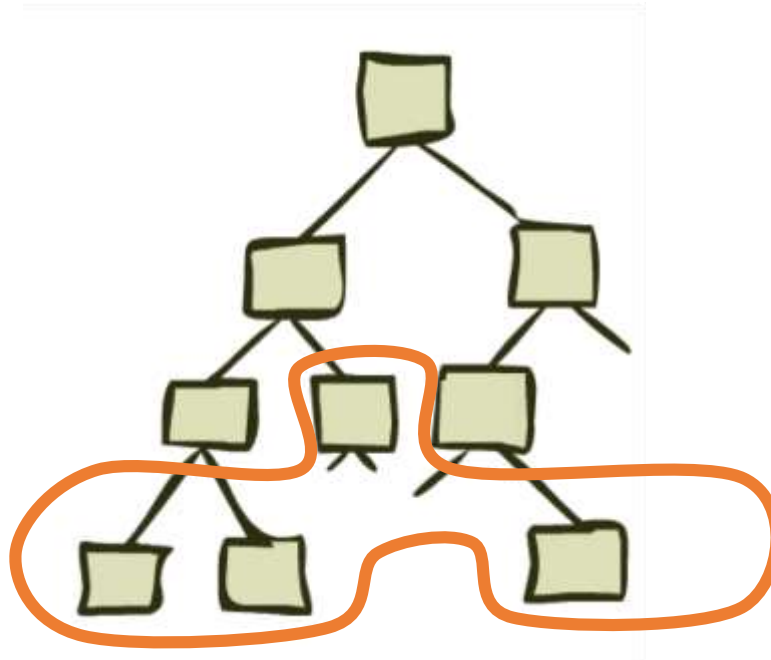


Uninformed Search-II

Depth First Search and its Variants



Previous Lecture

- Breadth First Search (BFS)
- Depth First Search (DFS)

Depth-Limited Search (DLS)

- **DFS with a depth bound**
 - Searching is not permitted beyond the depth bound.
- Works well if we know what is the depth of the solution.
- If the solution is beneath the depth bound, the search cannot find the goal (hence this search algorithm is **incomplete**).

Depth-Limited Search (DLS)

- **Main idea:**
 - *Expand node at the deepest level, but limit depth to D .*
- **Implementation:**
 - *Enqueue nodes in LIFO (last-in, first-out) order. But limit depth to D*
- *Complete?*
 - No
 - Yes: if there is a goal state at a depth less than D
- *Optimal?*
 - No
- **Time Complexity:**
 - $O(b^D)$, where D is the cutoff.
- **Space Complexity:**
 - $O(bD)$, where D is the cutoff.

Iterative Deepening Search

- To avoid the infinite depth problem of DFS:
 - Only search until depth L
 - i.e, don't expand nodes beyond depth L
 - Depth-Limited Search
- What if solution is deeper than L ?
 - Increase depth iteratively
 - Iterative Deepening Search
- IDS
 - Inherits the memory advantage of depth-first search
 - Has the completeness property of breadth-first search

