# ENCS3340 - Artificial Intelligence

Uninformed Search

# Uninformed Search Strategies

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

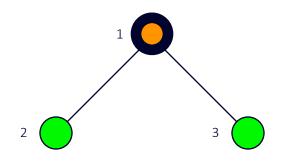
#### Breadth-First Search

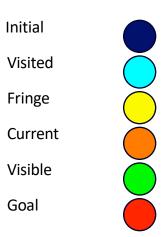
- all the nodes reachable from the current node are explored first
  - Expand shallowest unexpanded node
  - achieved by the TREE-SEARCH method by appending newly generated nodes at the end of the search queue

```
function GENERAL-SEARCH(problem) returns solution

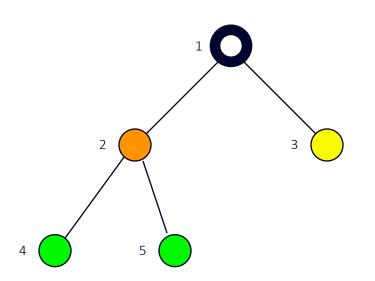
return TREE-SEARCH(problem, FIFO-QUEUE())
```

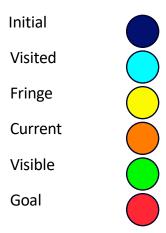
• Implementation: fringe is a FIFO queue, i.e., new successors go at end



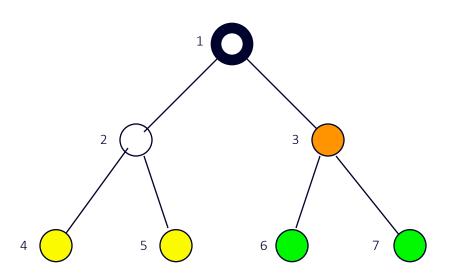


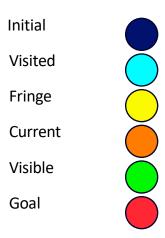
Fringe: [] + [2,3]



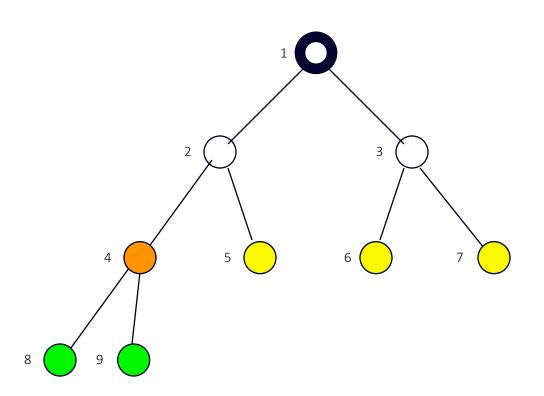


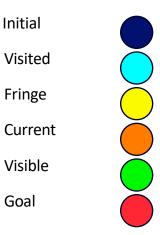
Fringe: [3] + [4,5]

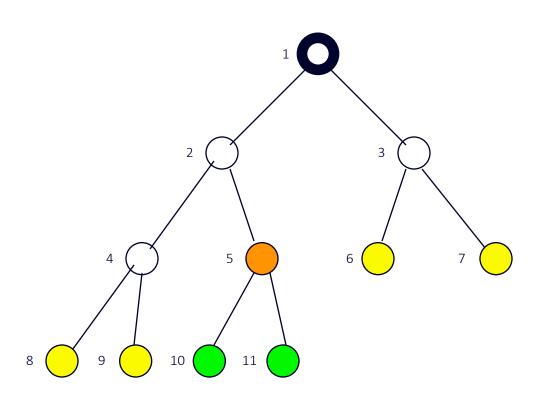


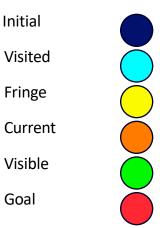


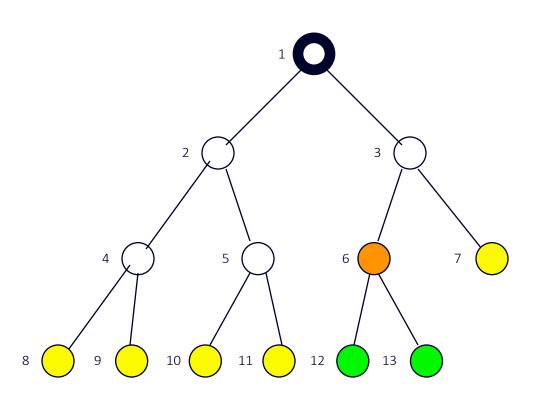
Fringe: [4,5] + [6,7]

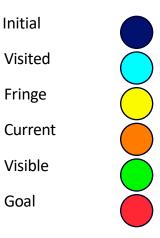




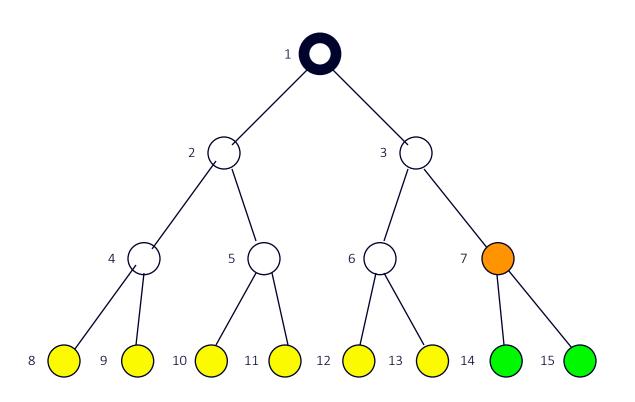




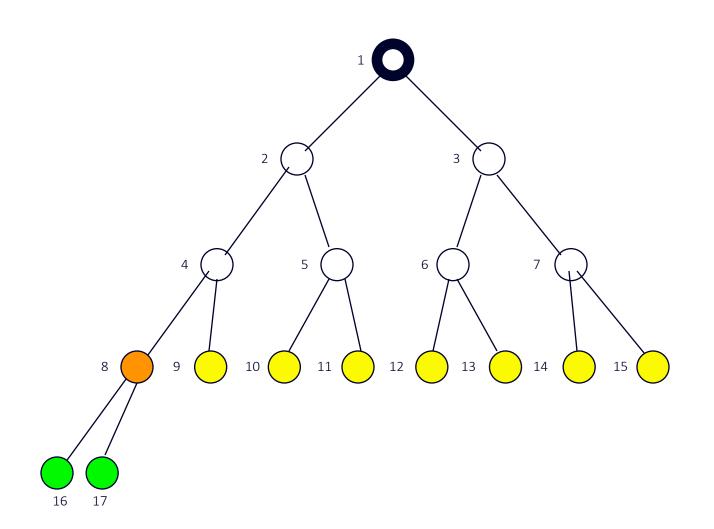


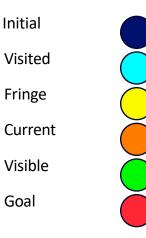


Fringe: [7,8,9,10,11] + [12,13]

