Blind (Uninformed) Search

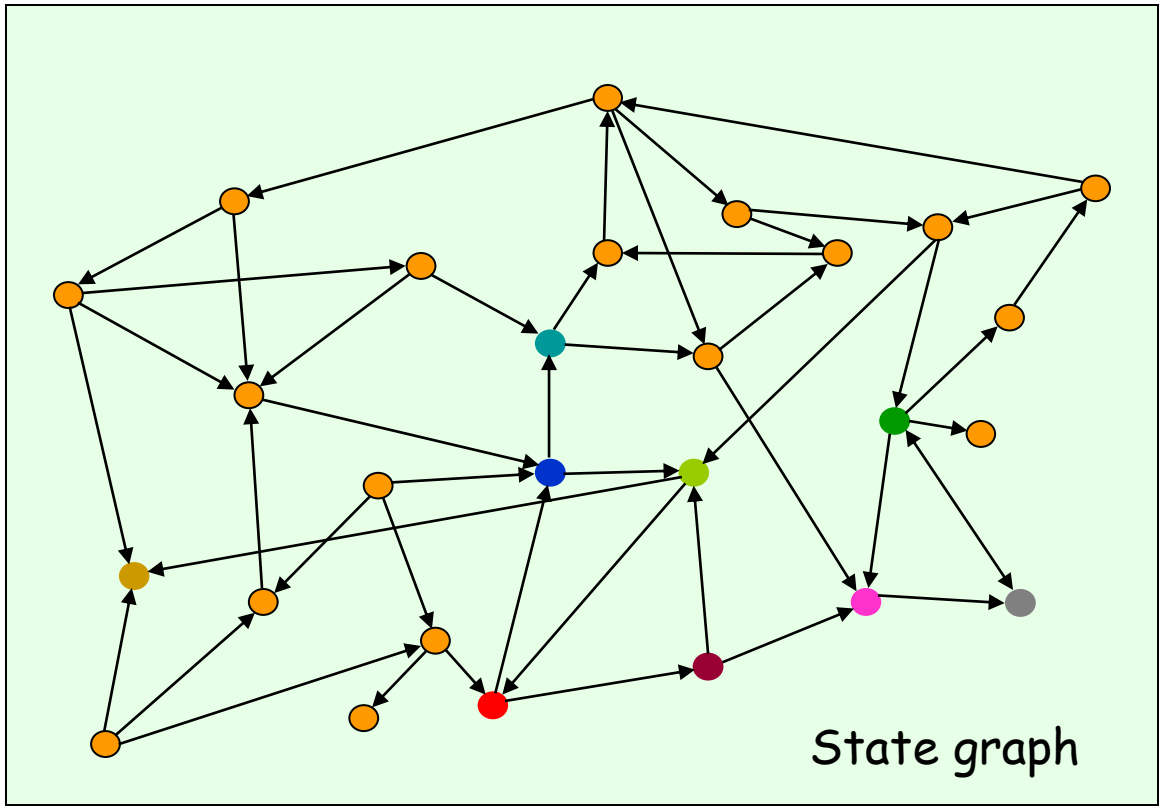
(Where we systematically explore alternatives)

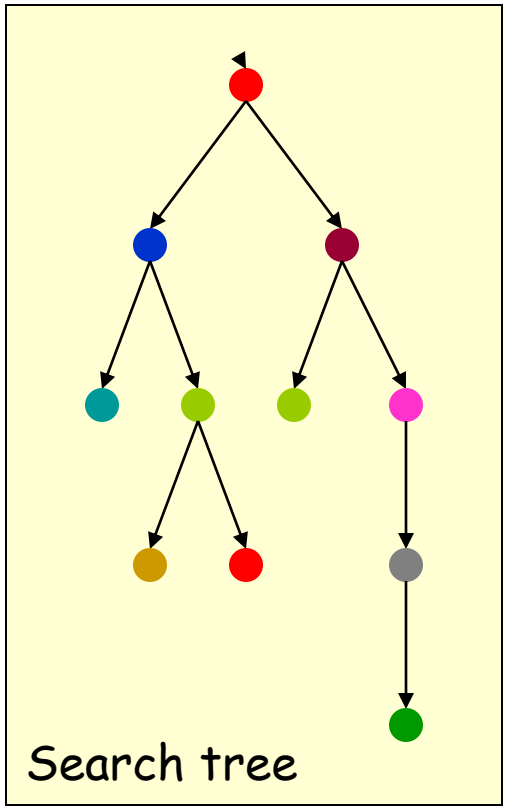
R&N: Chap. 3, Sect. 3.3-5

Simple Problem-Solving-Agent Agent Algorithm

1. $s_0 \leftarrow$ sense/read initial state
2. GOAL? $\leftarrow$ select/read goal test
3. Succ $\leftarrow$ read successor function
4. solution $\leftarrow$ **search**(s_0, GOAL?, Succ)
5. perform(solution)

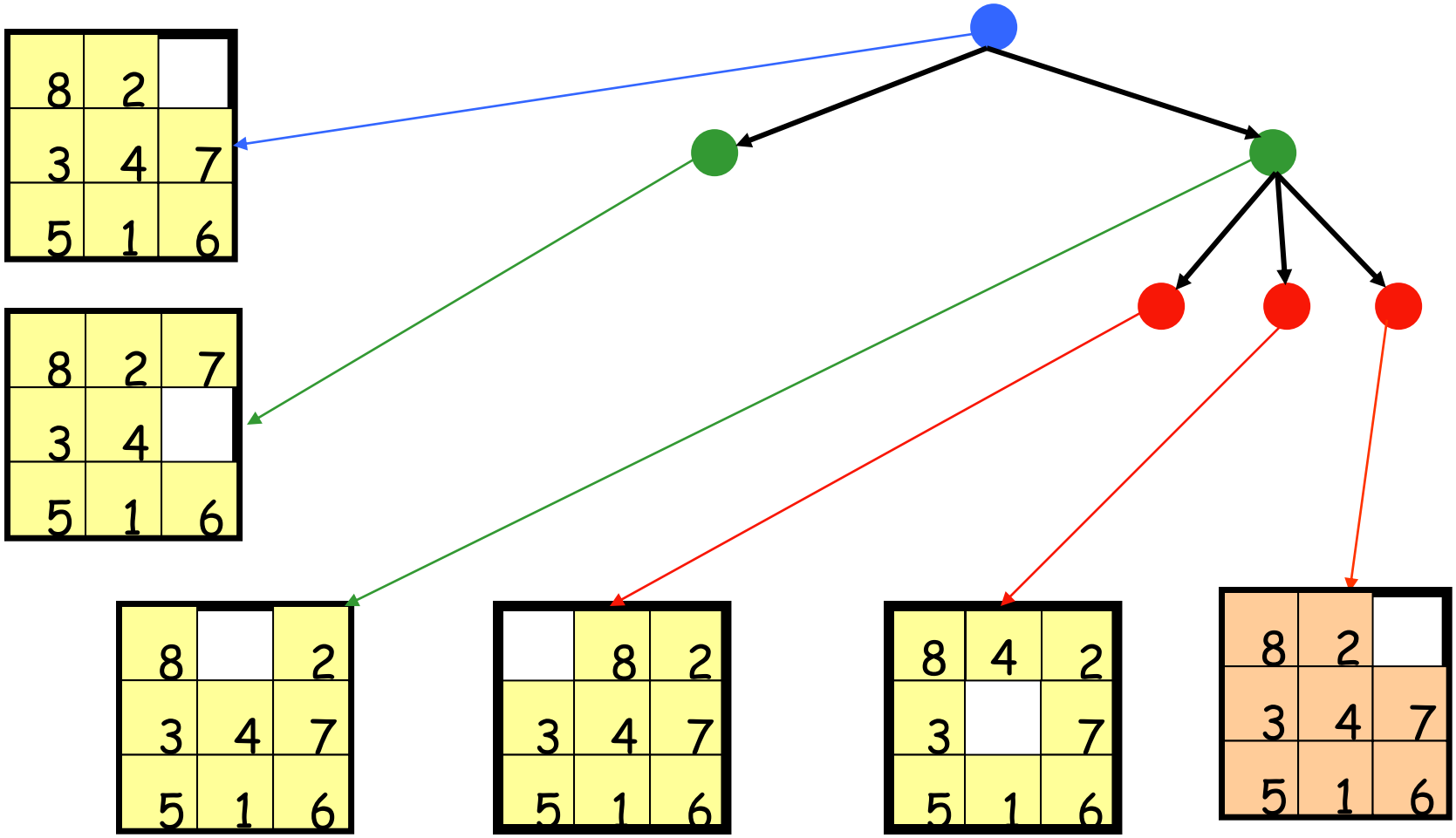
Search Tree



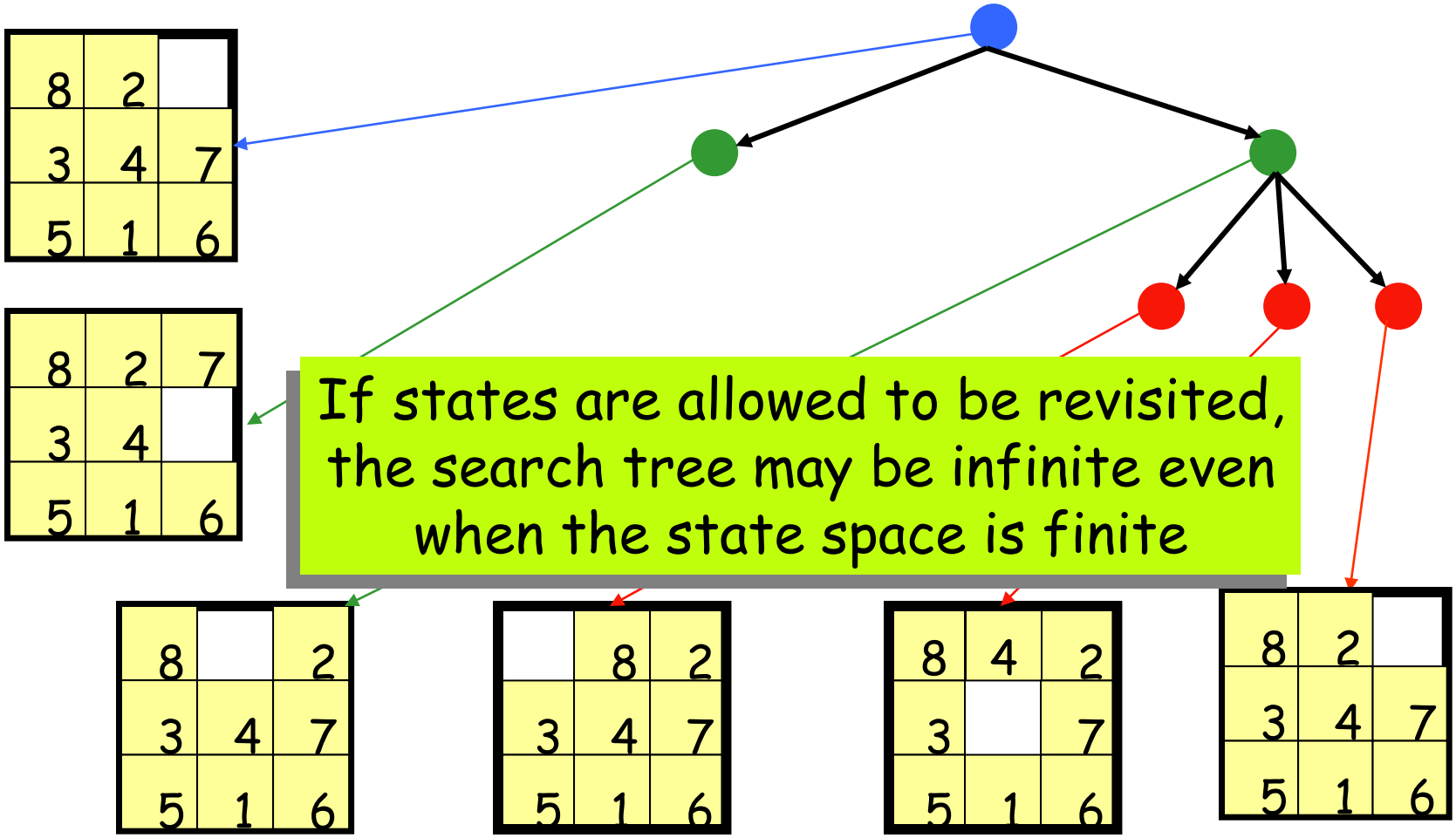


Note that some states may be visited multiple times

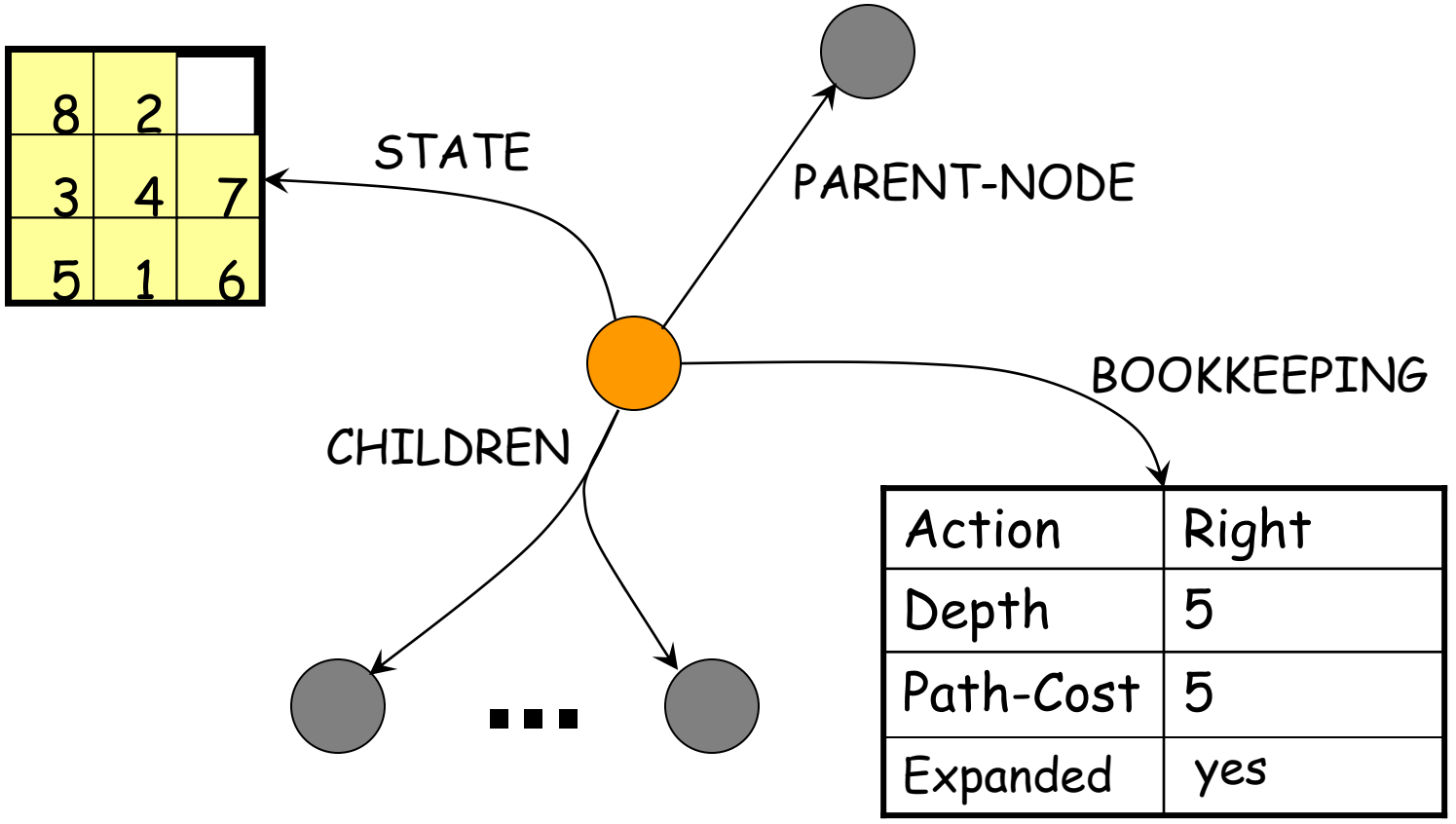
Search Nodes and States



Search Nodes and States



Data Structure of a Node



Action	Right
Depth	5
Path-Cost	5
Expanded	yes

Depth of a node N
= length of path from root to N

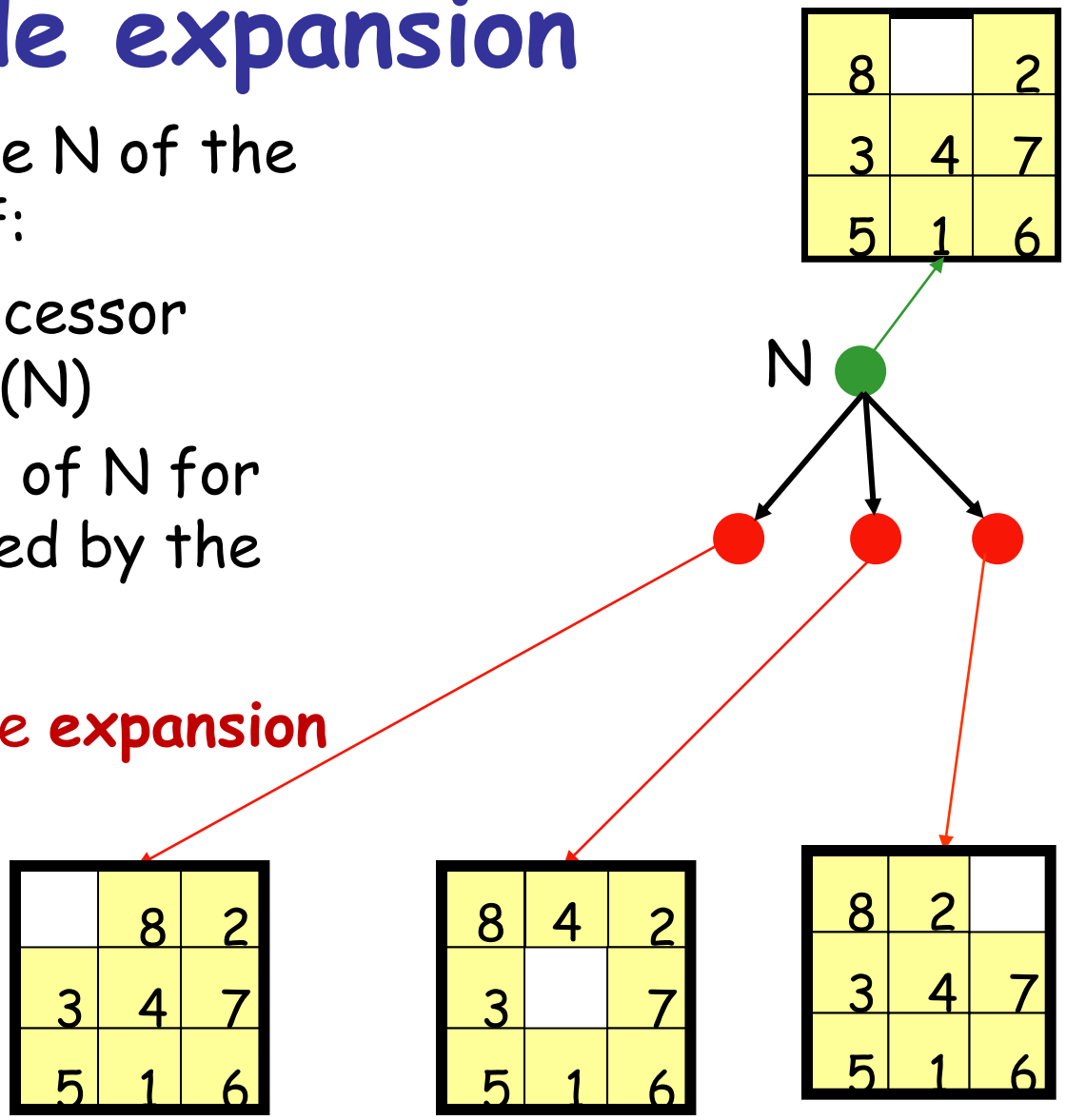
(depth of the root = 0)

Node expansion

The expansion of a node N of the search tree consists of:

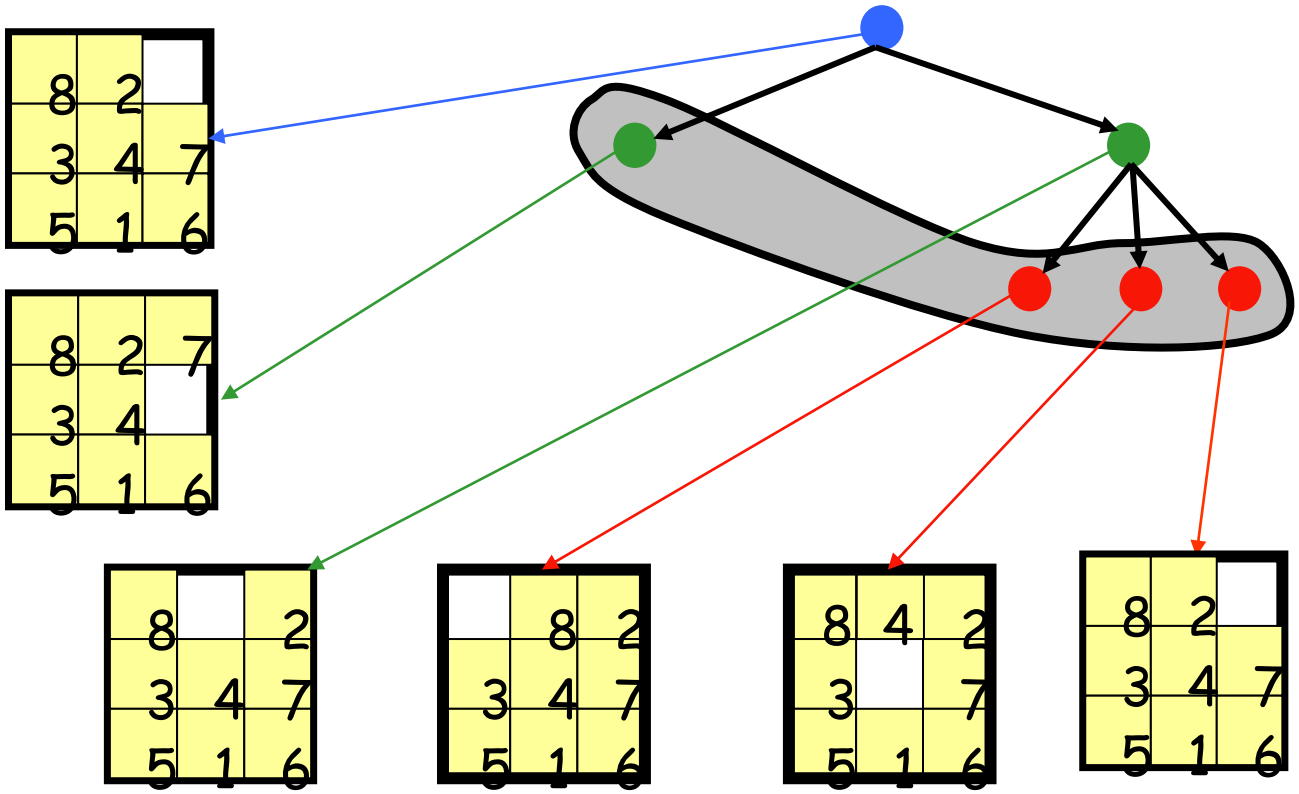
1) Evaluating the successor function on STATE(N)
2) Generating a child of N for each state returned by the function

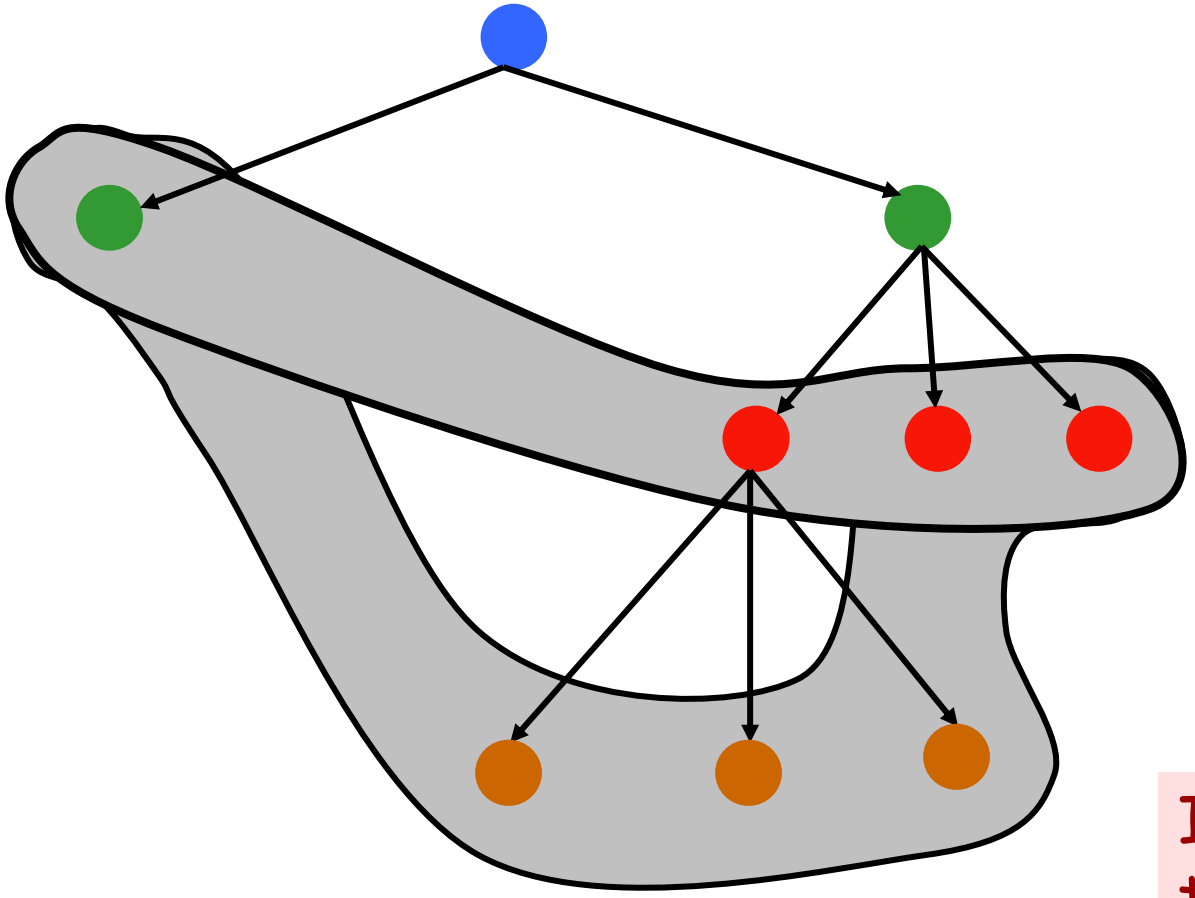
node **generation** ≠ node **expansion**



Fringe (Frontier) of Search Tree

- The fringe (Frontire) is the set of all search nodes that haven't been expanded yet





Is it identical to the set of leaves?