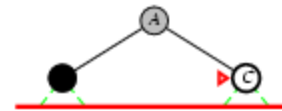
Iterative deepening search / =0

Limit = 0



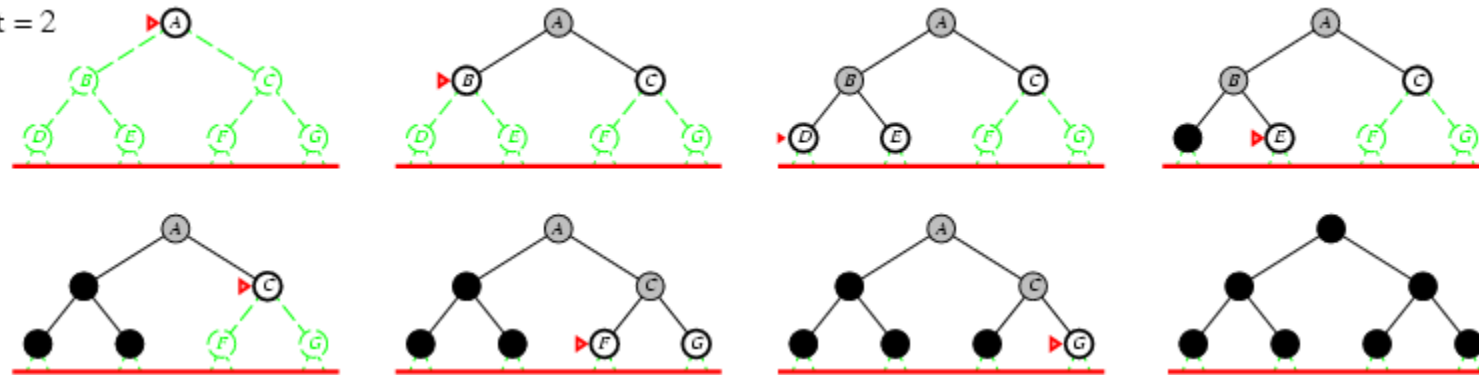
Iterative deepening search / =1

Limit = 1

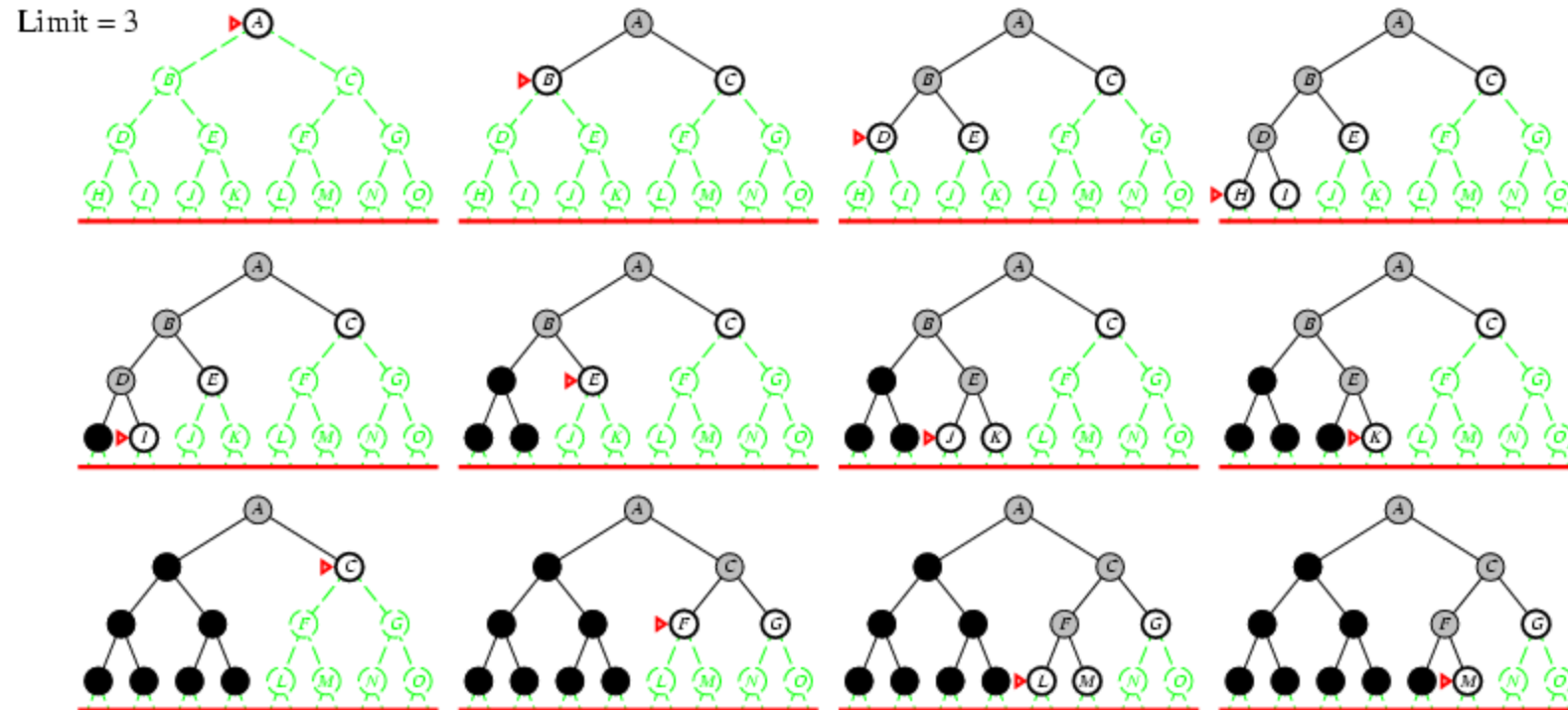


Iterative deepening search / =2

Limit = 2

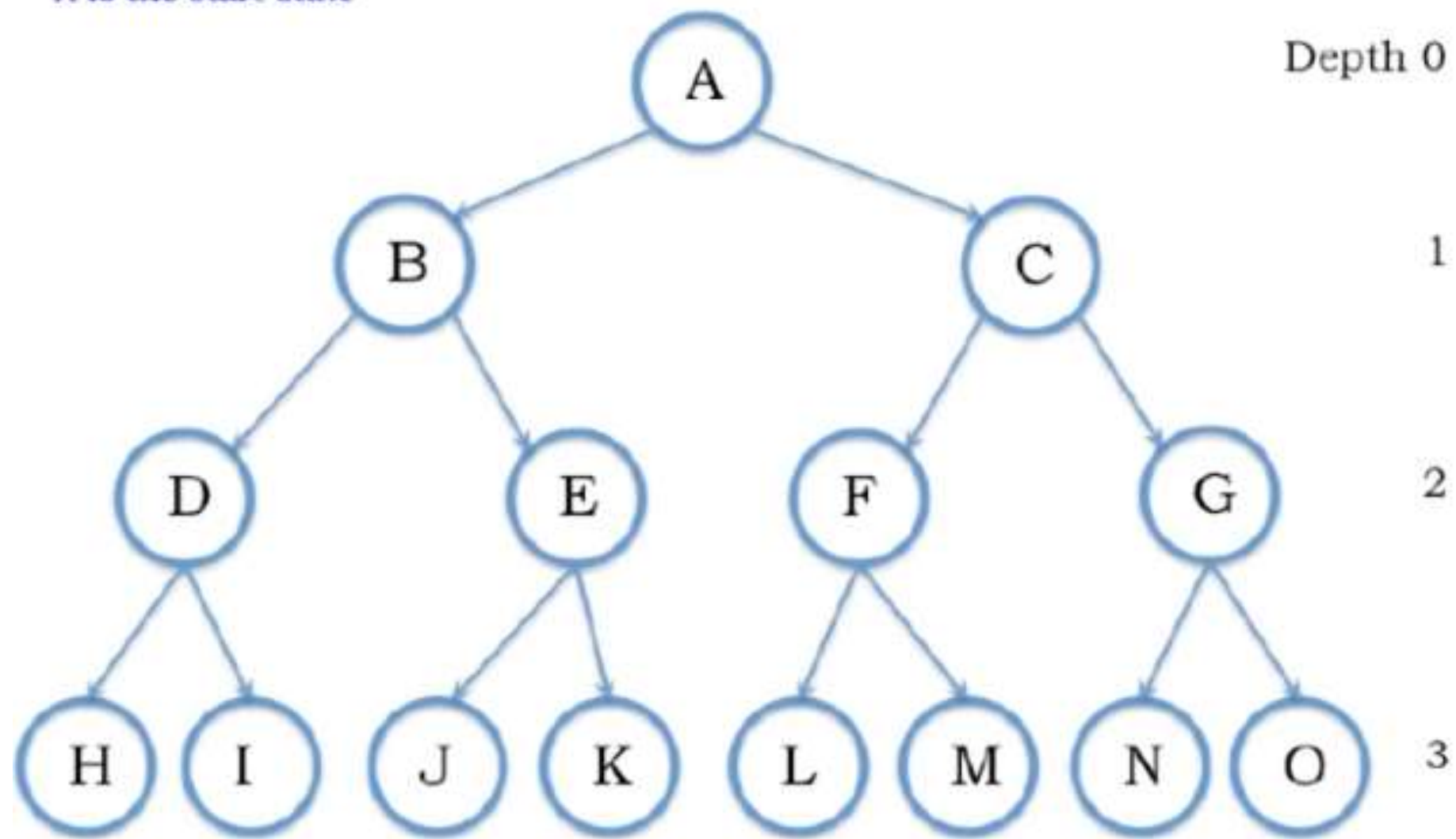


Iterative deepening search / =3

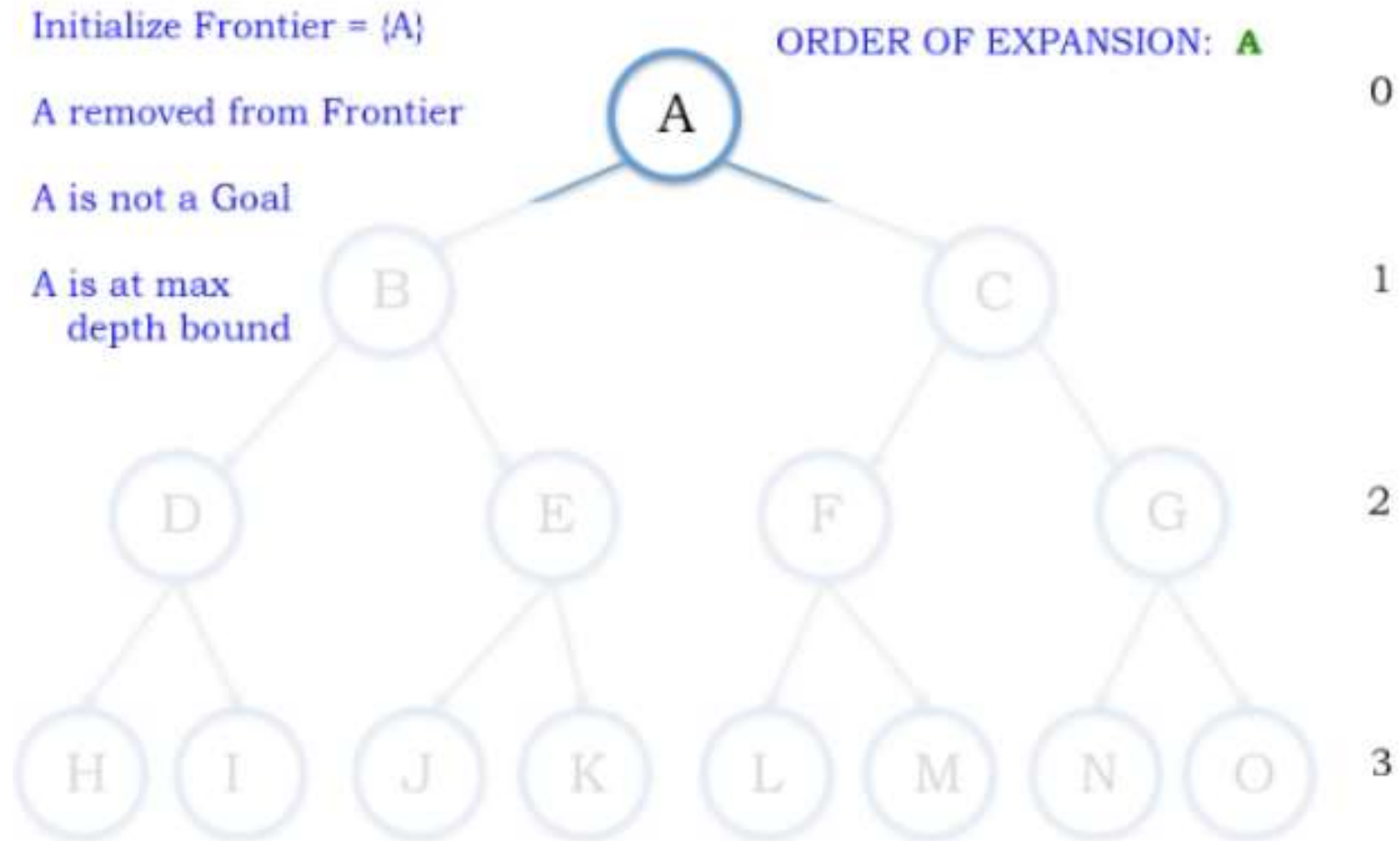


Operators (or actions) are not reversible
A is the start state

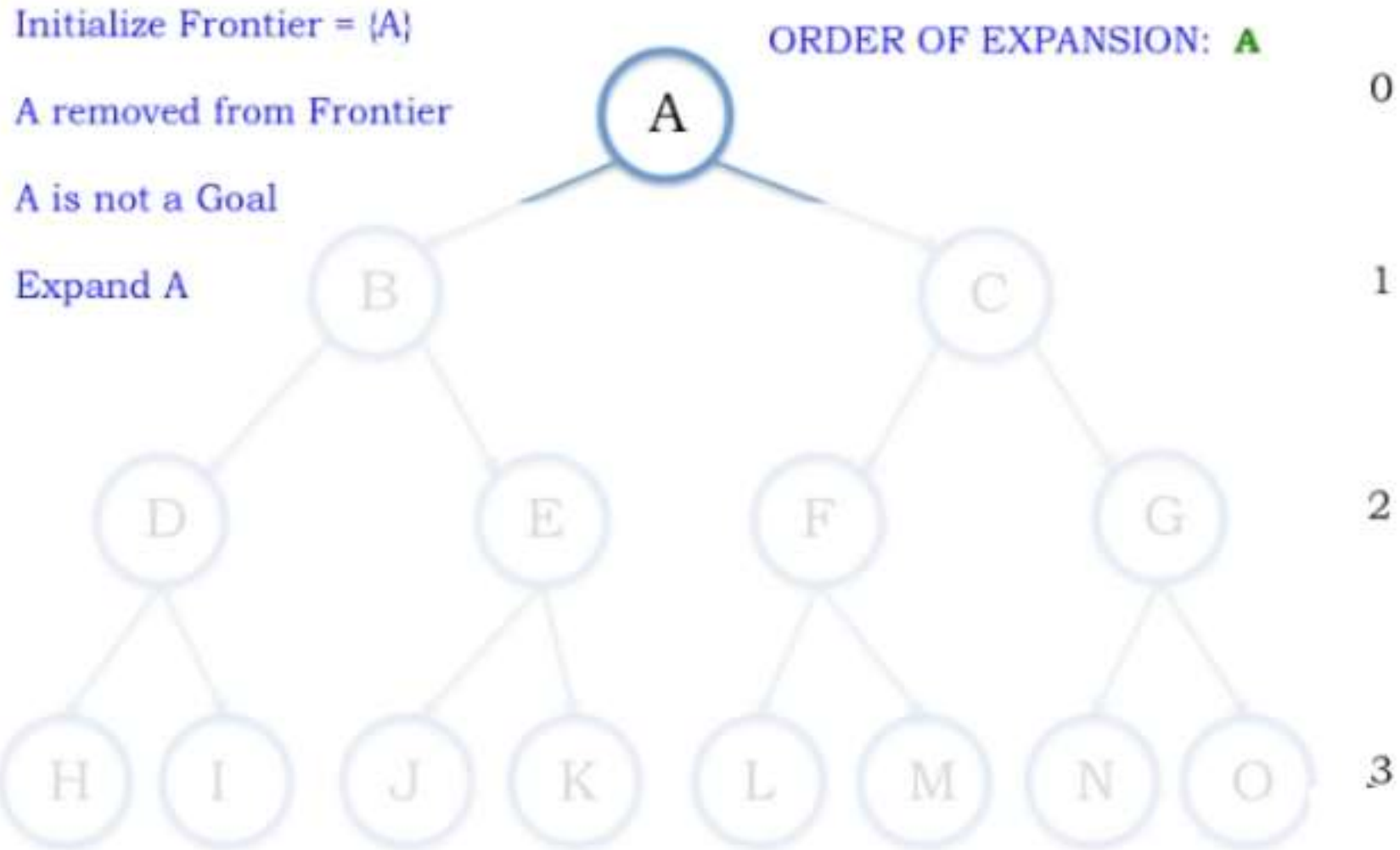
Repeated Depth Bounded searches



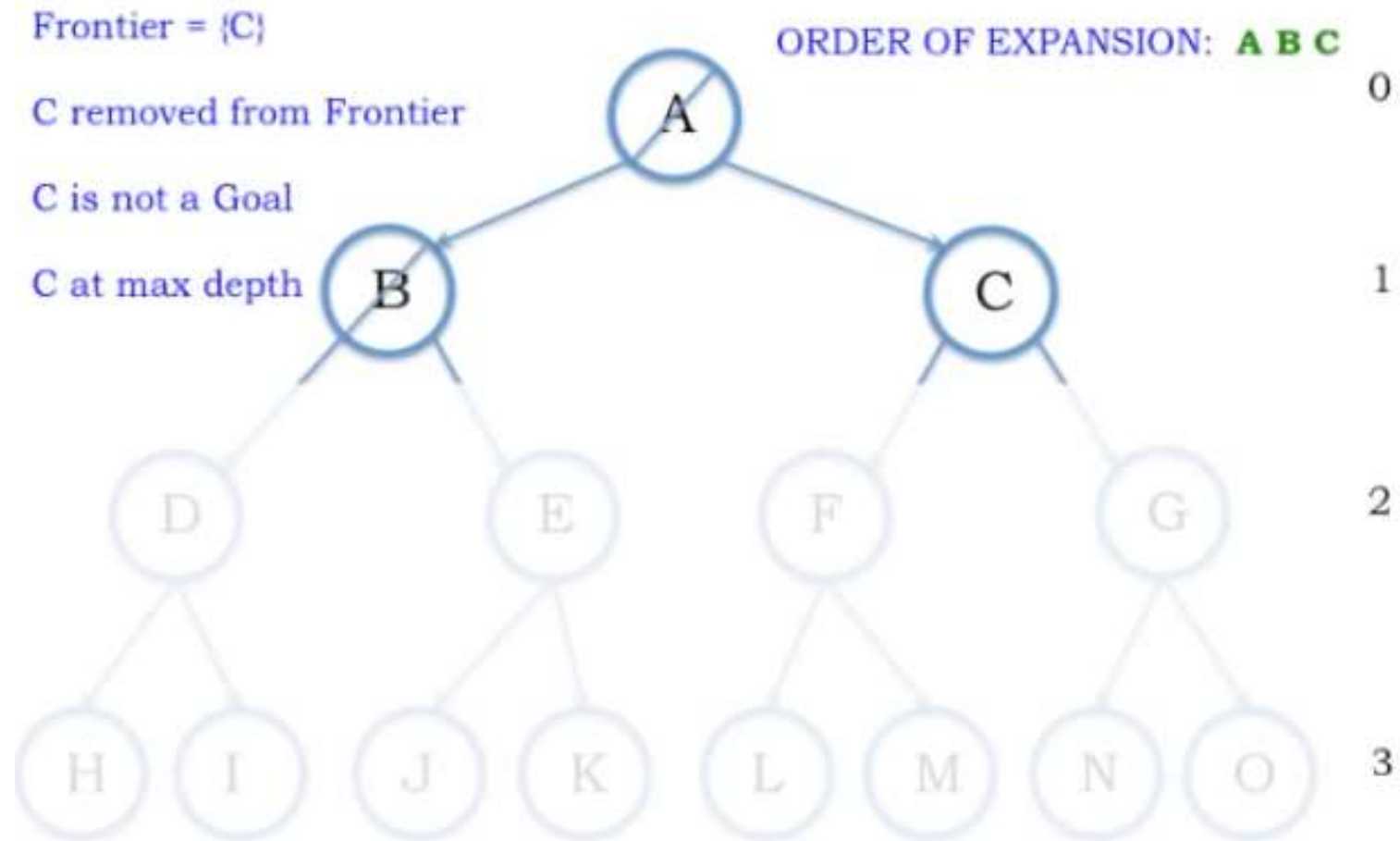
Depth Bound = 0