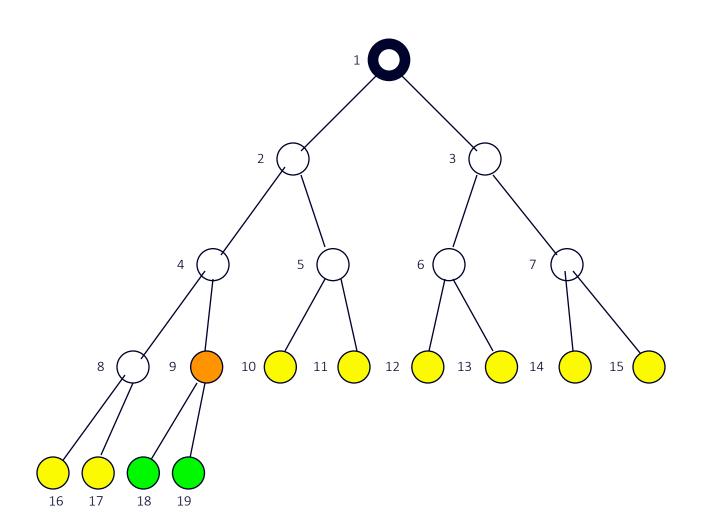


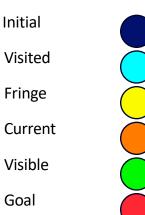
Initial
Visited
Fringe
Current
Visible
Goal



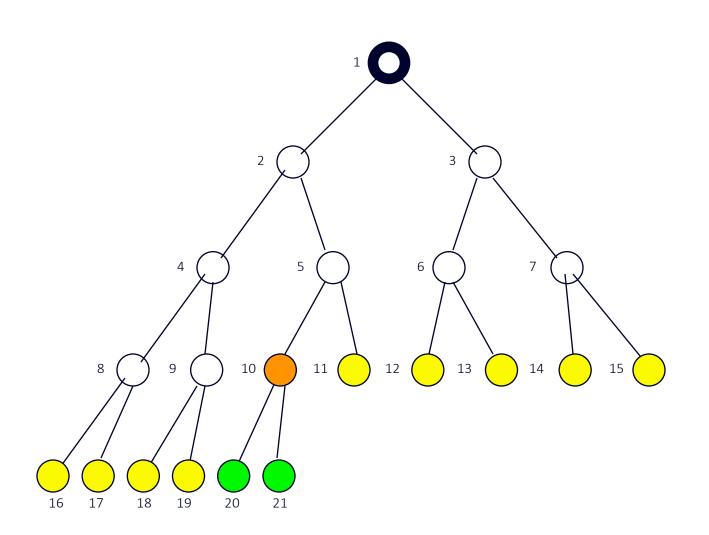


Fringe: [9,10,11,12,13,14,15] + [16,17]



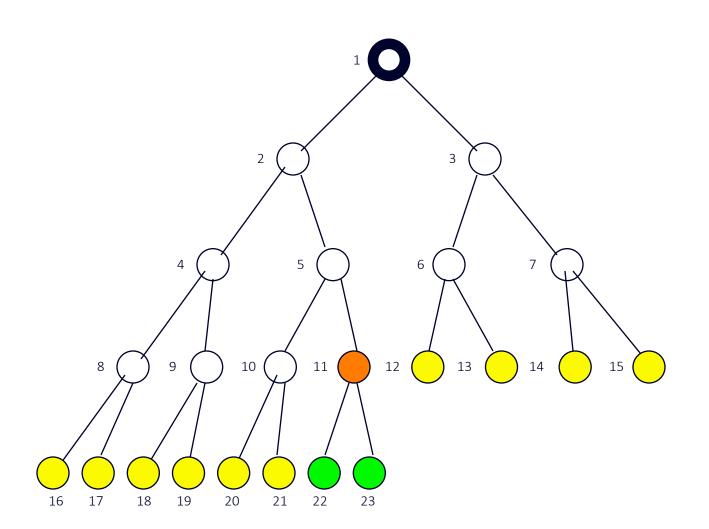


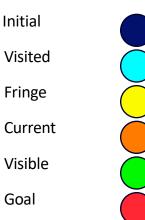
Fringe: [10,11,12,13,14,15,16,17] + [18,19]

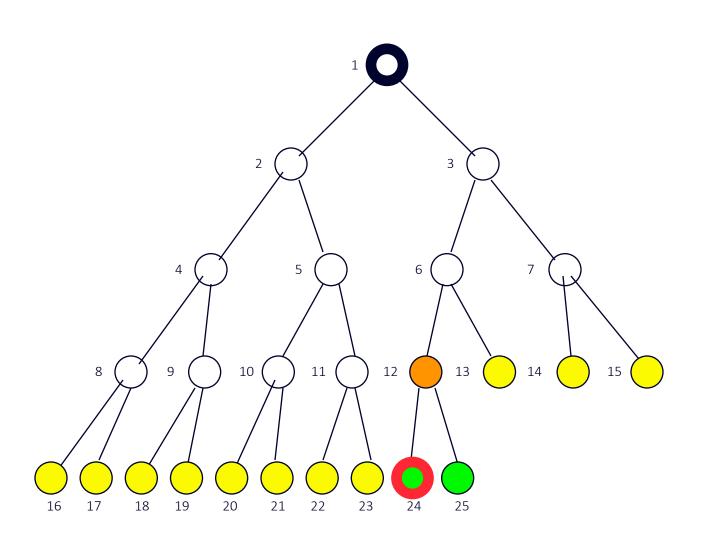


Initial
Visited
Fringe
Current
Visible
Goal

Fringe: [11,12,13,14,15,16,17,18,19] + [20,21]







