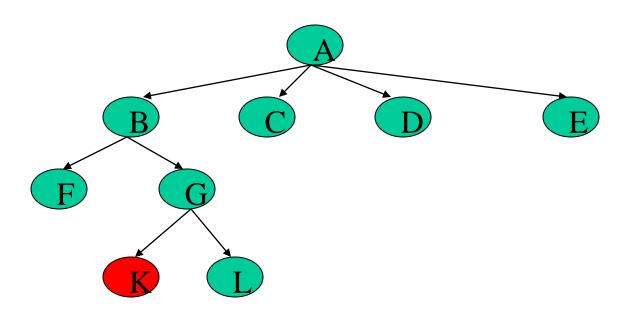
A,B,F,



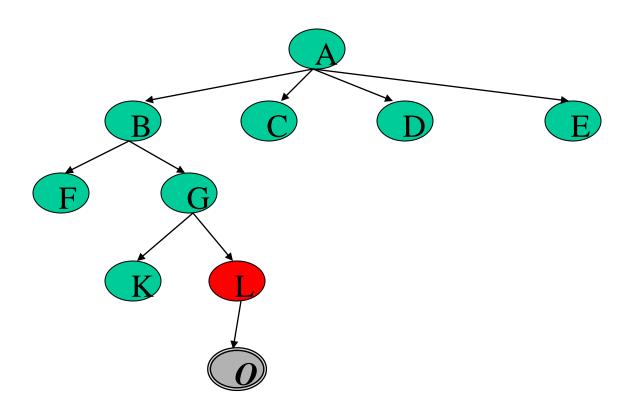
- A,B,F,
- **G**,



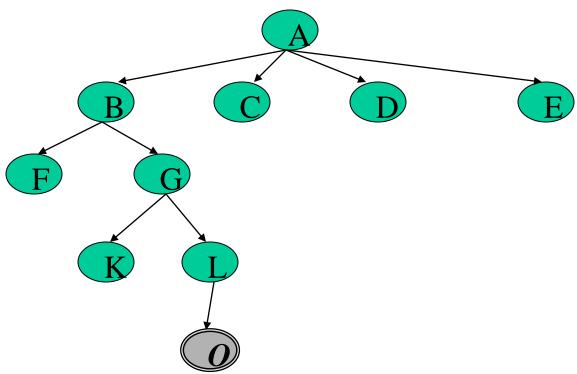
- A,B,F,
- G,K,



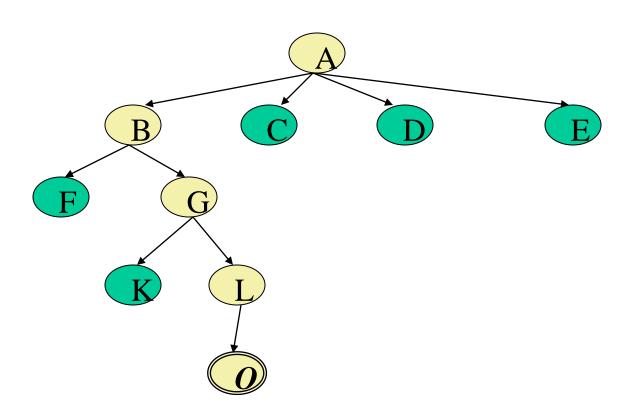
- A,B,F,
- G,K,
- L,



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



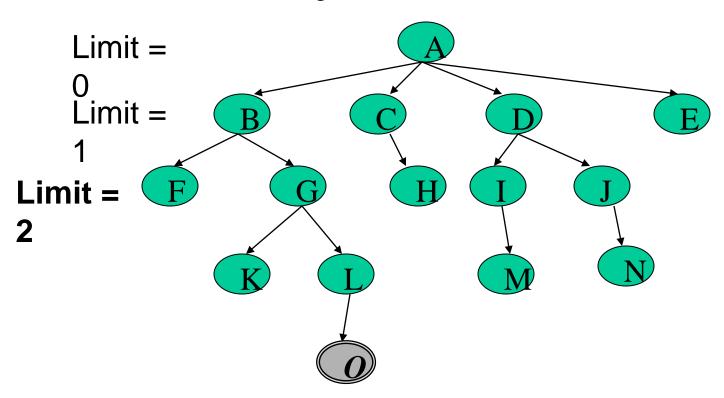
The returned solution is the sequence of operators in the path: A, B, G, L, O



# **Depth-Limited Search (DLS)**

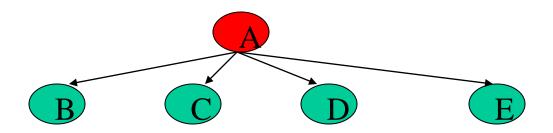
#### Application3:

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



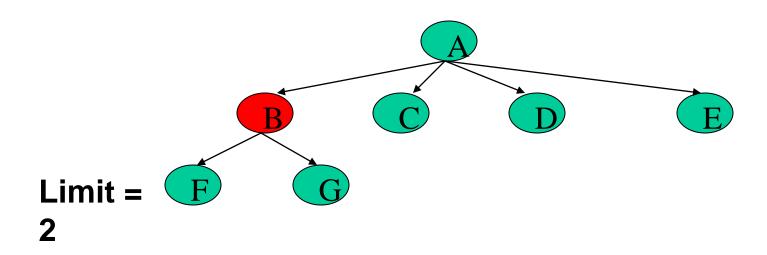
# **Depth-Limited Search (DLS)**

A,

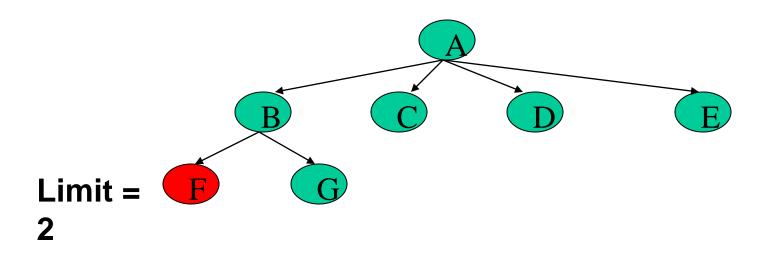


# **Depth-Limited Search (DLS)**

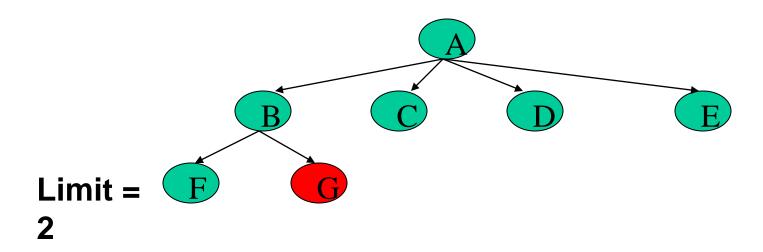
**A,B**,



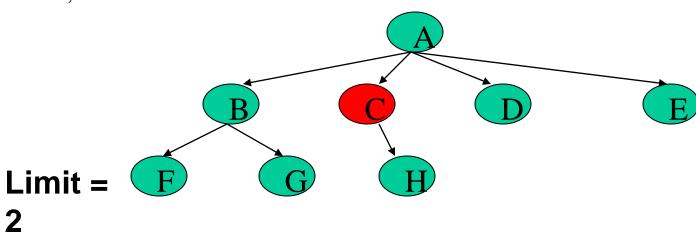
A,B,F,



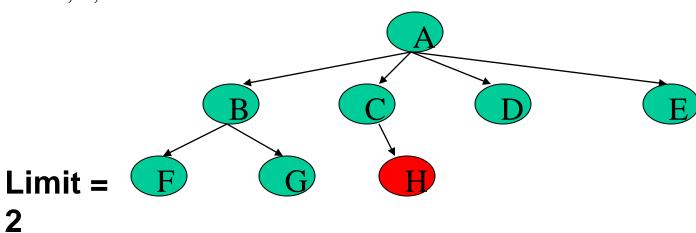
- A,B,F,
- **G**,

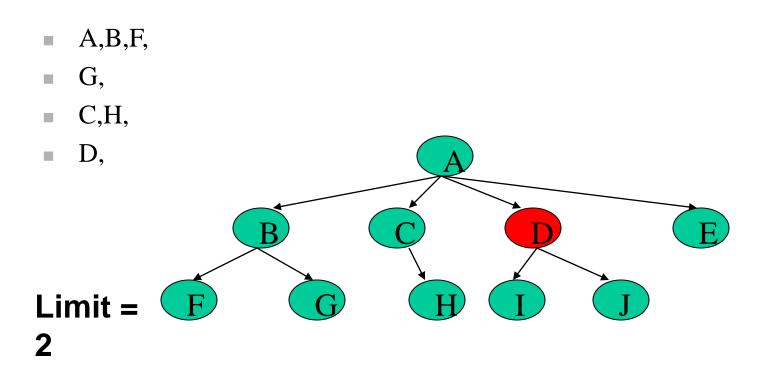


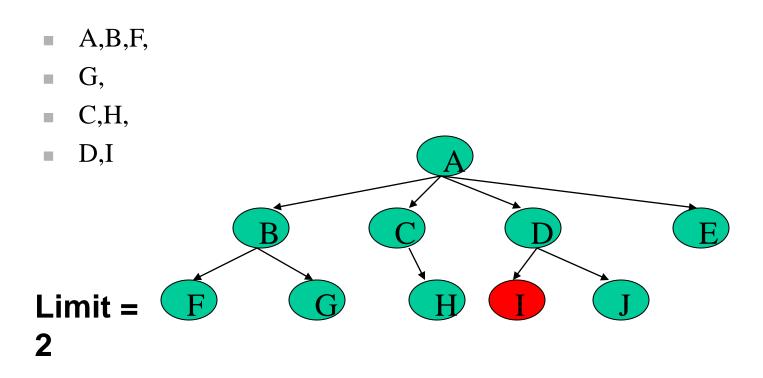
- A,B,F,
- **G**,
- C.

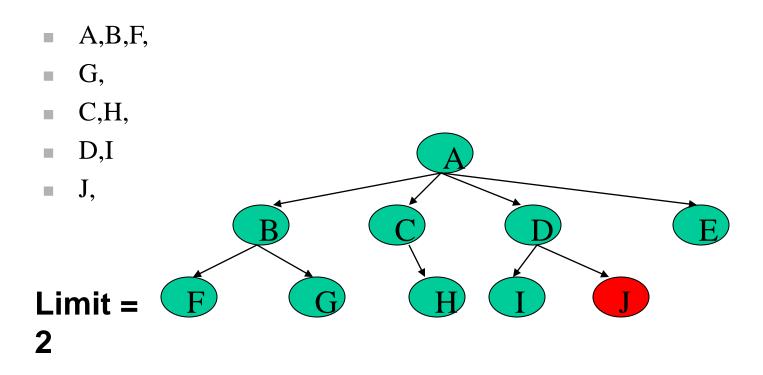


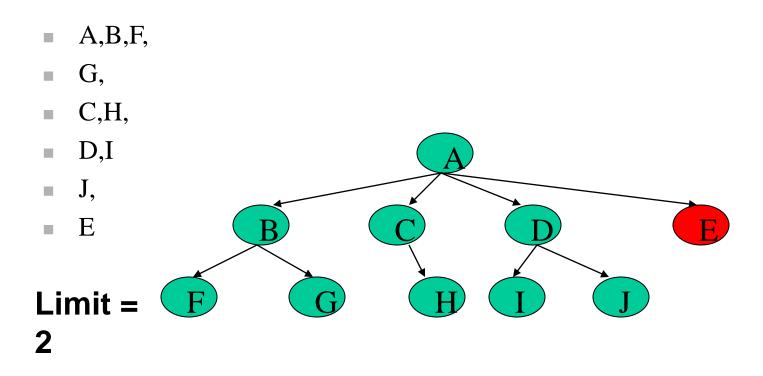
- A,B,F,
- **G**,
- **C,H**,

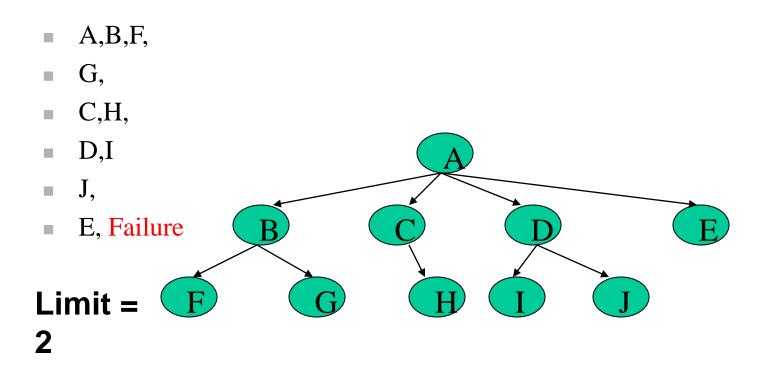




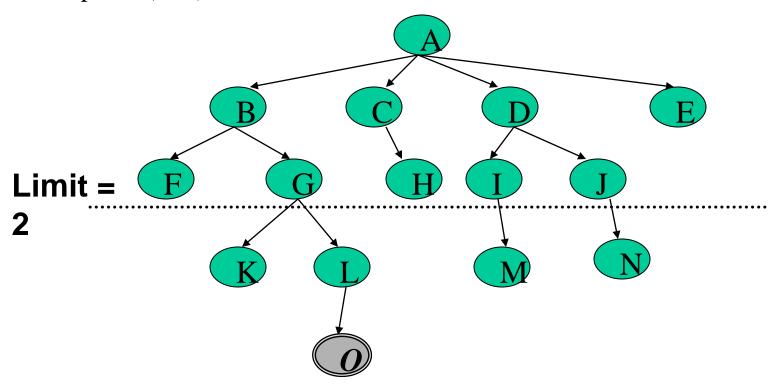






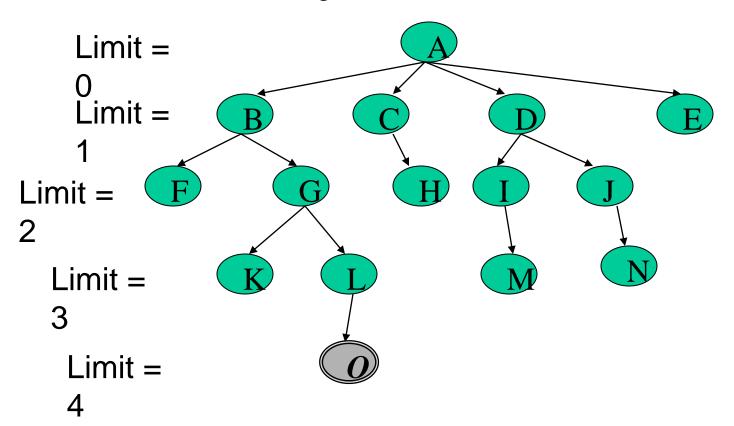


- DLS algorithm returns Failure (no solution)
- The reason is that the goal is beyond the limit (Limit =2): the goal depth is (d=4)