Search Strategy

- The fringe is the set of all search nodes that haven't been expanded yet
- The fringe is implemented as a priority queue FRINGE
  - INSERT(node,FRINGE)
  - REMOVE(FRINGE)
- The ordering of the nodes in FRINGE defines the search strategy

Search Algorithm #1

SEARCH#1

1. If GOAL?(initial-state) then return initial-state
2. INSERT(initial-node,FRINGE)
3. Repeat:
   a. If empty(FRINGE) then return failure
   b. N ← REMOVE(FRINGE)
   c. s ← STATE(N)
   d. For every state s' in SUCCESSORS(s)
      i. Create a new node N' as a child of N
      ii. If GOAL?(s') then return path or goal state
      iii. INSERT(N',FRINGE)

Expansion of N

Performance Measures

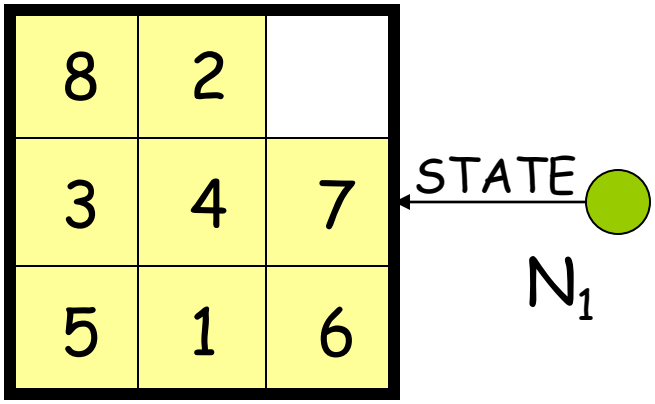
- **Completeness**
  A search algorithm is complete if it finds a solution whenever one exists
  [What about the case when no solution exists?]
- **Optimality**
  A search algorithm is optimal if it returns a minimum-cost path whenever a solution exists
- **Complexity**
  It measures the time and amount of memory required by the algorithm

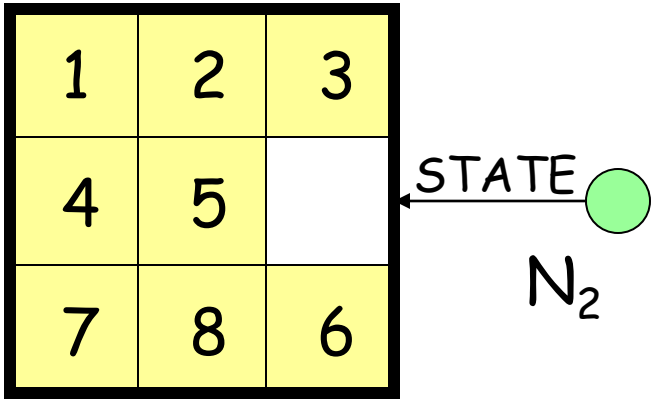
Blind vs. Heuristic Strategies

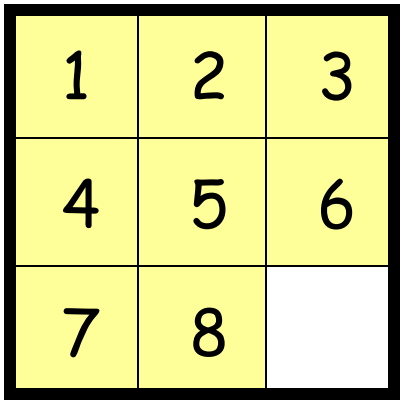
- Blind (or un-informed) strategies do not exploit state descriptions to order FRINGE. They only exploit the positions of the nodes in the search tree
- Heuristic (or informed) strategies exploit state descriptions to order FRINGE (the most "promising" nodes are placed at the beginning of FRINGE)

Example

For a blind strategy, N_1 and N_2 are just two nodes (at some position in the search tree)



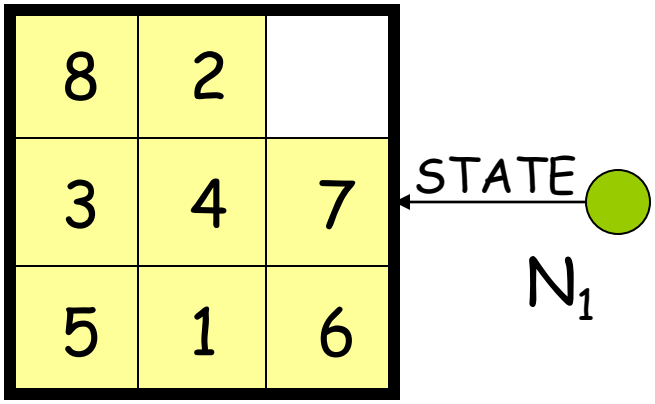


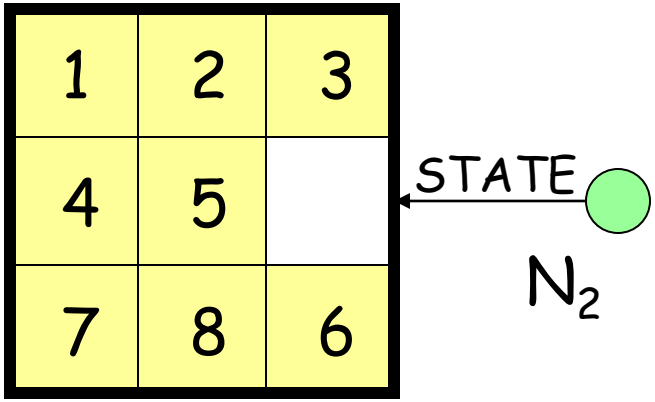


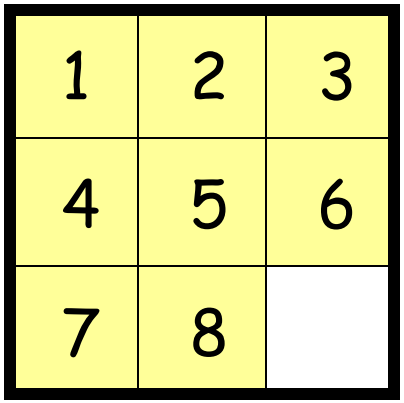
Goal state

Example

For a heuristic strategy counting the number of misplaced tiles, N_2 is more promising than N_1







Goal state

Remark

- Some search problems, such as the (n^2-1)-puzzle, are NP-hard
- One can't expect to solve all instances of such problems in less than exponential time (in n)
- One may still strive to solve each instance as efficiently as possible

→ This is the purpose of the search strategy

Blind Strategies

- Breadth-first
  - Bidirectional
- Depth-first
  - Depth-limited
  - Iterative deepening

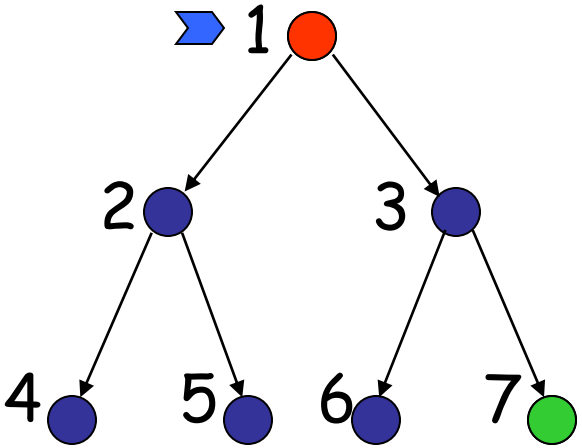
Arc cost = 1

- Uniform-Cost
  (variant of breadth-first)

Arc cost
= c(action) $\geq \varepsilon > 0$

Breadth-First Strategy

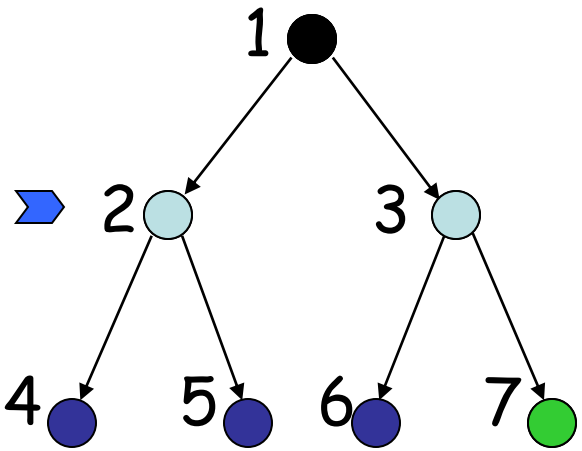
New nodes are inserted at the end of FRINGE



FRINGE = (1)

Breadth-First Strategy

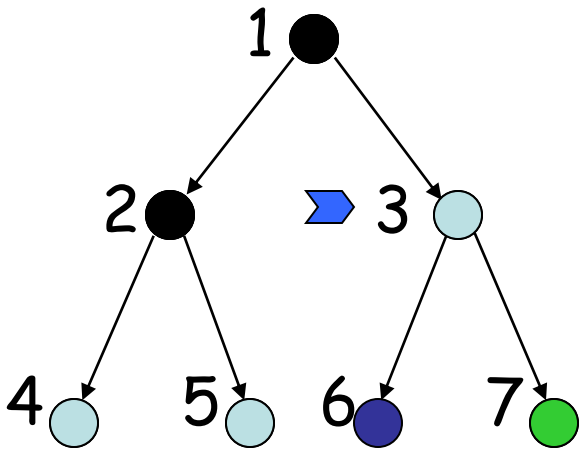
New nodes are inserted at the end of FRINGE



FRINGE = (2, 3)

Breadth-First Strategy

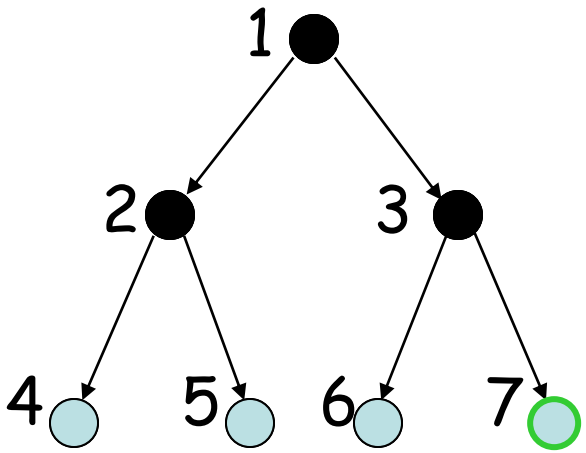
New nodes are inserted at the end of FRINGE



FRINGE = (3, 4, 5)

Breadth-First Strategy

New nodes are inserted at the end of FRINGE



FRINGE = (4, 5, 6, 7)

Important Parameters

1) Maximum number of successors of any state

   → branching factor b of the search tree

2) Minimal length (≠ cost) of a path between the initial and a goal state

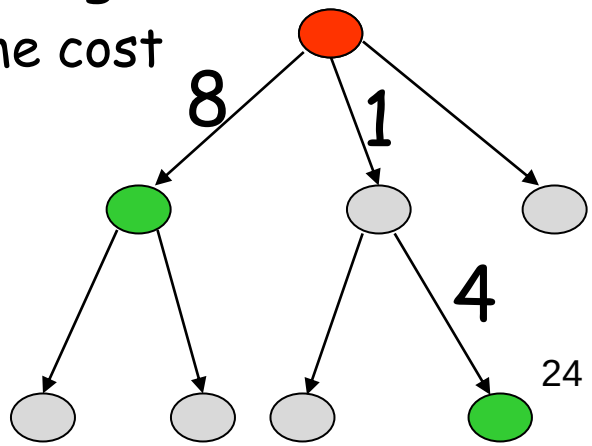
   → depth d of the shallowest goal node in the search tree

Evaluation

- b: branching factor
- d: depth of shallowest goal node
- Breadth-first search is:
  - Complete? Not complete?
  - Optimal? Not optimal?

Evaluation

- b: branching factor
- d: depth of shallowest goal node
- Breadth-first search is:
  - Complete (for finite b and d)
  - Optimal
    - if path cost is a non-decreasing function of d
      - e.g. all actions having the same cost



Evaluation

- b: branching factor
- d: depth of shallowest goal node
- Breadth-first search is:
  - Complete (for finite b and d)
  - Optimal
    - if path cost is a non-decreasing function of d
      - e.g. all actions having the same cost
- Number of nodes generated:

  ???

Evaluation

- b: branching factor
- d: depth of shallowest goal node
- Breadth-first search is:
  - Complete (for finite b and d)
  - Optimal
    - if path cost is a non-decreasing function of d
      - e.g. all actions having the same cost
- Number of nodes generated:
  $1 + b + b^2 + \ldots + b^d$ = ???

Evaluation

- b: branching factor
- d: depth of shallowest goal node
- Breadth-first search is:
  - Complete (for finite b and d)
  - Optimal
    - if path cost is a non-decreasing function of d
      - e.g. all actions having the same cost
- Number of nodes generated:
  $1 + b + b^2 + \ldots + b^d = (b^{d+1}-1)/(b-1) = O(b^d)$
- → Time and space complexity is $O(b^d)$

Big O Notation

$g(n) = O(f(n))$ if there exist two positive constants a and N such that:

for all $n > N$: $g(n) \leq a \times f(n)$

Time and Memory Requirements

d	# Nodes	Time	Memory
2	111	.01 msec	11 Kbytes
4	11,111	1 msec	1 Mbyte
6	~10^6	1 sec	100 Mb
8	~10^8	100 sec	10 Gbytes
10	~10^{10}	2.8 hours	1 Tbyte
12	~10^{12}	11.6 days	100 Tbytes
14	~10^{14}	3.2 years	10,000 Tbytes

Assumptions: b = 10; 1,000,000 nodes/sec; 100bytes/node

Time and Memory Requirements

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14	~10^{14}	3.2 years	10,000 Tbytes

Assumptions: b = 10; 1,000,000 nodes/sec; 100bytes/node

Remark

If a problem has no solution, breadth-first may run for ever (if the state space is infinite or states can be revisited arbitrary many times)

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

?

1	2	3	4
5	6	7	8
9	10	11	12
13	15	14	

Bidirectional Strategy

2 fringe queues: FRINGE1 and FRINGE2

Bidirectional Strategy

2 fringe queues: FRINGE1 and FRINGE2

Bidirectional Strategy

2 fringe queues: FRINGE1 and FRINGE2

Bidirectional Strategy

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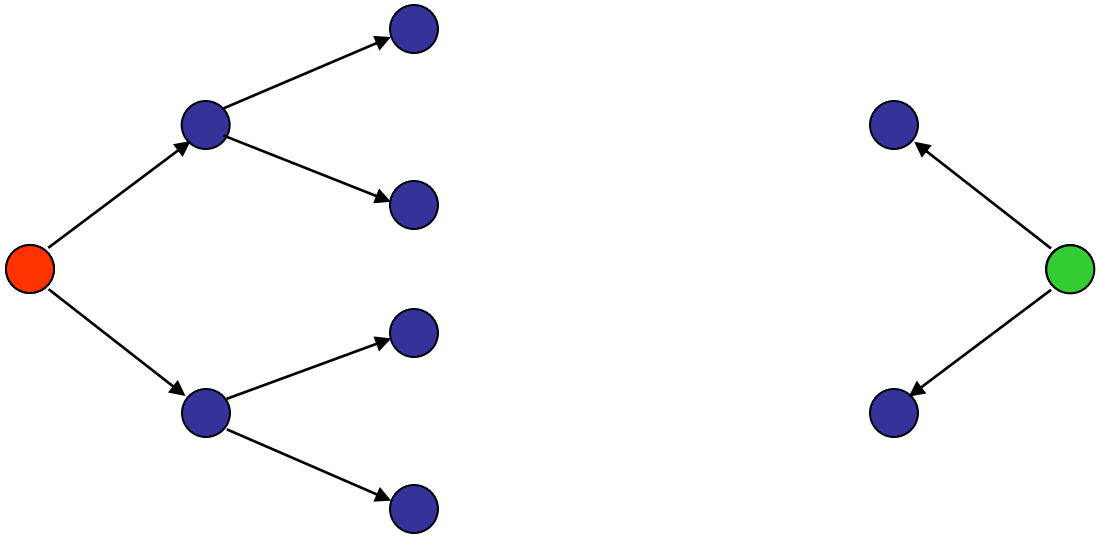
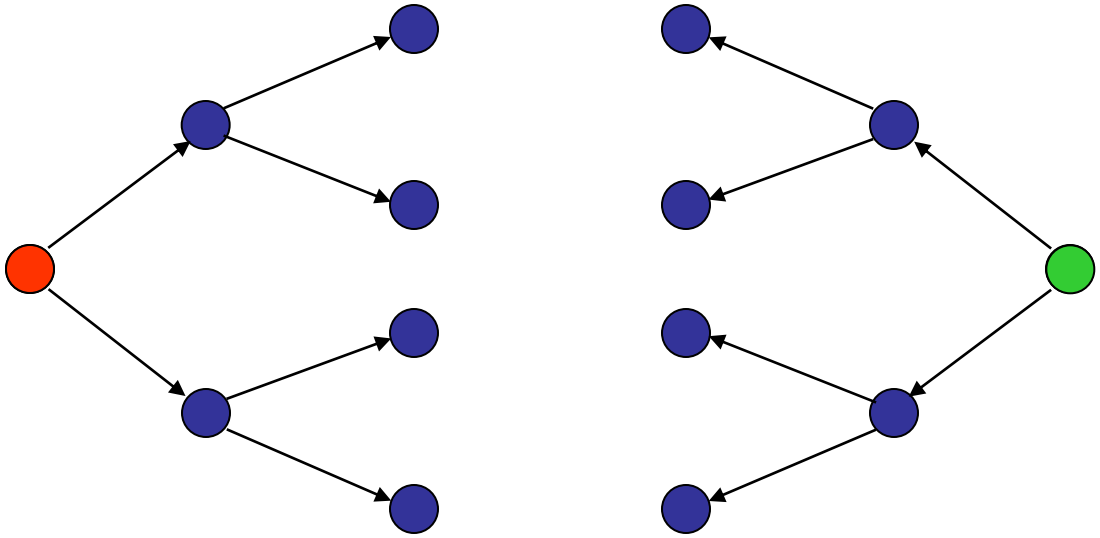


Bidirectional Strategy

2 fringe queues: FRINGE1 and FRINGE2



Bidirectional Strategy

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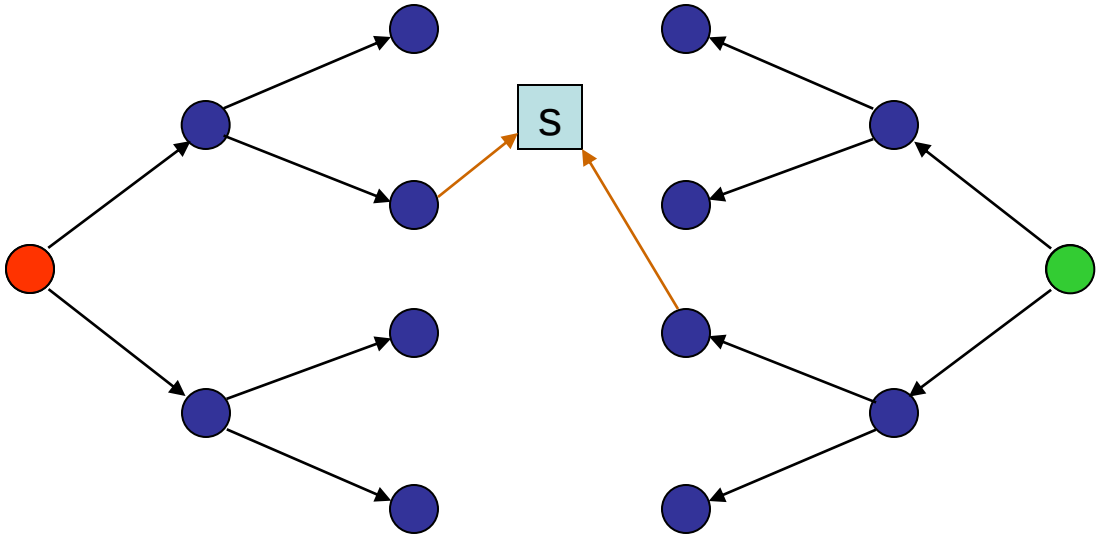

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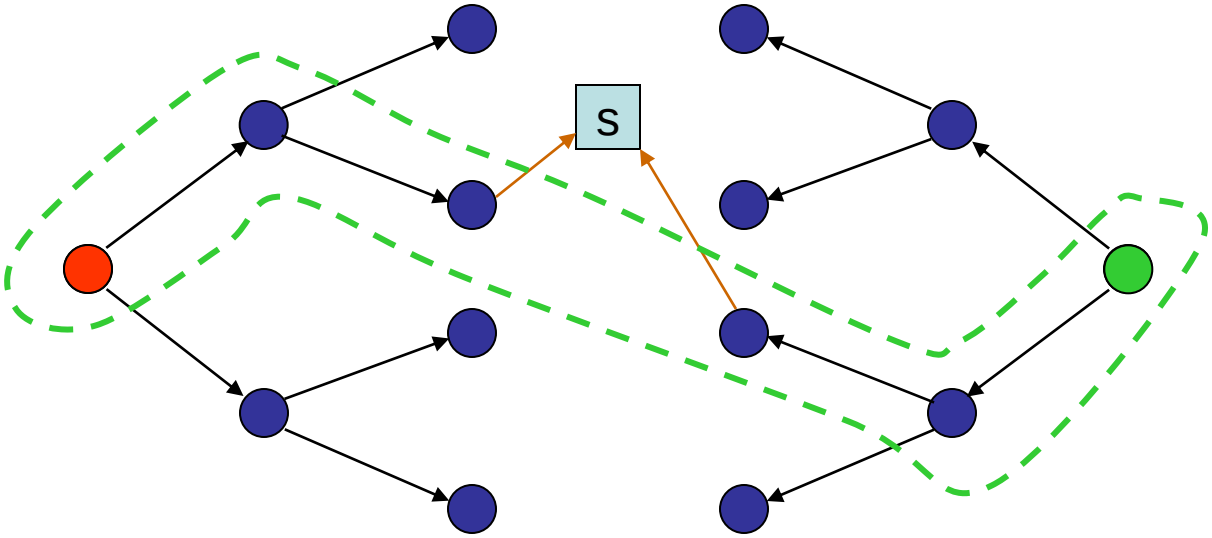

Bidirectional Strategy

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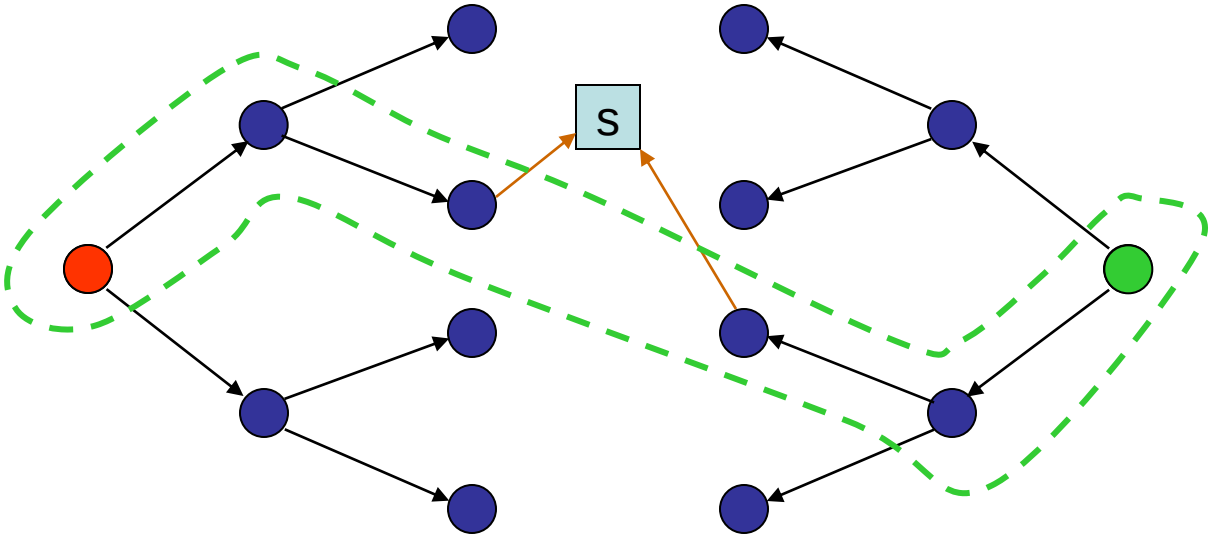
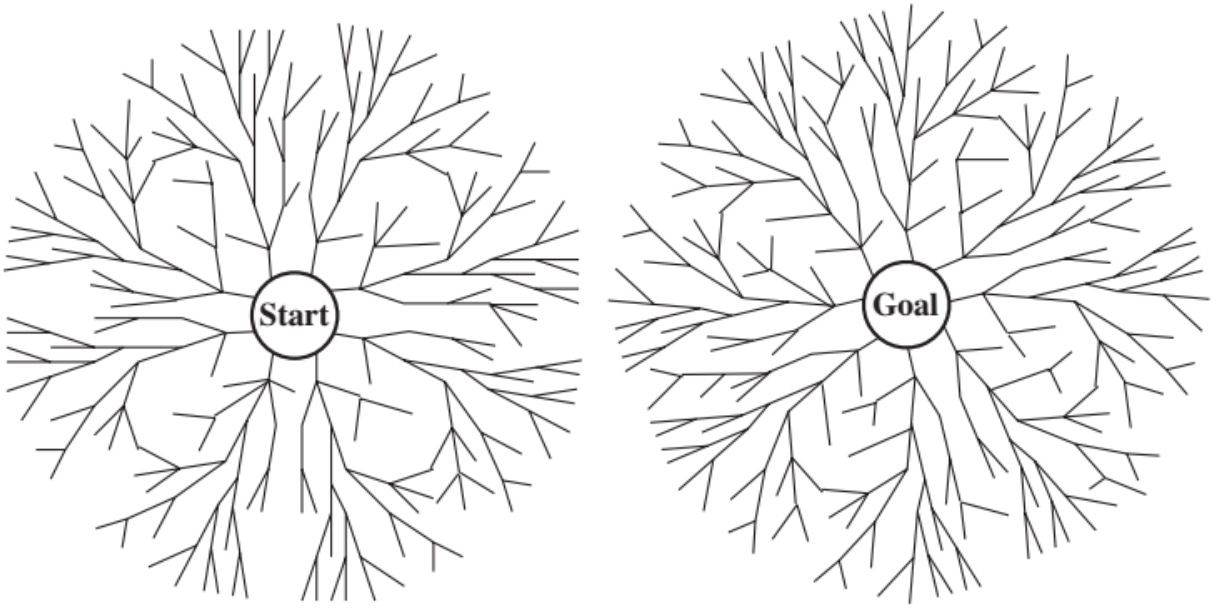


Time and space complexity is $O(b^{d/2}) << O(b^d)$
if both trees have the same branching factor b

Question: What happens if the branching factor is different in each direction?