Initial

Visited

Fringe

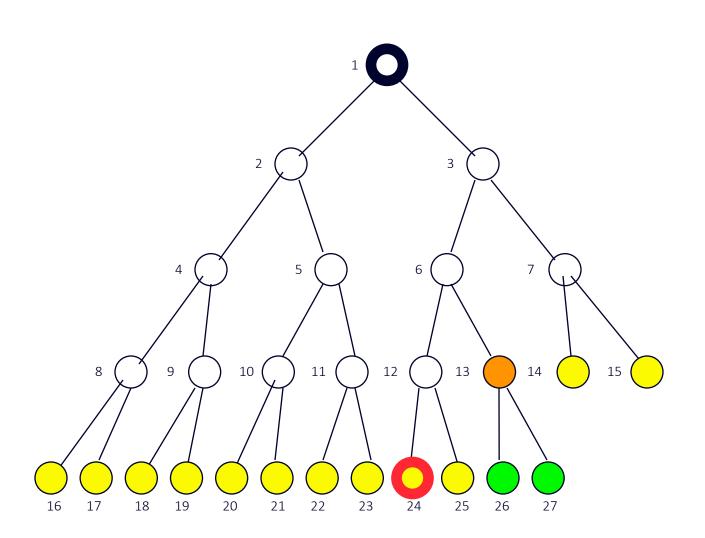
Current

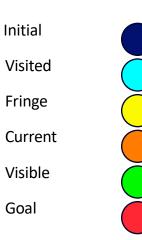
Visible

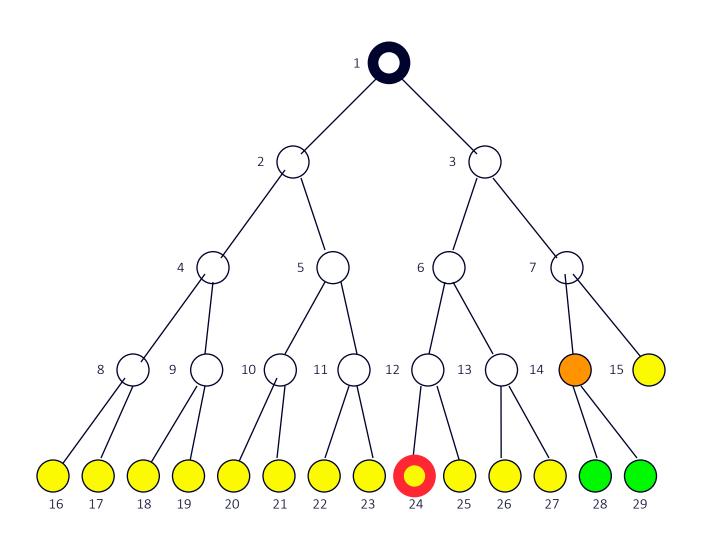
Goal

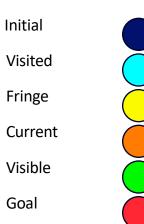
Note: The goal node is "visible" here, but we can not perform the goal test yet.

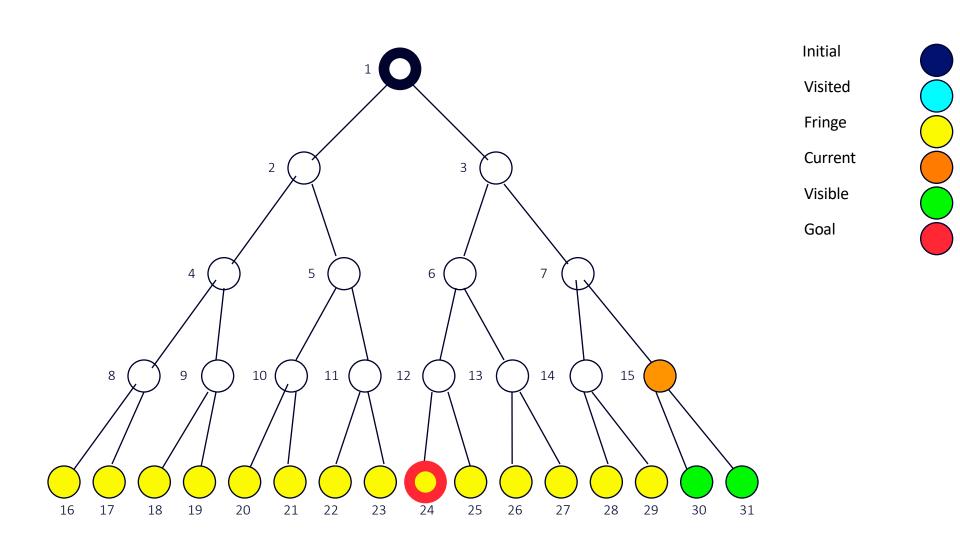
Fringe: [13,14,15,16,17,18,19,20,21, 22, 23] + [24,25]