#### Application4:

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



DLS with bound = 0

A,

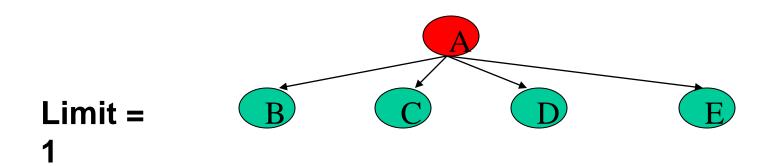


A, Failure

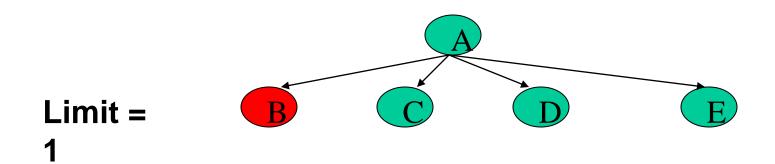


DLS with bound = 1

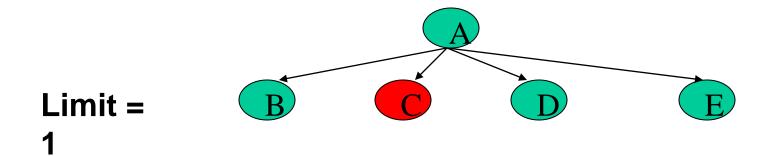
**A**,



A,B,



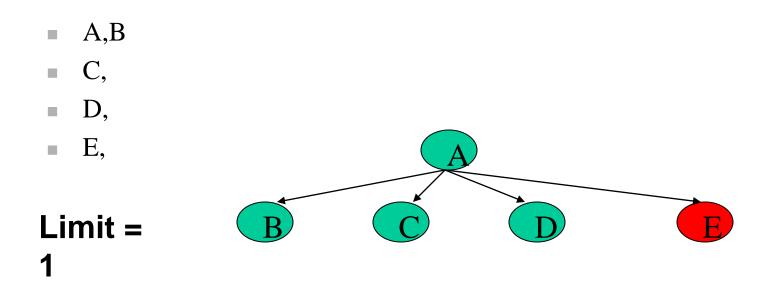
- A,B,
- C,



A,B,
 C,
 D,

Limit =

 B

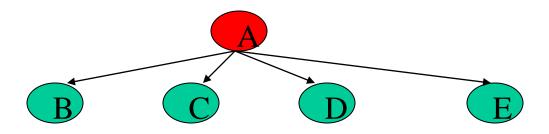


A,B,
 C,
 D,
 E, Failure

Limit =

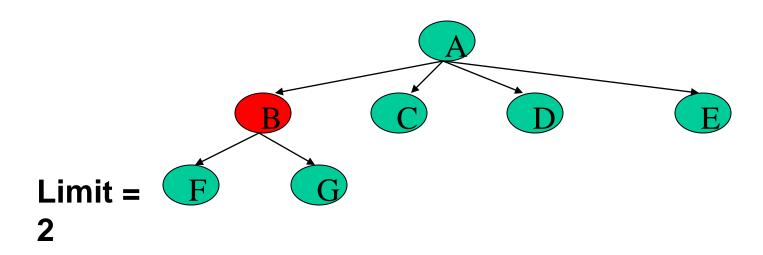
 B
 C
 D

A,

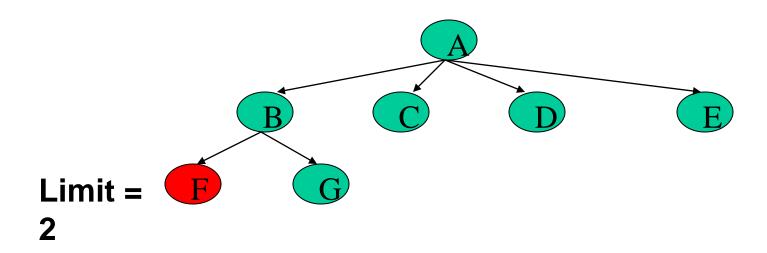


Limit = 2

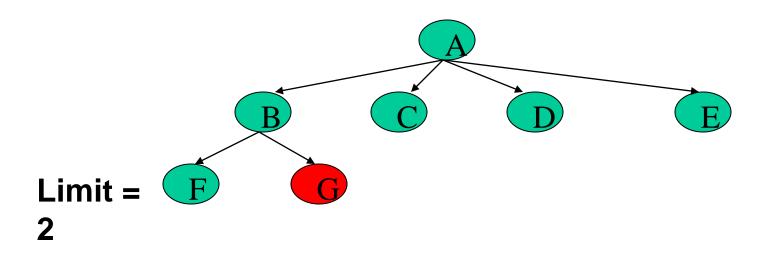
A,B,



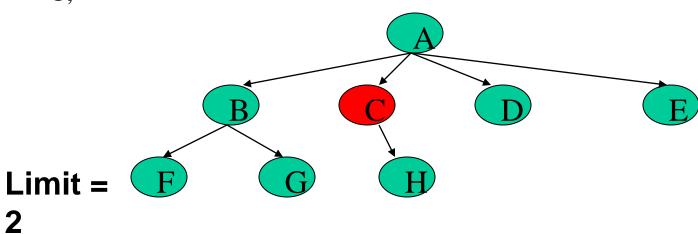
A,B,F,



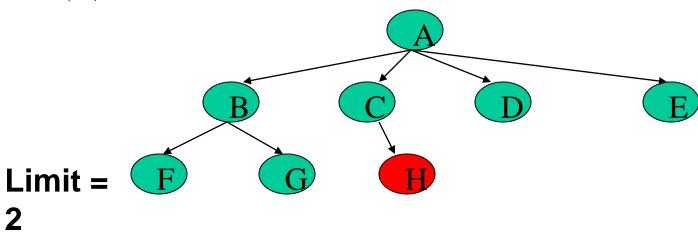
- A,B,F,
- **G**,

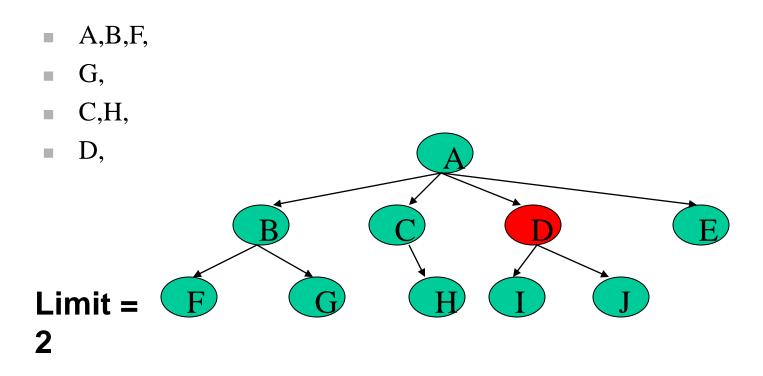


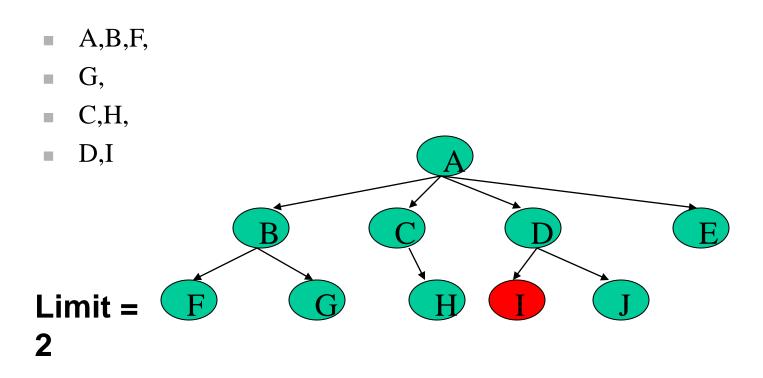
- A,B,F,
- **G**,
- C.