Depth Bound = 1



Depth Bound = 1



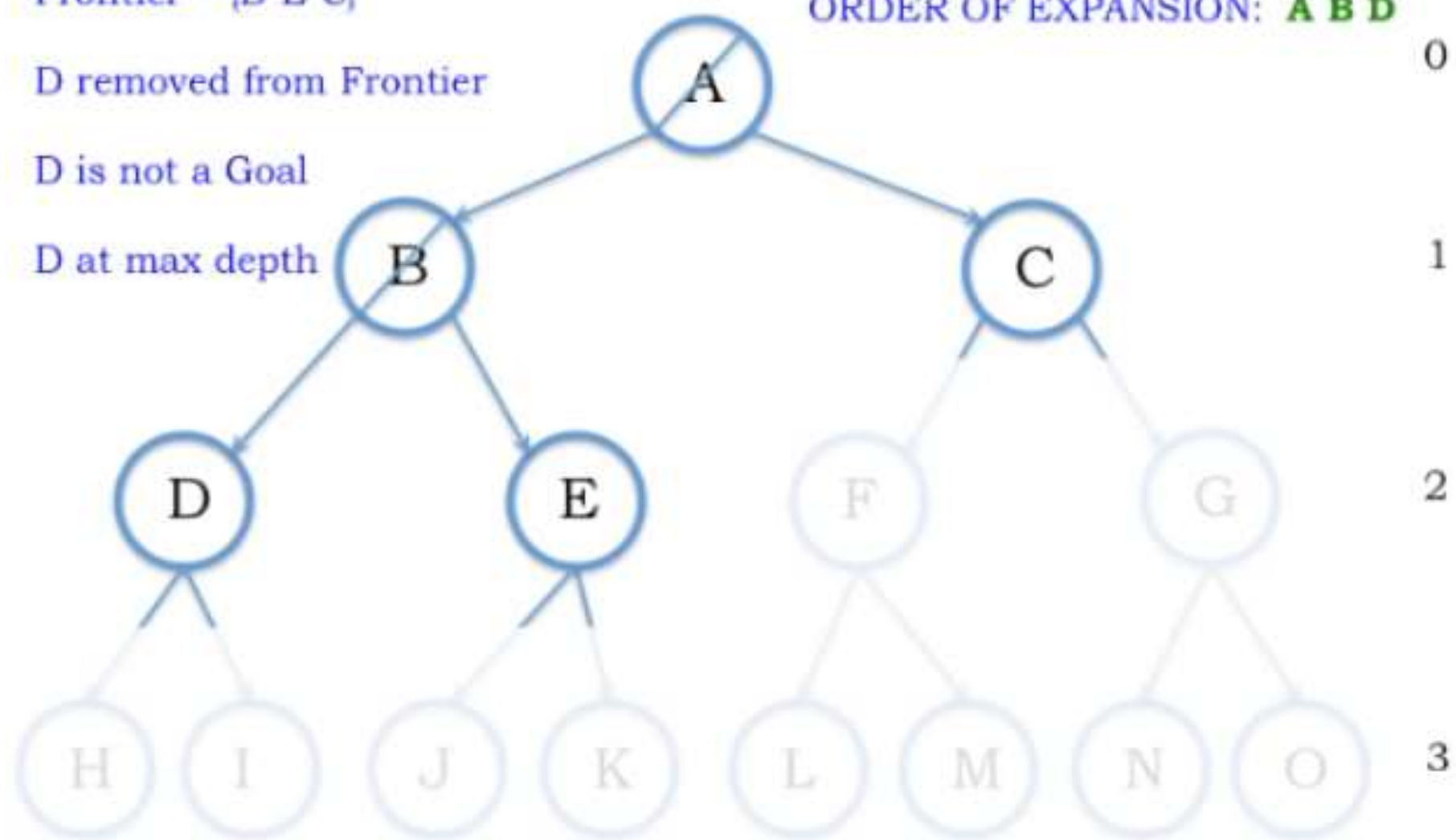
Frontier = {D E C}

D removed from Frontier

D is not a Goal

D at max depth

ORDER OF EXPANSION: **A B D**



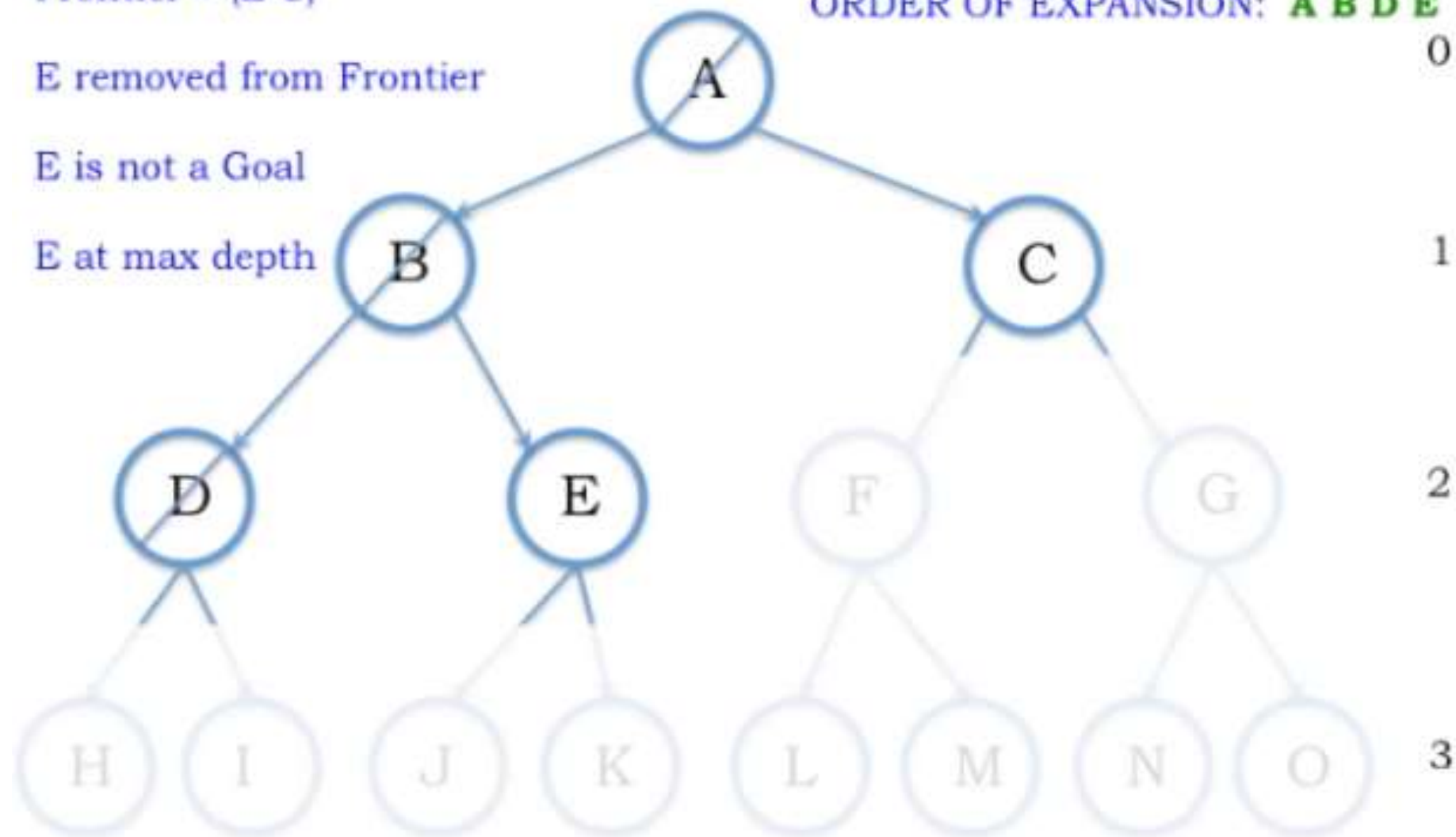
Frontier = {E C}

E removed from Frontier

E is not a Goal

E at max depth

ORDER OF EXPANSION: **A B D E**



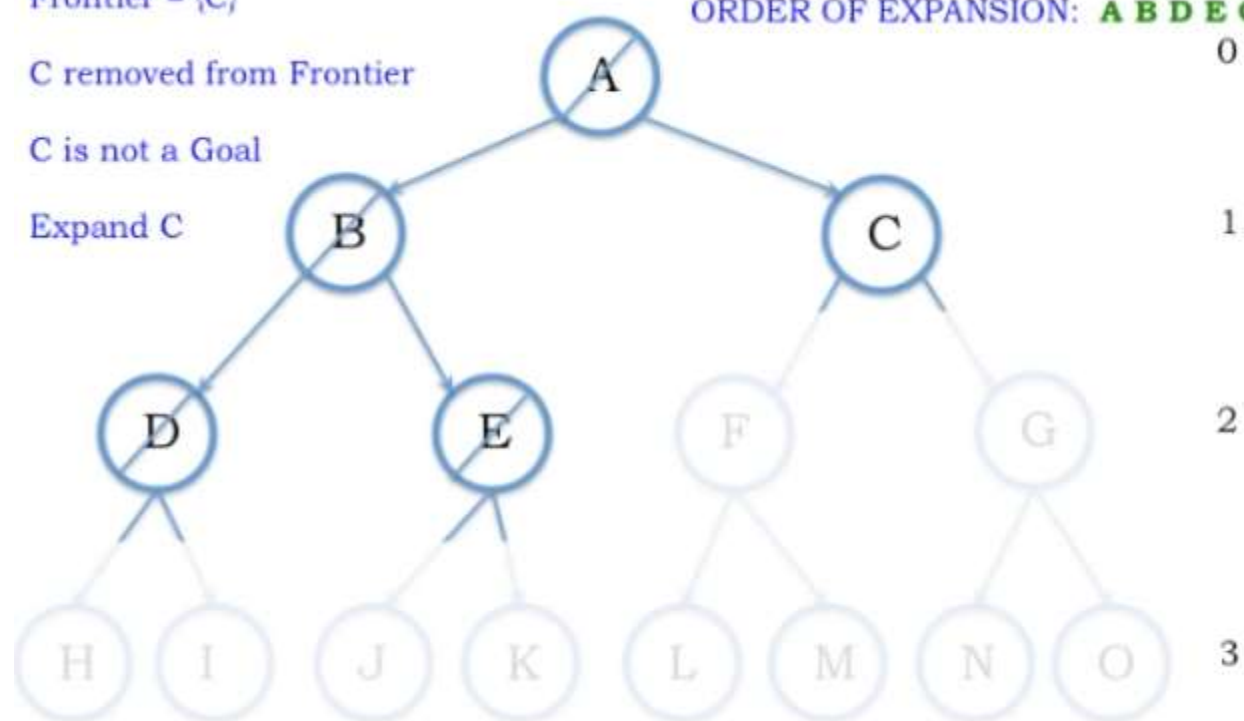
Frontier = {C}

C removed from Frontier

C is not a Goal

Expand C

ORDER OF EXPANSION: **A B D E C**



Frontier = {F G}

F removed from Frontier

F is not a Goal

F at max depth

ORDER OF EXPANSION:

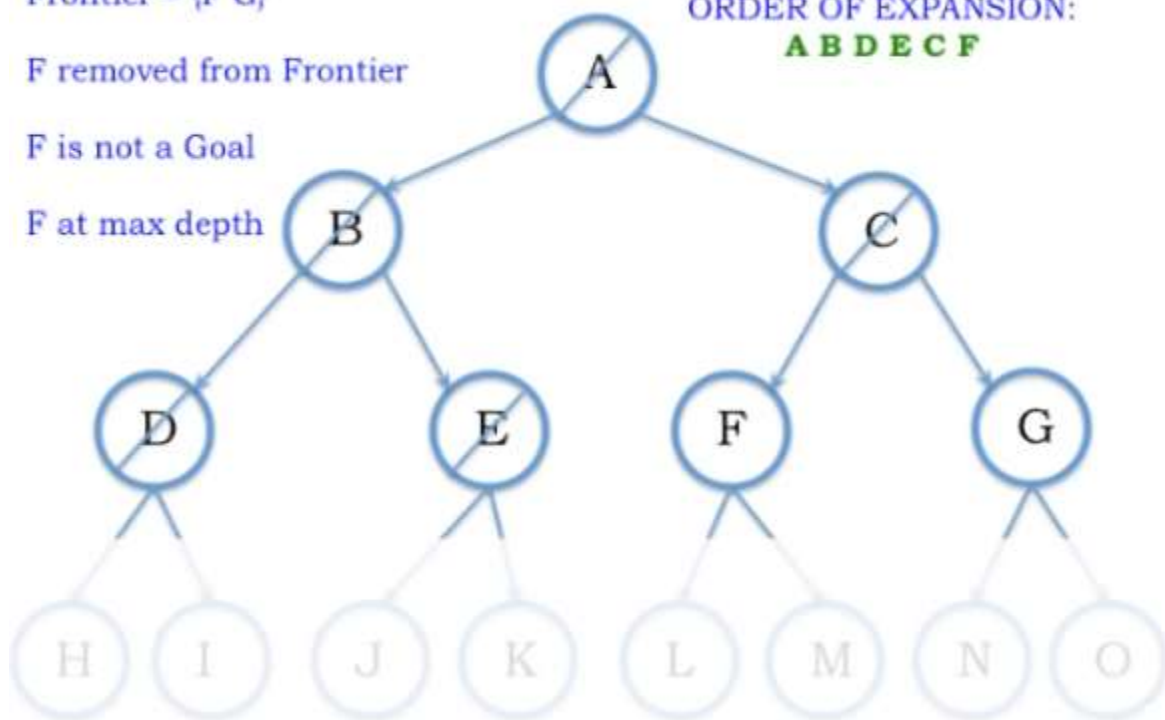
A B D E C F

0

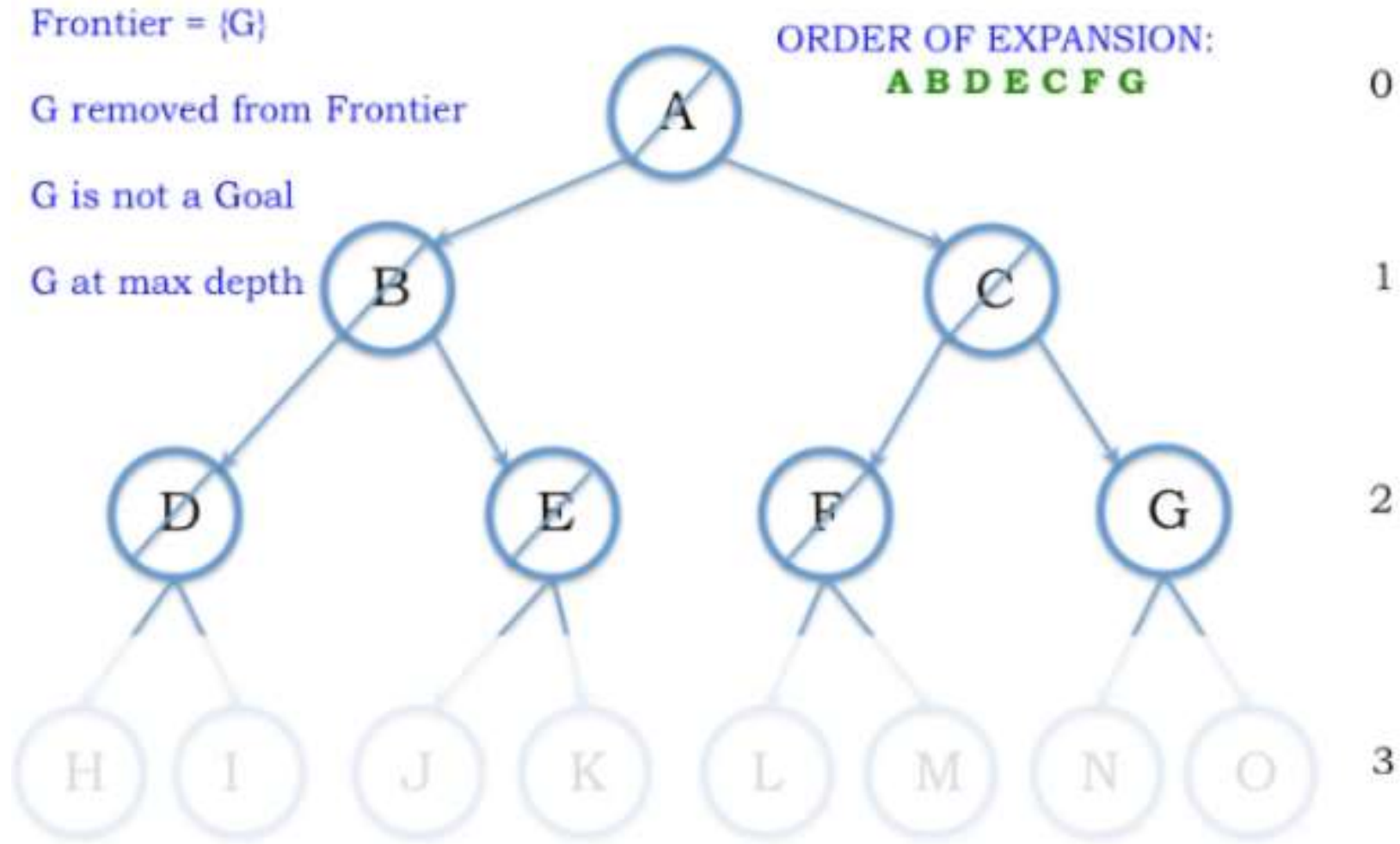
1

2

3

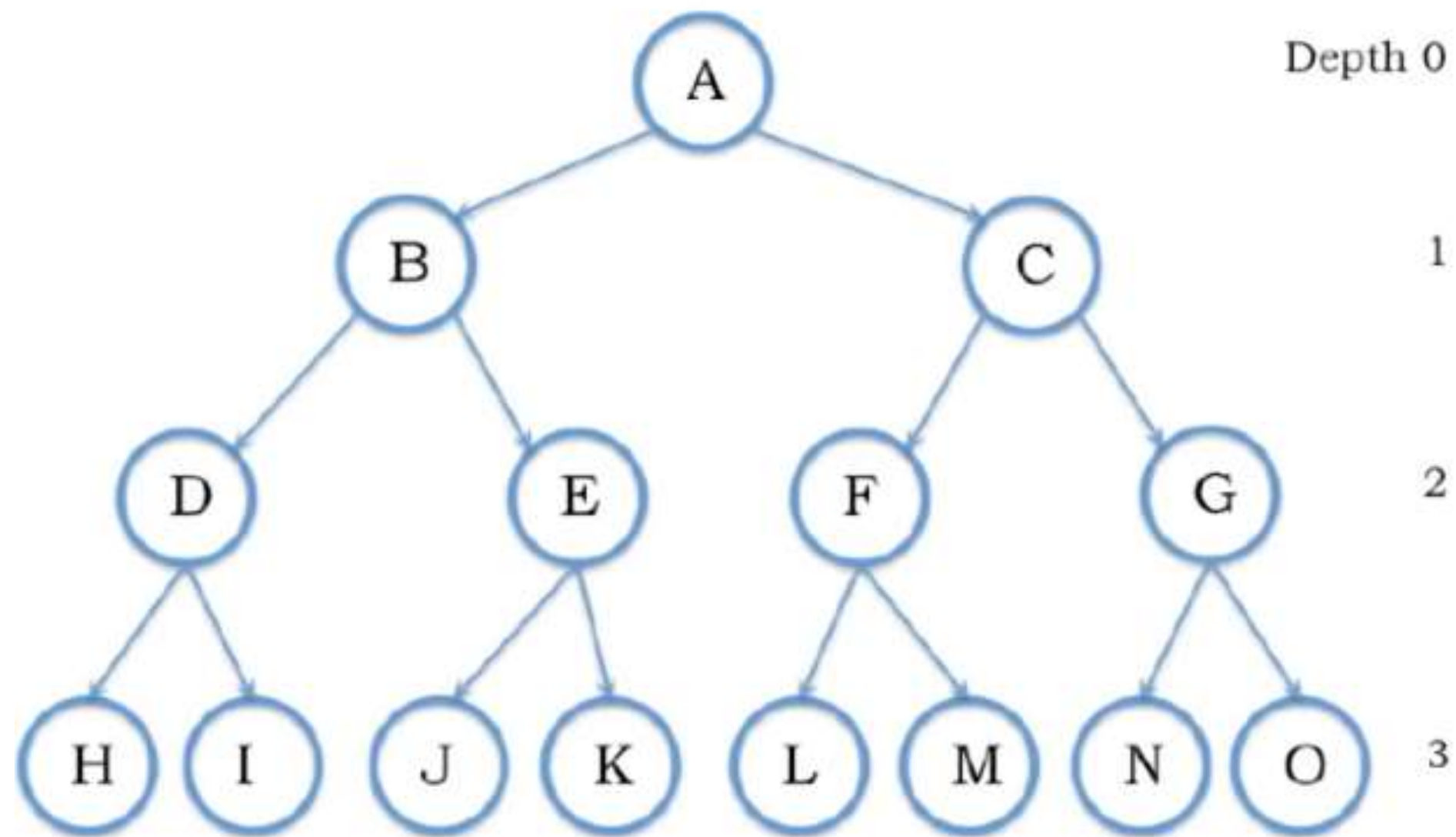


Depth 2



Over ALL the iterations, from depth bound 0 to 3, the order in which nodes removed from the frontier is:

A ABC ABDECFG ABDHIEJKCFLMGNO



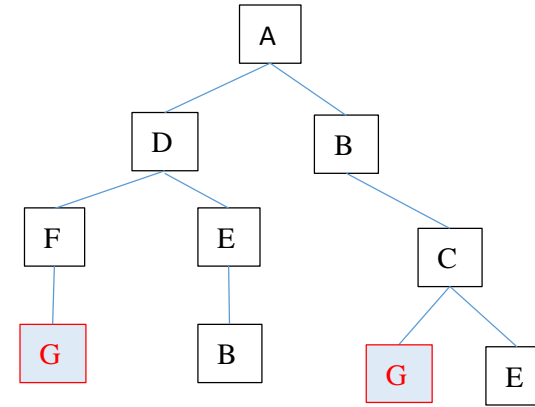
Properties of Iterative Deepening Search

- Complete? Yes (in finite spaces)
- Time? $O(b^d)$
- Space? $O(bd)$
- Optimal? Yes, (if step cost = 1 i.e. identical step cost)