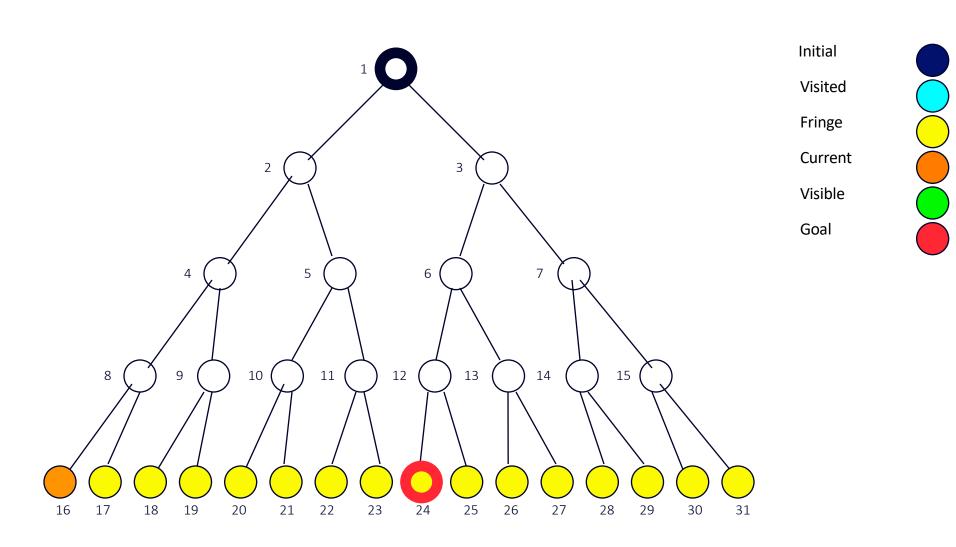




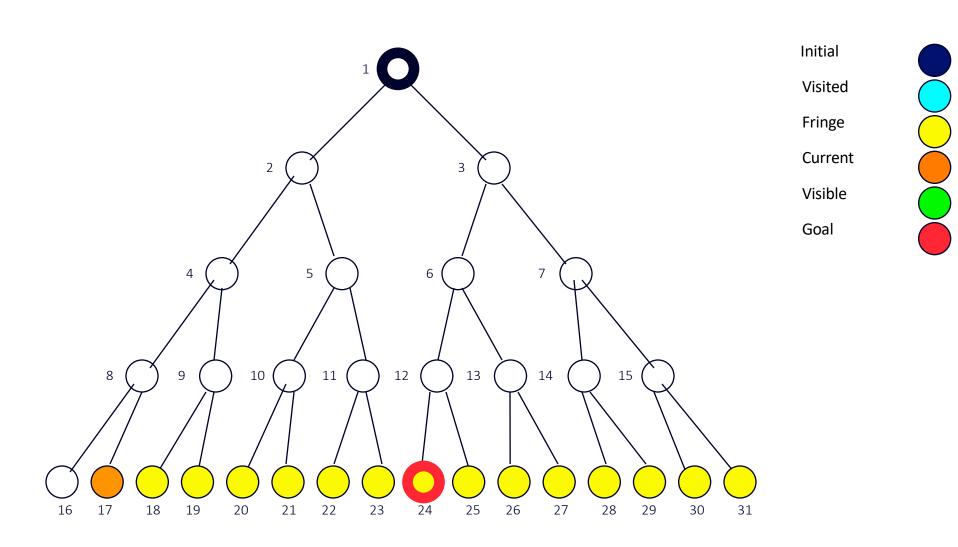




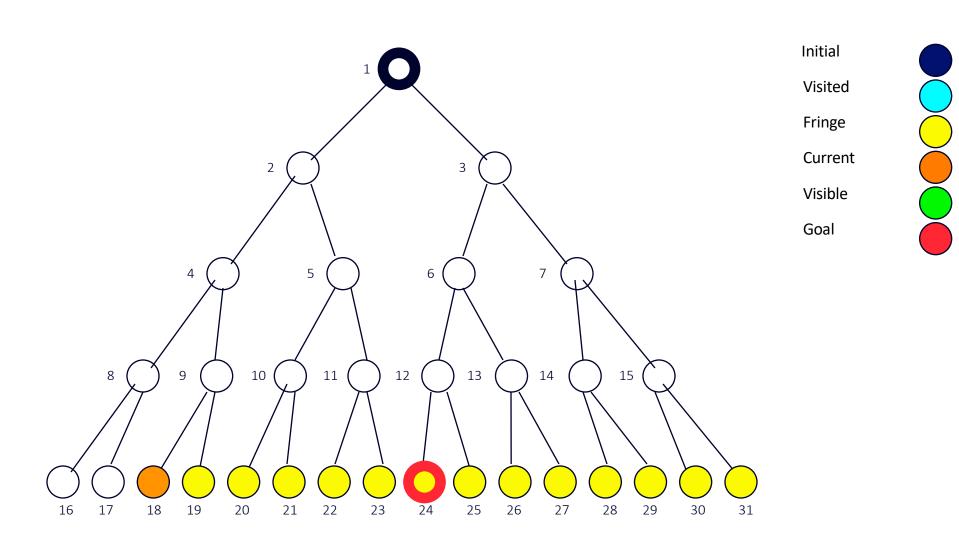




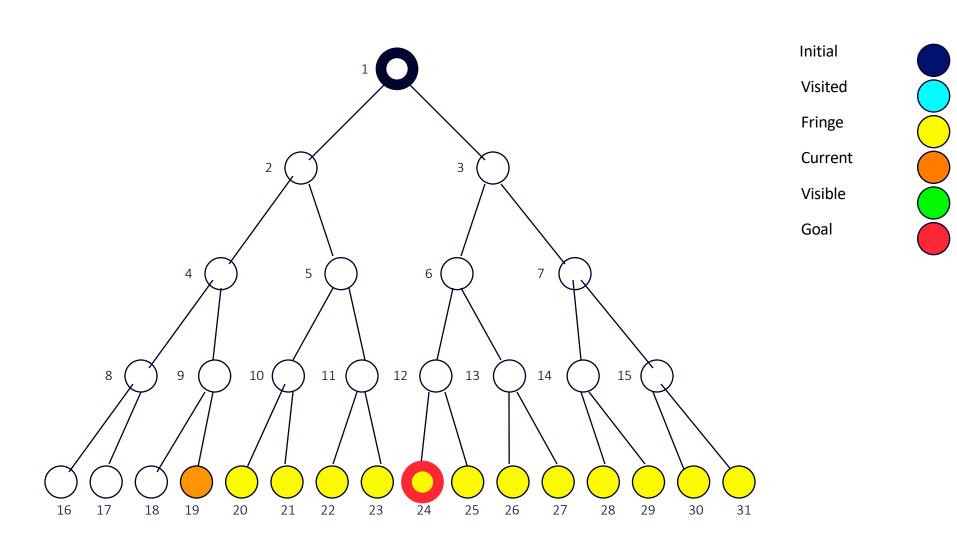
Fringe: [17,18,19,20,21,22,23,24,25,26,27,28,29,30,31]



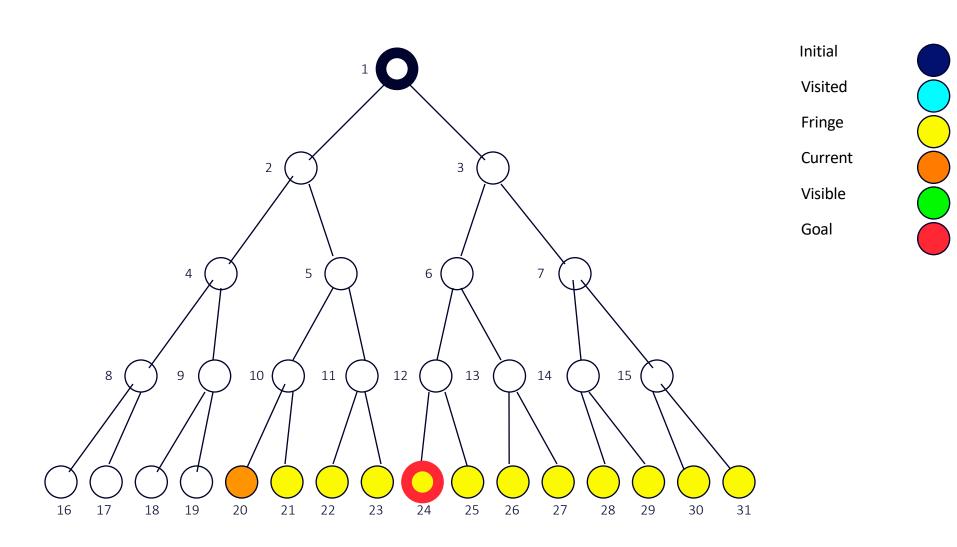
Fringe: [18,19,20,21,22,23,24,25,26,27,28,29,30,31]



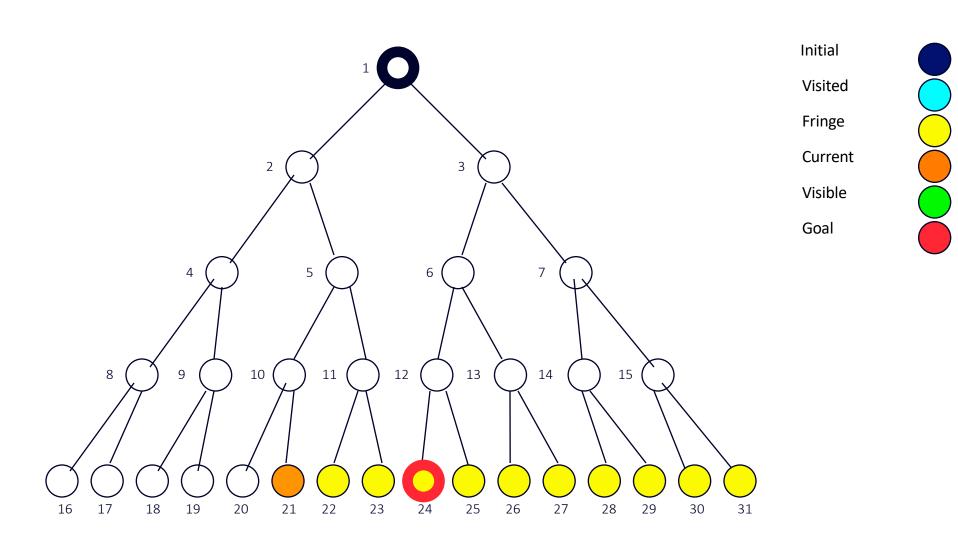
Fringe: [19,20,21,22,23,24,25,26,27,28,29,30,31]