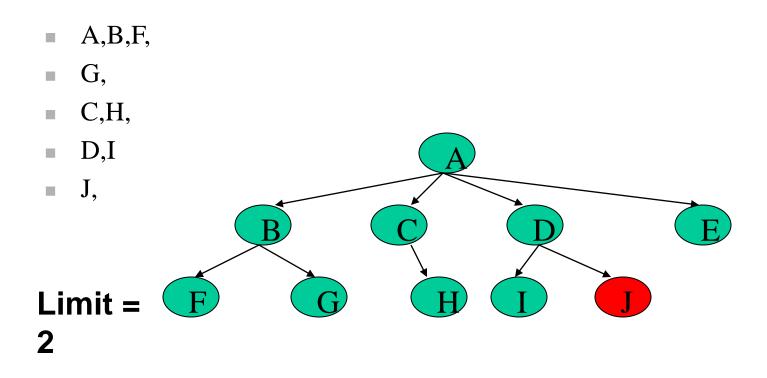


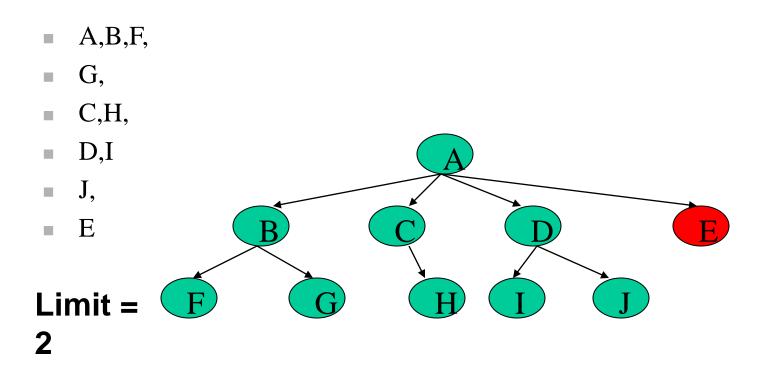
- A,B,F,
- **G**,
- **C,H**,

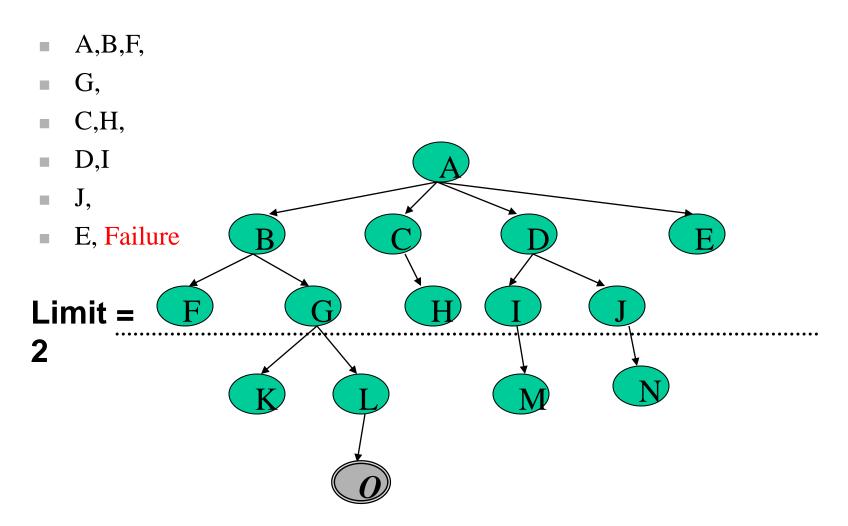






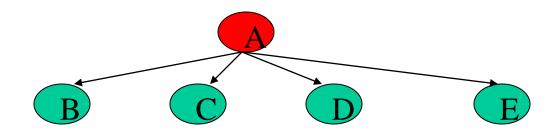




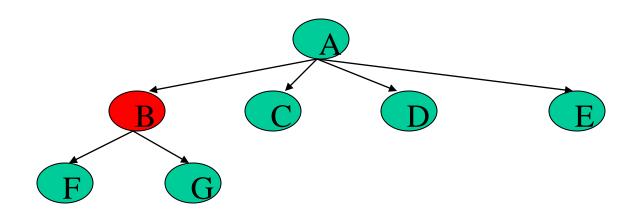


DLS with bound = 3

A,

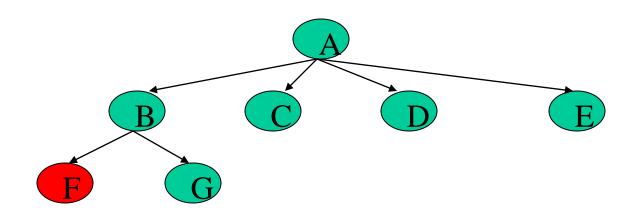


**A,B**,



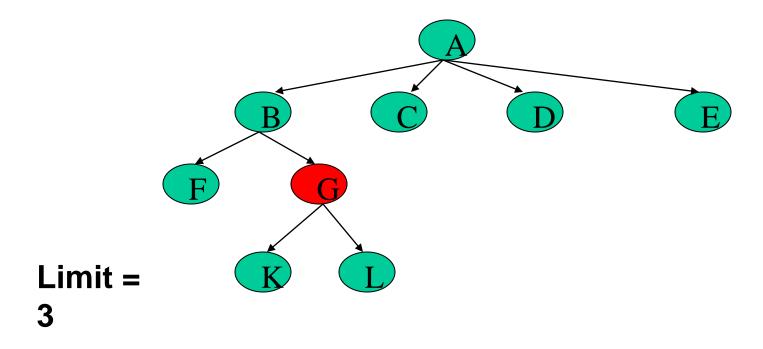
Limit = 3

A,B,F,

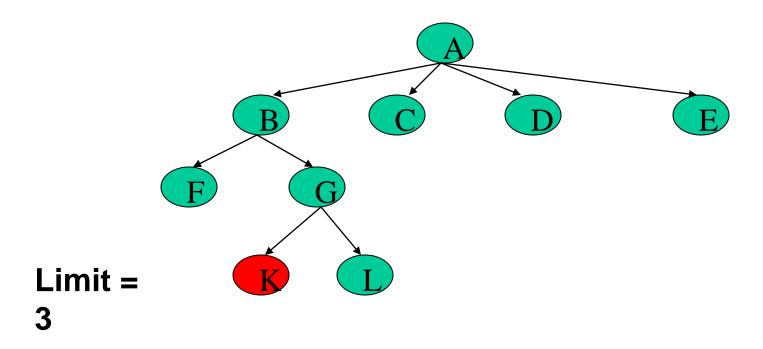


Limit = 3

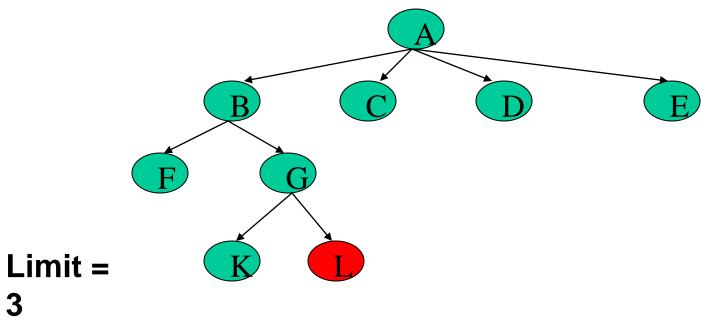
- A,B,F,
- **G**,

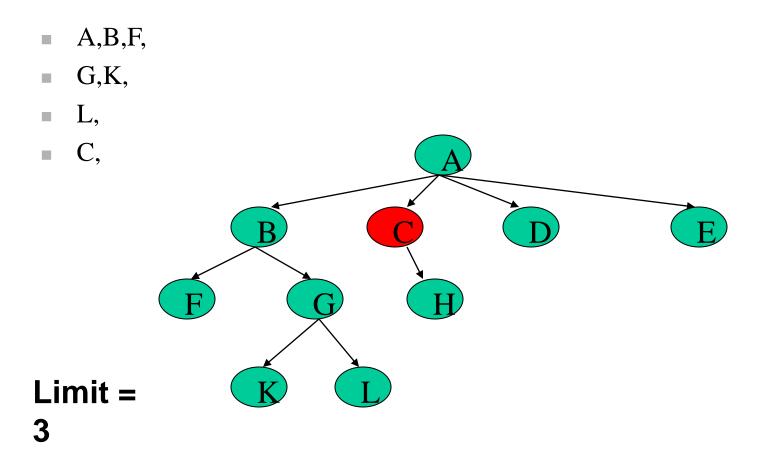