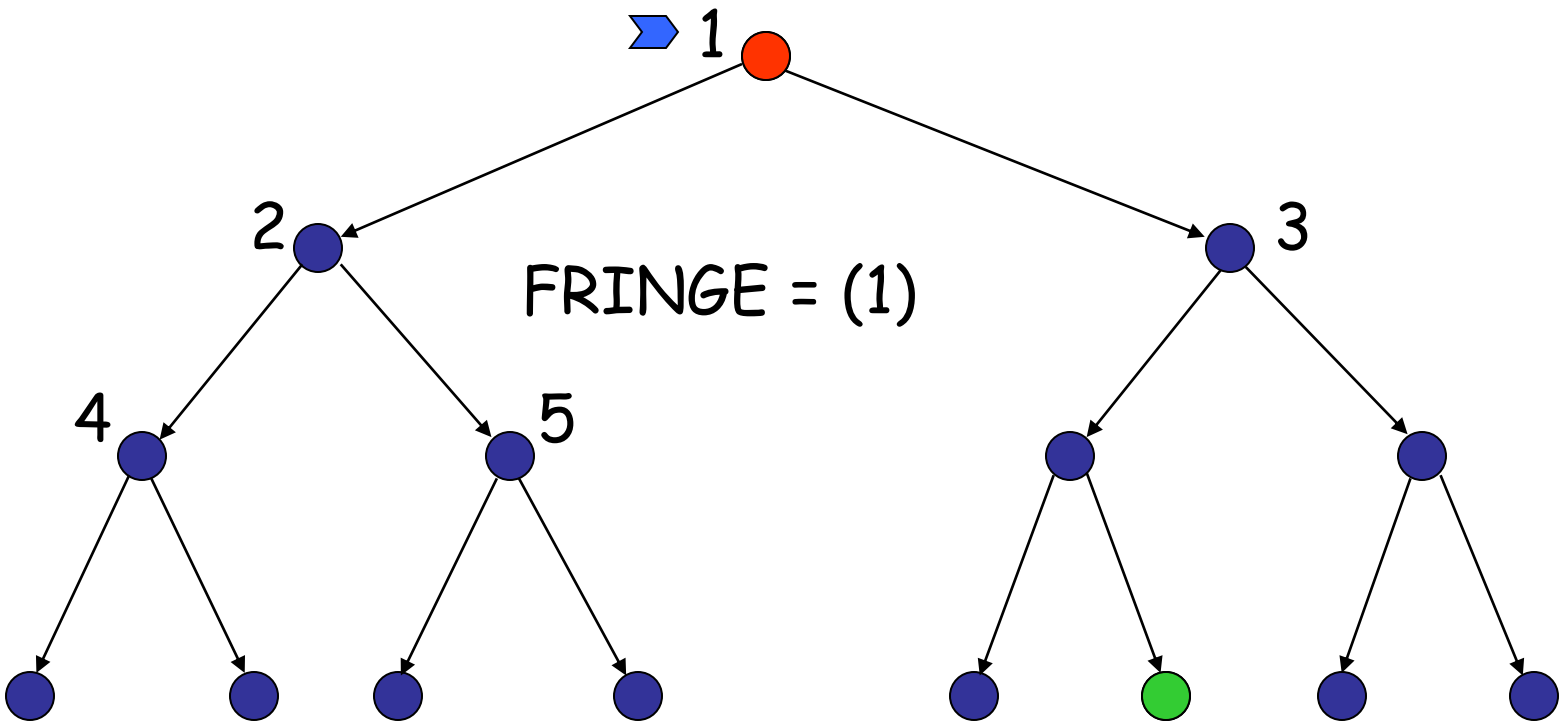
Bidirectional Strategy

- Implementation
  - Hash table for one of the fringe lists
  - Computing predecessors?
    - May be difficult
  - List of goals? a new dummy goal
    - Abstract goal (checkmate)?!



Depth-First Strategy

New nodes are inserted at the front of FRINGE



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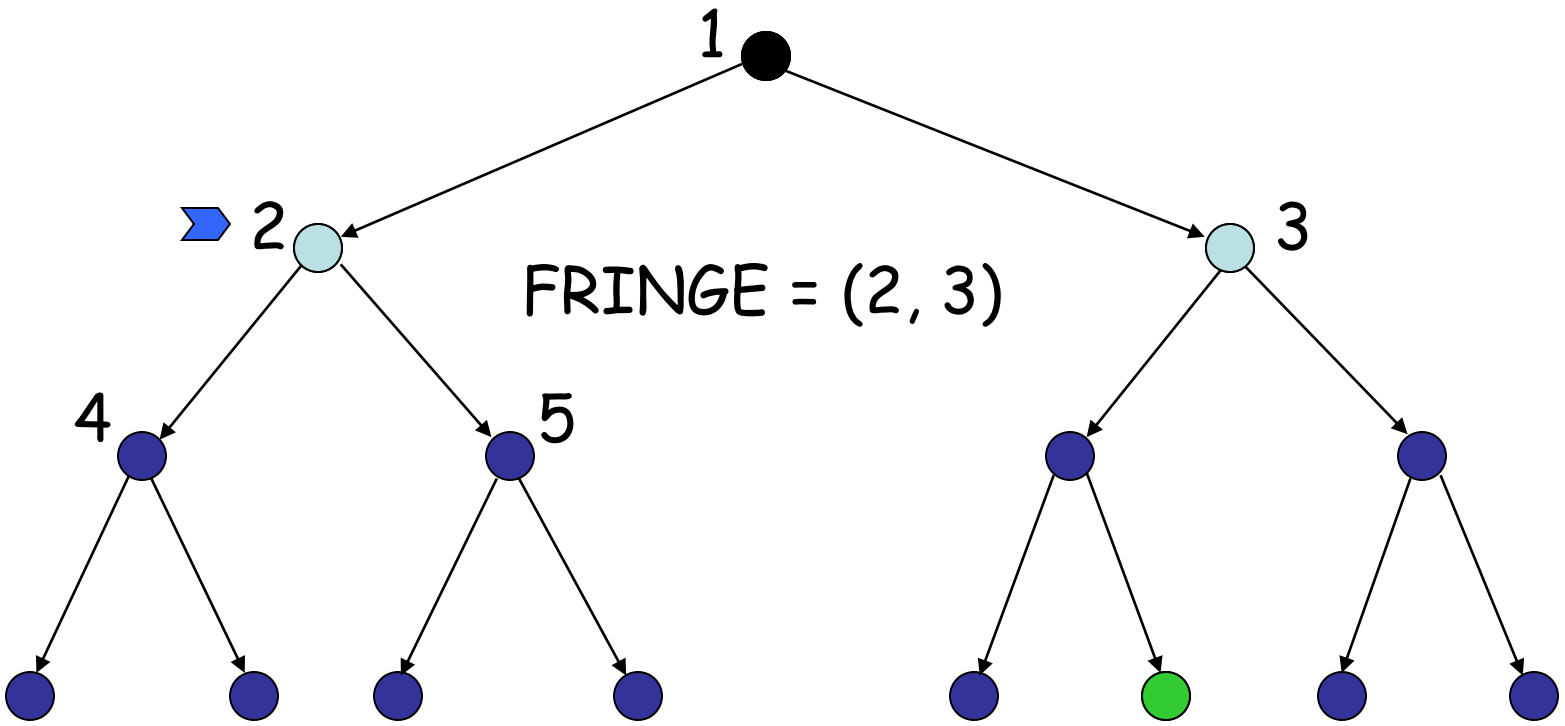

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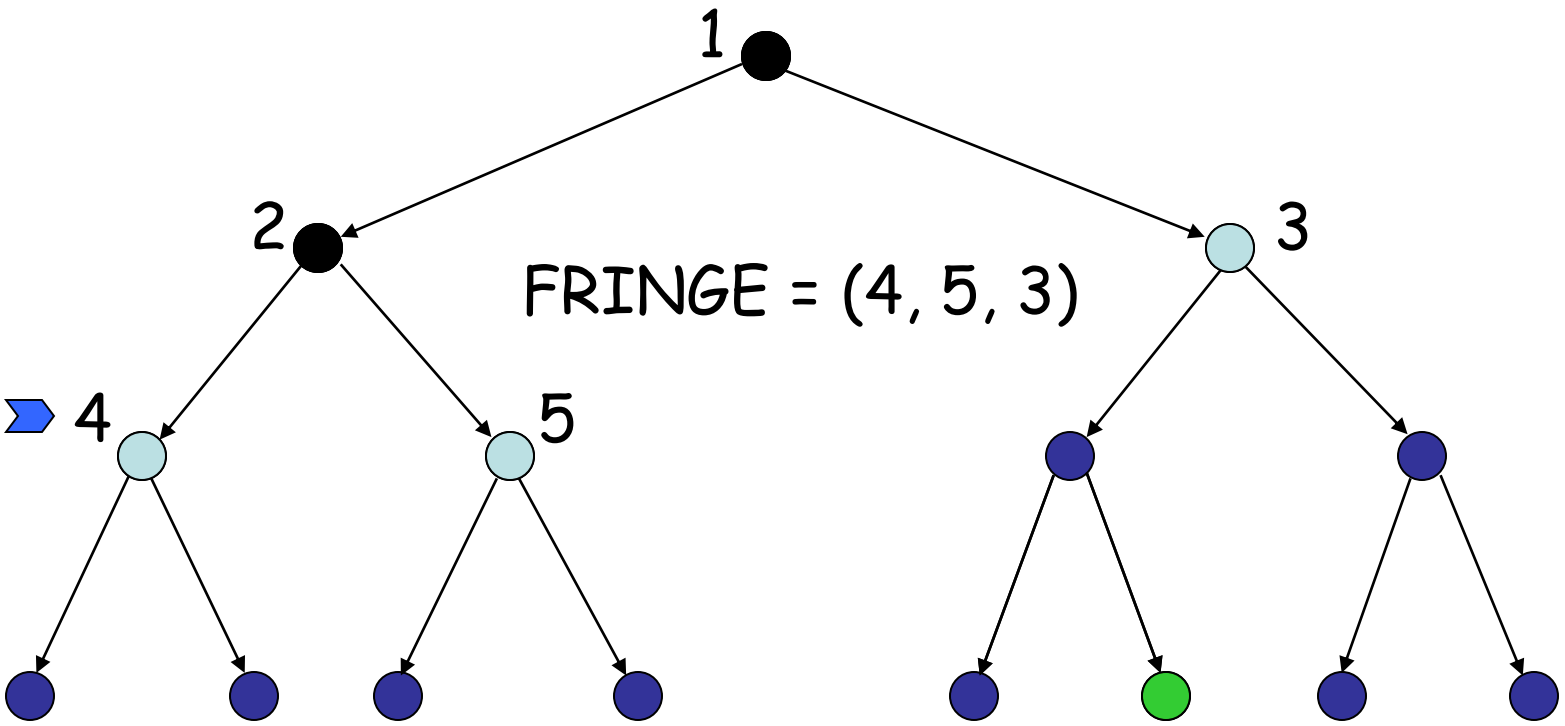

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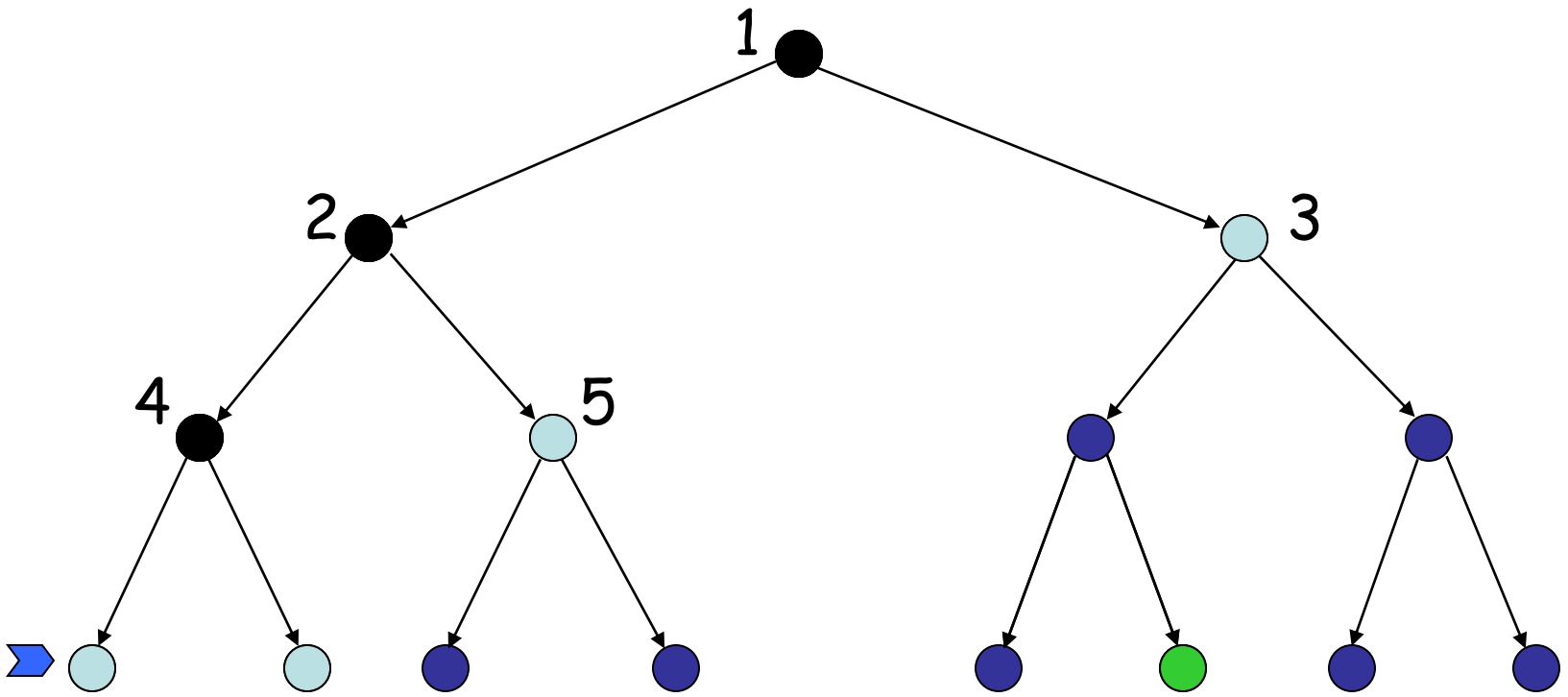

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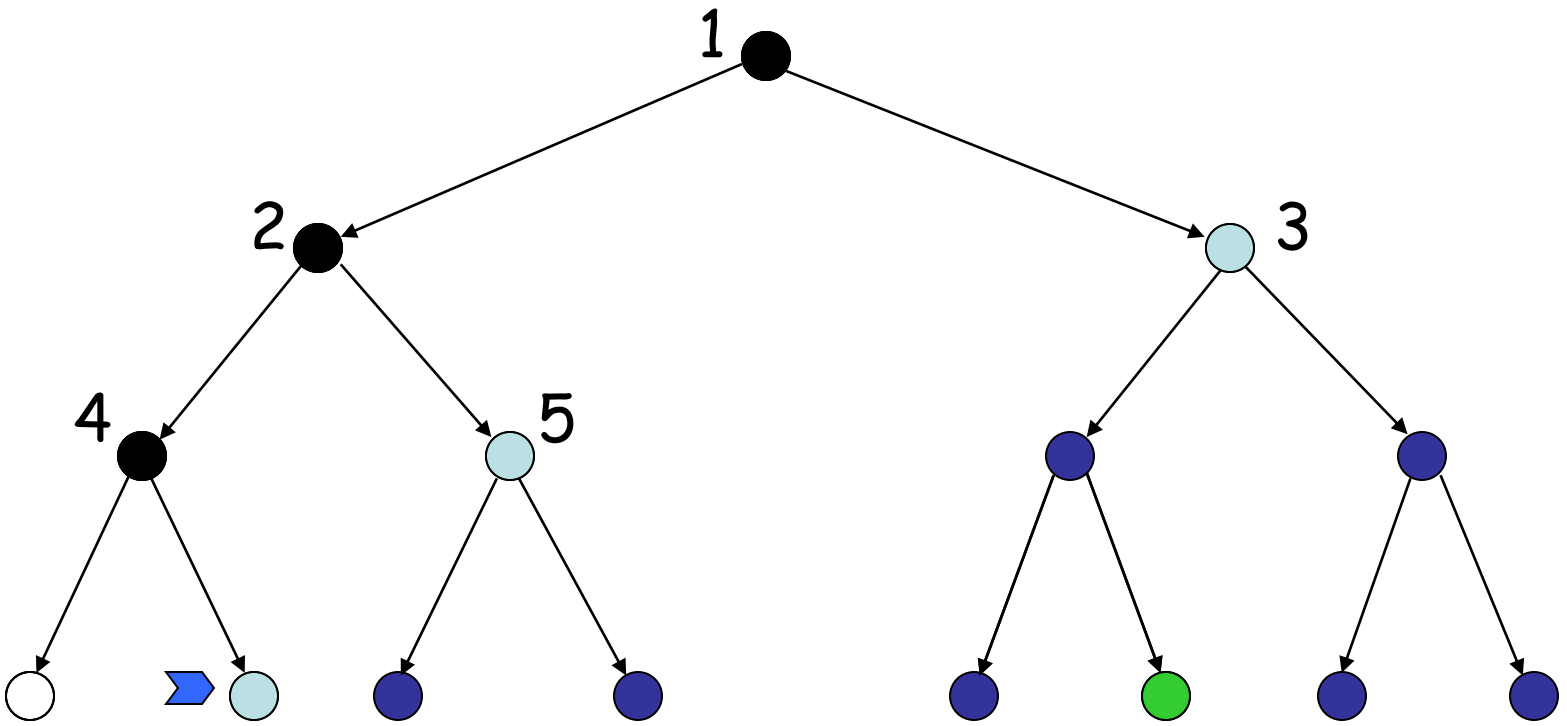

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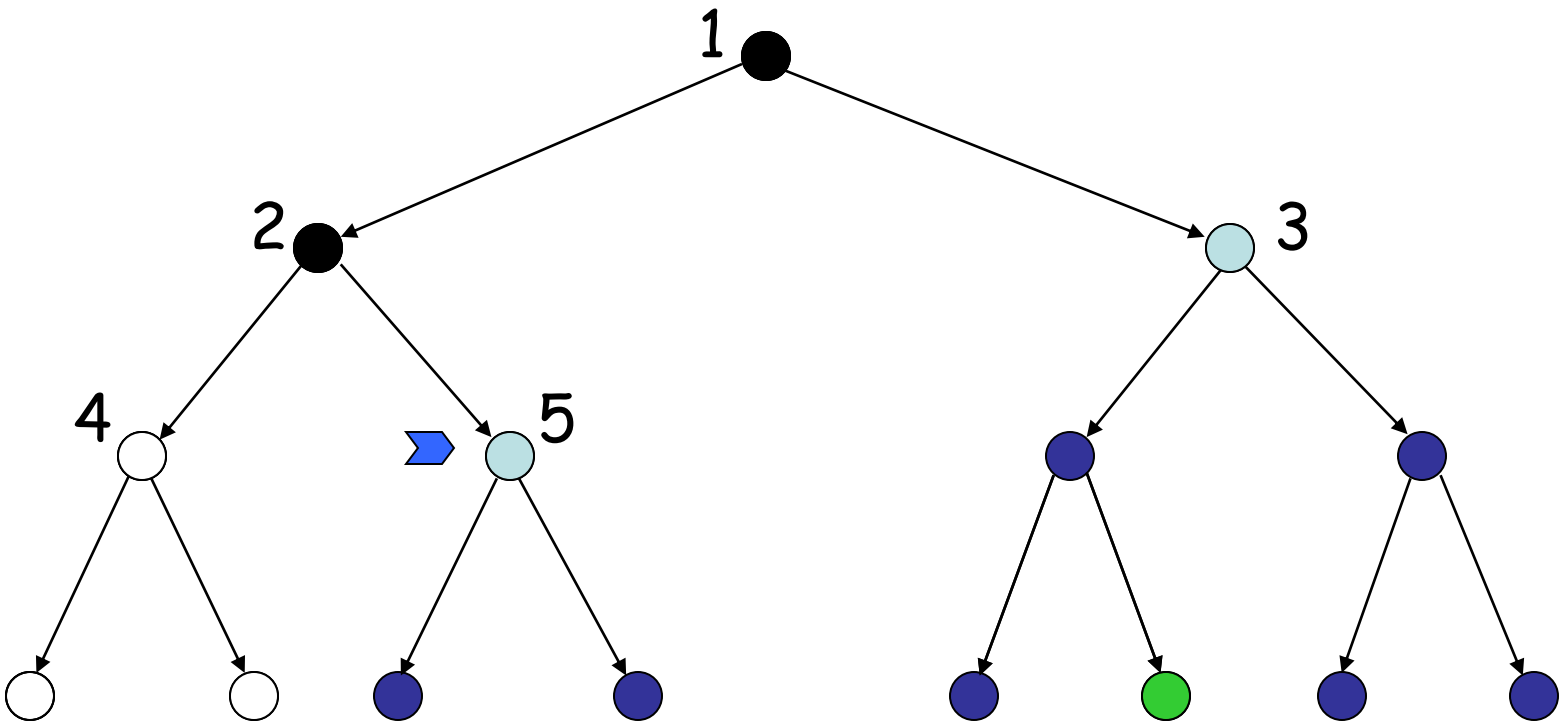

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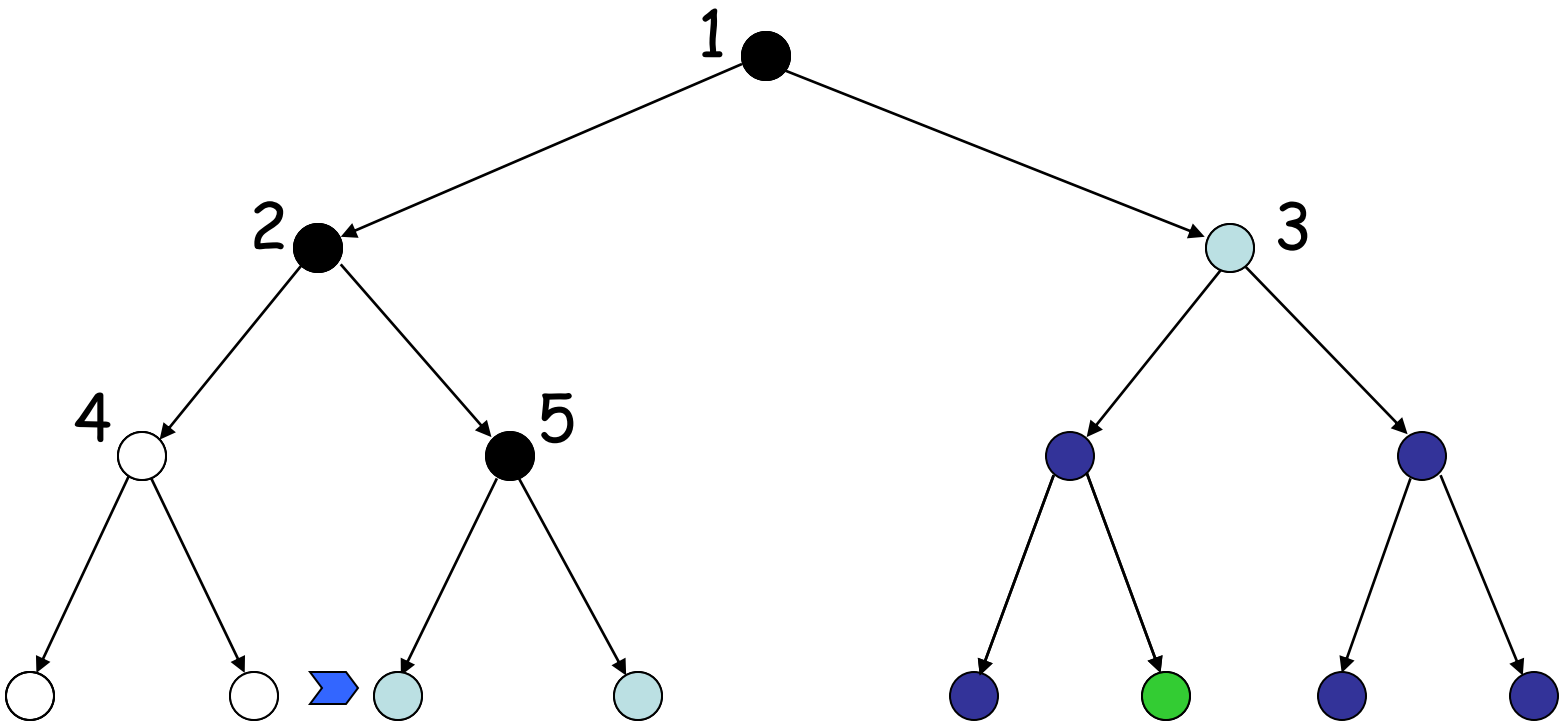

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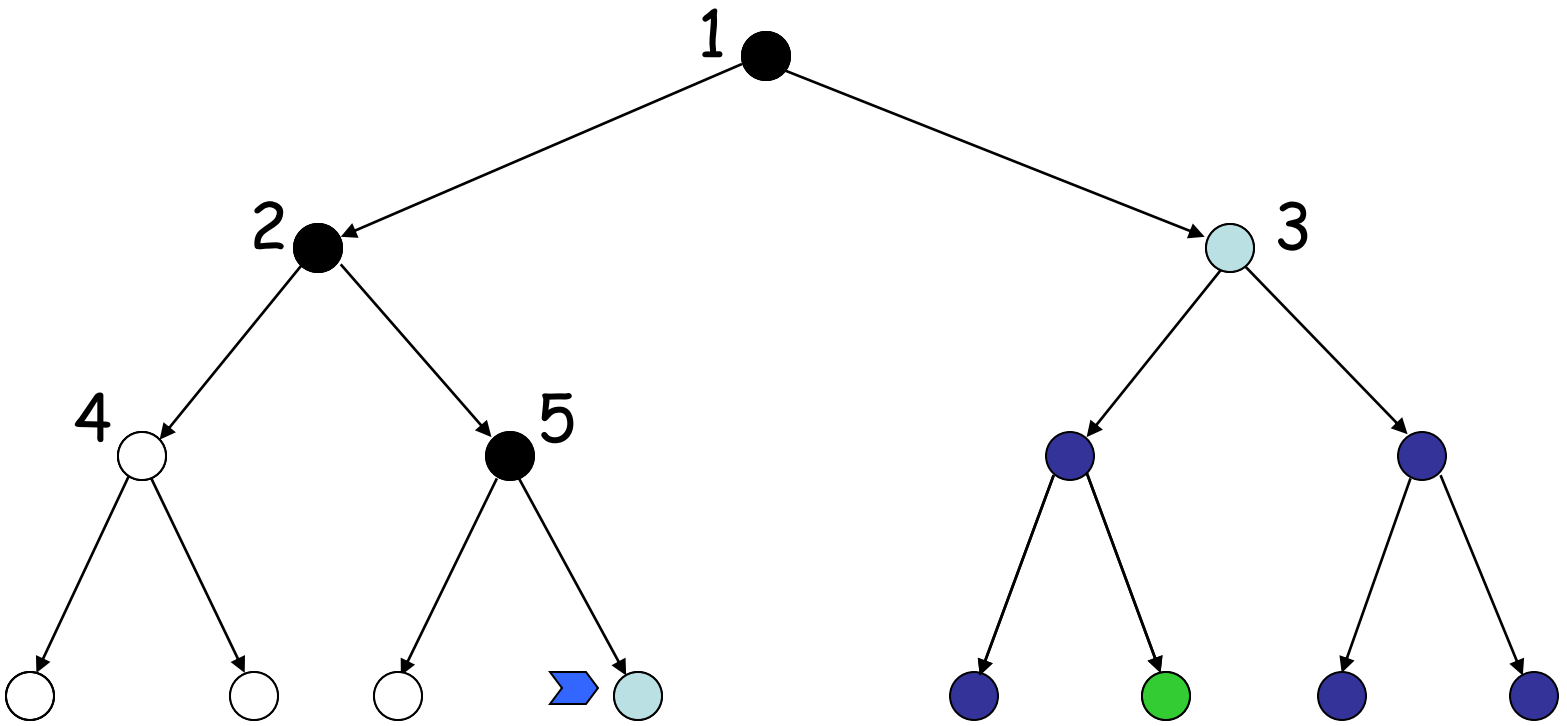