IDS-Example- Assignment 1

Do it on notebook

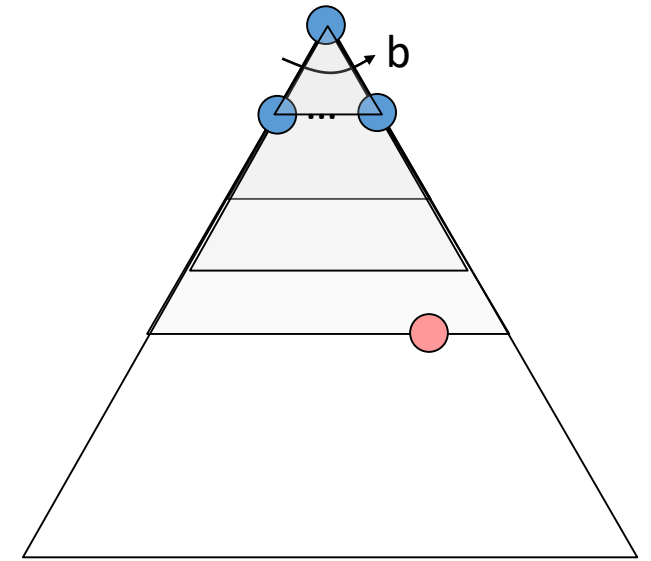
- Start Node: A
- Goal Node: G



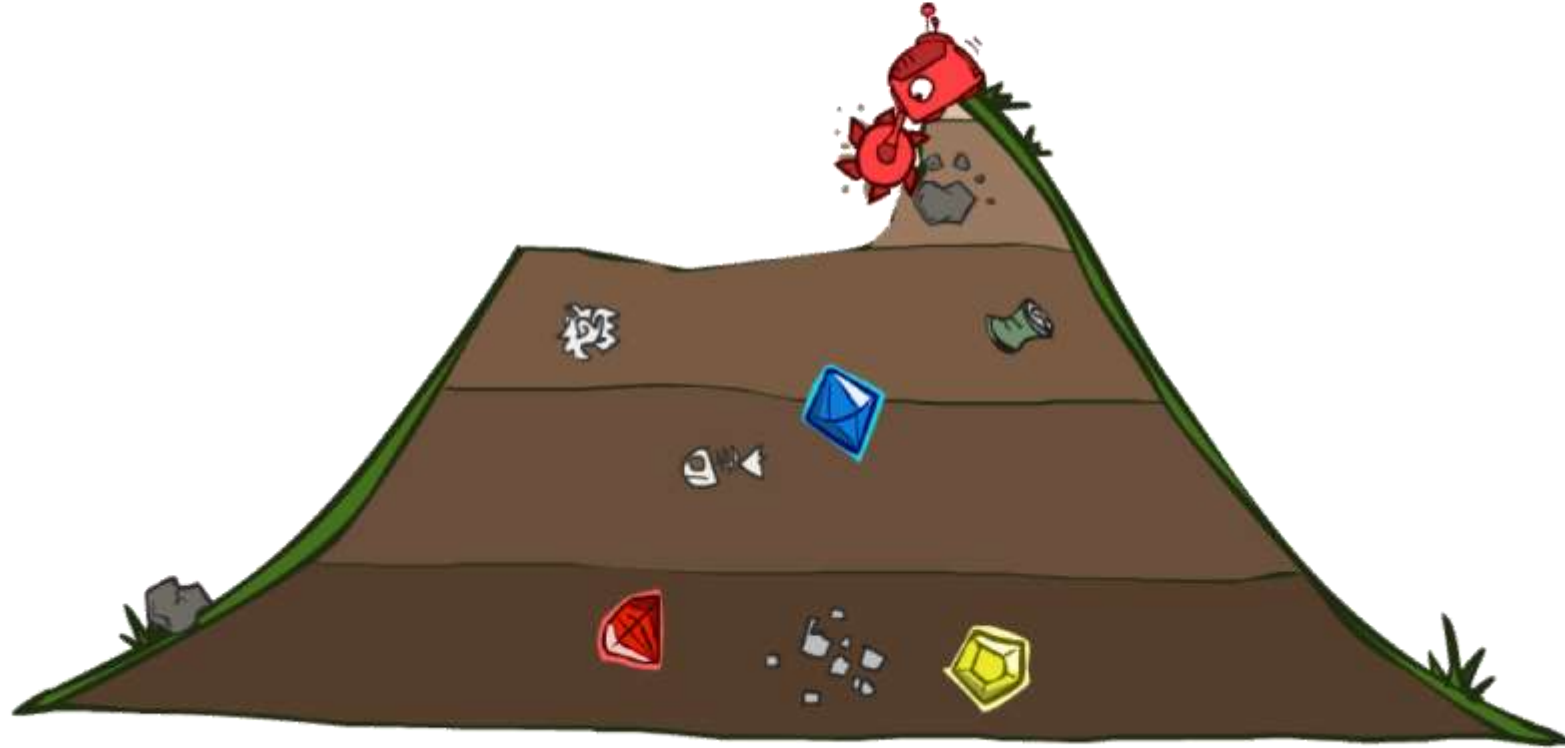
Step	Frontier	Expand[*]	Explored: a set of nodes
------	----------	-----------	--------------------------

IDS Redundancy

- Get DFS's space advantage with BFS's time / shallow-solution advantages
 - Run a DFS with depth limit 1. If no solution...
 - Run a DFS with depth limit 2. If no solution...
 - Run a DFS with depth limit 3.
- Isn't that wastefully redundant?
 - Generally most work happens in the lowest level searched, so not so bad!



Uniform Cost Search



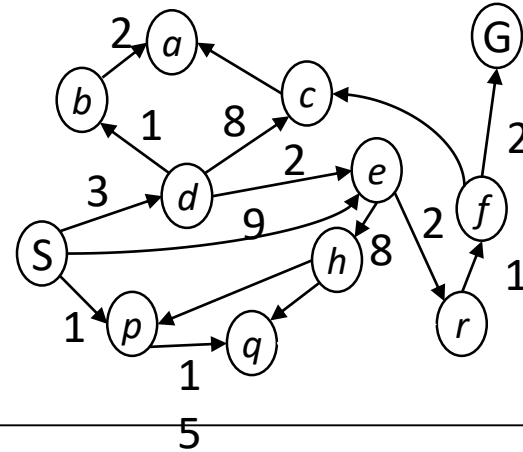
Uniform Cost Search

- Breadth-First Search find the shallowest goal state, but this may not always be the least-cost solution.
- Uniform-Cost Search modifies the Breadth-First Search strategy by always expanding the lowest path cost $g(n)$ node on the fringe.
- *Frontier* is a priority queue, i.e., new successors are merged into the queue sorted by $g(n)$.
 - Can remove successors already on queue w/higher $g(n)$.
 - Saves memory and time cost.

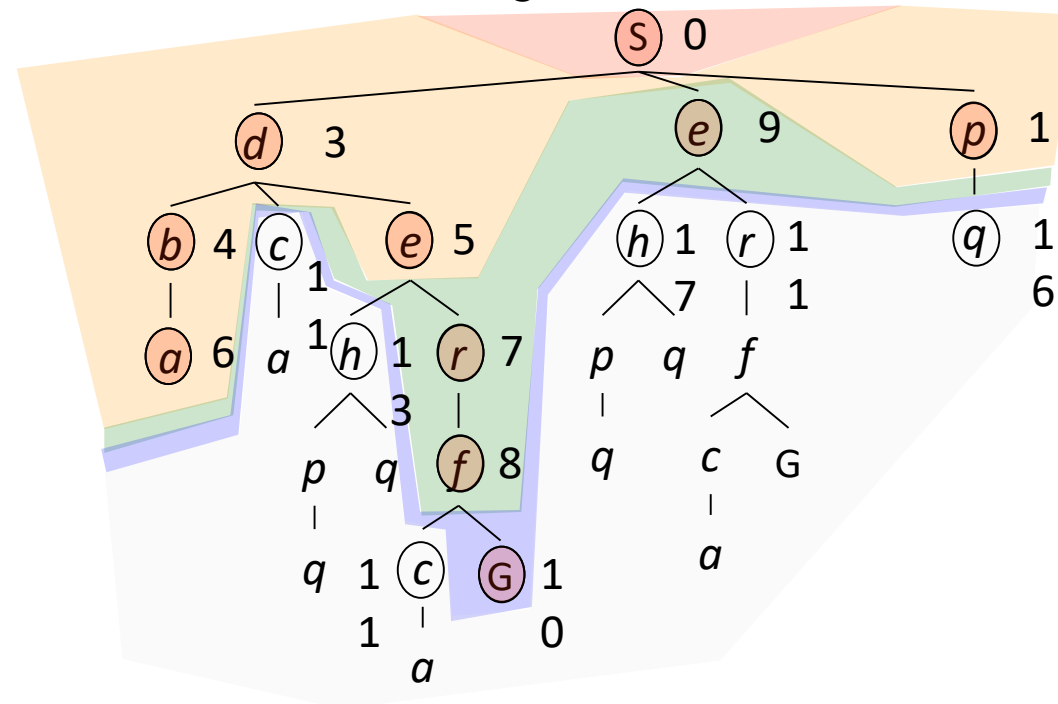
Uniform Cost Search

*Strategy: expand a
cheapest node first:*

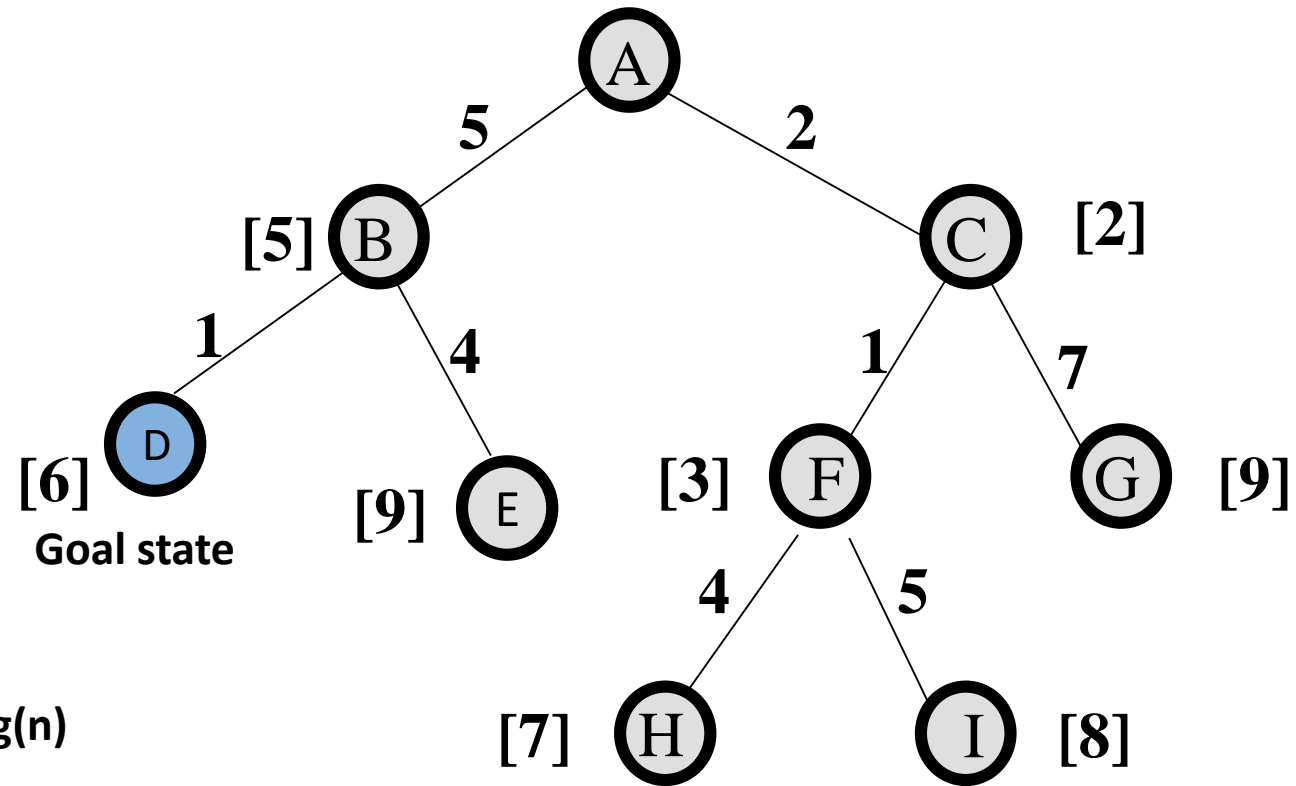
*Fringe is a priority
queue (priority:
cumulative cost)*