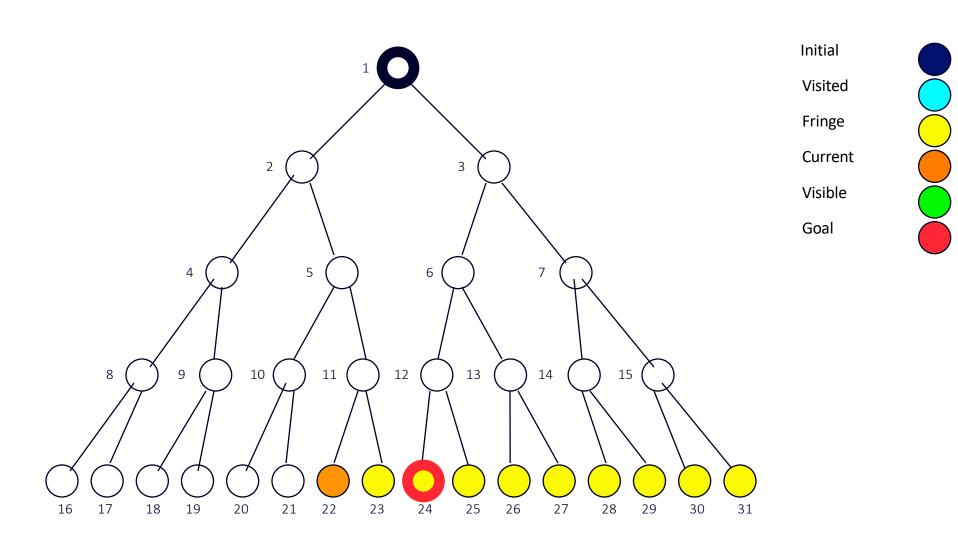
Fringe: [20,21,22,23,24,25,26,27,28,29,30,31]



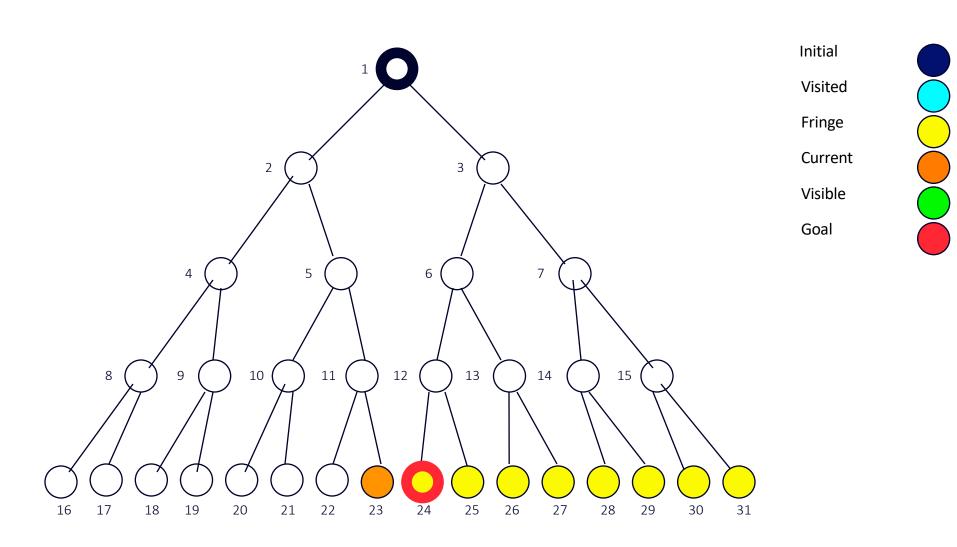
Fringe: [21,22,23,24,25,26,27,28,29,30,31]



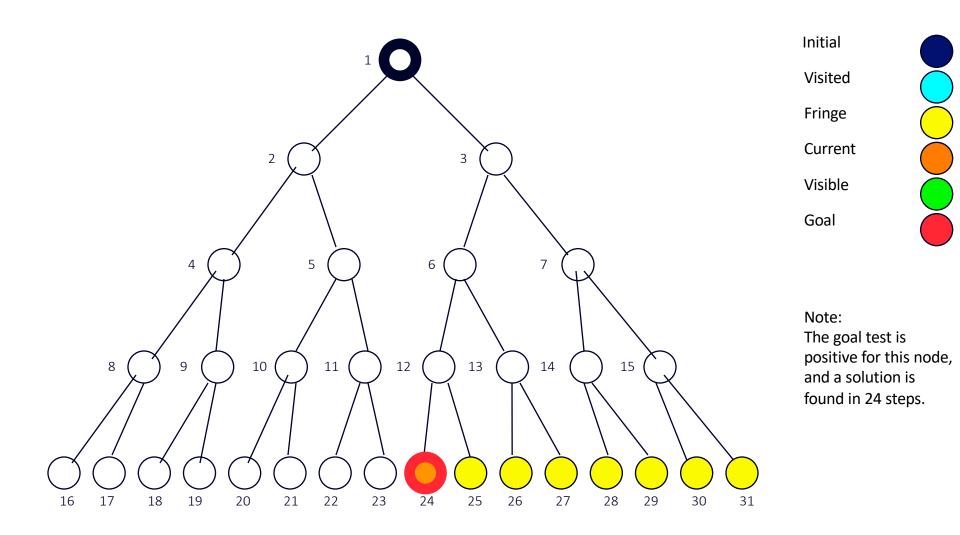
Fringe: [22,23,24,25,26,27,28,29,30,31]



Fringe: [23,24,25,26,27,28,29,30,31]



Fringe: [24,25,26,27,28,29,30,31]



Fringe: [25,26,27,28,29,30,31]

# Properties of Breadth-First Search

Time Complexity	O(b <sup>d+1</sup> )
Space Complexity	O(b <sup>d+1</sup> )
Completeness	yes (for finite b)
Optimality	yes (for non-negative path costs)

b branching factor

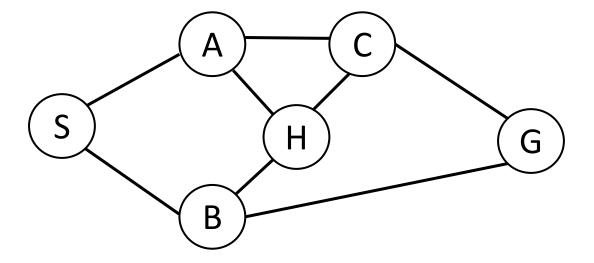
d depth of the optimal solution

### Breadth-First Search for Graphs

- Keep a set of explored (visited) nodes.
- When a node is expanded, add the generated nodes to frontier only if not in the frontier or explored set

#### Example: BFS for Graphs

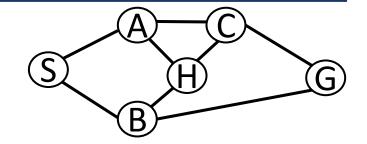
• For the following state space, list all the visited nodes and draw the search tree to go from S to G using BFS. What is the returned solution?