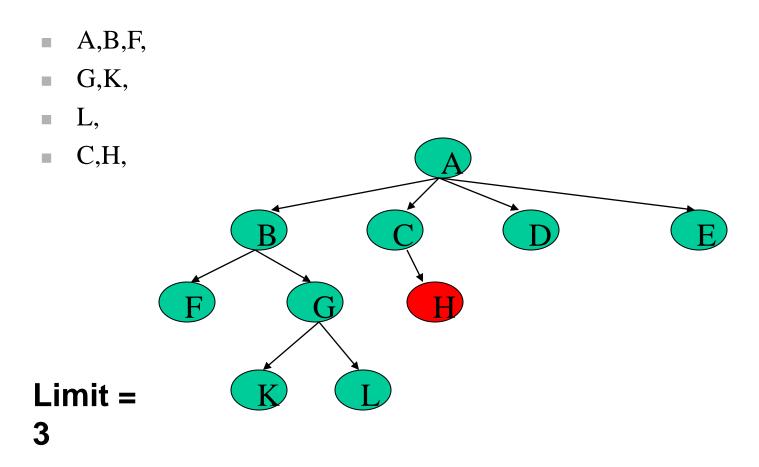
- A,B,F,
- **■** G,K,

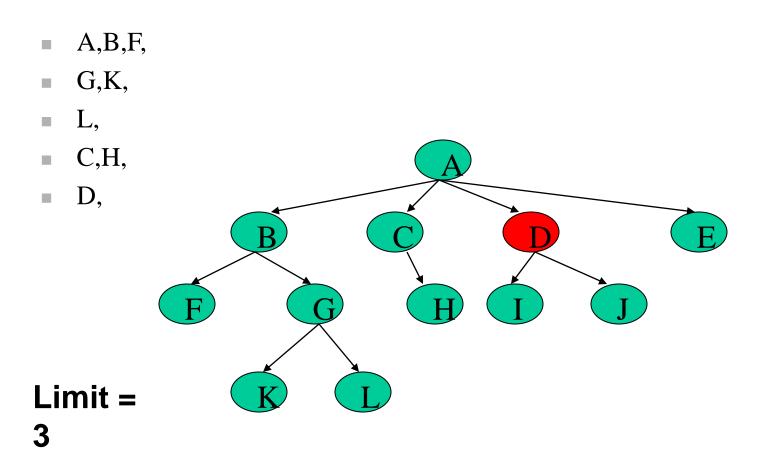


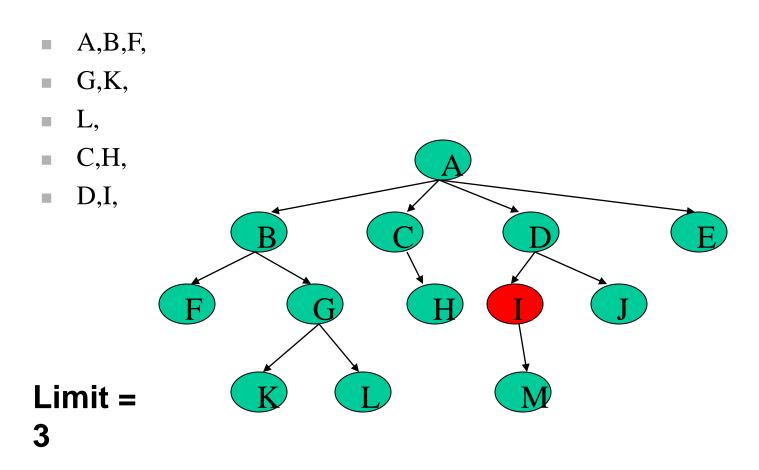
- A,B,F,
- G,K,
- L,

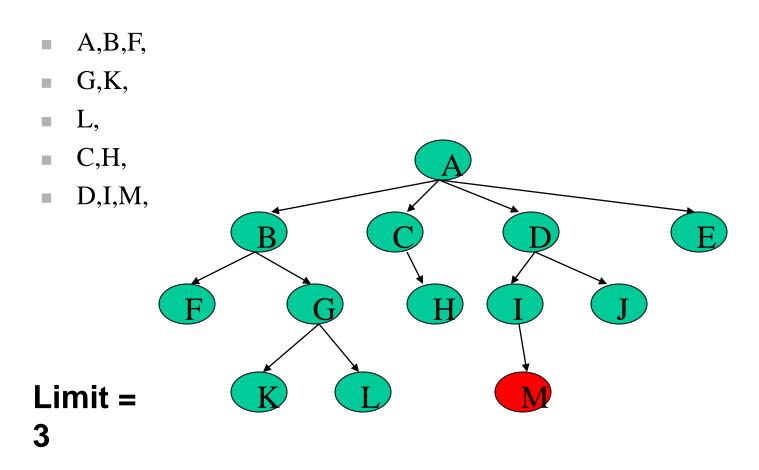


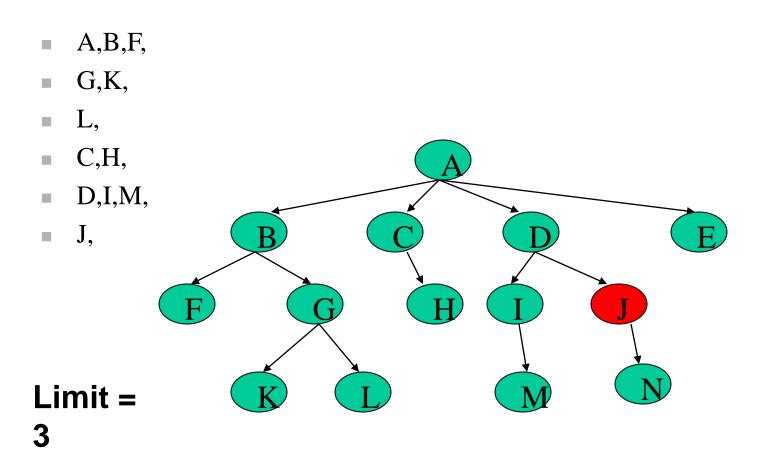


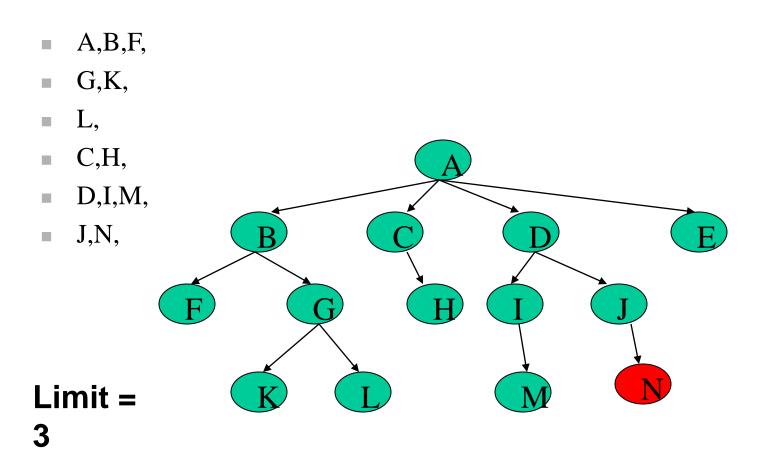


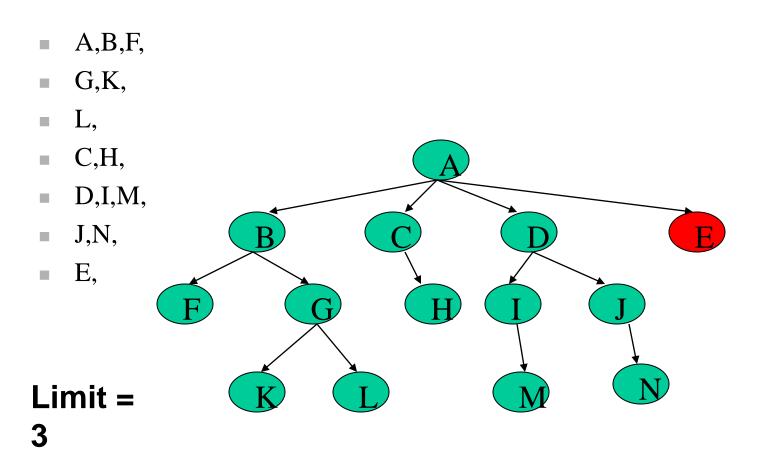


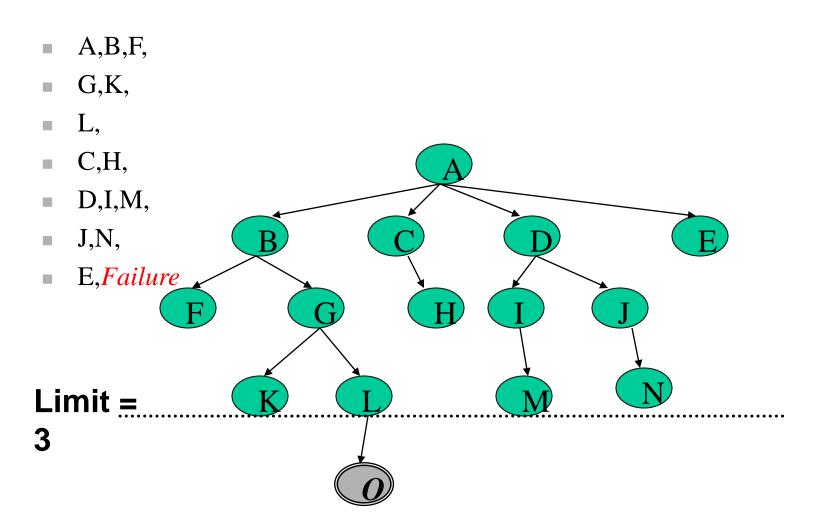






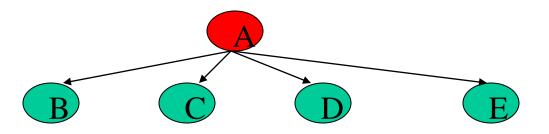




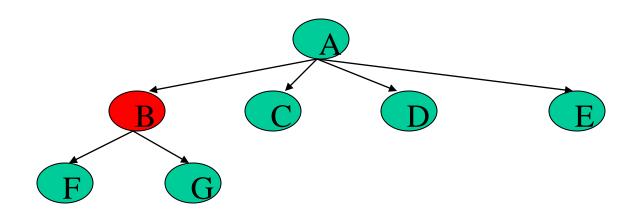


DLS with bound = 4

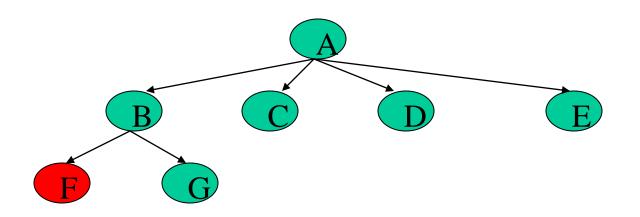
A,



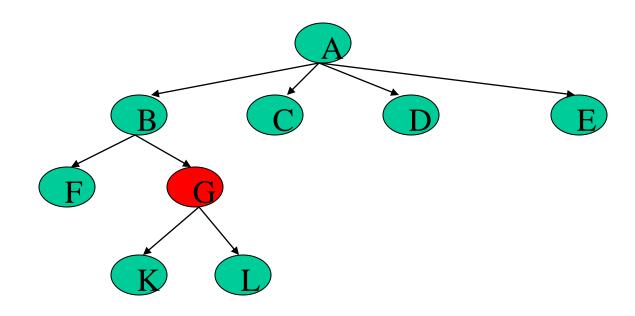