Cost
contours



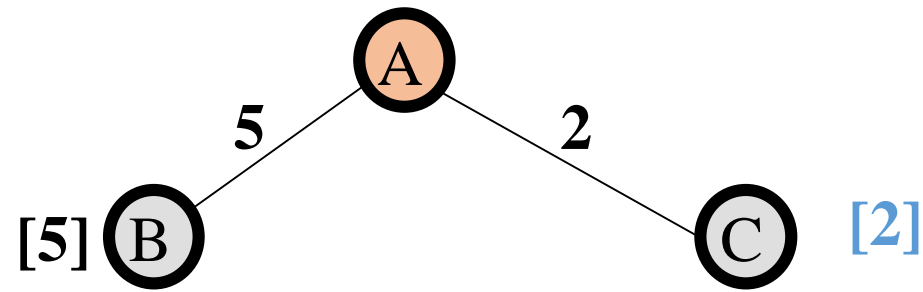
Uniform Cost Search (UCS)



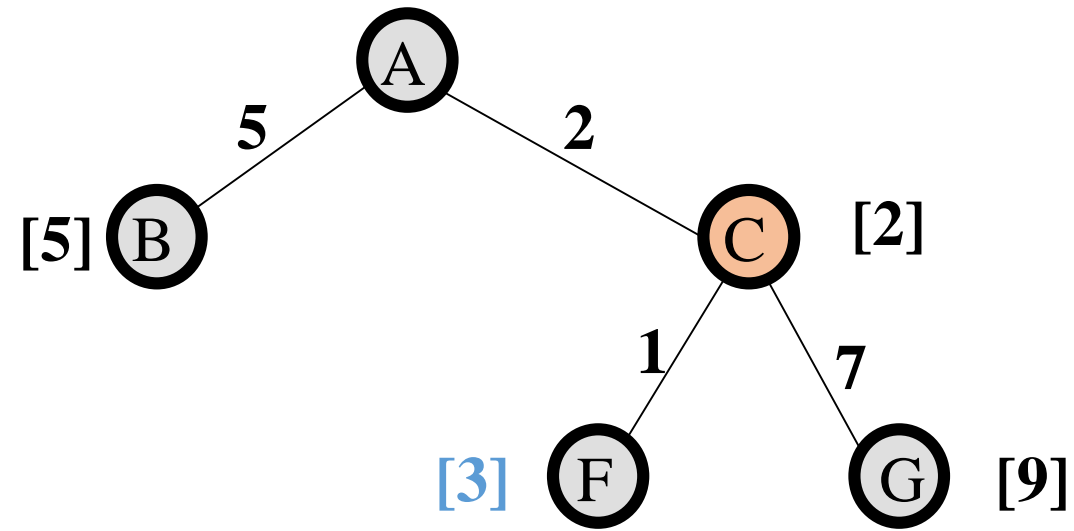
$[x] = g(n)$

path cost of node n

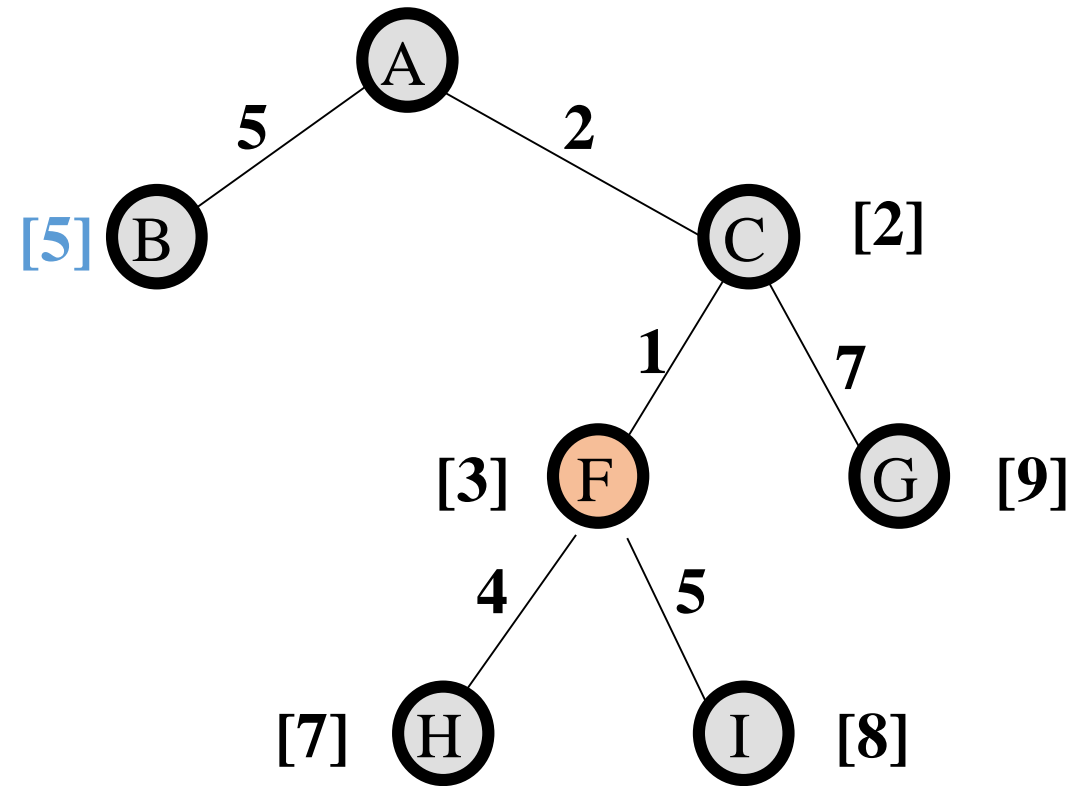
Uniform Cost Search (UCS)



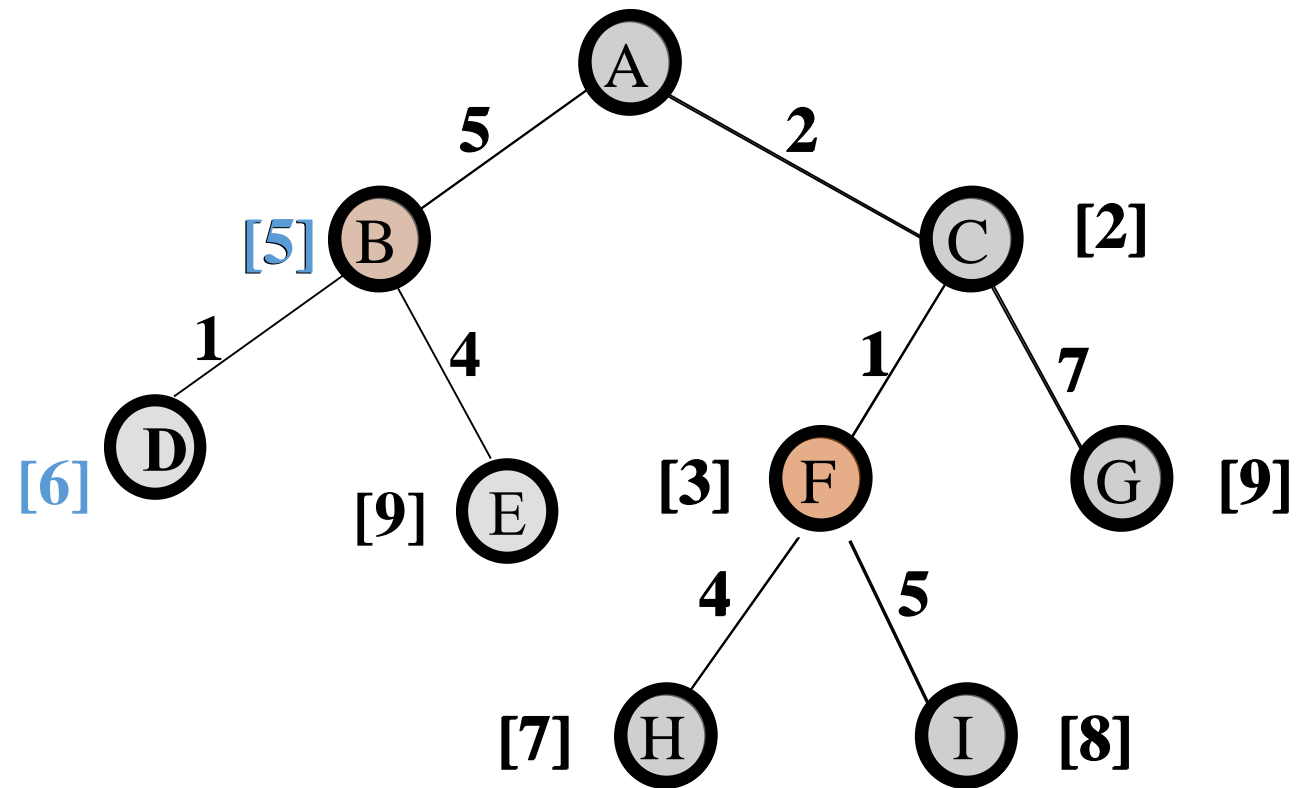
Uniform Cost Search (UCS)



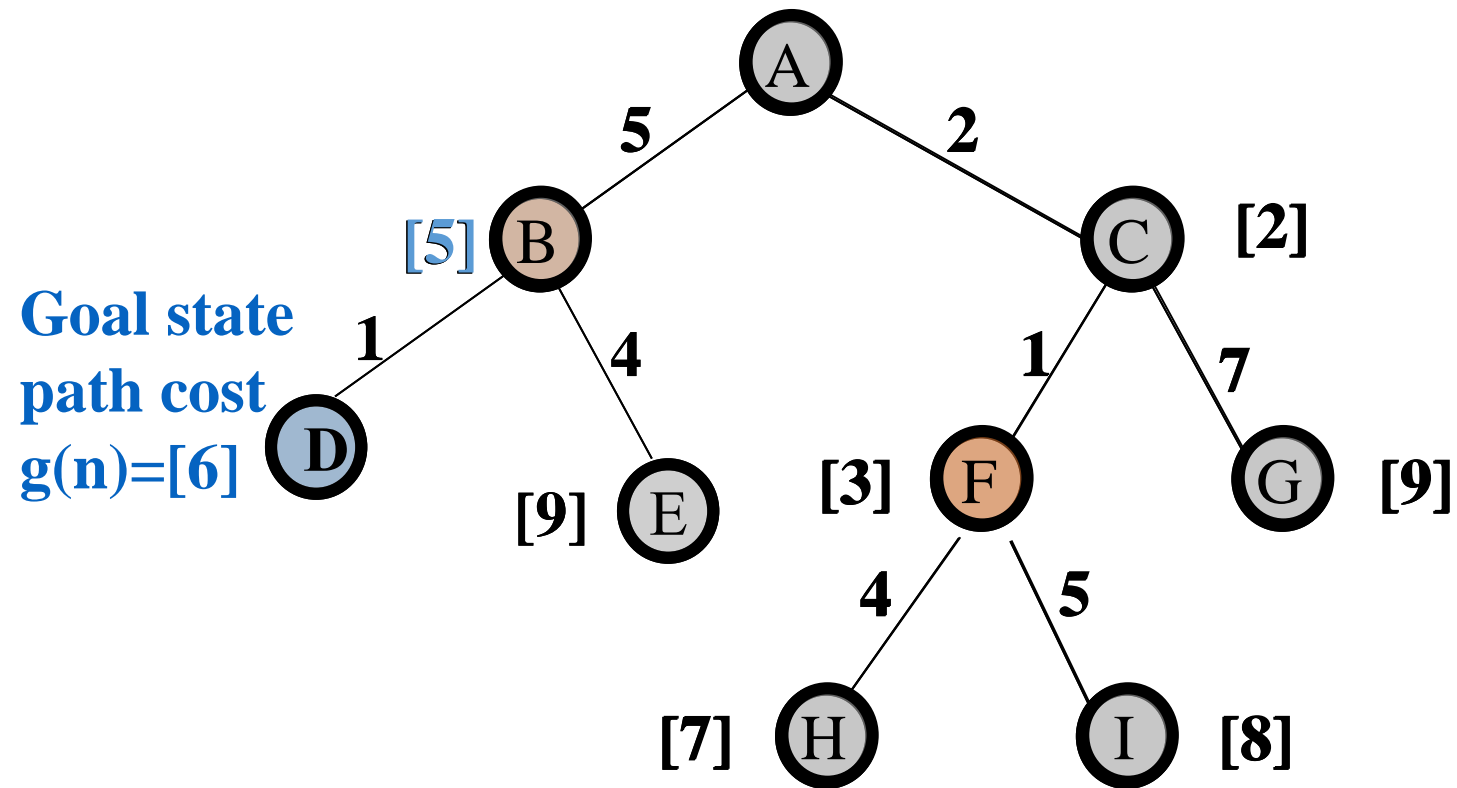
Uniform Cost Search (UCS)