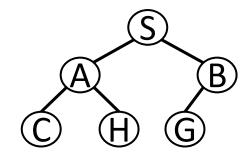


# Example: BFS for Graphs (Cont.)

• Visited Nodes: S, A, B, C, H, G

Solution (Path to goal): S, B, G





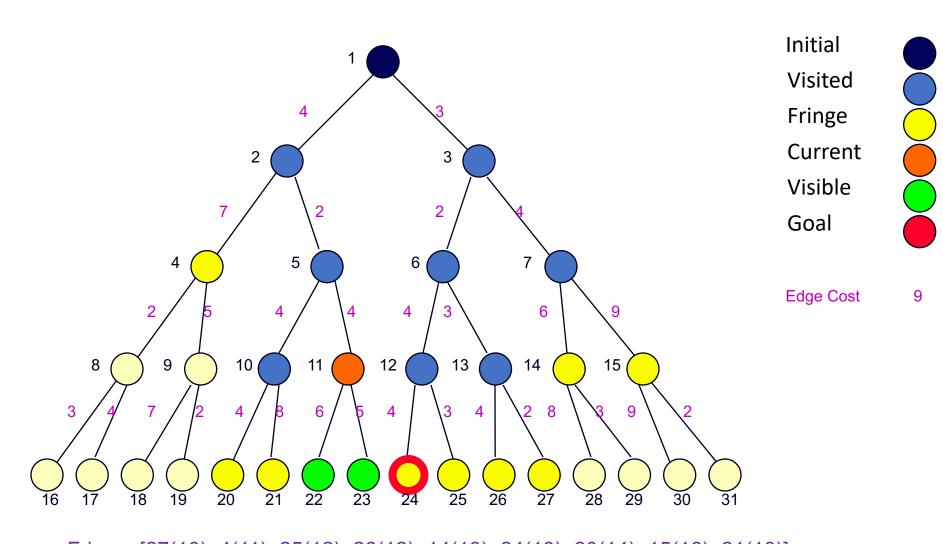
#### Uniform-Cost Search

- the nodes with the lowest cost are explored first
  - similar to BREADTH-FIRST, but with an evaluation of the cost for each reachable node
  - g(n) = path cost(n) = sum of individual edge costs to reach the current node
- Implementation: fringe is a queue ordered by path cost (priority queue)
- Equivalent to breadth-first if step costs all equal

```
function UNIFORM-COST-SEARCH(problem) returns solution

return TREE-SEARCH(problem, COST-FN, FIFO-QUEUE())
```

### Uniform-Cost Snapshot



Fringe: [27(10), 4(11), 25(12), 26(12), 14(13), 24(13), 20(14), 15(16), 21(18)] STUDENTS-HUB.com + [22(16), 23(15)] Uploaded By: Jibreel Bornat

#### Uniform Cost Fringe Trace

```
1.
       [1(0)]
       [3(3), 2(4)]
2.
       [2(4), 6(5), 7(7)]
3.
4.
       [6(5), 5(6), 7(7), 4(11)]
5.
       [5(6), 7(7), 13(8), 12(9), 4(11)]
6.
       [7(7), 13(8), 12(9), 10(10), 11(10), 4(11)]
7.
       [13(8), 12(9), 10(10), 11(10), 4(11), 14(13), 15(16)]
       [12(9), 10(10), 11(10), 27(10), 4(11), 26(12), 14(13), 15(16)]
8.
9.
       [10(10), 11(10), 27(10), 4(11), 26(12), 25(12), 14(13), 24(13), 15(16)]
10.
       [11(10), 27(10), 4(11), 25(12), 26(12), 14(13), 24(13), 20(14), 15(16), 21(18)]
11.
       [27(10), 4(11), 25(12), 26(12), 14(13), 24(13), 20(14), 23(15), 15(16), 22(16), 21(18)]
12.
       [4(11), 25(12), 26(12), 14(13), 24(13), 20(14), 23(15), 15(16), 23(16), 21(18)]
13.
       [25(12), 26(12), 14(13), 24(13), 8(13), 20(14), 23(15), 15(16), 23(16), 9(16), 21(18)]
       [26(12), 14(13), 24(13), 8(13), 20(14), 23(15), 15(16), 23(16), 9(16), 21(18)]
14.
15.
       [14(13), 24(13),8(13), 20(14), 23(15), 15(16), 23(16), 9(16), 21(18)]
       [24(13),8(13), 20(14), 23(15), 15(16), 23(16), 9(16), 29(16), 21(18), 28(21)]
16.
       Goal reached!
```

Notation: [Bold+Yellow: Current Node; White: Old Fringe Node; Green+Italics: New Fringe Node].

Assumption: New nodes with the same cost as existing nodes are added after the existing node.

# Properties of Uniform-Cost Search

Time Complexity	O(b <sup>C*/e</sup> )
Space Complexity	O(b <sup>C*/e</sup> )
Completeness	yes (finite b, step costs >= e)
Optimality	yes

b branching factor

C\* cost of the optimal solution

e minimum cost per action

# Uniform-Cost Search for Graphs

- Keep a set of explored (visited) nodes.
- When a node is expanded, add the generated nodes to frontier only if not in the frontier or explored set
  - If the state of the generated node is in the frontier but with higher cost, replace that frontier node with newly generated node

#### Breadth-First vs. Uniform-Cost

- breadth-first always expands the shallowest node
  - only optimal if all step costs are equal
- uniform-cost considers the overall path cost
  - optimal for any (reasonable) cost function
    - non-zero, positive
  - gets bogged down in trees with many fruitless, short branches
    - · low path cost, but no goal node
- both are complete for non-extreme problems
  - finite number of branches
  - strictly positive search function

- continues exploring newly generated nodes
  - achieved by the TREE-SEARCH method by appending newly generated nodes at the beginning of the search queue
- Implementation: fringe = LIFO queue, i.e., put successors at front

function DEPTH-FIRST-SEARCH(problem) returns solution

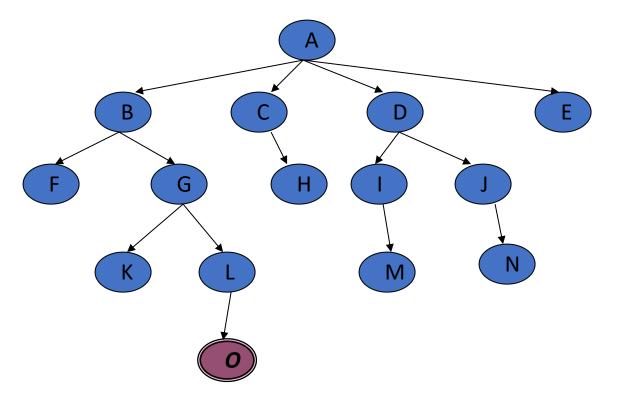
return TREE-SEARCH(problem, LIFO-QUEUE())