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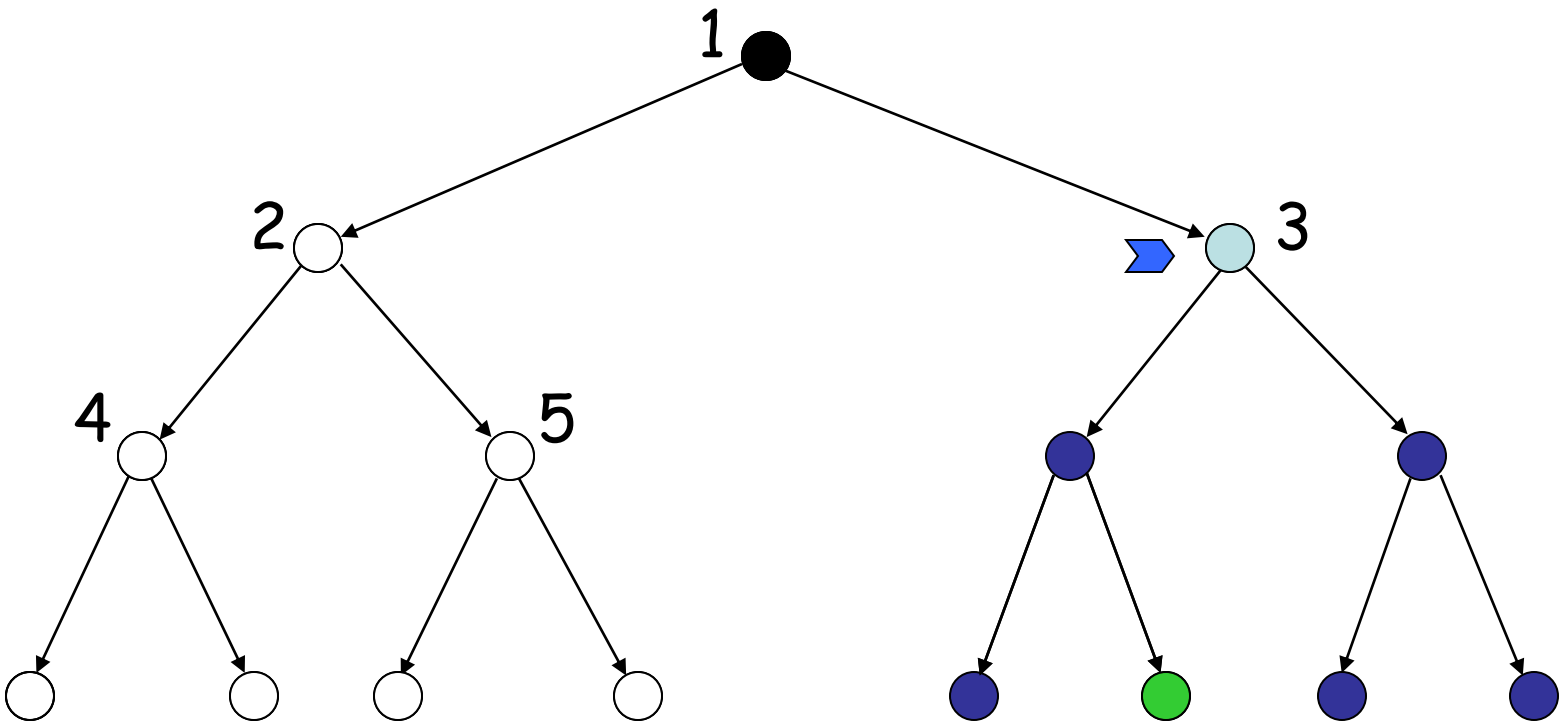

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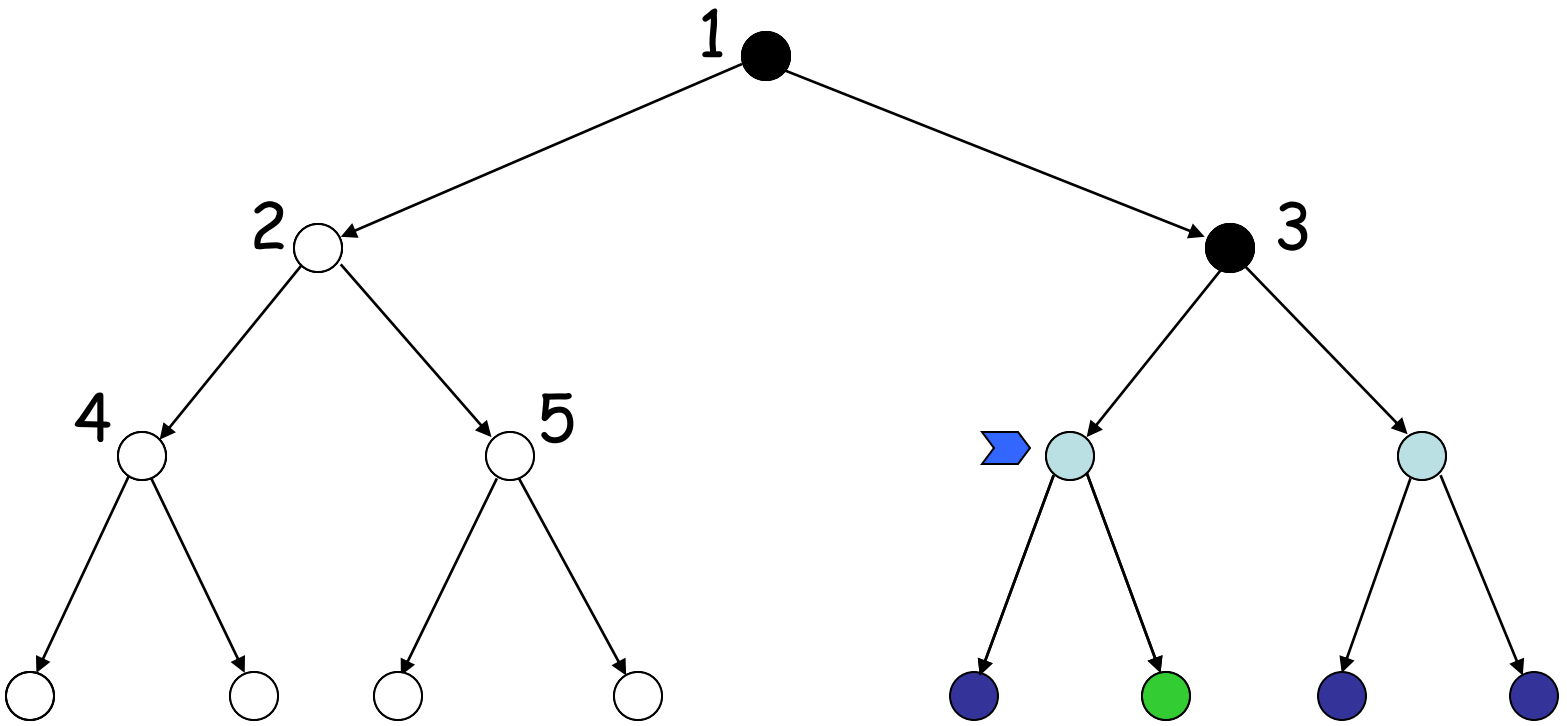

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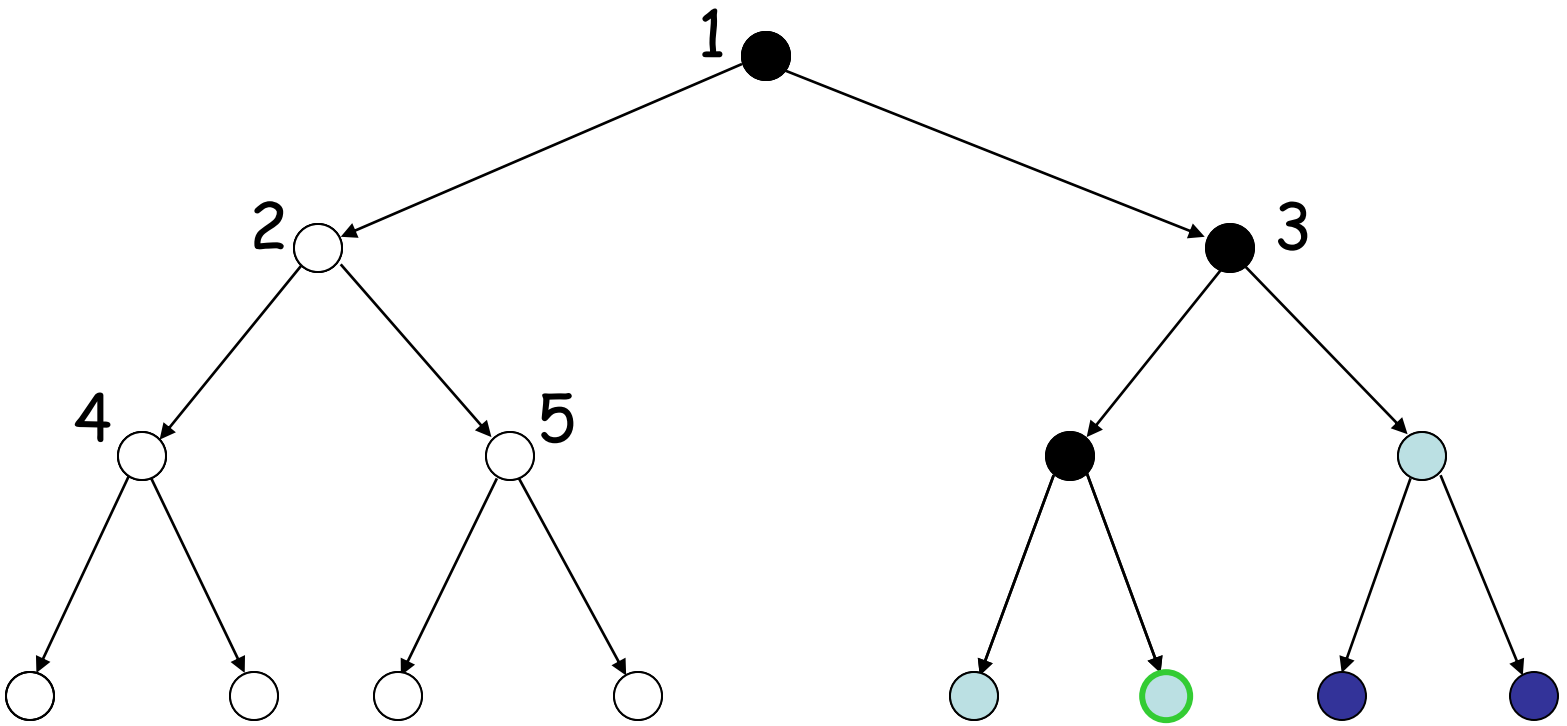
New nodes are inserted at the front of FRINGE

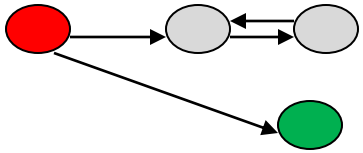


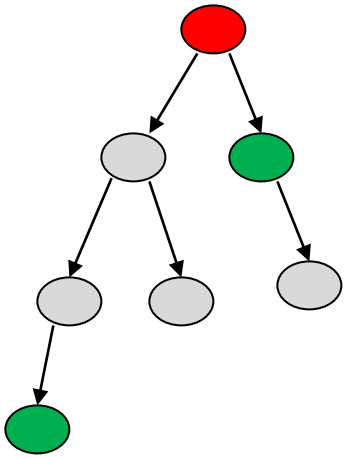
Evaluation

- b: branching factor
- d: depth of shallowest goal node
- m: maximal depth of a leaf node
- Depth-first search is:
  - Complete?
  - Optimal?

Evaluation

- b: branching factor
- d: depth of shallowest goal node
- m: maximal depth of a leaf node
- Depth-first search is:
  - Complete only for finite search tree (if we can avoid infinite loops)
  - Not optimal





Evaluation

- b: branching factor
- d: depth of shallowest goal node
- m: maximal depth of a leaf node
- Depth-first search is:
  - Complete only for finite search tree (if we can avoid infinite loops)
  - Not optimal
- Number of nodes generated (worst case):
  $1 + b + b^2 + \ldots + b^m = O(b^m)$
- Time complexity is $O(b^m)$
- Space complexity is $O(bm)$ [or $O(m)$]

[Reminder: Breadth-first requires $O(b^d)$ time and space]

Depth-Limited Search

- Depth-first with depth cutoff k (depth at which nodes are not expanded)
  - Solves the infinite-path problem

- Three possible outcomes:
  - Solution
  - Failure (no solution)
  - Cutoff (no solution within cutoff)

Depth-Limited Search

- Complete? If k > d, it is complete
- Time? $O(b^k)$
- Space? $O(bk)$
- Optimal? No

Depth-Limited Search

```
function DEPTH-LIMITED-SEARCH(problem, limit) returns a solution, or failure/cutoff
    return RECURSIVE-DLS(MAKE-NODE(problem.INITIAL-STATE), problem, limit)
```

```
function RECURSIVE-DLS(node, problem, limit) returns a solution, or failure/cutoff
    if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
    else if limit = 0 then return cutoff
    else
        cutoff _occurred?←false
        for each action in problem.ACTIONS(node.STATE) do
            child ←CHILD-NODE(problem, node, action)
            result ←RECURSIVE-DLS(child , problem, limit – 1)
            if result = cutoff then cutoff_occurred?←true
            else if result ≠ failure then return result
        if cutoff_occurred? then return cutoff else return failure
```

Iterative Deepening Search

Provides the best of both breadth-first and depth-first search

Main idea: Totally horrifying !

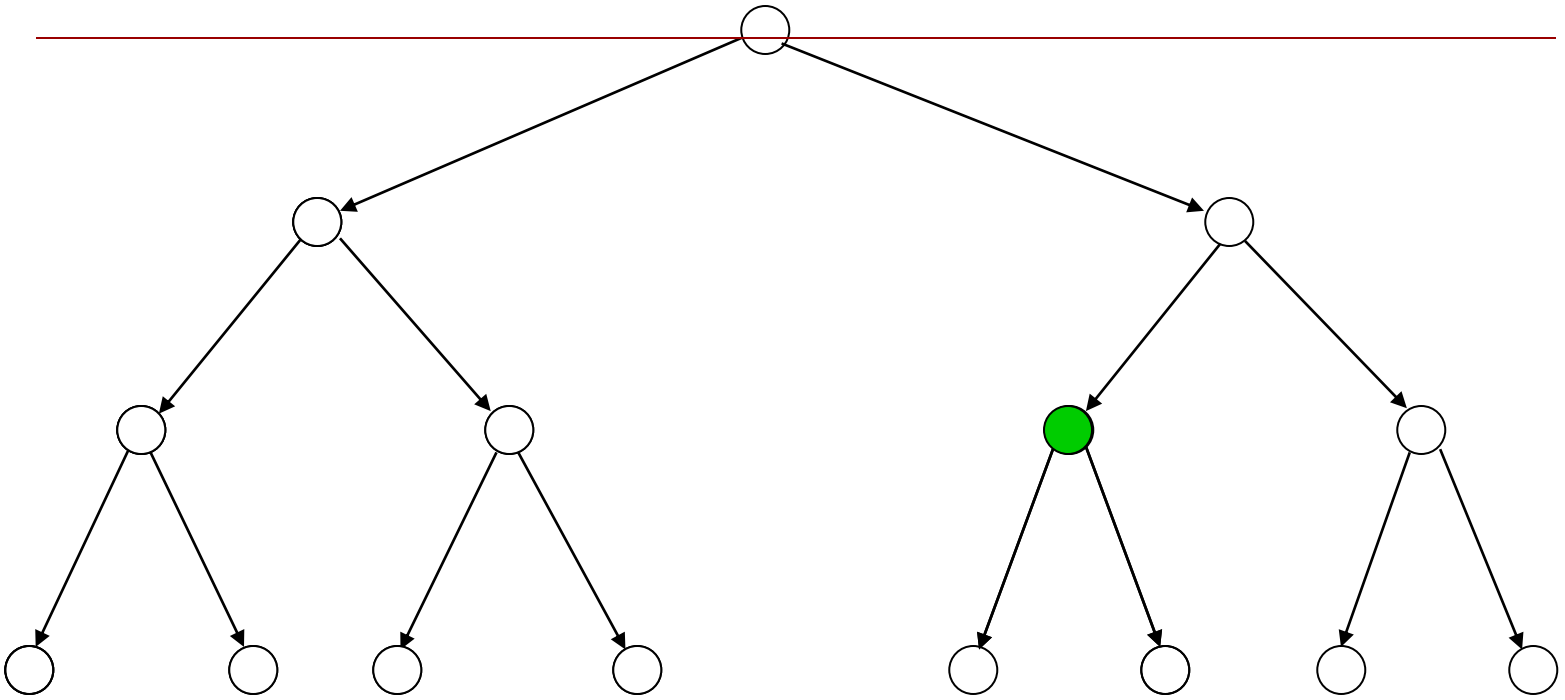
Iterative Deepening Search

Provides the best of both breadth-first and depth-first search

Main idea: Totally horrifying !

IDS
For k = 0, 1, 2, … do:
    Perform depth-first search with depth cutoff k
    (i.e., only generate nodes with depth ≤ k)

Iterative Deepening



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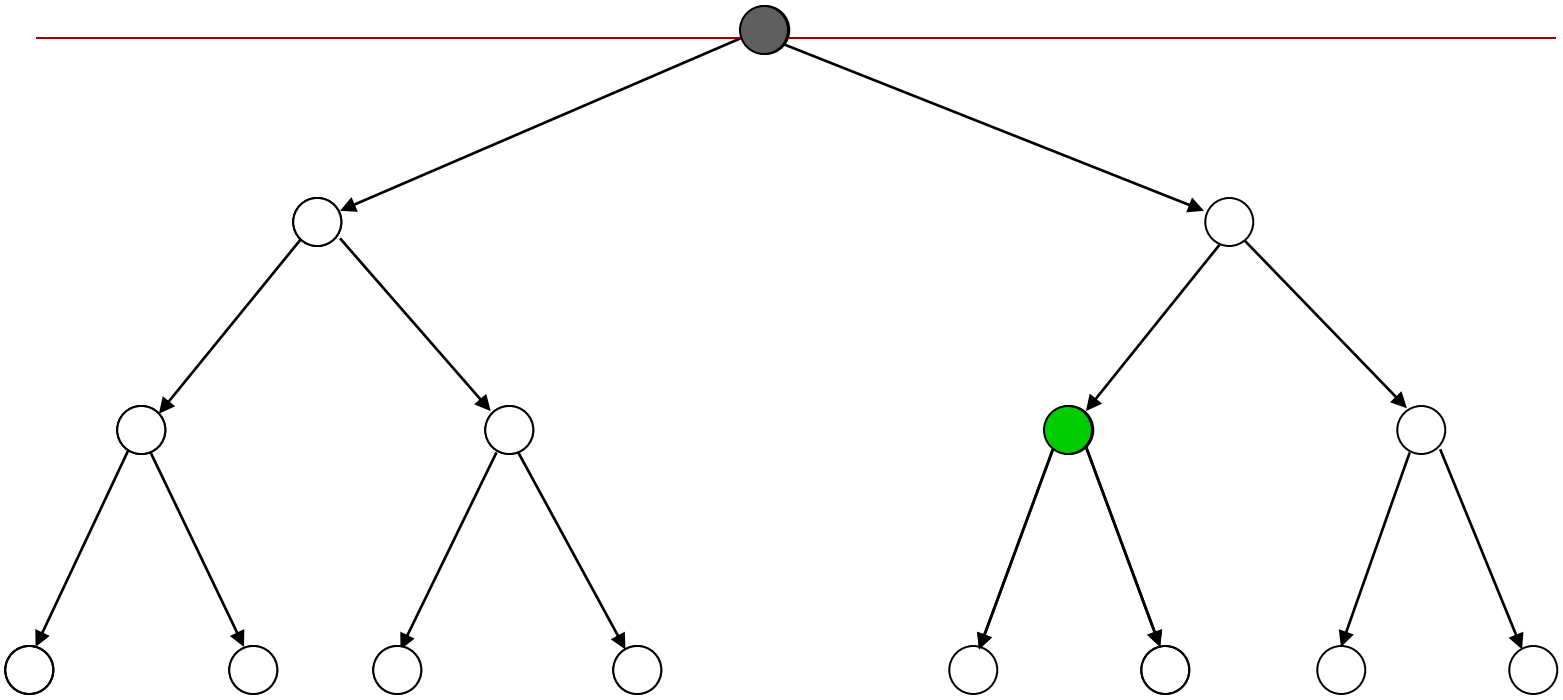

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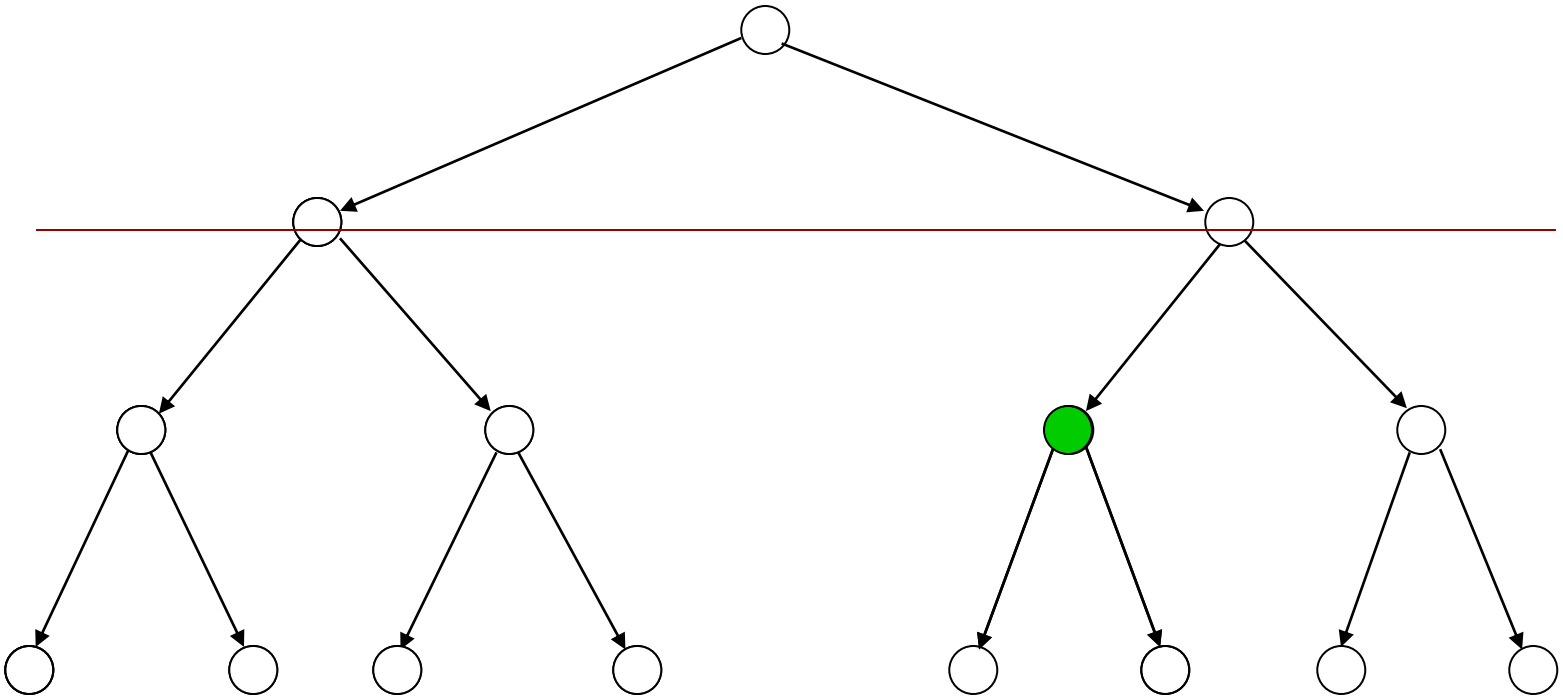

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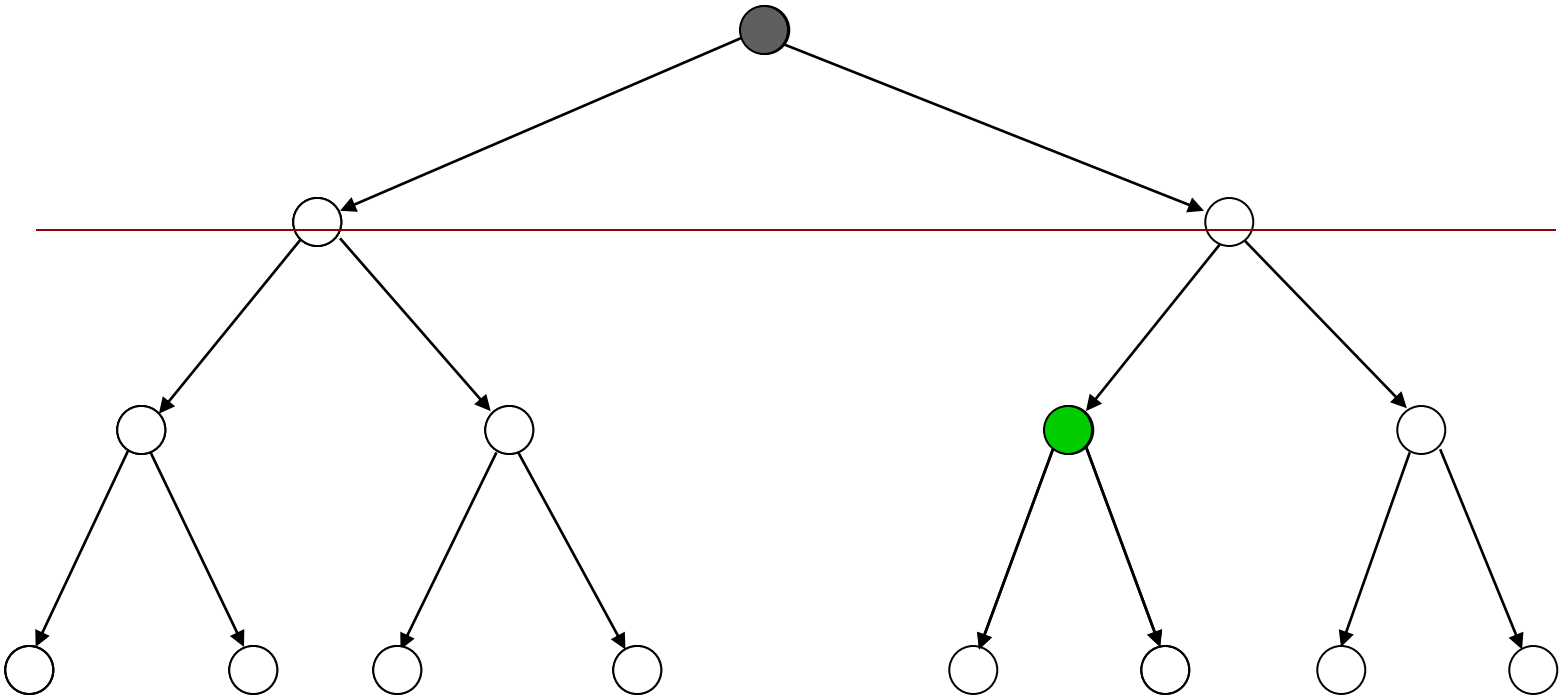
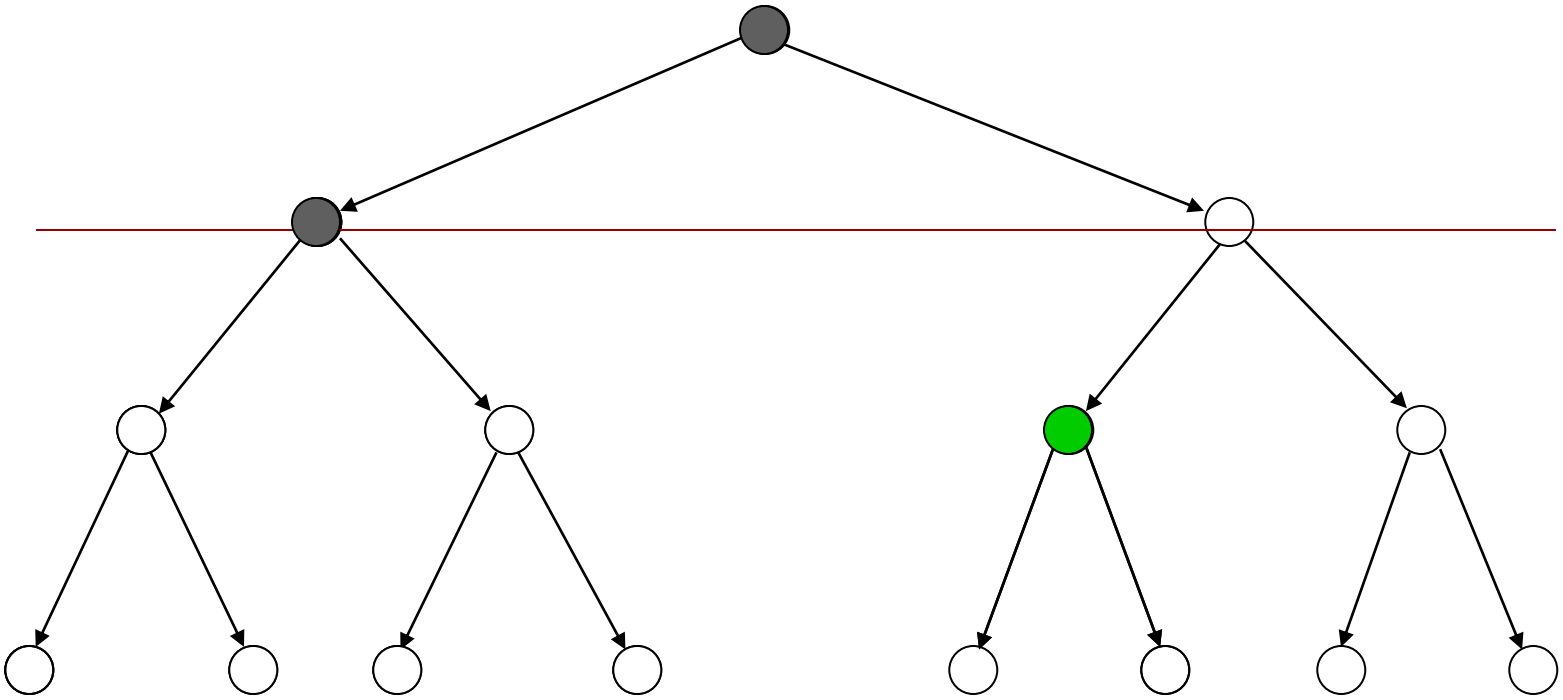