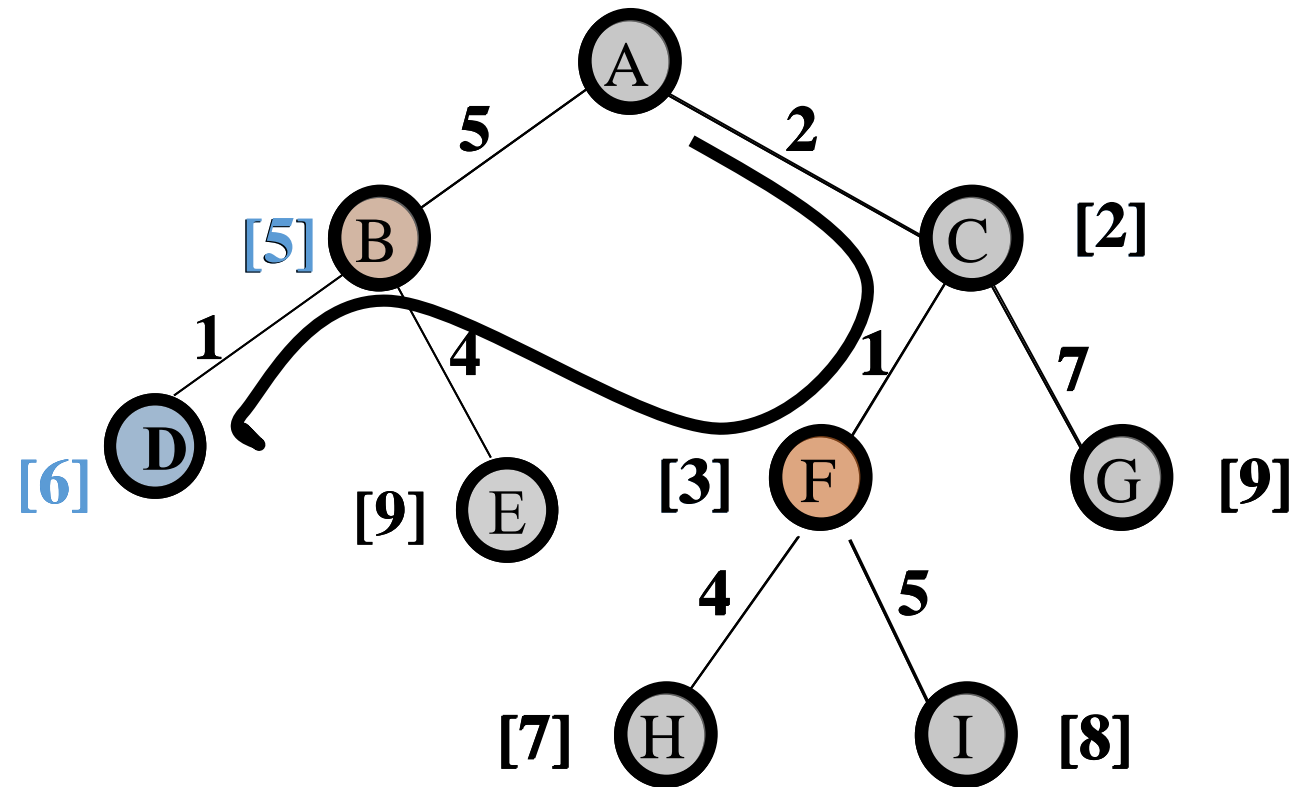
Uniform Cost Search (UCS)

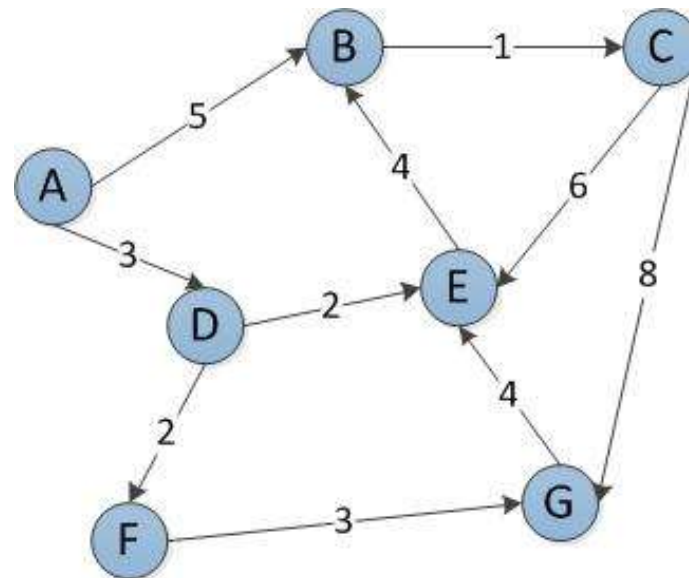


Uniform Cost Search (UCS)



Uniform Cost Search (UCS)





- Start Node: A
- Goal Node: G

Step	Frontier	Expand[*]	Explored: a set of nodes
1	{(A,0)}	A	∅
2	{(A-D,3),(A-B,5)}	D	{A}
3	{(A-B,5),(A-D-E,5),(A-D-F,5)}	B	{A,D}
4	{(A-D-E,5),(A-D-F,5),(A-B-C,6)}	E	{A,D,B}
5	{(A-D-F,5),(A-B-C,6)}[*]	F	{A,D,B,E}
6	{(A-B-C,6),(A-D-F-G,8)}	C	{A,D,B,E,F}
7	(A-D-F-G,8)[+]	G	{A,D,B,E,F,C}
8	∅		

- Found the path: A -> D -> F -> G.
- *B is not added to the frontier because it is found in the explored set.
- +G and E are not added to the frontier because they are found in explored and frontier

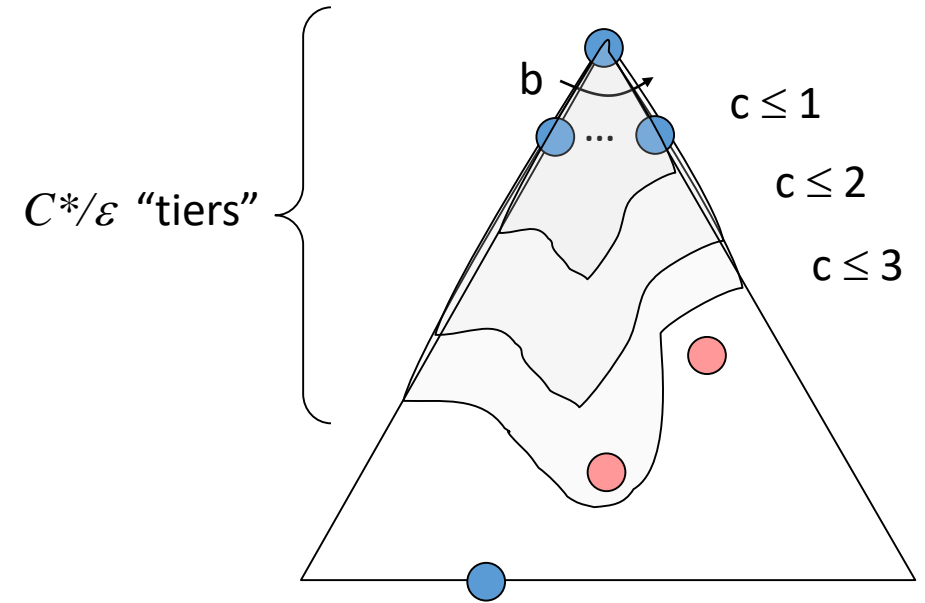
Uniform Cost Search

```
function UNIFORM-COST-SEARCH(problem) returns a solution, or failure
  node  $\leftarrow$  a node with STATE = problem.INITIAL-STATE, PATH-COST = 0
  frontier  $\leftarrow$  a priority queue ordered by PATH-COST, with node as the only element
  explored  $\leftarrow$  an empty set
  loop do
    if EMPTY?(frontier) then return failure
    node  $\leftarrow$  POP(frontier) /* chooses the lowest-cost node in frontier */
    if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
    add node.STATE to explored
    for each action in problem.ACTIONS(node.STATE) do
      child  $\leftarrow$  CHILD-NODE(problem, node, action)
      if child.STATE is not in explored or frontier then
        frontier  $\leftarrow$  INSERT(child, frontier)
      else if child.STATE is in frontier with higher PATH-COST then
        replace that frontier node with child
```

Figure 3.14 Uniform-cost search on a graph. The algorithm is identical to the general graph search algorithm in Figure 3.7, except for the use of a priority queue and the addition of an extra check in case a shorter path to a frontier state is discovered. The data structure for *frontier* needs to support efficient membership testing, so it should combine the capabilities of a priority queue and a hash table.