Time Complexity	O(b <sup>m</sup> )
Space Complexity	O(b*m)
Completeness	no (for infinite branch length)
Optimality	no

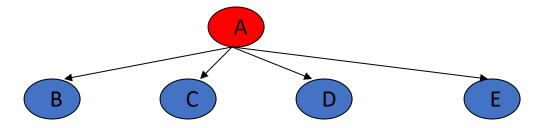
b branching factorm maximum path length

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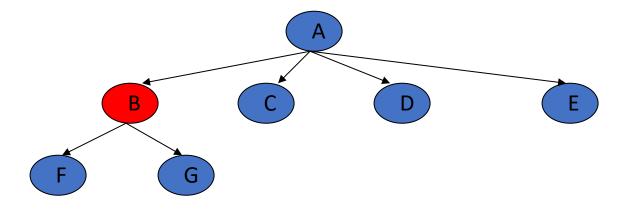
• Given the following state space (tree search), give the sequence of visited nodes when using DFS (assume that the node O is the goal state):



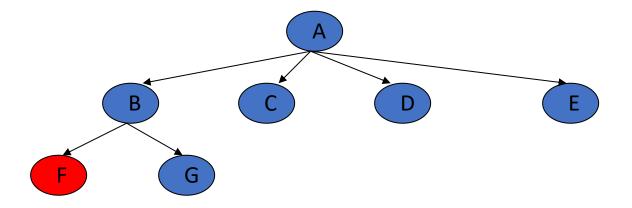
• A,



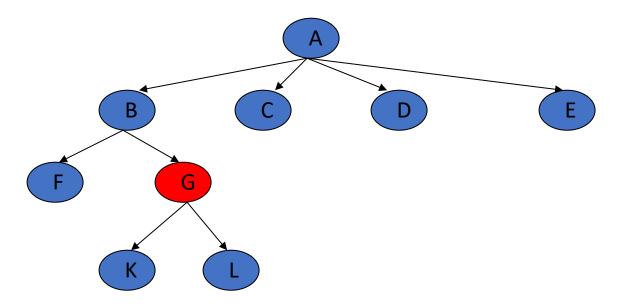
• A, B,



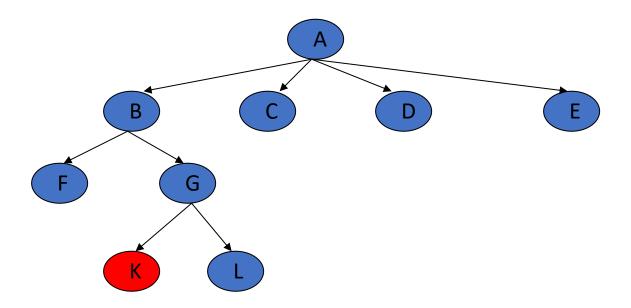
• A, B, F,



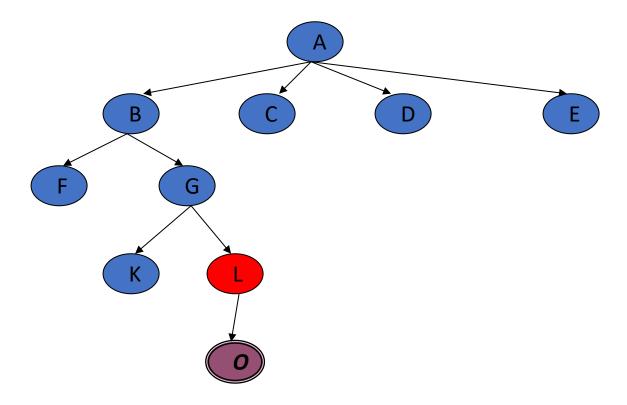
- A, B, F,
- G,



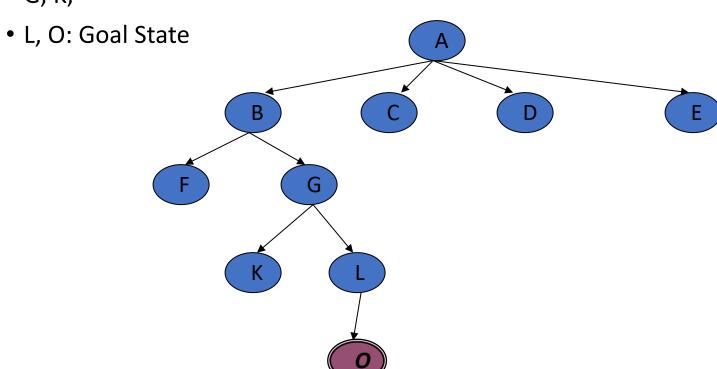
- A, B, F,
- G, K,



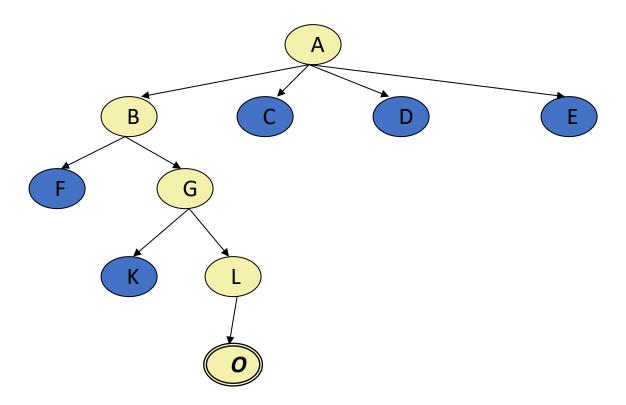
- A, B, F,
- G, K,
- L,



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

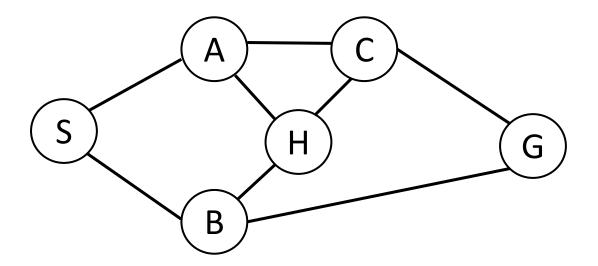


• The returned solution is the sequence of operators in the path: A, B, G, L, O



#### Example: DFS for Graphs

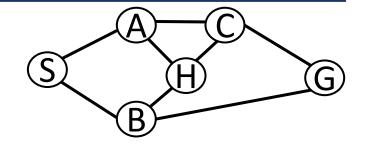
• For the following state space, list all the visited nodes and draw the search tree to go from S to G using DFS. What is the returned solution?