Iterative Deepening



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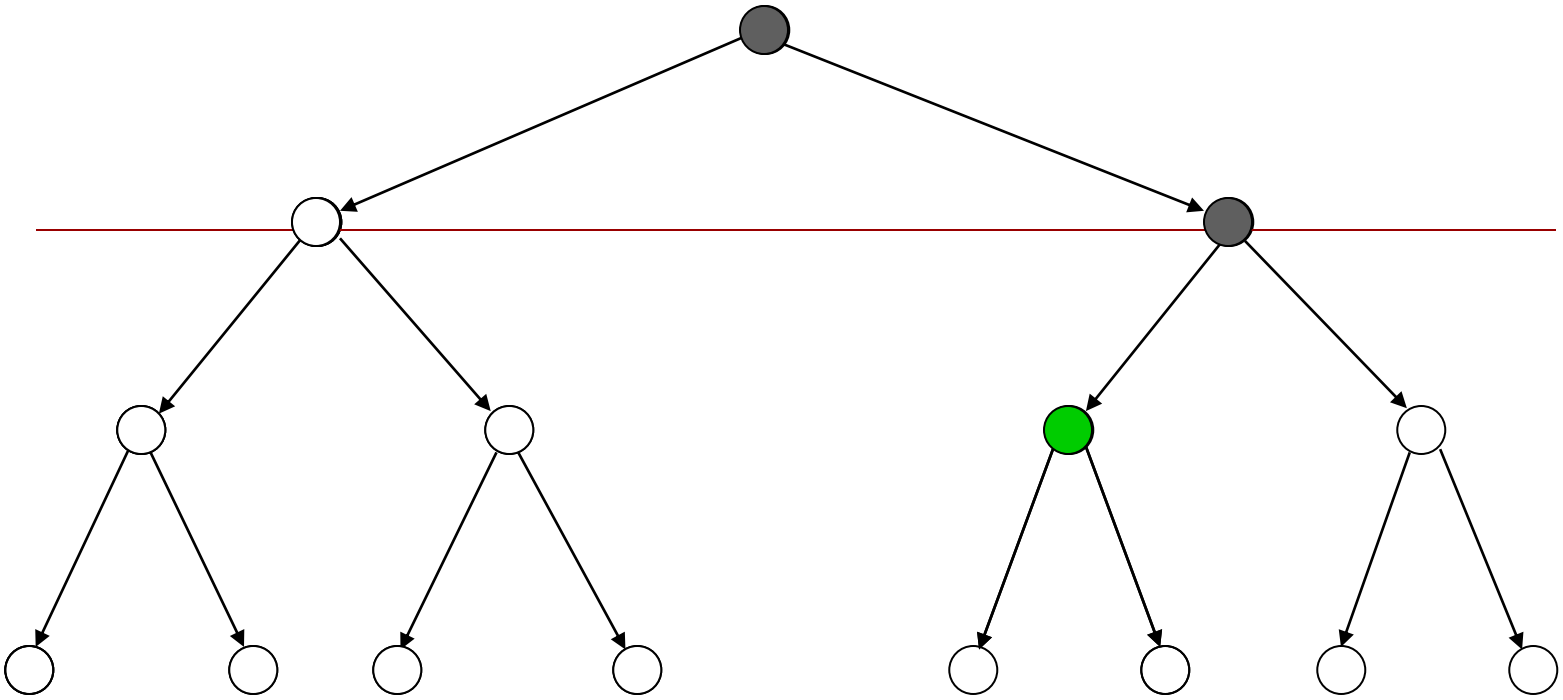

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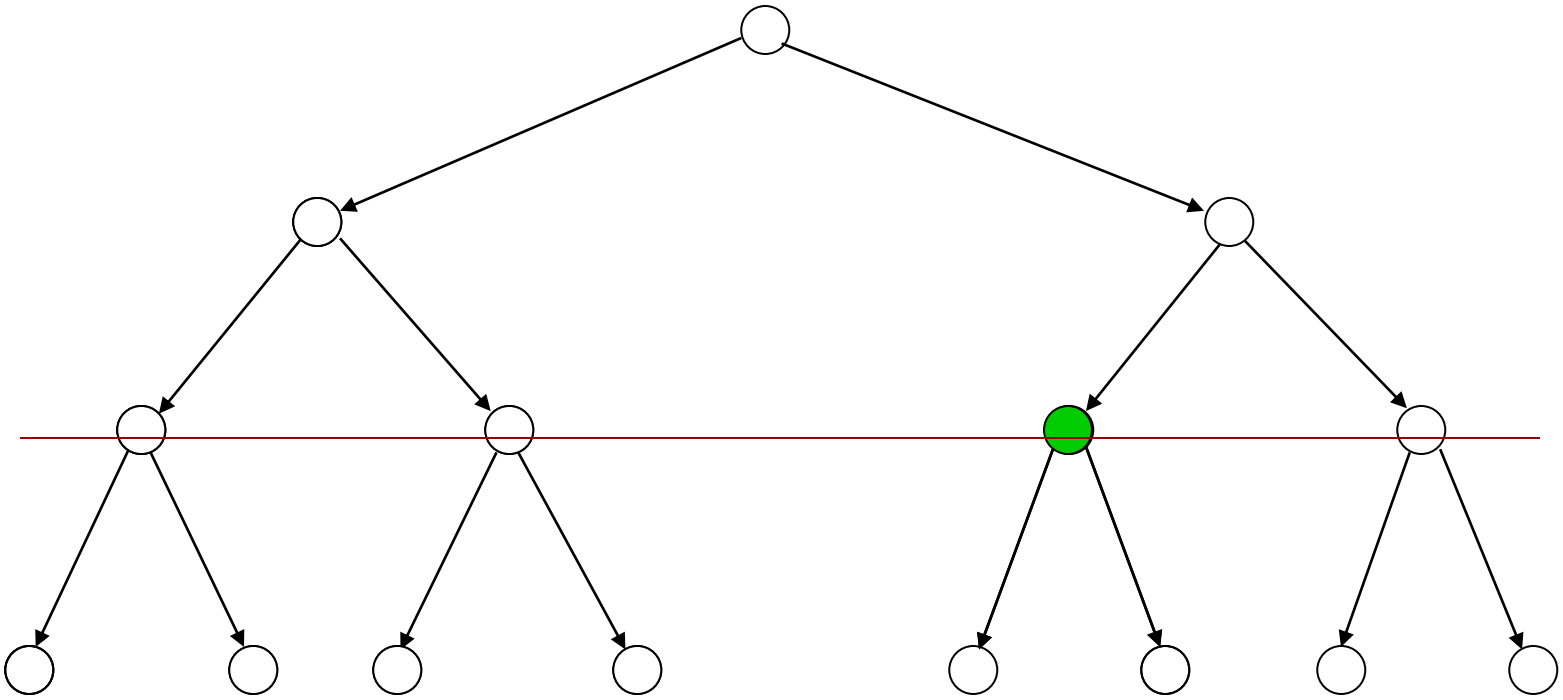

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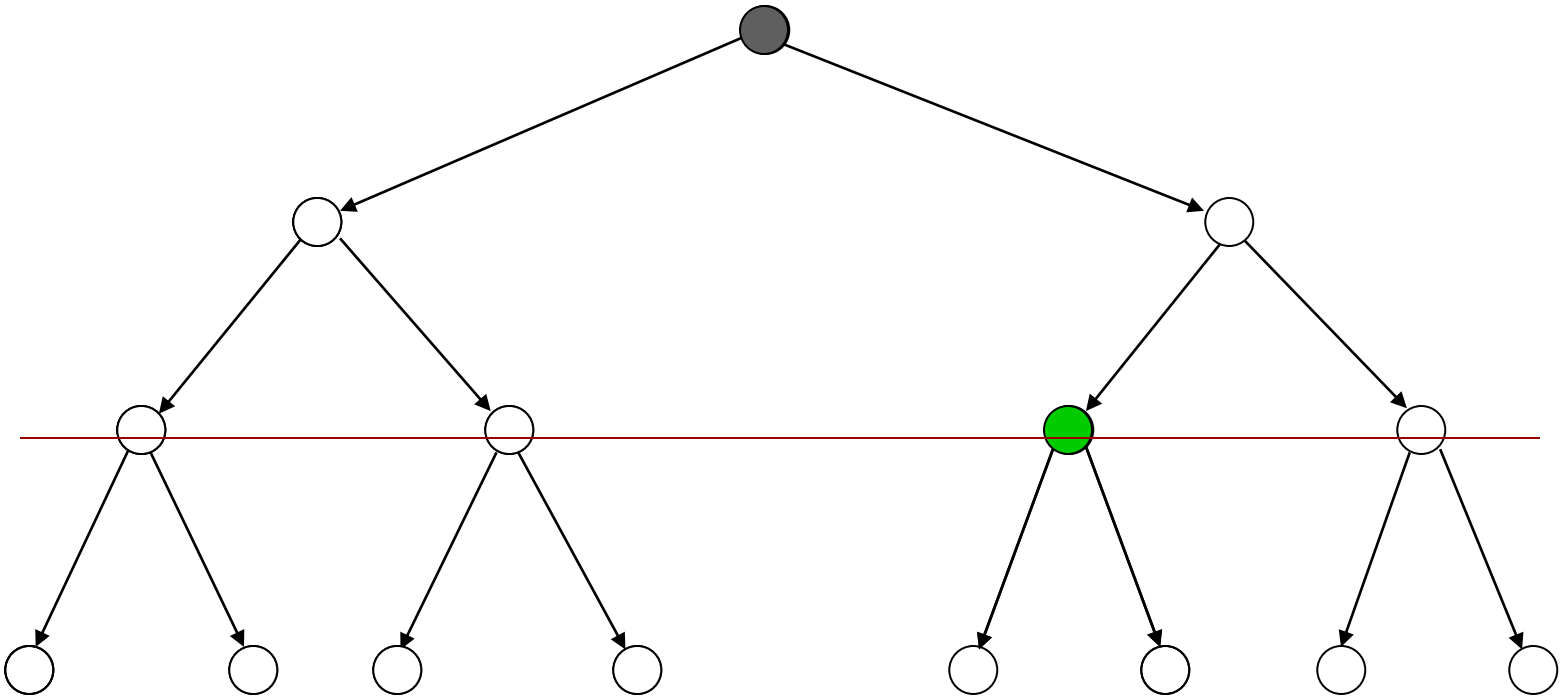

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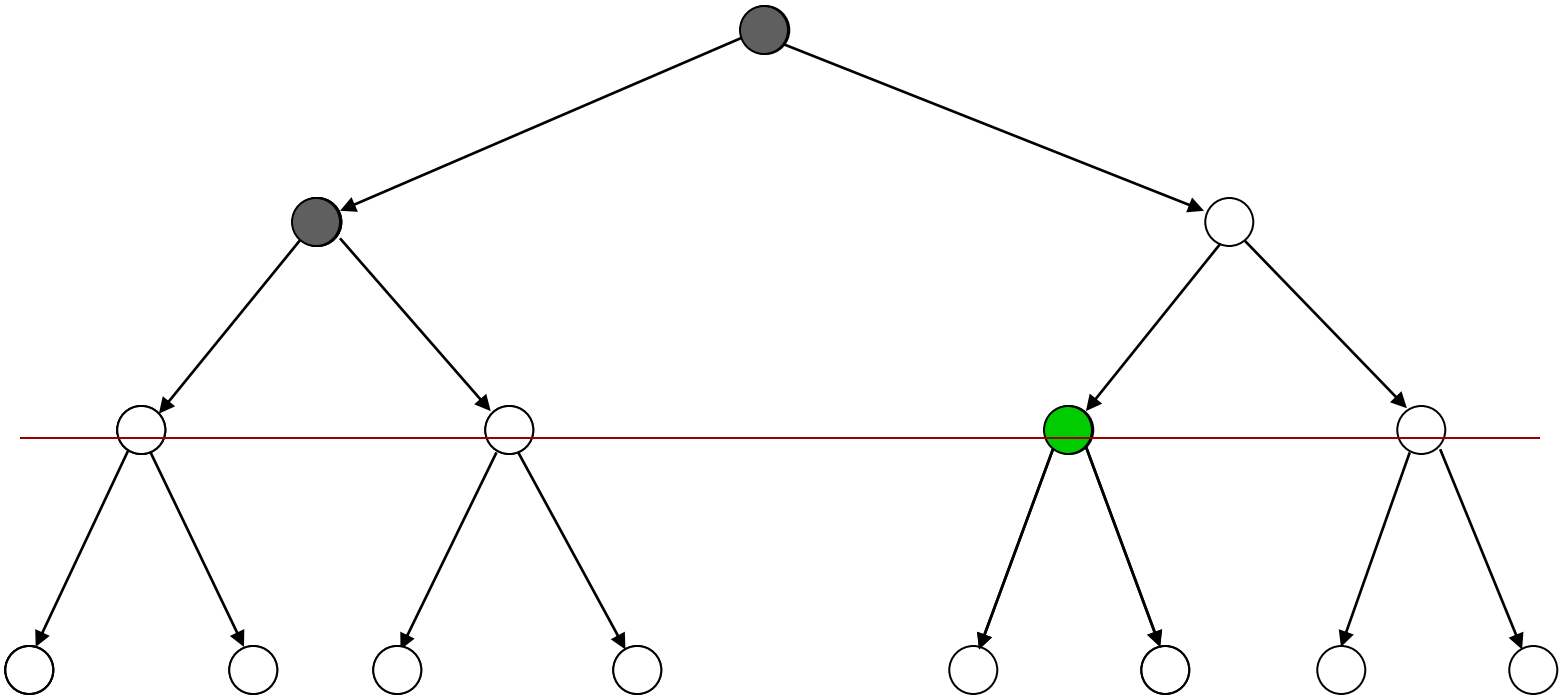

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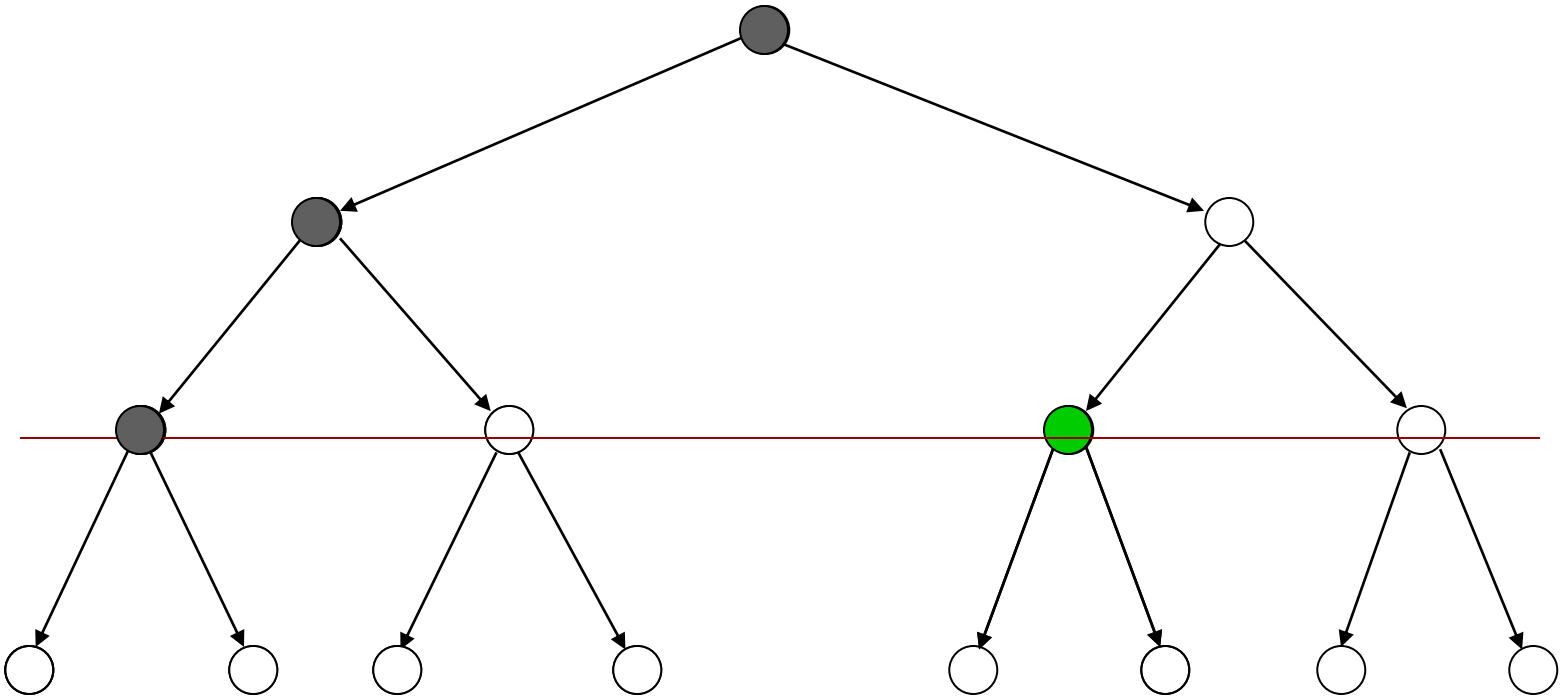

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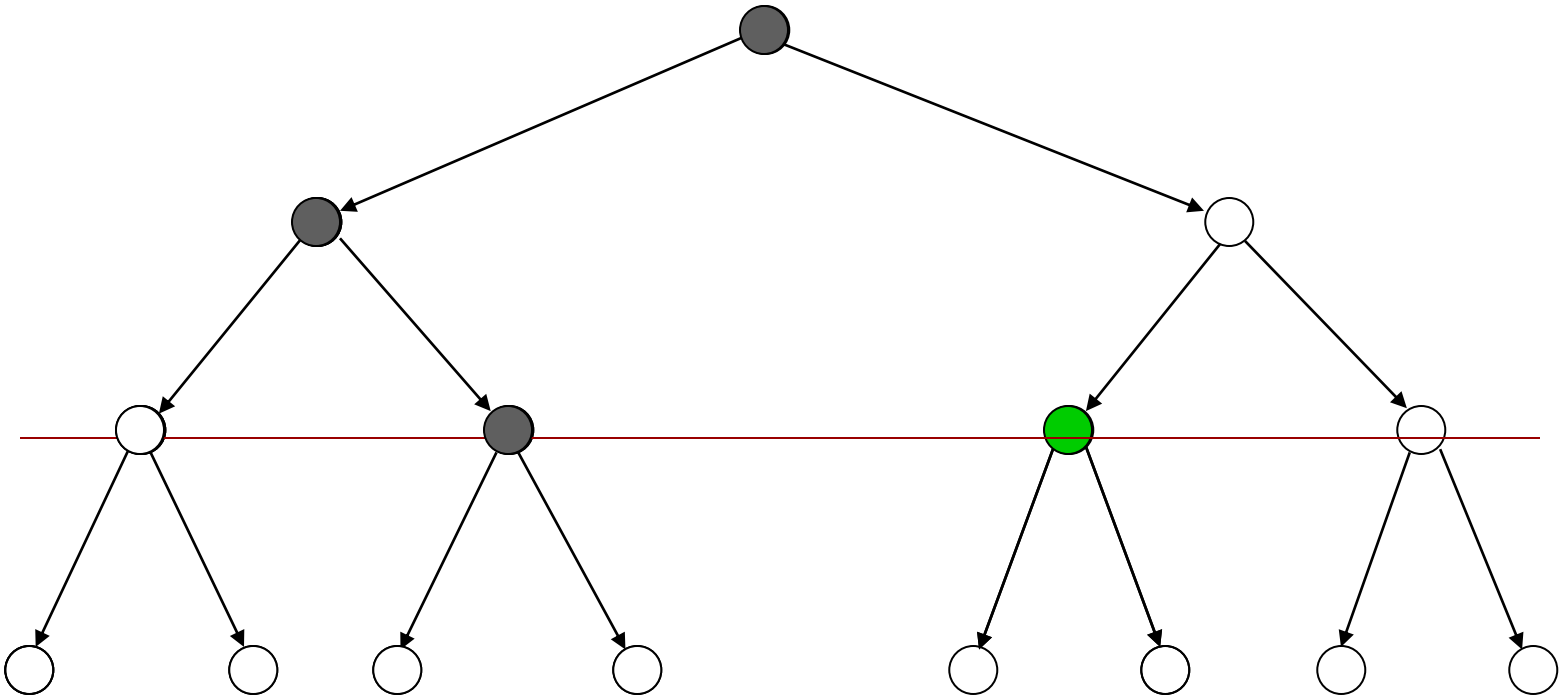
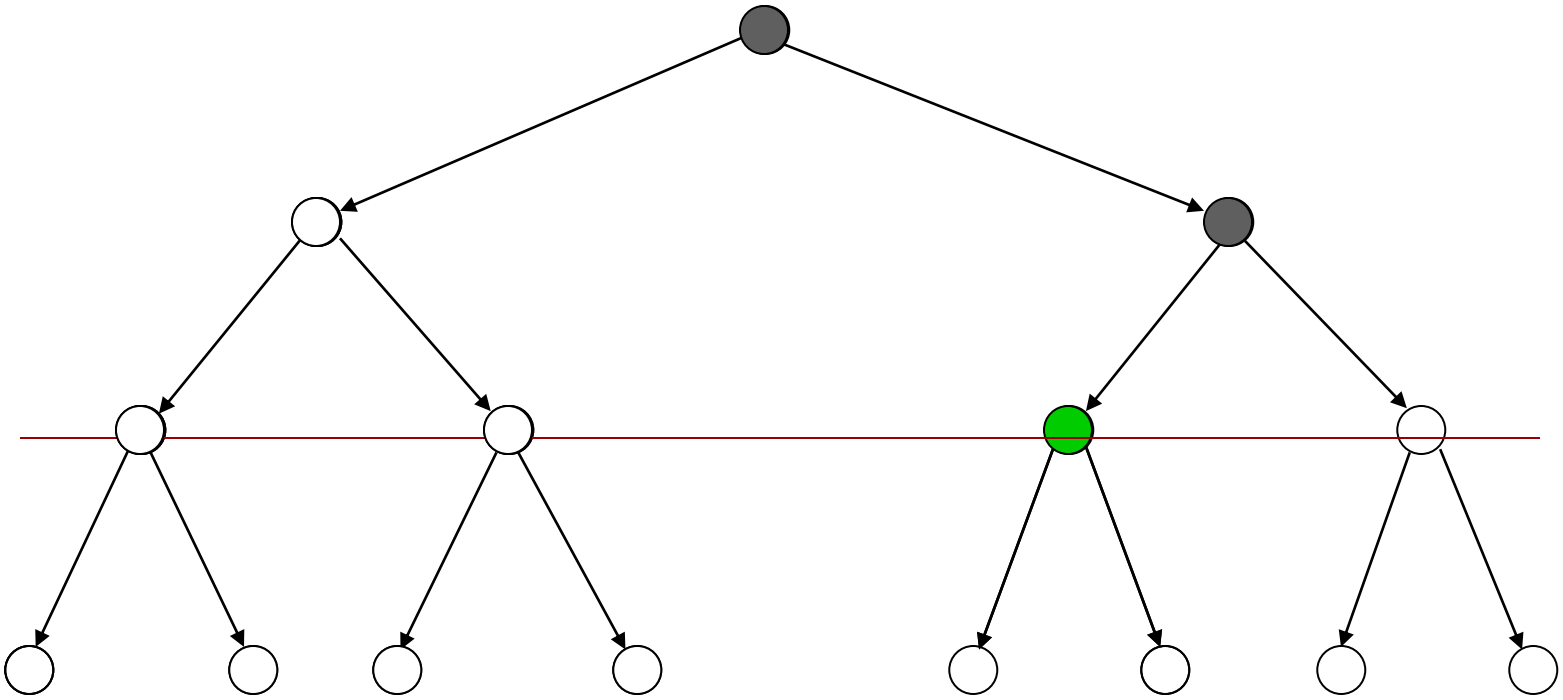
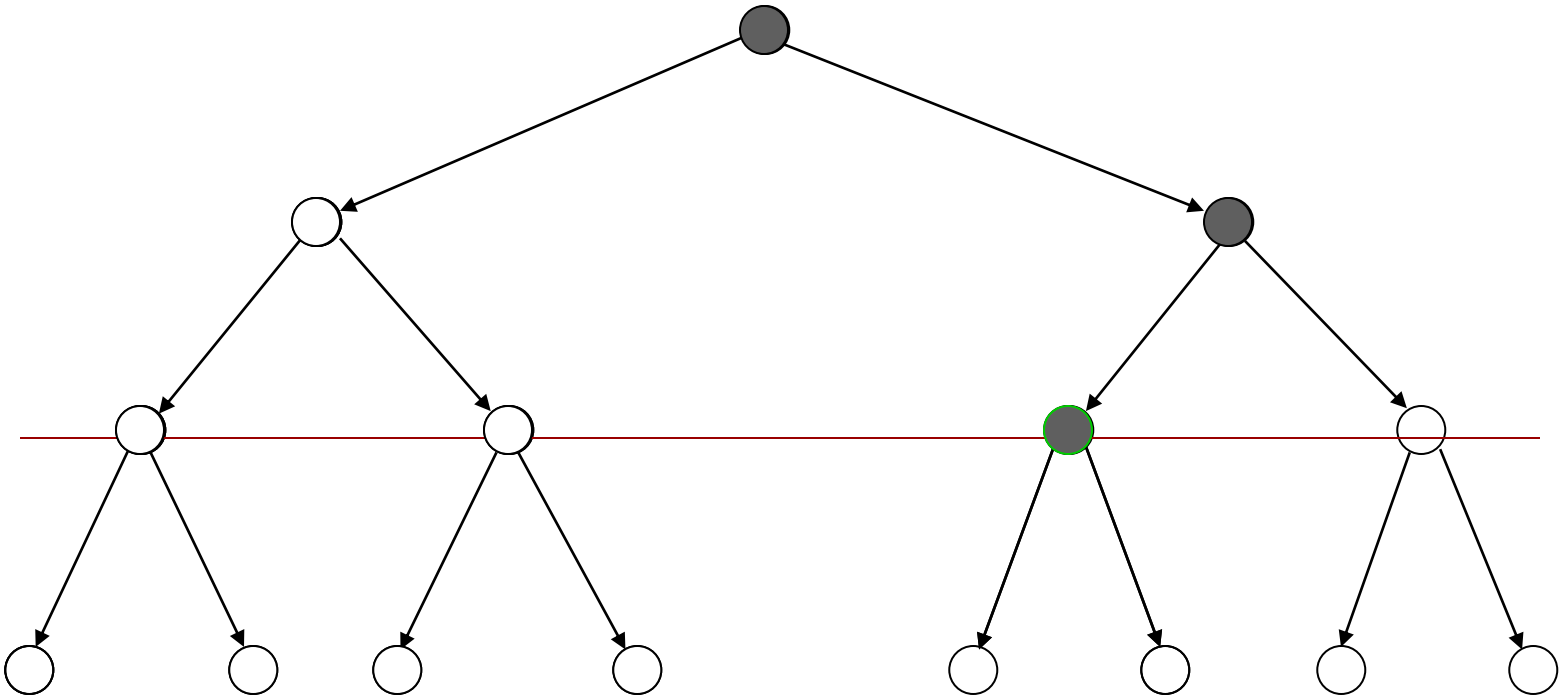


Iterative Deepening



Iterative Deepening



Performance

- Iterative deepening search is:
  - Complete (for finite b and d)
  - Optimal
    - if path cost is a non-decreasing function of the node depth. For example if step cost =1
- Time complexity is:

  $db + (d-1)b^2 + \ldots + (1)\, b^d = O(b^d)$
- Space complexity is: $O(bd)$ or $O(d)$

IDS is the **preferred method** when search space is large and the depth of solution is unknown

Number of Generated Nodes (Breadth-First & Iterative Deepening)

d = 5 and b = 2

BF	ID
1	1 x 6 = 6
2	2 x 5 = 10
4	4 x 4 = 16
8	8 x 3 = 24
16	16 x 2 = 32
32	32 x 1 = 32
63	120

120/63 ~ 2

Number of Generated Nodes (Breadth-First & Iterative Deepening)

d = 5 and b = 10

BF	ID
1	6
10	50
100	400
1,000	3,000
10,000	20,000
100,000	100,000
111,111	123,456

123,456/111,111 ~ 1.111

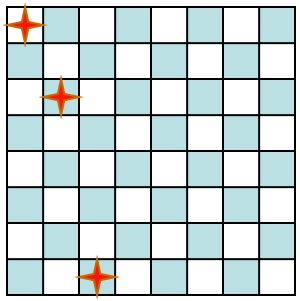
Comparison of Strategies

- Breadth-first is complete and optimal, but has high space complexity
- Depth-first is space efficient, but is neither complete, nor optimal
- Iterative deepening is complete and optimal, with the same space complexity as depth-first and almost the same time complexity as breadth-first

Quiz: Would IDS + bi-directional search be a good combination?