A,B,



A,B,F,

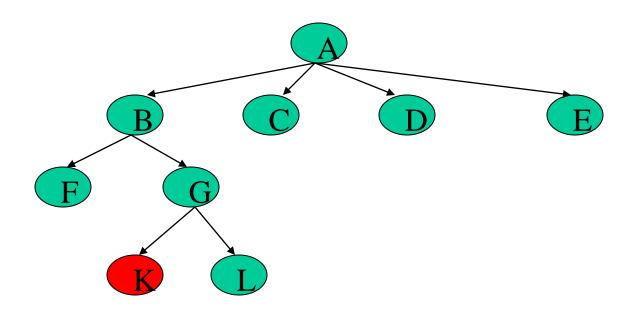


- A,B,F,
- **G**,



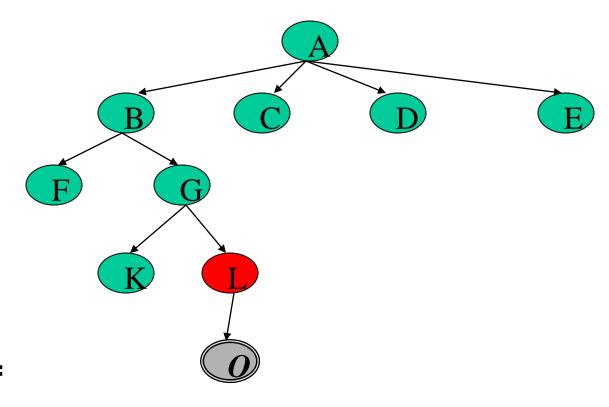
Limit = 4

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



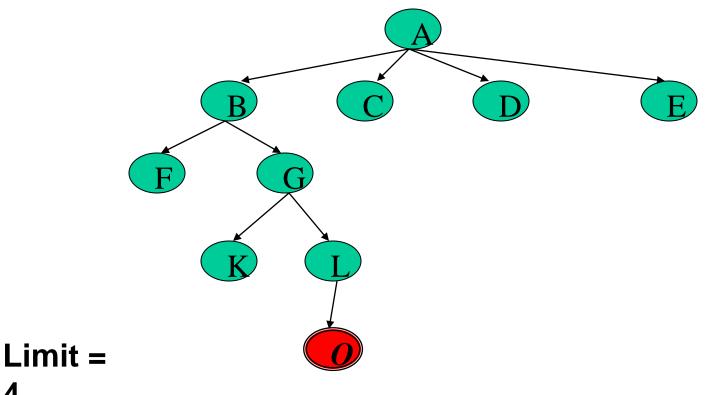
Limit = 4

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



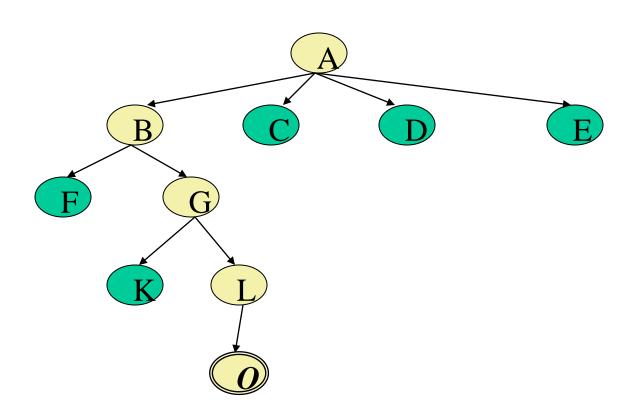
Limit =

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



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



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

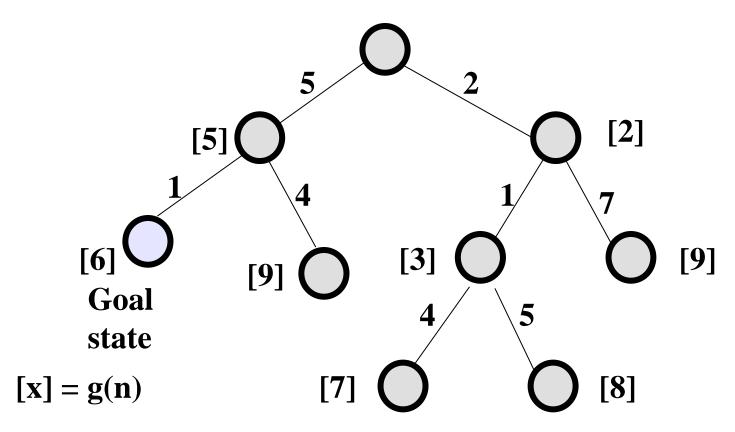