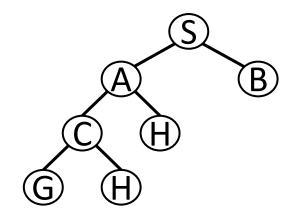


### Example: DFS for Graphs (Cont.)

• Visited Nodes: S, A, C, G

• Solution (Path to goal): S, A, C, G





#### Depth-First vs. Breadth-First

- depth-first goes off into one branch until it reaches a leaf node
  - not good if the goal is on another branch
  - neither complete nor optimal
  - uses much less space than breadth-first

- breadth-first is more careful by checking all alternatives
  - complete and optimal
  - very memory-intensive

#### Depth-Limited Search

- similar to depth-first, but with a limit
  - overcomes problems with infinite paths
  - sometimes a depth limit can be inferred or estimated from the problem description
  - based on the TREE-SEARCH method
  - must keep track of the depth

```
function DEPTH-LIMITED-SEARCH(problem, depth-limit) returns solution
    return TREE-SEARCH(problem, depth-limit, LIFO-QUEUE())
```

Time Complexity	O(b <sup>l</sup> )
Space Complexity	O(b*I)
Completeness	no (goal beyond I, or infinite branch length)
Optimality	no

b branching factorl depth limit

#### Iterative Deepening

- applies LIMITED-DEPTH with increasing depth limits
- combines advantages of BREADTH-FIRST and DEPTH-FIRST methods
- many states are expanded multiple times
  - doesn't really matter because the number of those nodes is small
- in practice, one of the best uninformed search methods
  - for large search spaces, unknown depth

```
function ITERATIVE-DEEPENING-SEARCH(problem) returns solution
  for depth := 0 to unlimited do
    result := DEPTH-LIMITED-SEARCH(problem, depth-limit)
    if result != cutoff then return result
```

Time Complexity	O(b <sup>d</sup> )
Space Complexity	O(b*d)
Completeness	yes (finite b)
Optimality STUDENTS-HUB.com	yes (all step costs identical)

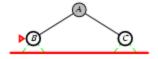
b branching factor

d depth of optimal solution

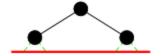
Uploaded By: Jibreel Bornat

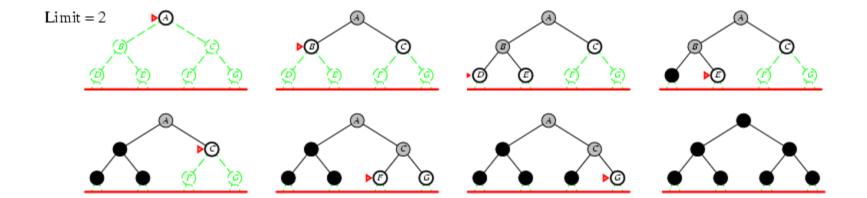


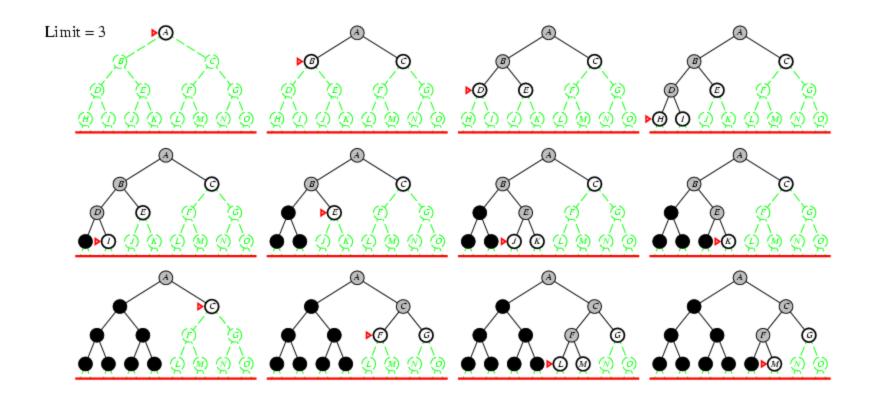




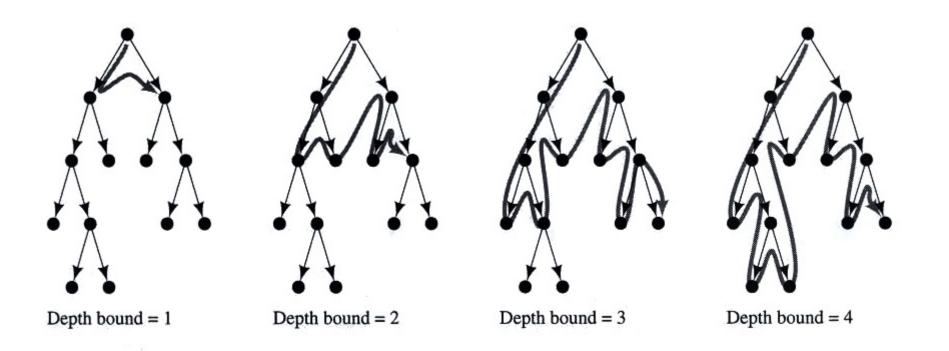








# Example IDS



Stages in Iterative-Deepening Search

#### Bi-directional Search

- search simultaneously from two directions
  - forward from the initial and backward from the goal state
- may lead to substantial savings if it is applicable
- has severe limitations
  - predecessors must be generated, which is not always possible
  - search must be coordinated between the two searches
  - one search must keep all nodes in memory

Time Complexity	O(b <sup>d/2</sup> )
Space Complexity	O(b <sup>d/2</sup> )
Completeness	yes (b finite, breadth-first for both directions)
Optimality	yes (all step costs identical, breadth-first for both directions)

b branching factor

d depth of optimal solution

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#### Improving Search Methods

- assumption for improvements
  - remember information about the search so far
    - all nodes visited so far
    - path to the current node
- make algorithms more efficient
  - avoiding repeated states
  - utilizing memory efficiently
- use additional knowledge about the problem => informed search
  - properties ("shape") of the search space
    - more interesting areas are investigated first
  - pruning of irrelevant areas
    - · areas that are guaranteed not to contain a solution can be discarded

### Comparing Uninformed Search Strategies

Criterion	Breadth- First	Uniform- Cost	Depth- First	Depth- Limited	Iterative Deepening	Bidirectional (if applicable)
Complete? Time	$egin{aligned} \operatorname{Yes}^a \ O(b^d) \end{aligned}$	$\operatorname{Yes}^{a,b} O(b^{1+\lfloor C^*/\epsilon  floor})$	$egin{aligned} No \ O(b^m) \end{aligned}$	No $O(b^\ell)$	$egin{aligned} \operatorname{Yes}^a \ O(b^d) \end{aligned}$	$\operatorname{Yes}^{a,d} O(b^{d/2})$
Space Optimal?	$egin{aligned} O(b^d) \ \operatorname{Yes}^c \end{aligned}$	$O(b^{1+\lfloor C^*/\epsilon  floor})$ Yes	O(bm) No	$O(b\ell)$ No	$O(bd)$ Yes $^c$	$O(b^{d/2}) \ \operatorname{Yes}^{c,d}$

Evaluation of tree-search strategies. b is the branching factor; d is the depth of the shallowest solution; m is the maximum depth of the search tree; l is the depth limit. Superscript caveats are as follows: a complete if b is finite; b complete if step costs b for positive b; b optimal if step costs are all identical; d if both directions use breadth-first search.