Revisited States

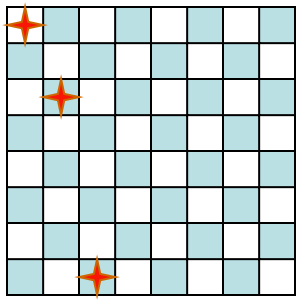
No



8-queens

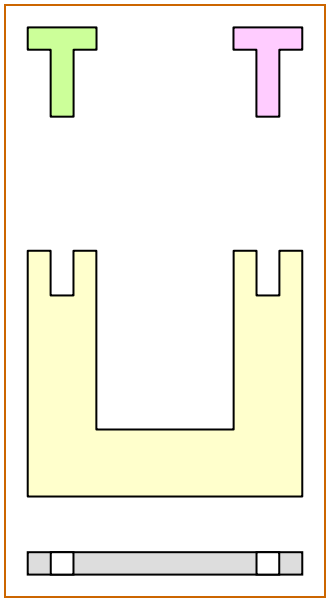
Revisited States

No



8-queens

Few

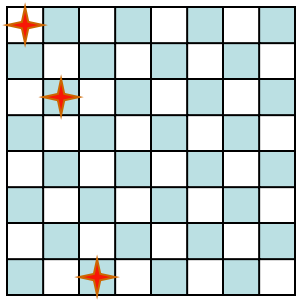


assembly planning

Revisited States

No Few

search tree is finite



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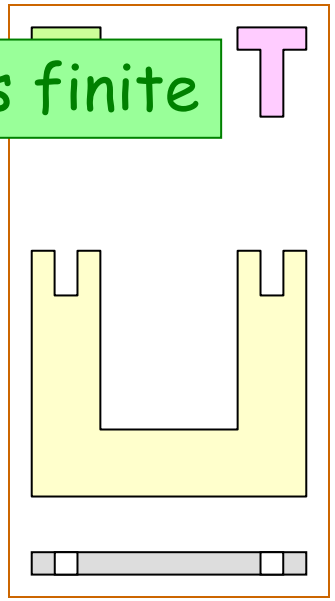
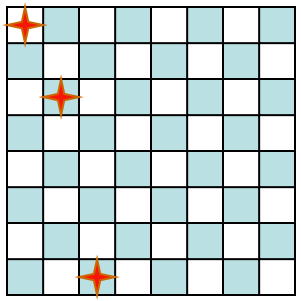


assembly planning

Revisited States

No Few Many

search tree is finite



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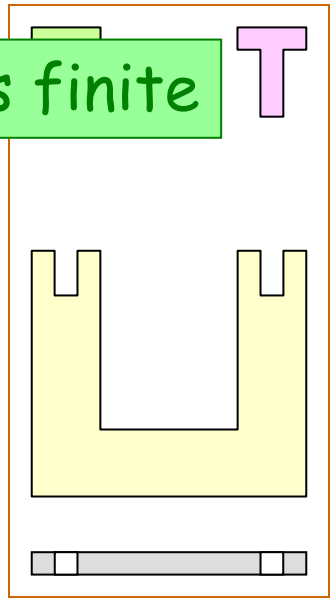

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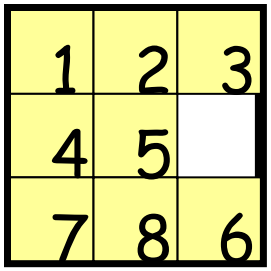
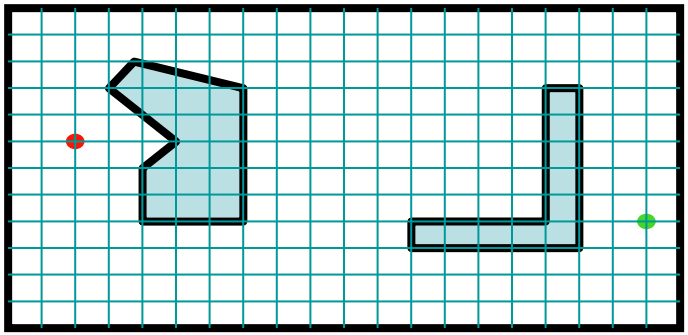




8-puzzle and robot navigation

Revisited States

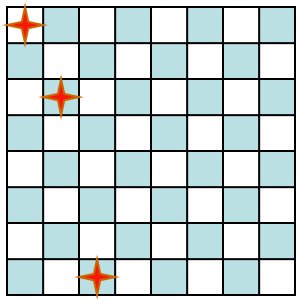
No

Few

Many

search tree is finite

search tree is infinite



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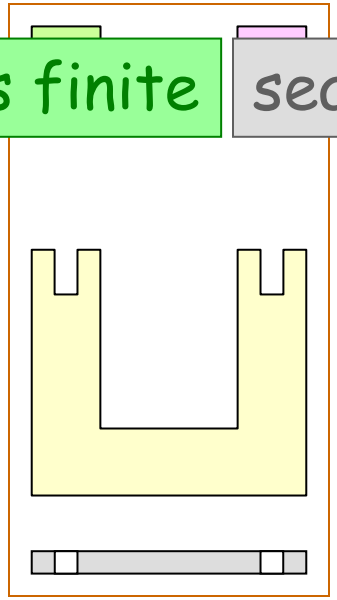
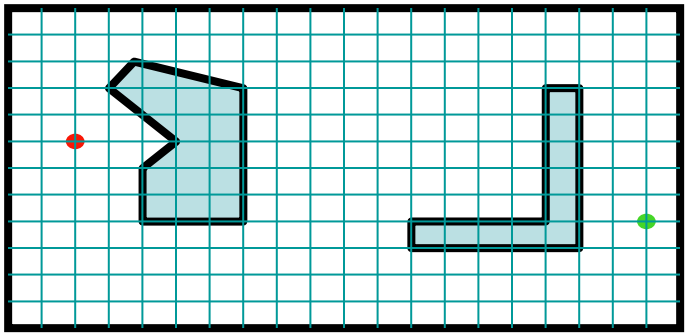


assembly planning



8-puzzle and robot navigation

Avoiding Revisited States

- Requires comparing state descriptions
- Breadth-first search:
  - Store all states associated with **generated** nodes in VISITED
  - If the state of a new node is in VISITED, then discard the node

Avoiding Revisited States

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Implemented as hash-table
or as explicit data structure with flags

Avoiding Revisited States

- Depth-first search:

  Solution 1:

  - Store all states associated with nodes in current path in VISITED
  - If the state of a new node is in VISITED, then discard the node

  → ??

Avoiding Revisited States

- Depth-first search:

  Solution 1:

  - Store all states associated with nodes in current path in VISITED
  - If the state of a new node is in VISITED, then discard the node

  → Only avoids loops

  Solution 2:

  - Store all generated states in VISITED
  - If the state of a new node is in VISITED, then discard the node

  → Same space complexity as breadth-first !

Uniform-Cost Search

- Each arc has some cost $c \geq \varepsilon > 0$
- The cost of the path to each node N is

  $g(N) = \Sigma$ costs of arcs
- The goal is to generate a solution path of minimal cost
- The nodes N in the queue FRINGE are sorted in increasing $g(N)$

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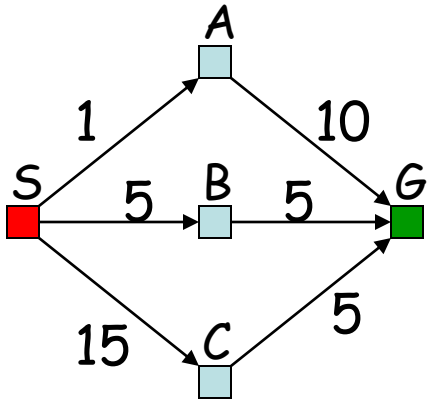

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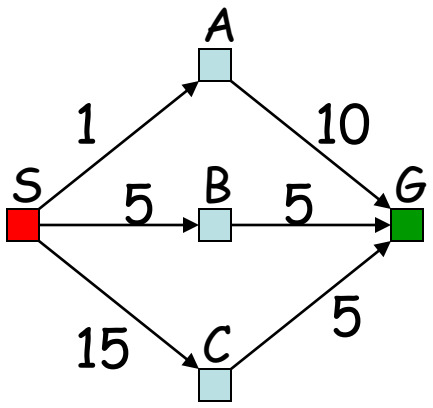

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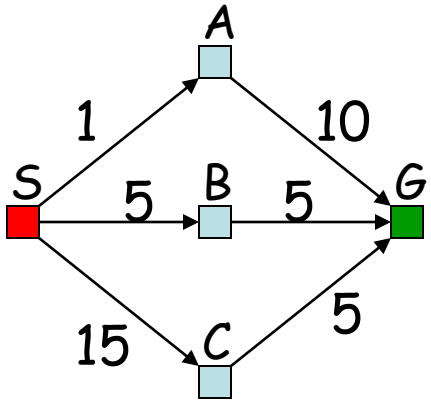

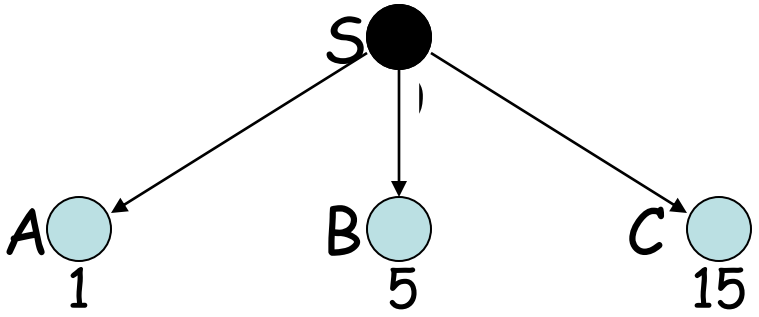

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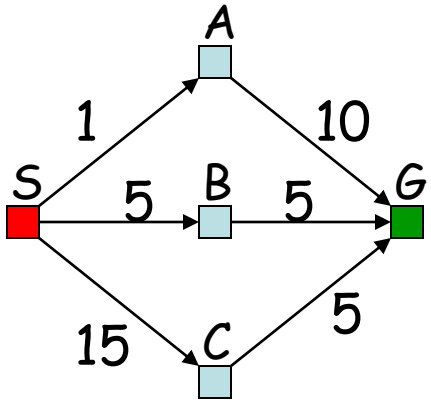

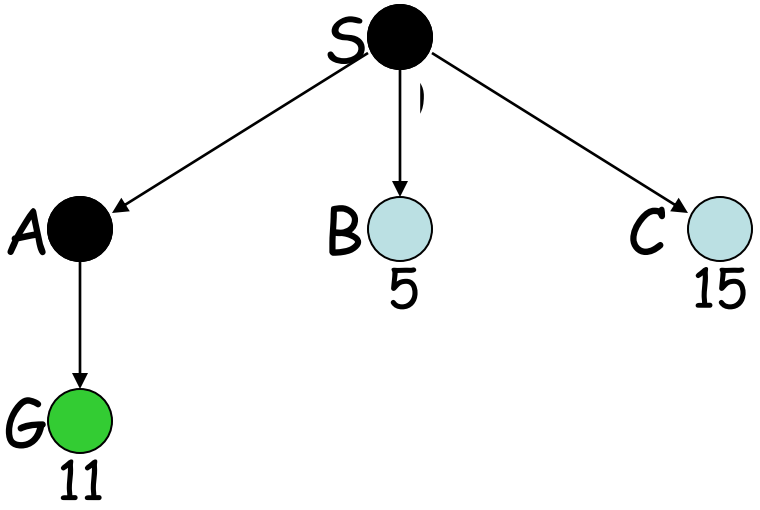

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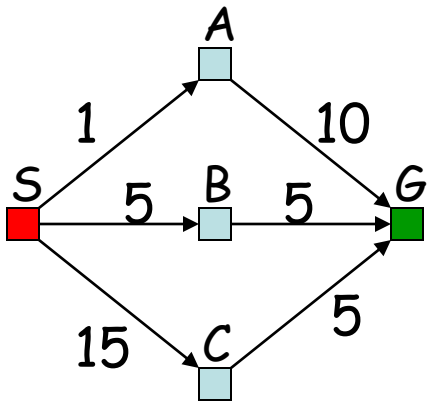
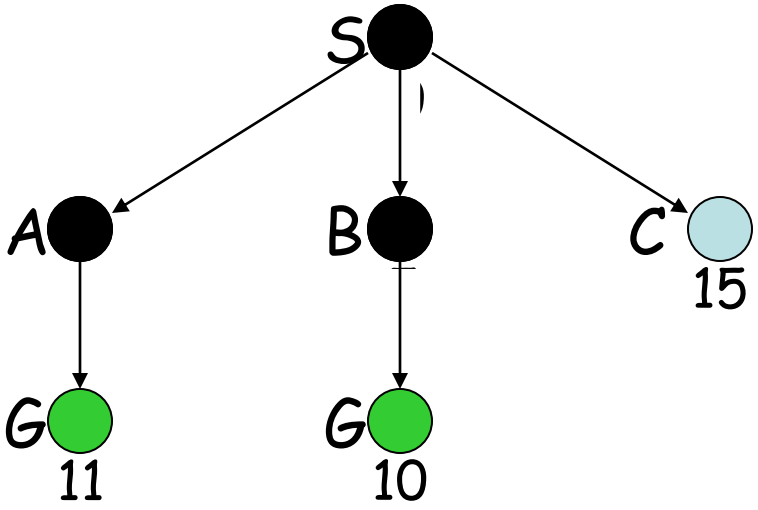




- Need to modify search algorithm

Search Algorithm #2

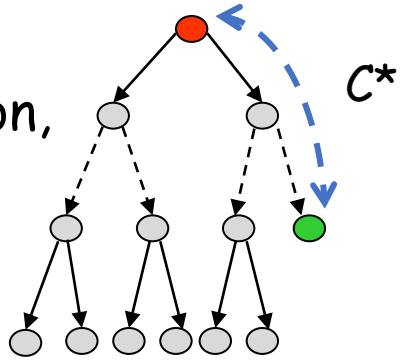
SEARCH#2

1. INSERT(initial-node,FRINGE)
2. Repeat:
   a. If empty(FRINGE) then return failure
   b. N ← REMOVE(FRINGE)
   c. s ← STATE(N)
   d. If GOAL?(s) then return path or goal state
   e. For every state s' in SUCCESSORS(s)
      i. Create a node N' as a successor of N
      ii. INSERT(N',FRINGE)

The goal test is applied to a node when this node is **expanded**, not when it is generated.

Properties of UCS

- Complete? Yes, if step-cost ≥ ε > 0
  - to avoid infinit sequence of zero-cost actions
- Time?
  - Number of nodes with $g \leq$ cost of optimal solution,
    - $O(b^{1+[C^*/\varepsilon]})$ where C^* is the optimal solution cost
    - $O(b^{d+1})$ where all step costs are equal



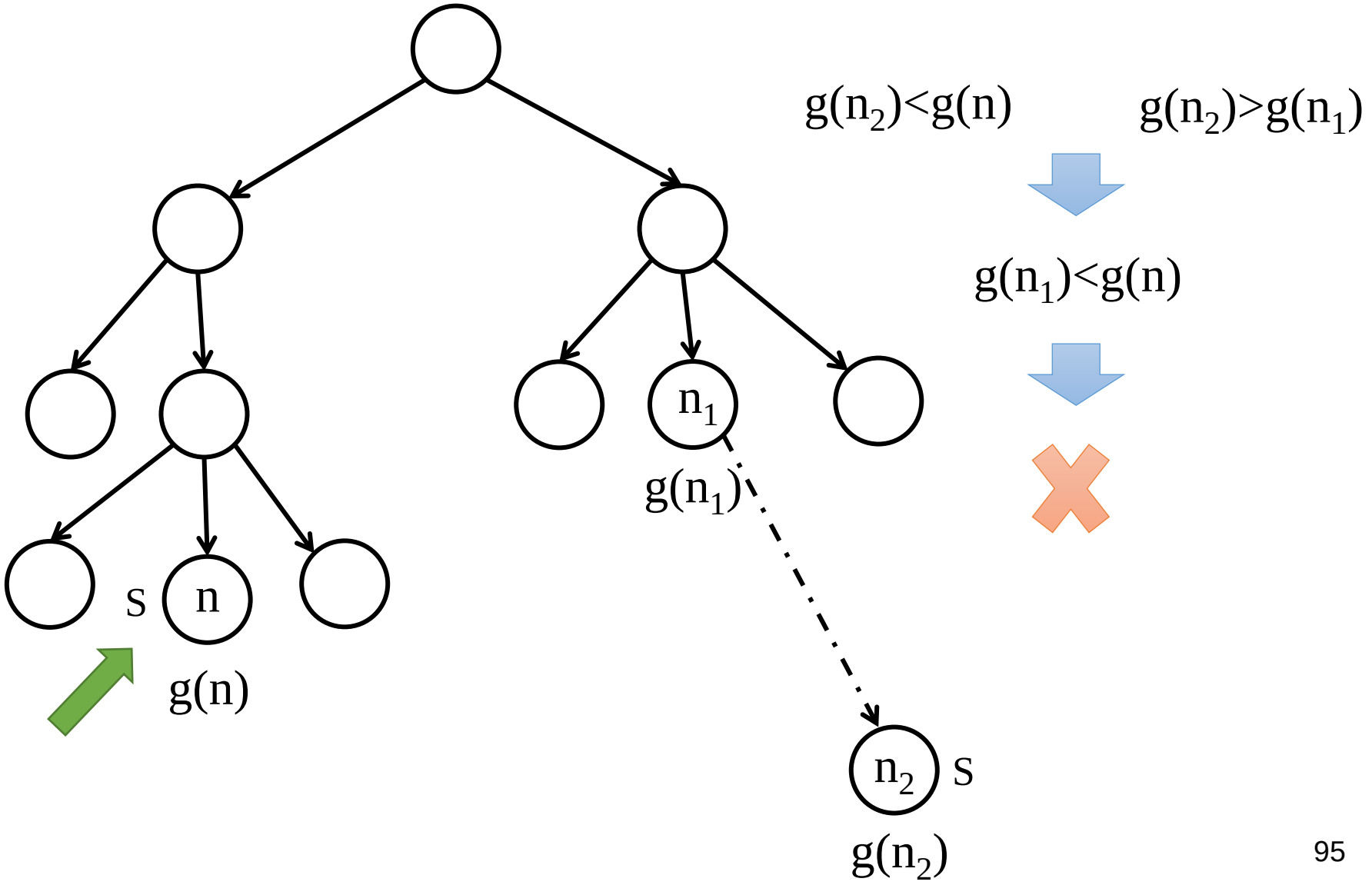
- Space
  - Number of nodes with $g \leq$ cost of optimal solution
    - $O(b^{1+[C^*/\varepsilon]})$ where C^* is the optimal solution cost
- Optimal?
  - Yes – nodes expanded in increasing order of g(n)
- Difficulty: many long paths may exist with cost ≤ C^*

UCS (proof of optimality)

- Lemma:
  - If UCS selects a node n for expansion, the optimal solution to that node has been found.
- Proof
  - Proof by contradiction: Another fringe (frontier) node n′ must exist on the optimal path from initial node to n (using graph separation property). Moreover, based on definition of path cost (due to non-negative step costs, paths never get shorter as nodes are added), we have $g(n') \leq g(n)$ and thus n′ would have been selected first.

Nodes are expanded in order of their optimal path cost

UCS (proof of optimality)



Avoiding Revisited States in Uniform-Cost Search

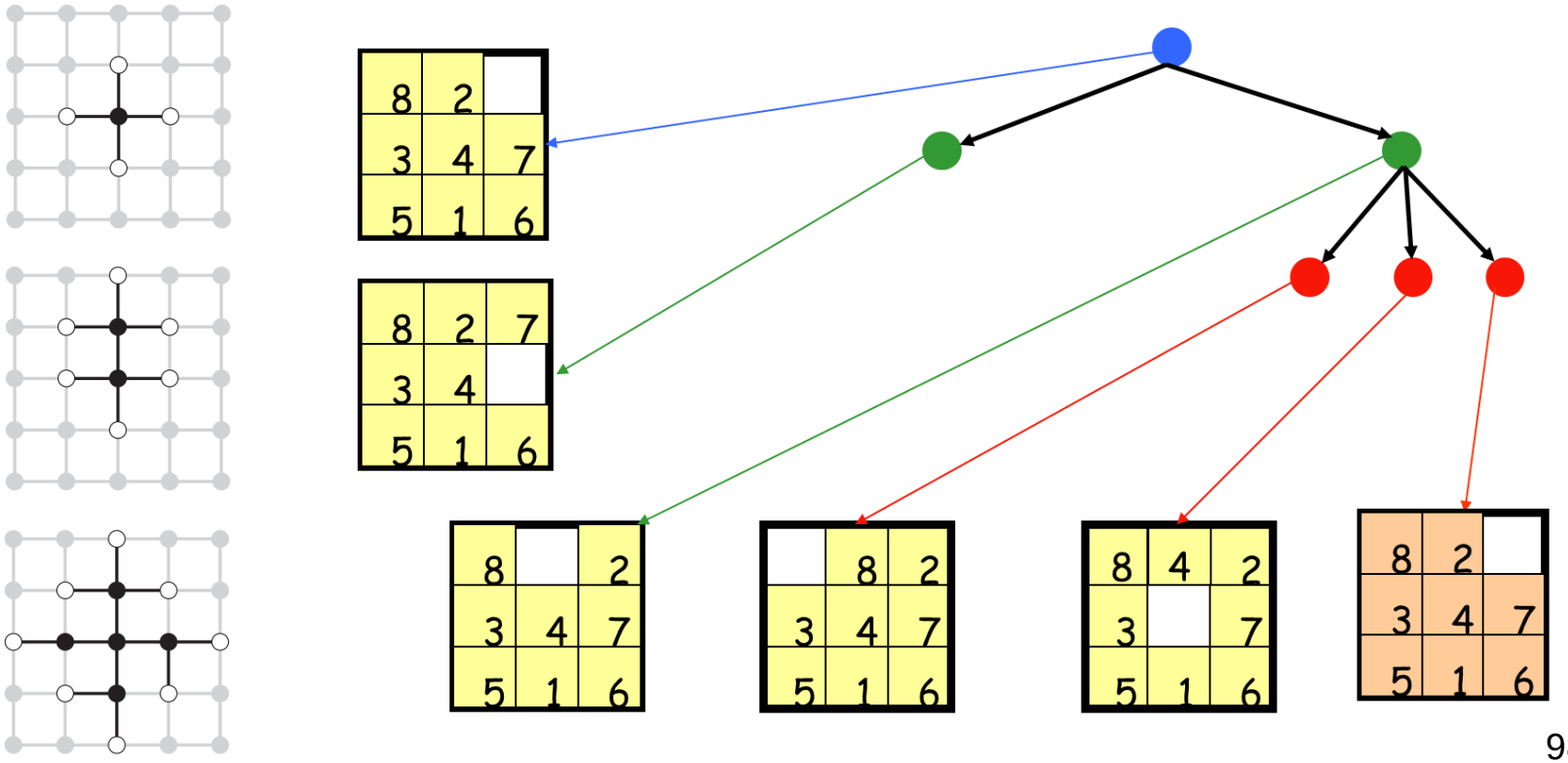
- For any state S, when the first node N such that STATE(N) = S is expanded, the path to N is the best path from the initial state to S
- So:
  - When a node is **expanded**, store its state into CLOSED (Explored)
  - When a new node N is generated:
    - If STATE(N) is in CLOSED (Explored), discard N
    - If there exits a node N' in the fringe such that STATE(N') = STATE(N), discard the node -- N or N' -- with the highest-cost path