#### Implementation:

Enqueue nodes in order of cost g(n).

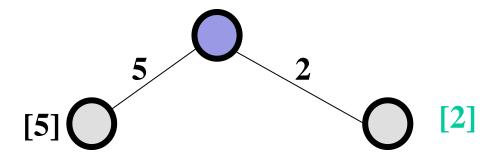
QUEUING-FN:- insert in order of increasing path cost.

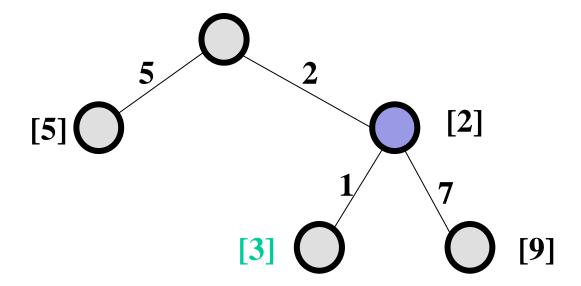
Enqueue new node at the appropriate position in the queue so that we dequeue the cheapest node.

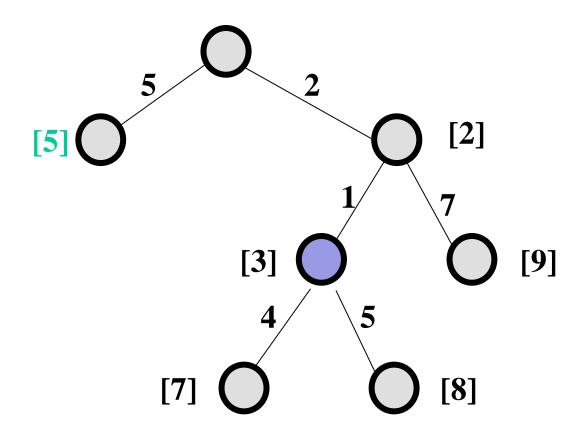
- Complete? Yes.
- Optimal? Yes, if path cost is nondecreasing function of depth
- Time Complexity: O(b<sup>d</sup>)
- Space Complexity: O(b<sup>d</sup>), note that every node in the fringe keep in the queue.