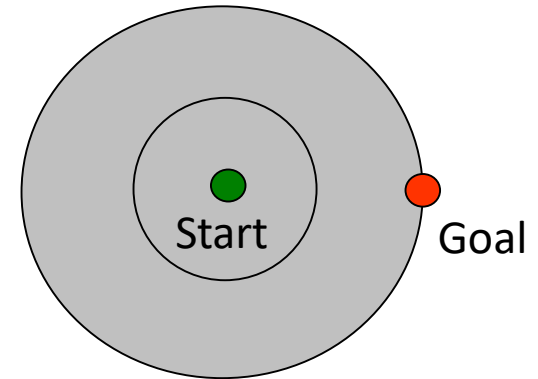
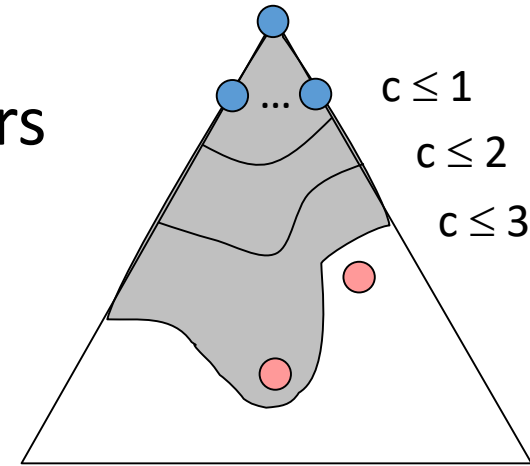
Uniform Cost Search (UCS) Properties

- What nodes does UCS expand?
 - Processes all nodes with cost less than cheapest solution!
 - If that solution costs C^* and arcs cost at least ε , then the “effective depth” is roughly C^*/ε
 - Takes time $O(b^{C^*/\varepsilon})$ (exponential in effective depth)
- How much space does the fringe take?
 - Has roughly the last tier, so $O(b^{C^*/\varepsilon})$
- Is it complete?
 - Assuming best solution has a finite cost and minimum arc cost is positive, yes!
- Is it optimal?
 - Yes! (Proof next lecture via A^*)

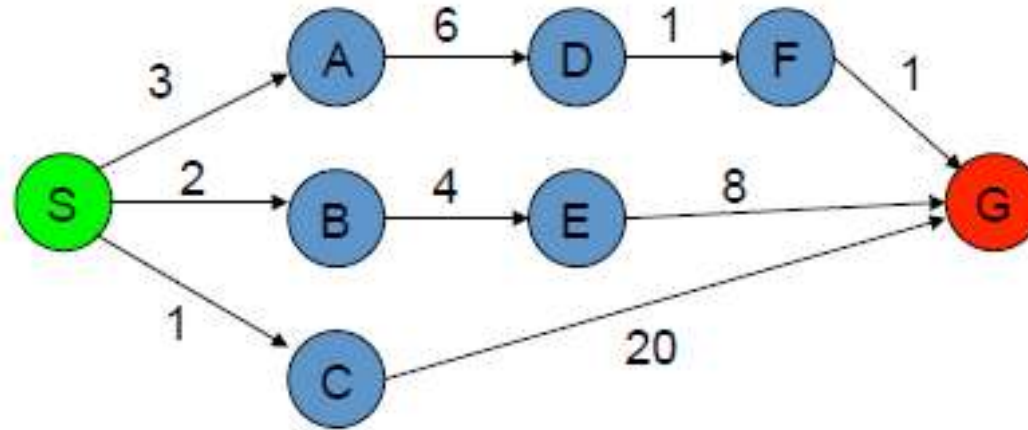


Uniform Cost Issues

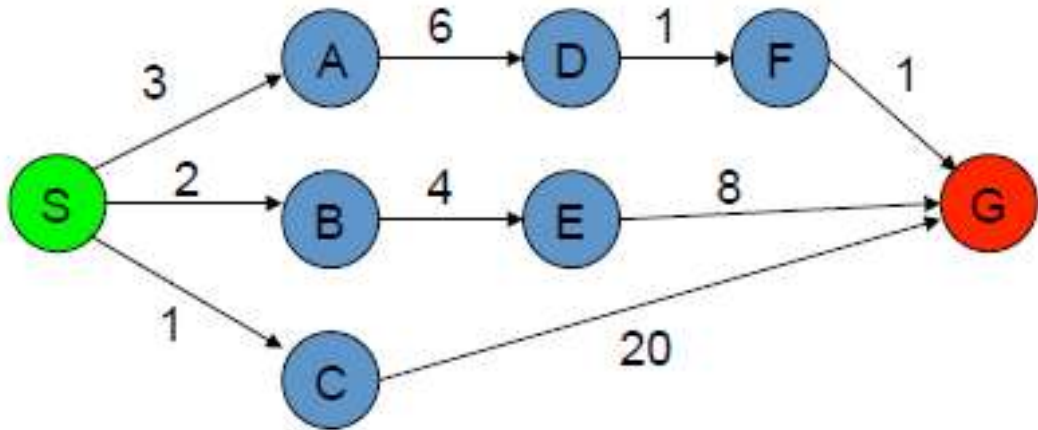
- Remember: UCS explores increasing cost contours
- The good: UCS is complete and optimal!
- The bad:
 - Explores options in every “direction”
 - No information about goal location
- We’ll fix that soon!



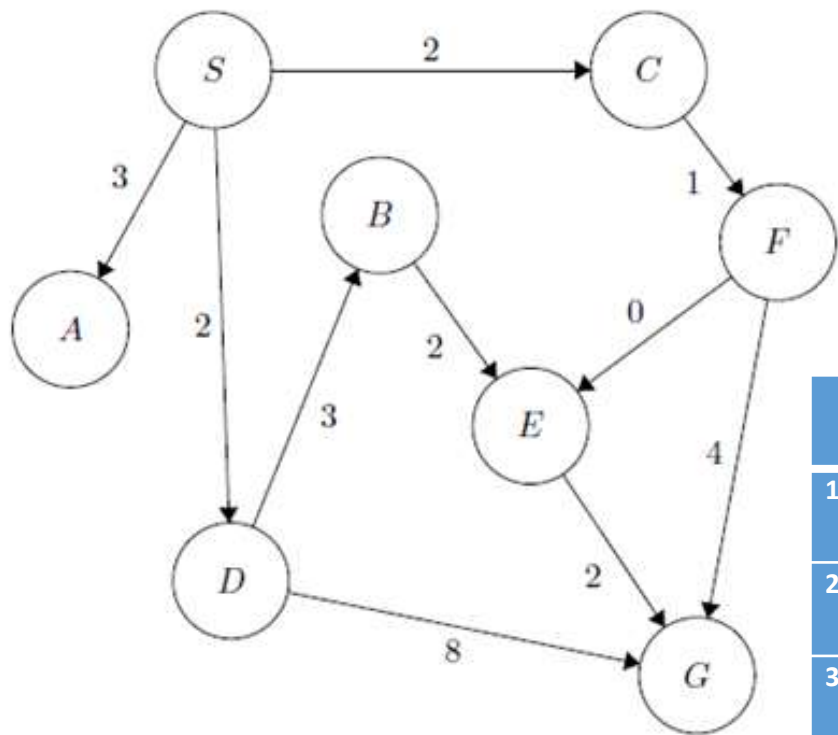
Uniform Cost Search



- The graph above shows the step-costs for different paths going from the start (S) to the goal (G).
- Use uniform cost search to find the optimal path to the goal.



	Frontier	Expand	Explored
1	S	S	Empty
2	(S-C,1) (S-B,2)(S-A,3)	C	S
3	(S-B,2)(S-A,3)(S-C-G,21)	B	S,C
4	(S-A,3)(S-C-G,21)(S-B-E,6)	A	S,C,B
5	(S-C-G,21)(S-B-E,6)(S-A-D,9)	E	S,C,B,A
6	(S-C-G,21) (S-A-D,9) (S-B-E-G,14)	D	S,C,B,A,E
7	(S-B-E-G,14) (S-A-D-F,10)	F	S,C,B,A,E,D
8	(S-B-E-G,14) (S-A-D-F-G,11)	G	S,C,B,A,E,D,F
9	S-A-D-F-G,11 Goal Found		S,C,B,A,E,D,F,G

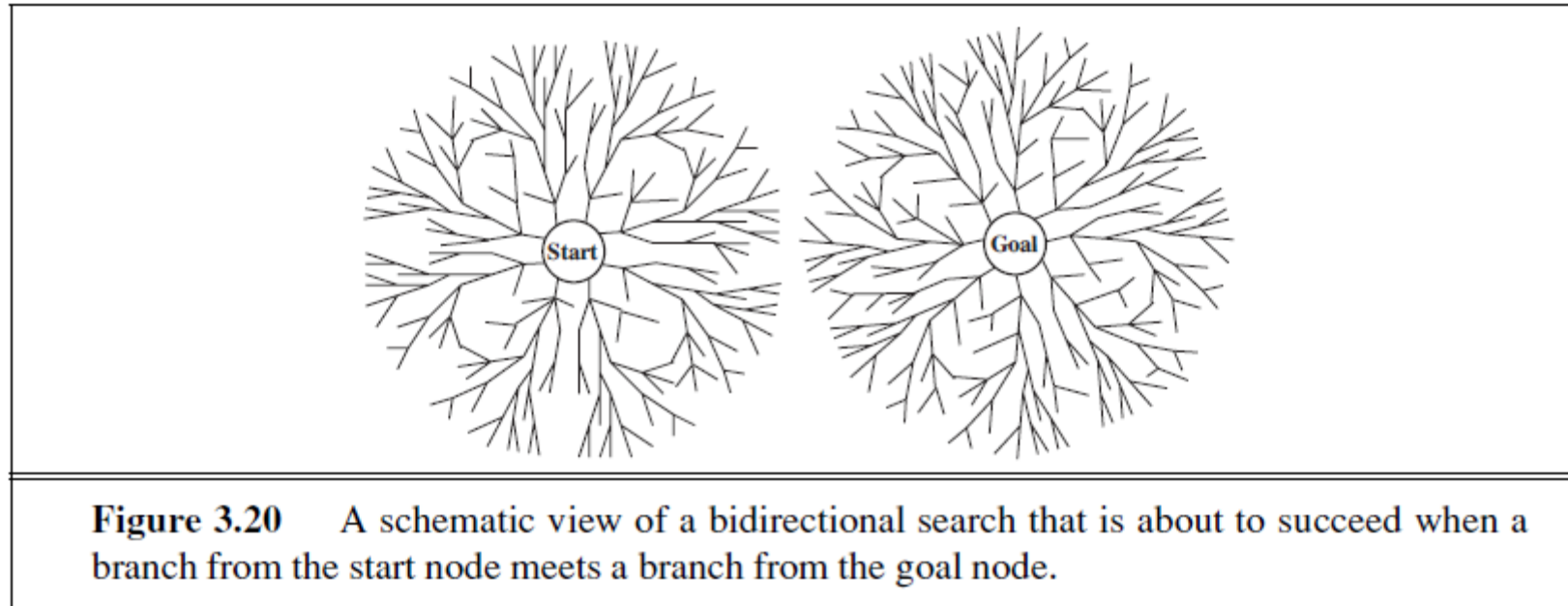


Step	Frontier	Expand	Explored: a set of Nodes
1	(S,0)	S	∅
2	(S-C,2) (S-D,2) (S-A,3)	C	S
3	(S-D,2) (S-A,3)(S-C-F,3)	D	S,C
4	(S-A,3)(S-C-F,3)(S-D-B,5)(S-D-G,10)	A	S,C,D
5	(S-C-F,3)(S-D-B,5)(S-D-G,10)	F	S,C,D,A
6	(S-D-B,5)(S-D-G,10)(S-C-F-E,3)(S-C-F-G,7)	E	S,C,D,A,F
7	(S-D-B,5)(S-C-F-G,7)(S-C-F-E-G,5)	B	S,C,D,A,F,E
8	(S-C-F-E-G,5)*	G	S,C,D,A,F,E,B
9	(S-C-F-E-G,5) Goal Found		

Bidirectional Search

- Idea
 - simultaneously search forward from S and backwards from G (Data driven Vs Goal Driven)
 - 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
- what if there are multiple goal states?
 - what if there is only a goal test, no explicit list?
- Complexity
 - Time complexity is best: $O\left(2b^{(d/2)}\right) = O\left(b^{(d/2)}\right)$
 - memory complexity is the same

Bidirectional Search



Summary of Algorithms

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

b branching factor

d depth of the shallowest solution

m maximum depth of the search tree

ℓ depth limit

Superscripts:

a complete if b is finite

b complete if step costs \geq epsilon for +ve epsilon

c optimal if step costs are all identical

Summary of Algorithms

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

Figure 3.21 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 $\geq \epsilon$ for positive ϵ ; ^c optimal if step costs are all identical; ^d if both directions use breadth-first search.