Tree search

```
function TREE-SEARCH(problem) returns a solution, or failure
    initialize the frontier using the initial state of problem
    loop do
        if the frontier is empty then return failure
        choose a leaf node and remove it from the frontier
        if the node contains a goal state then return the
            corresponding solution
        expand the chosen node, adding the resulting nodes
            to the frontier
```

Graph search

- Redundant paths in tree search: more than one way to get from one state to another

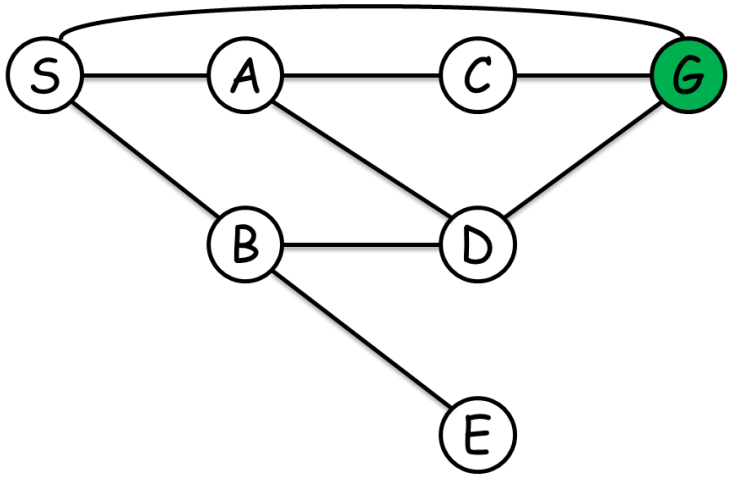


Graph search

```
function GRAPH-SEARCH(problem) returns a solution, or failure
    initialize the frontier using the initial state of problem
    initialize the explored set to be empty
    loop do
        if the frontier is empty then return failure
        choose a leaf node and remove it from the frontier
        if the node contains a goal state then return the corresponding solution
        add the node to the explored set
        expand the chosen node, adding the resulting nodes to the frontier
         only if not in the frontier or explored set
```

BFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order



Frontier	Explored

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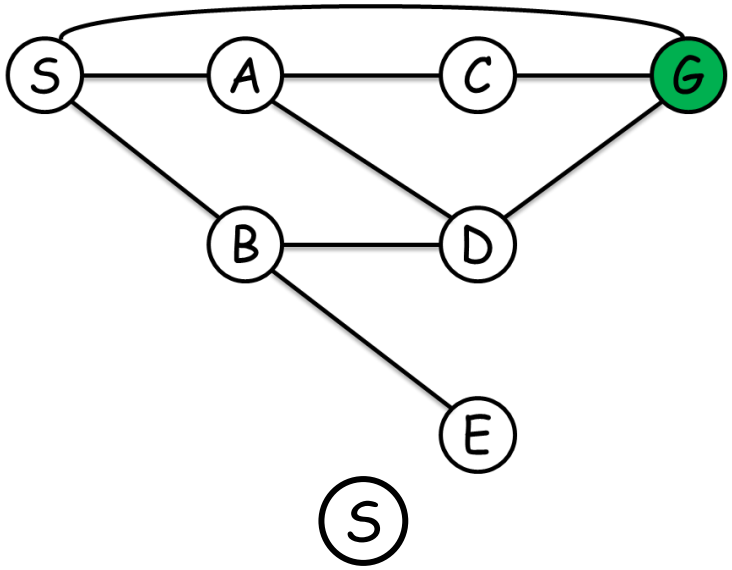

Frontier	Explored
S	

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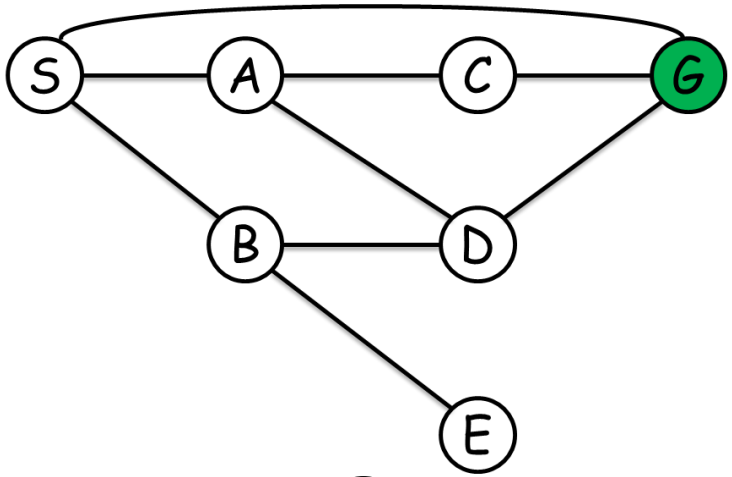
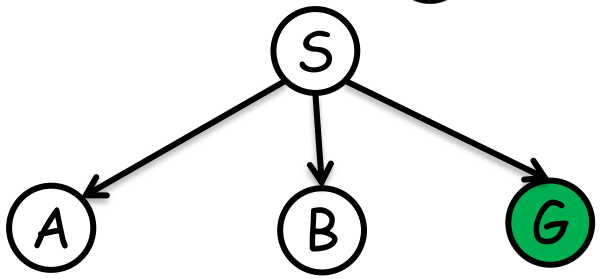


Frontier	Explored
S	
A, B, G	S



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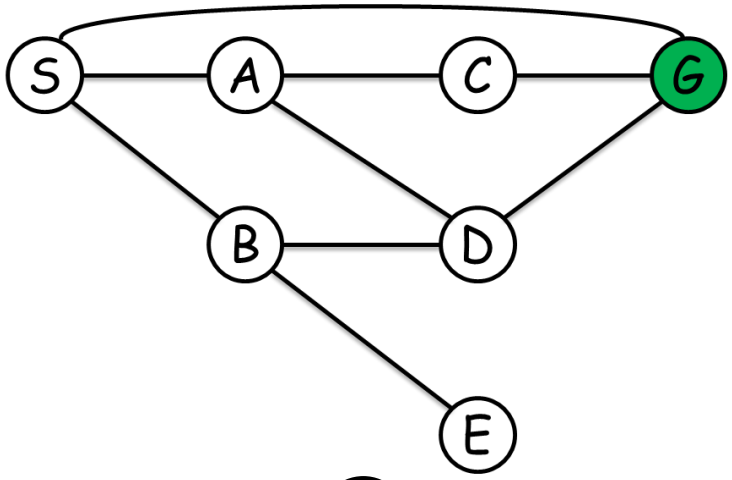
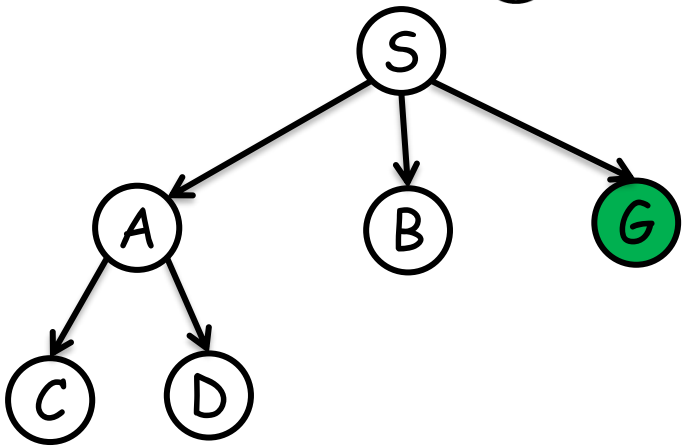
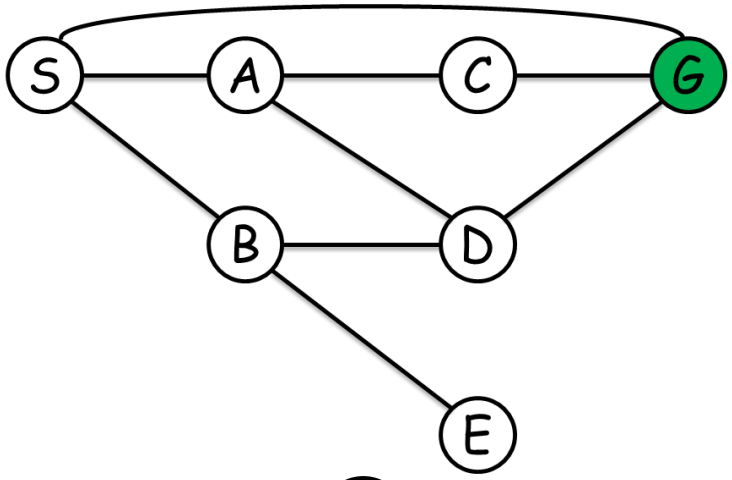


Frontier	Explored
S	
A, B, G	S
B, G, C, D	S, A

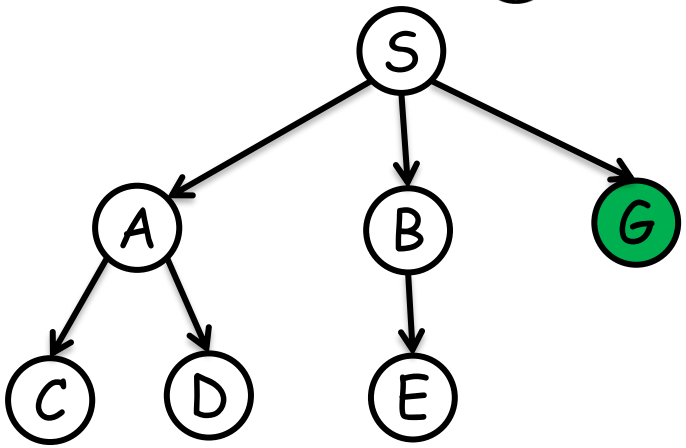


BFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order

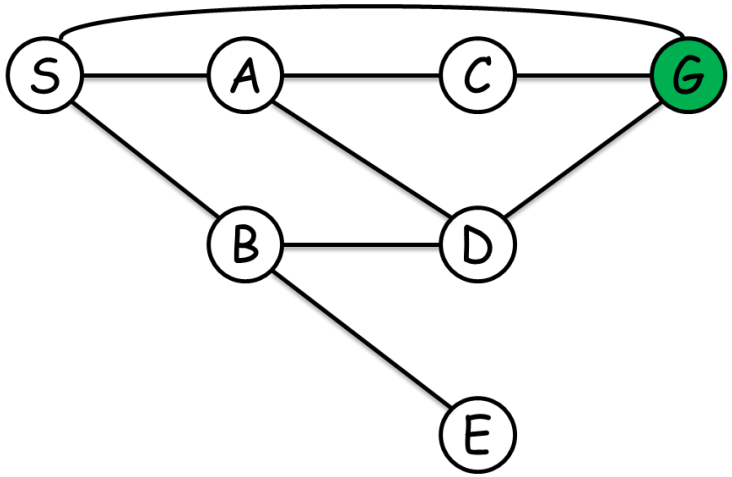


Frontier	Explored
S	
A, B, G	S
B, G, C, D	S, A
G, C, D, E	S, A, B



BFS Example

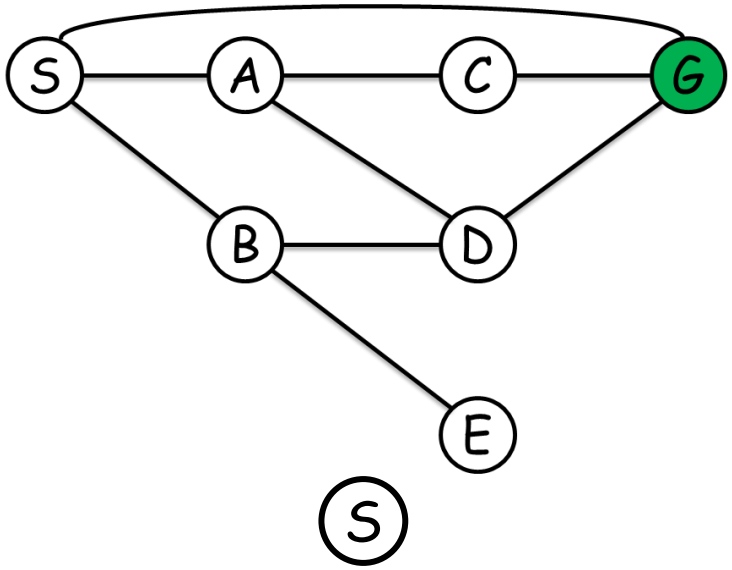
- Tree search
- Goal test when expanding a node
- Alphabetical order



Frontier

BFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

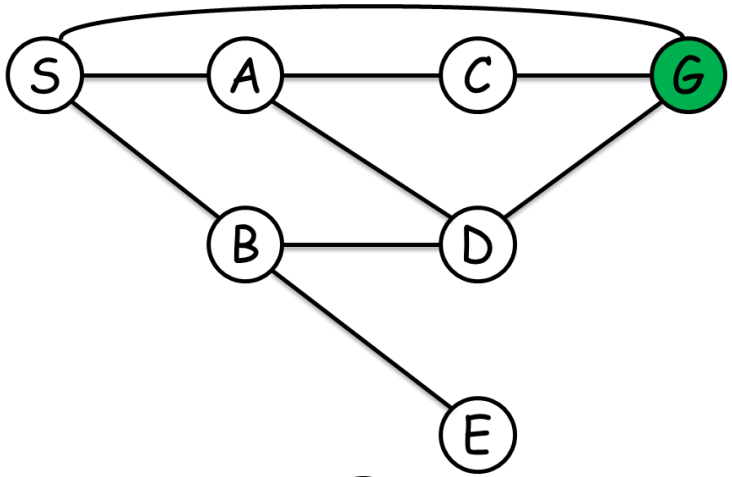


Frontier

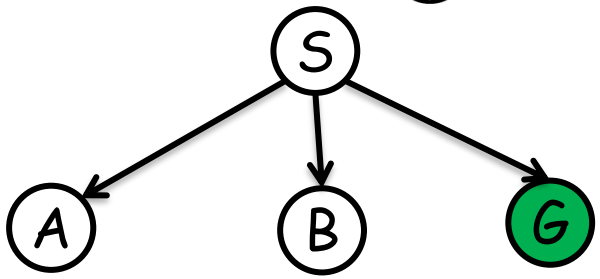
S

BFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

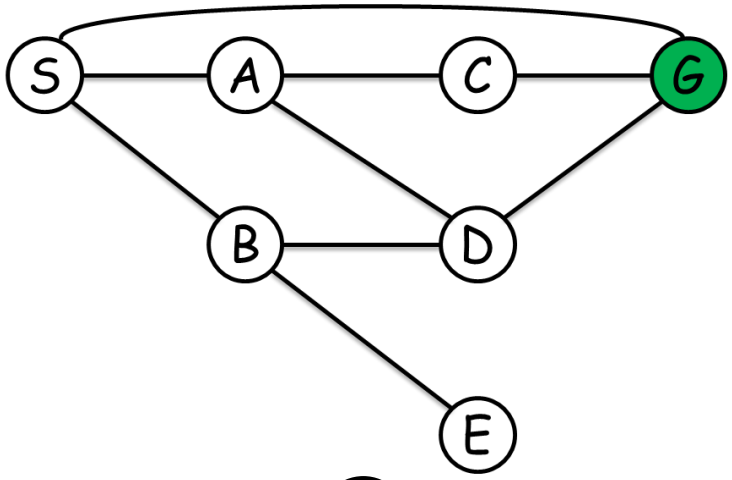


Frontier
S
A, B, G

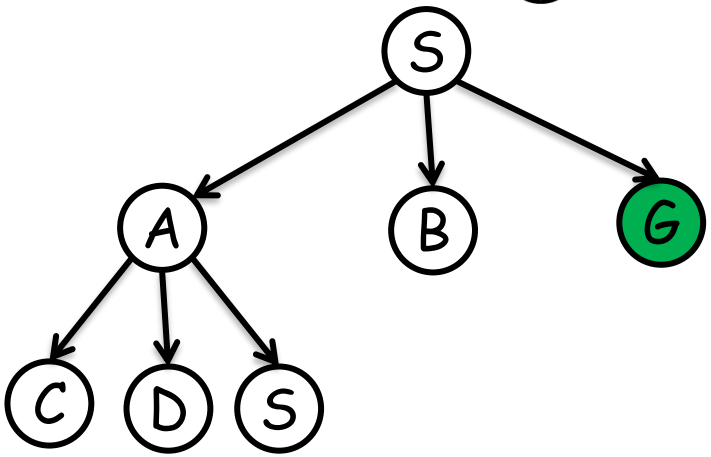


BFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

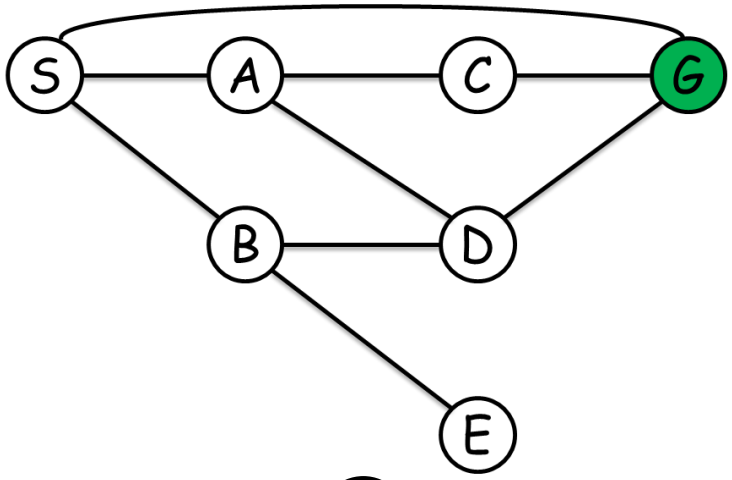


Frontier
S
A, B, G
B, G, C, D, S

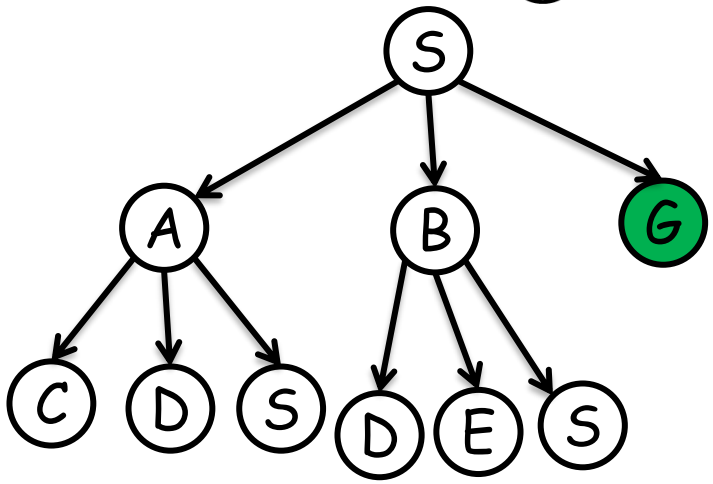


BFS Example

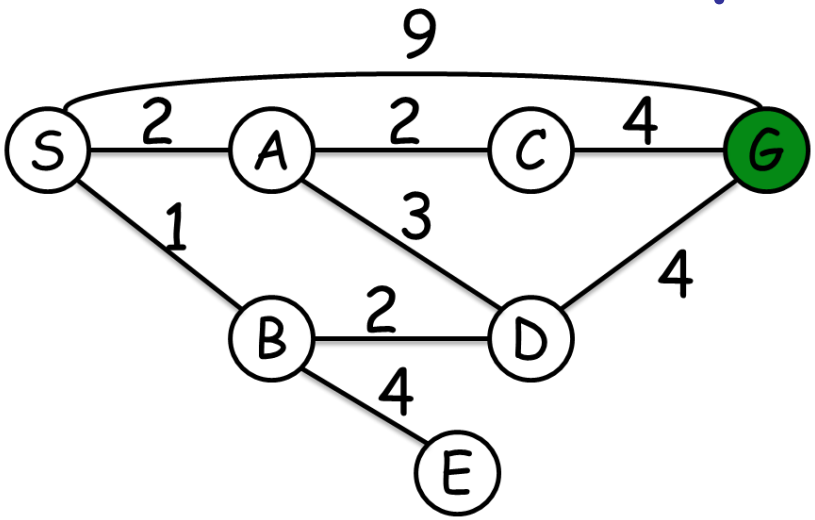
- Tree search
- Goal test when expanding a node
- Alphabetical order



Frontier
S
A, B, G
B, G, C, D, S
G, C, D, S, D, E, S

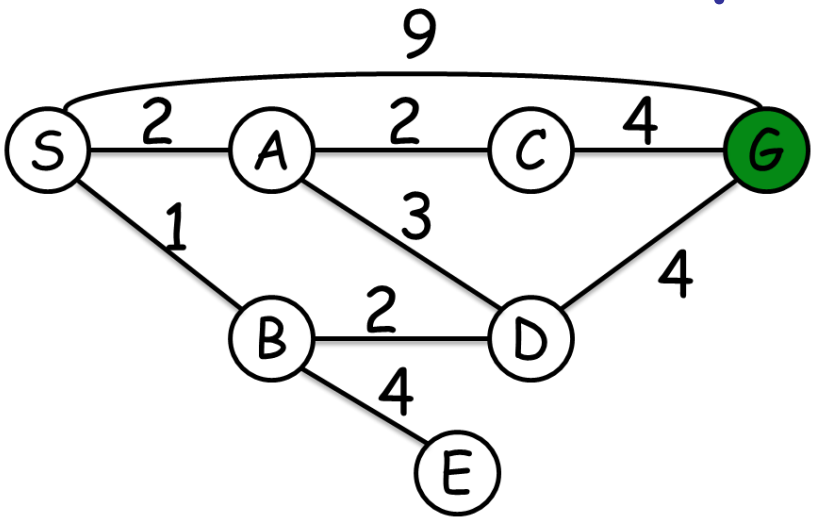


UCS Example (Graph search)



Frontier	Explored

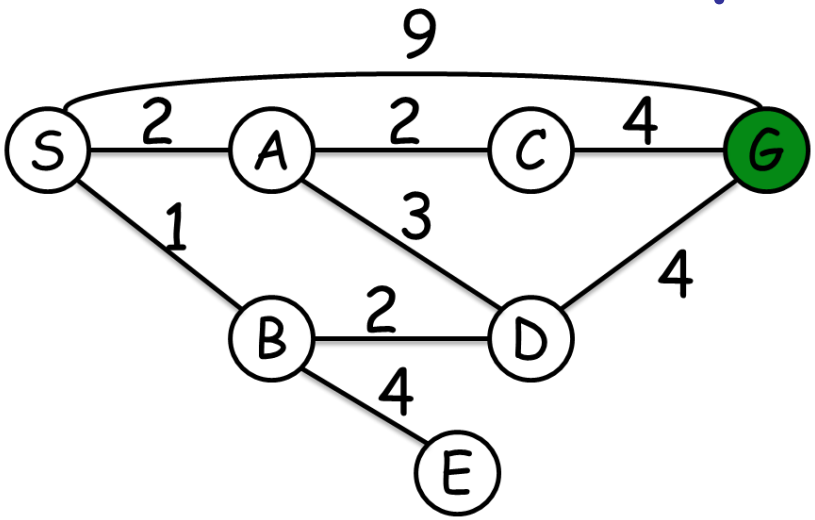
UCS Example (Graph search)

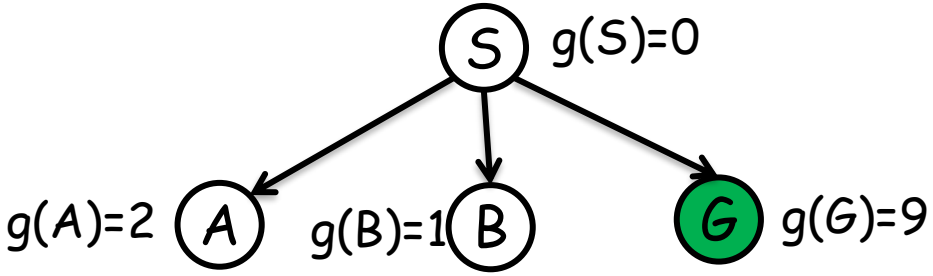


S g(S)=0

Frontier	Explored
S(0)	

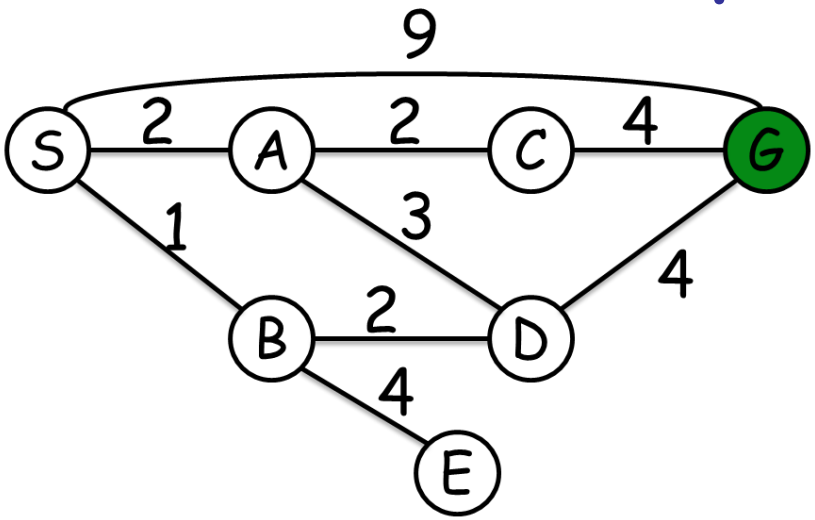
UCS Example (Graph search)



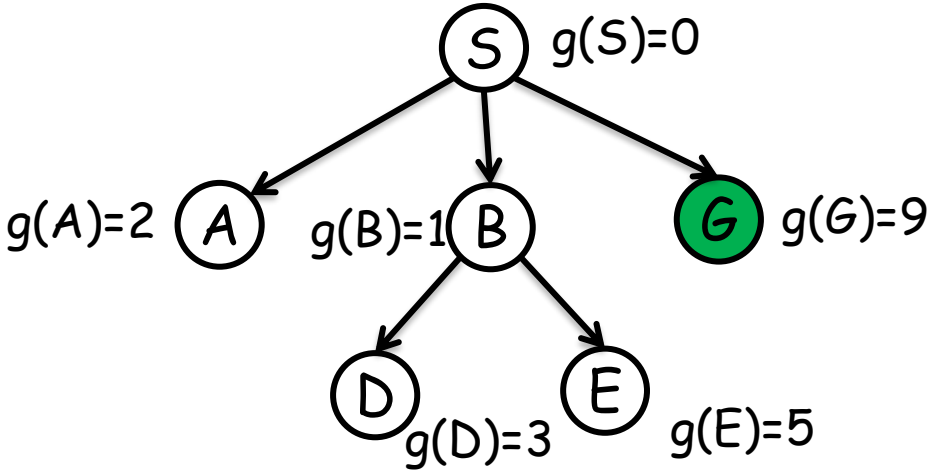


Frontier	Explored
S(0)	
B(1), A(2), G(9)	S

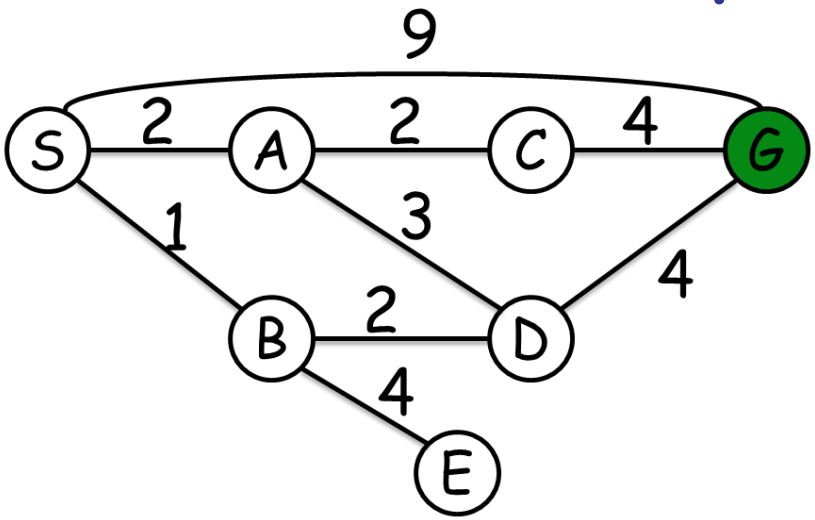
UCS Example (Graph search)