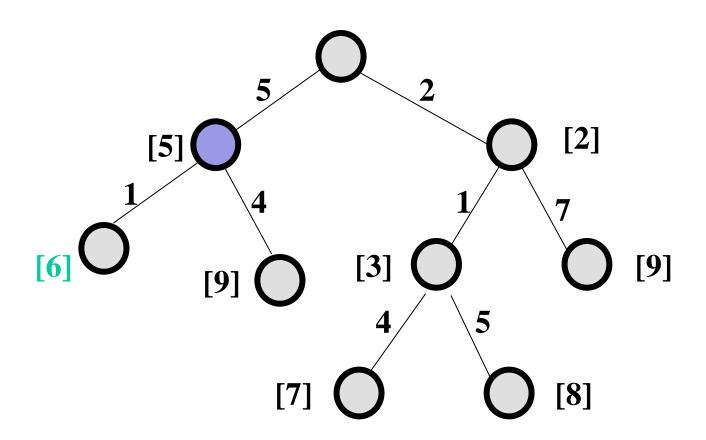


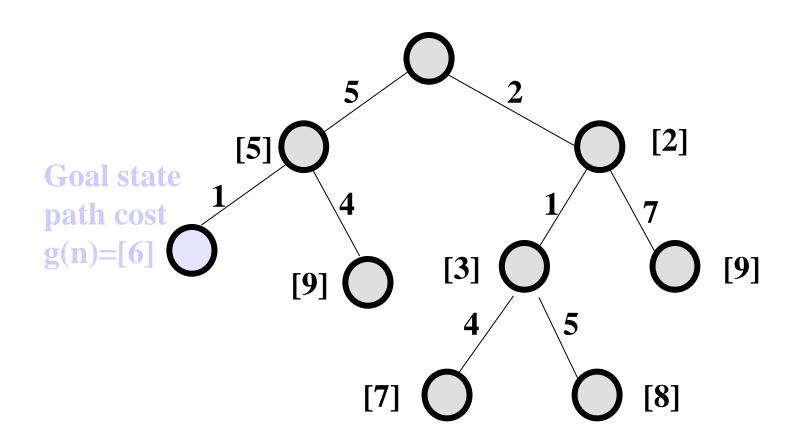
path cost of node n

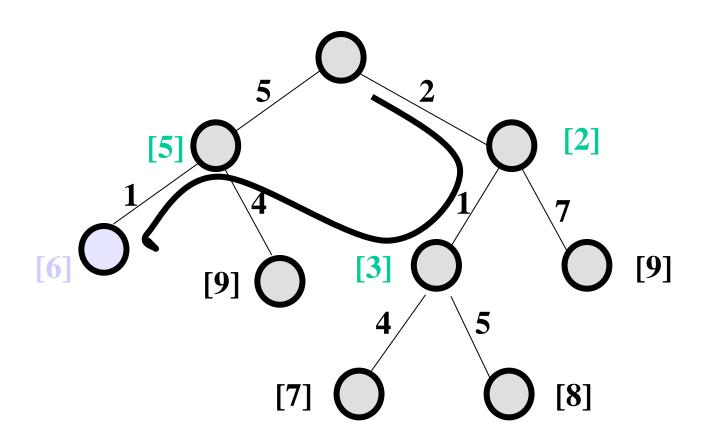












## **Uniform-cost search**

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

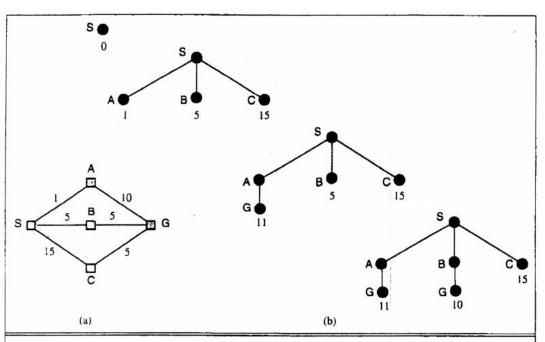
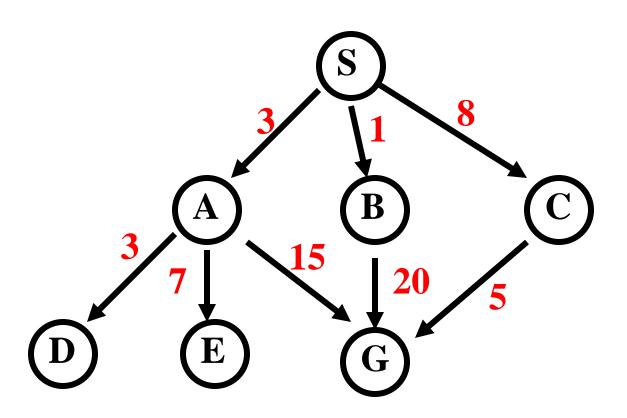
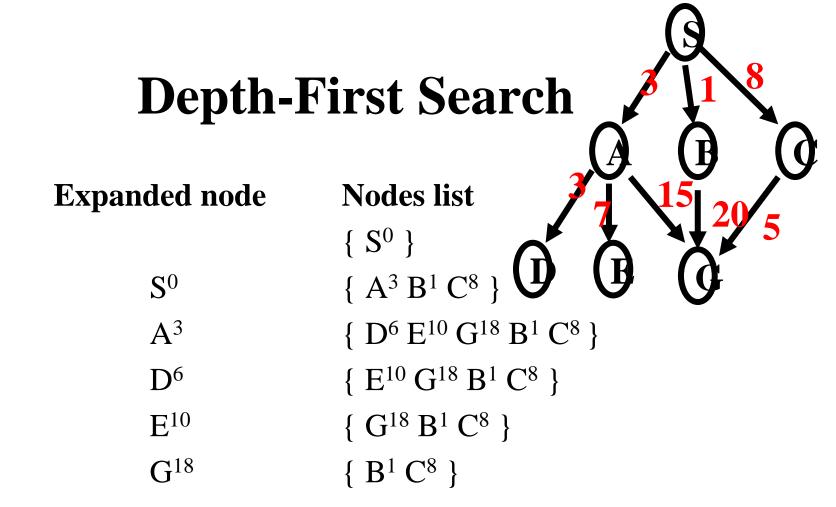


Figure 3.13 A route-finding problem. (a) The state space, showing the cost for each operator. (b) Progression of the search. Each node is labelled with g(n). At the next step, the goal node with g = 10 will be selected.

#### **Example for Illustrating Search Strategies**





Solution path found is S A G, cost 18 Number of nodes expanded (including goal node) = 5 **Breadth-First Search** 

| <b>Expanded node</b> | Nodes list  |
|----------------------|---|
|                      | $\{S^0\}$   |
| $\mathbf{S}^0$       | $\{ A^3 B^1 C^8 \} $                                    |
| $A^3$                | $\{ B^1 C^8 D^6 E^{10} G^{18} \}$                       |
| $\mathbf{B}^1$       | $\{\ C^8\ D^6\ E^{10}\ G^{18}\ G^{21}\ \}$              |
| $\mathbb{C}_8$       | $\{\ D^6\ E^{10}\ G^{18}\ G^{21}\ G^{13}\ \}$           |
| $D^6$                | $\{ \ E^{10} \ G^{18} \ G^{21} \ G^{13} \ \}$           |
| $\mathrm{E}^{10}$    | $\{ \mathbf{G}^{18} \mathbf{G}^{21} \mathbf{G}^{13} \}$ |
| $G^{18}$             | $\{ \mathbf{G}^{21} \mathbf{G}^{13} \}$                 |

Solution path found is S A G, cost 18

Number of nodes expanded (including goal node) = 7

**Uniform-Cost Search** 

| Expanded node  | <b>Nodes list</b>                  |  |  |
|----------------|------------------------------------|--|--|
|                | $\{S^0\}$                          |  |  |
| $\mathbf{S}^0$ | $\{ B^1 A^3 C^8 \}$                |  |  |
| $\mathbf{B}^1$ | $\{ A^3 C^8 G^{21} \}$             |  |  |
| $A^3$          | $\{\ D^6\ C^8\ E^{10}\ G^{18}\ \}$ |  |  |
| $D^6$          | $\{ C^8 E^{10} G^{18} \}$          |  |  |
| $\mathbb{C}^8$ | $\{ E^{10} G^{13} \}$              |  |  |
| $E^{10}$       | $\{G^{13}\}$                       |  |  |
| $G^{13}$       | {}                                 |  |  |

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?  $1+b+b^2+b^3+...+b^d+(b^{d+1}-b)$  = O(b<sup>d+1</sup>) (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

- <u>Complete?</u> 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?  $(d+1)b^0 + db^1 + (d-1)b^2 + ... + b^d = O(b^d)$
- <u>Space?</u> *O*(*bd*)
- Optimal? Yes, if step cost = 1 or increasing function of depth.

#### **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  $\geq \varepsilon$  (otherwise it can get stuck in infinite loops)

<u>Time?</u> # of nodes with  $path cost \le cost of optimal solution.$ 

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

Optimal? Yes, for any step cost.

#### **Bi-Directional Search**

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

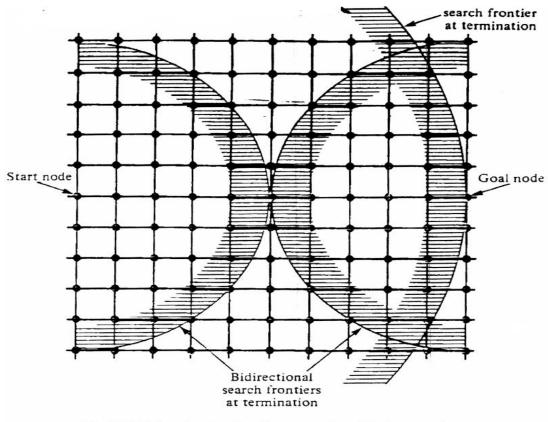


Fig. 2.10 Bidirectional and unidirectional breadth-first searches.

# **Summary of algorithms**

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