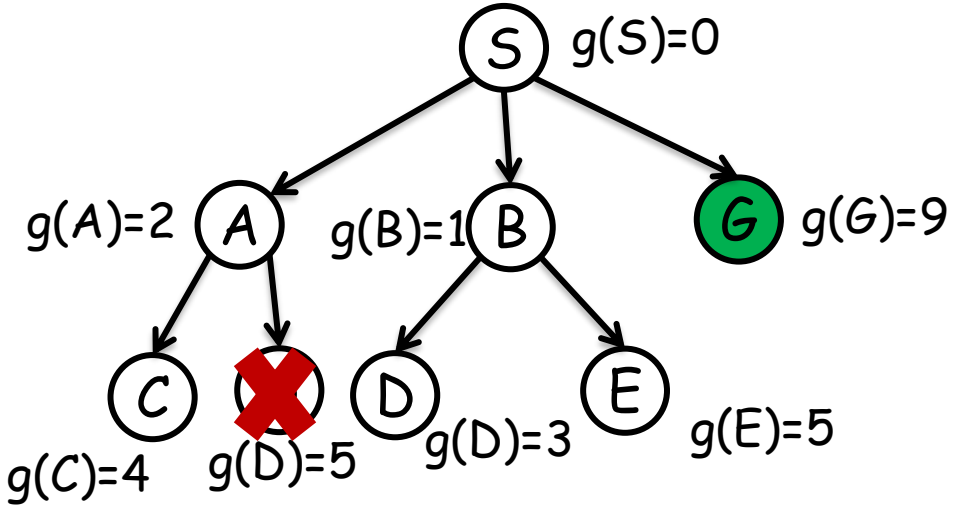
Frontier	Explored
S(0)	
B(1), A(2), G(9)	S
A(2), D(3), E(5), G(9)	S, B



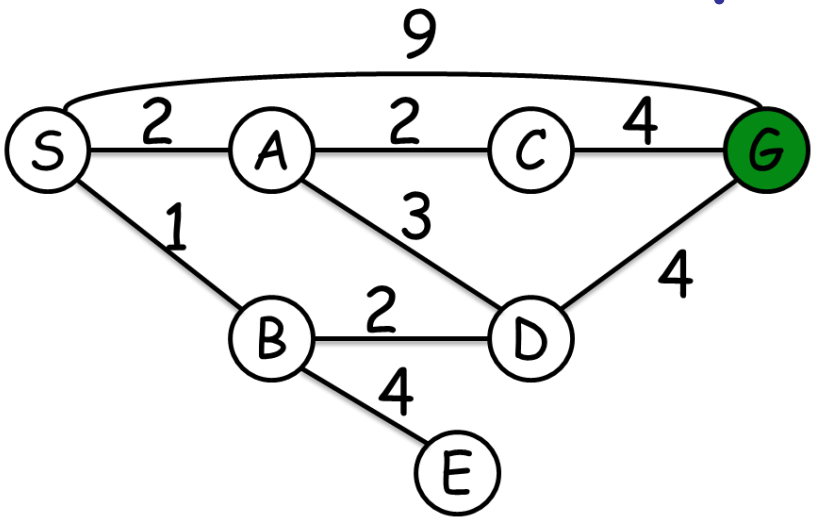
UCS Example (Graph search)



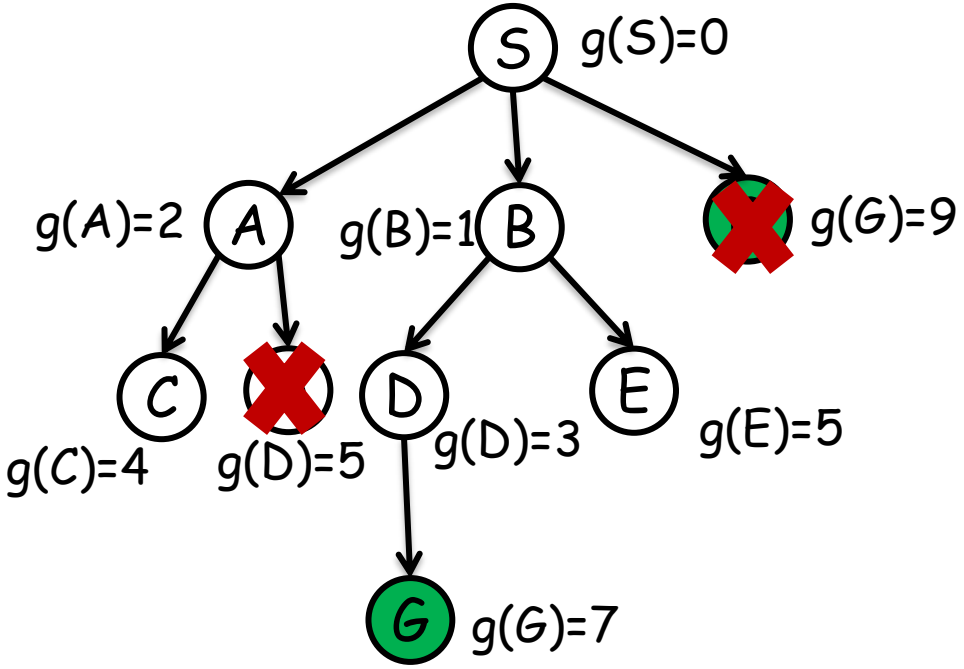
Frontier	Explored
S(0)	
B(1), A(2), G(9)	S
A(2), D(3), E(5), G(9)	S, B
D(3), C(4), E(5), G(9)	S, B, A



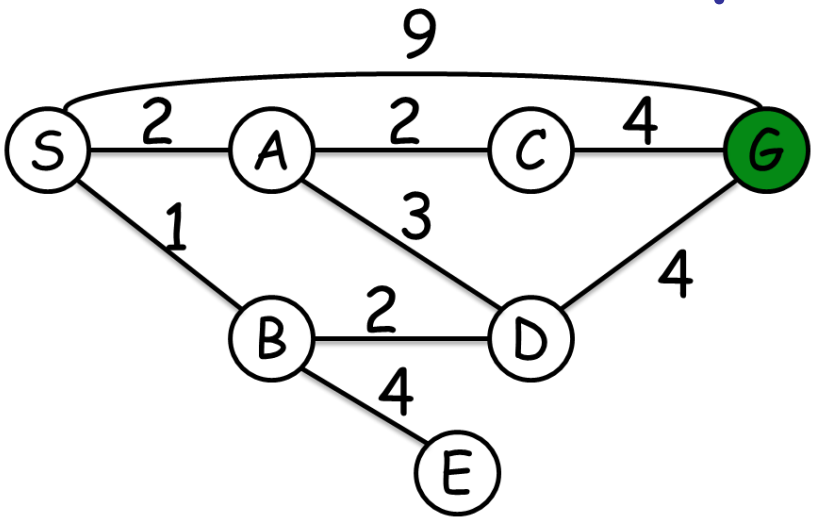
UCS Example (Graph search)



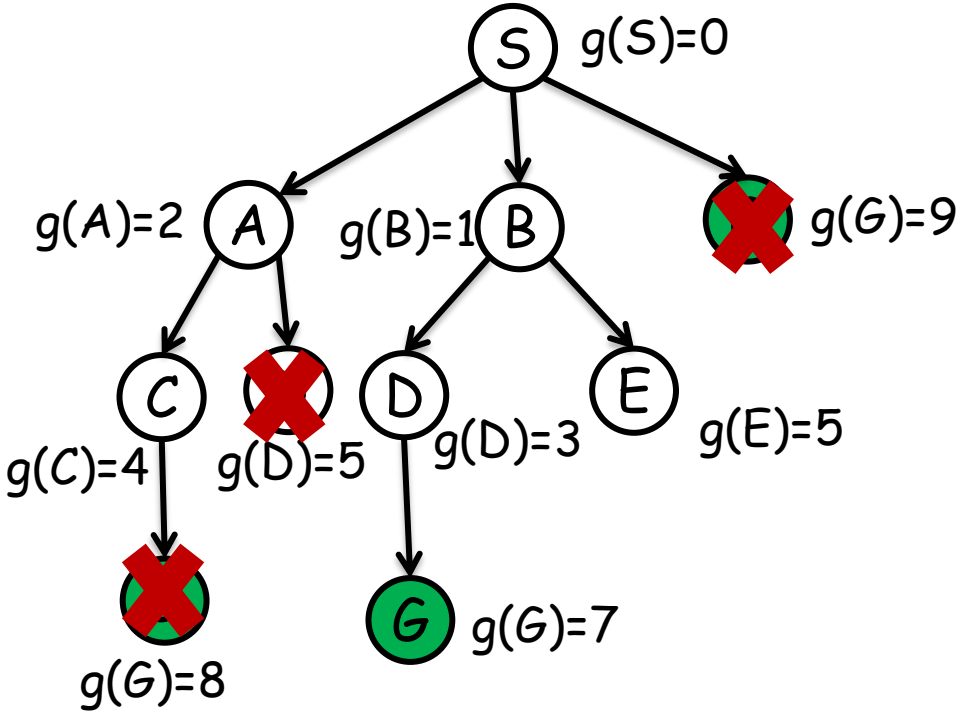
Frontier	Explored
S(0)	
B(1), A(2), G(9)	S
A(2), D(3), E(5), G(9)	S, B
D(3), C(4), E(5), G(9)	S, B, A
C(4), E(5), G(7)	S, B, A, D



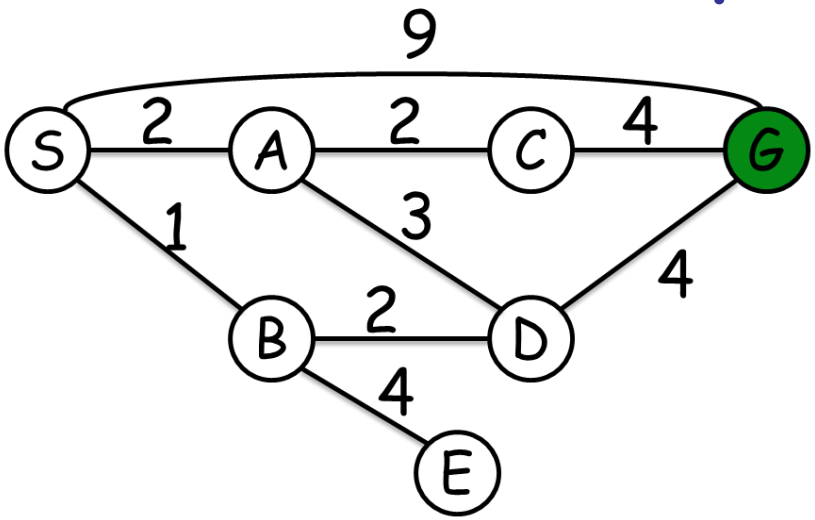
UCS Example (Graph search)



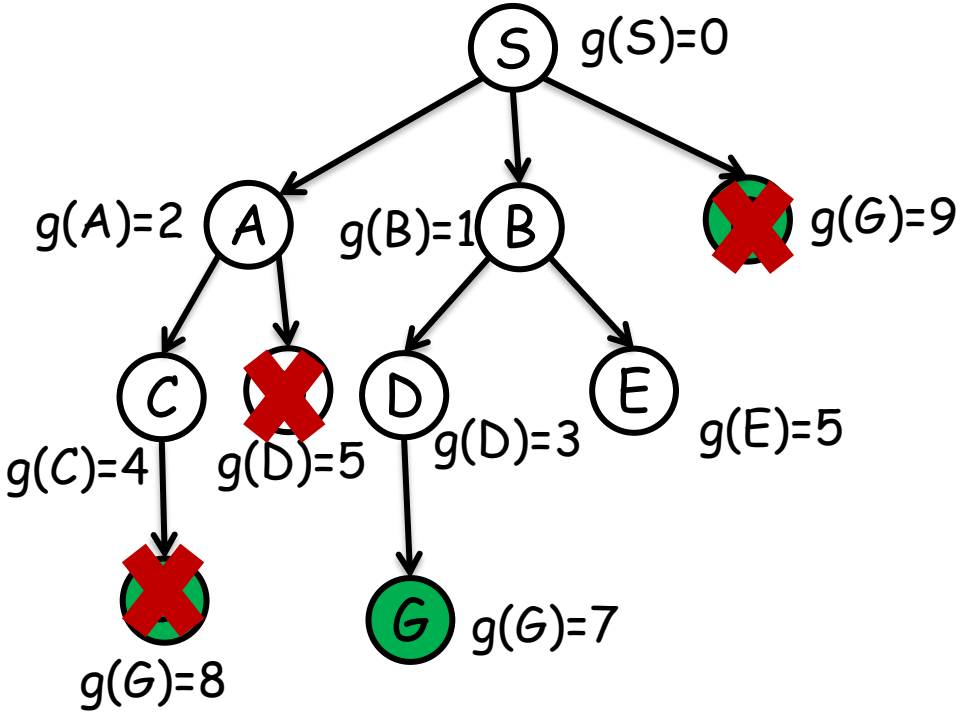
Frontier	Explored
S(0)	
B(1), A(2), G(9)	S
A(2), D(3), E(5), G(9)	S, B
D(3), C(4), E(5), G(9)	S, B, A
C(4), E(5), G(7)	S, B, A, D
E(5), G(7)	S, B, A, D, C



UCS Example (Graph search)

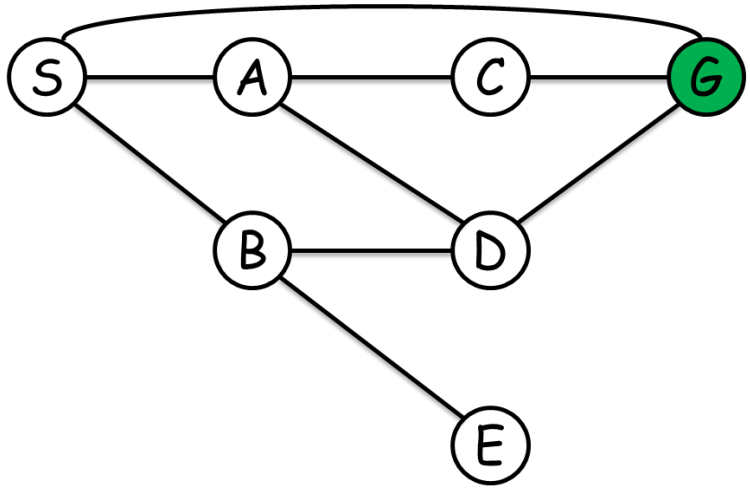


Frontier	Explored
S(0)	
B(1), A(2), G(9)	S
A(2), D(3), E(5), G(9)	S, B
D(3), C(4), E(5), G(9)	S, B, A
C(4), E(5), G(7)	S, B, A, D
E(5), G(7)	S, B, A, D, C
G(7)	S, B, A, D, C, E



DFS Example

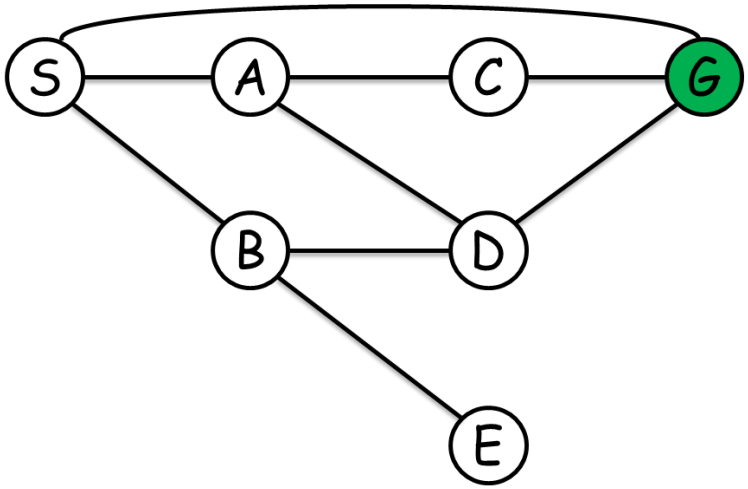
- Graph search
- Goal test when expanding a node
- Alphabetical order



Frontier	Explored

DFS Example

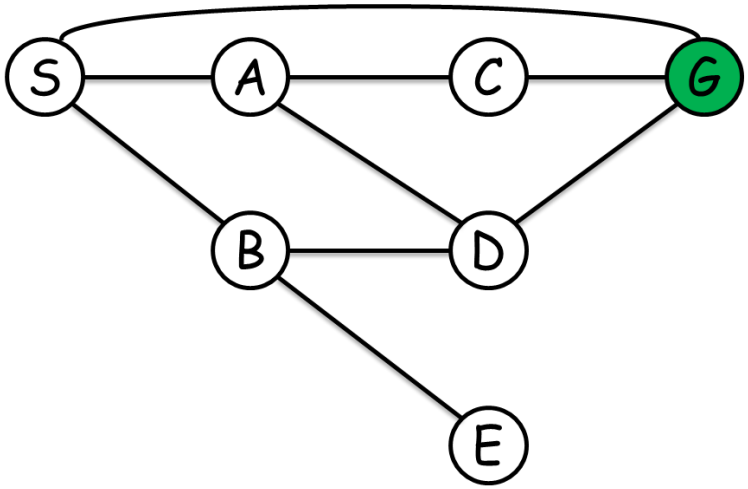
- Graph search
- Goal test when expanding a node
- Alphabetical order



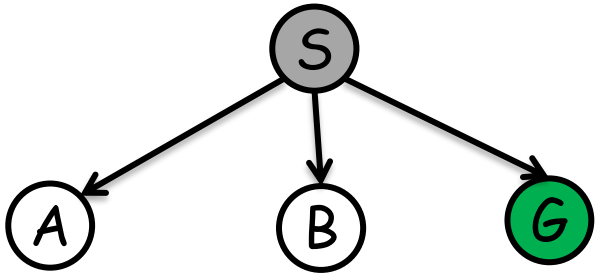
Frontier	Explored
S	

DFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order

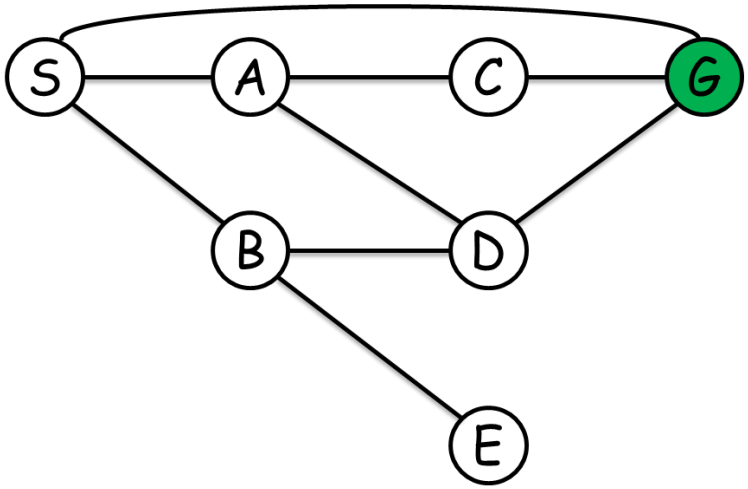


Frontier	Explored
S	
A, B, G	S

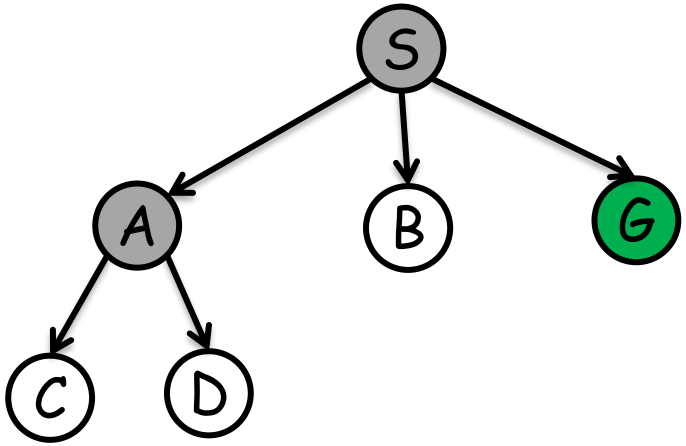


DFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order

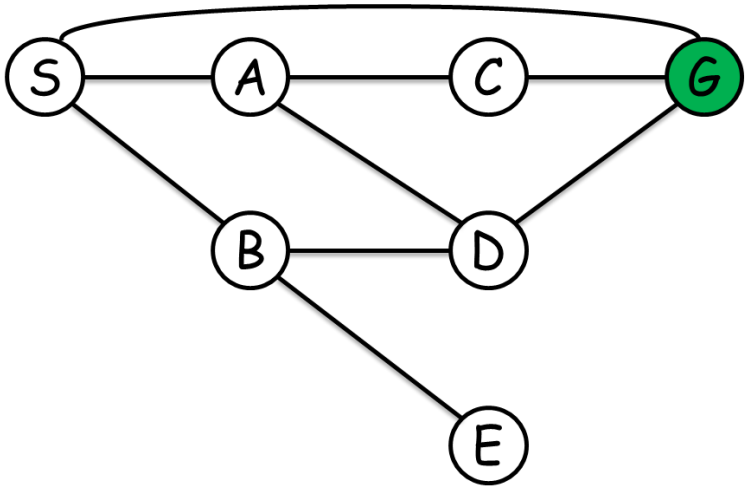


Frontier	Explored
S	
A, B, G	S
C, D, B, G	S, A

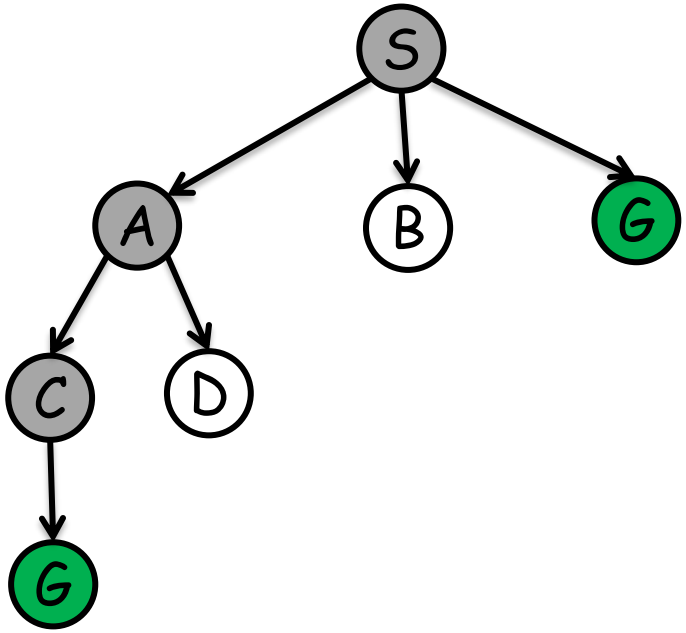


DFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order

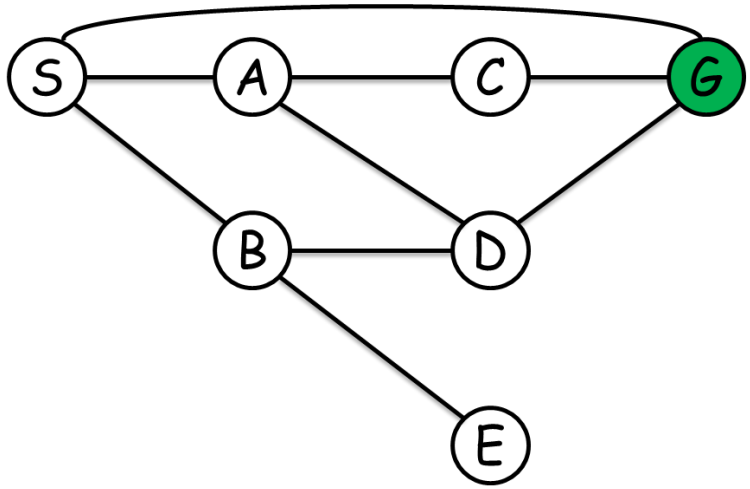


Frontier	Explored
S	
A, B, G	S
C, D, B, G	S, A
G, D, B	S, A, C

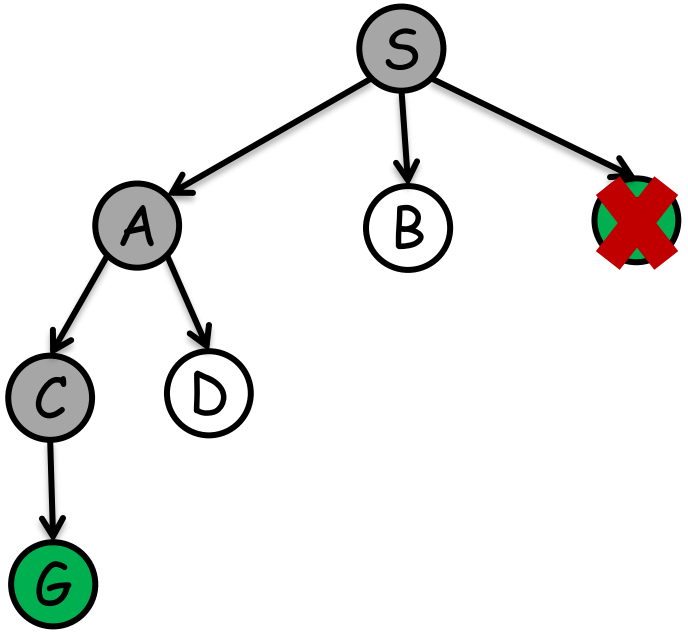


DFS Example

- Graph search
- Goal test when expanding a node
- Alphabetical order

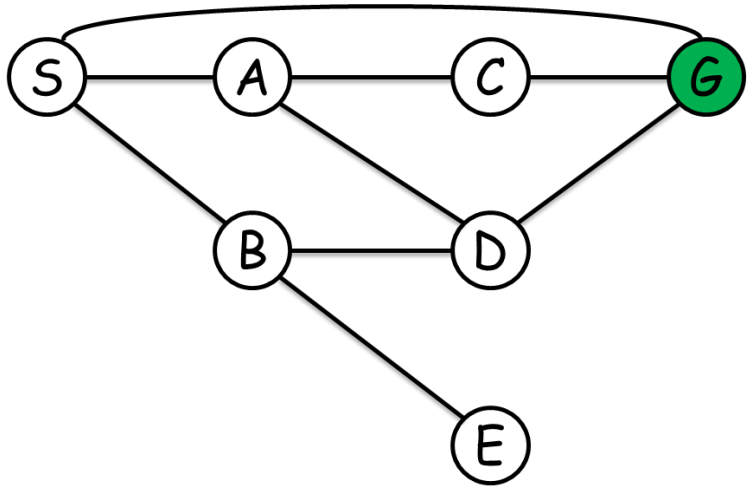


Frontier	Explored
S	
A, B, G	S
C, D, B, G	S, A
G, D, B	S, A, C



DFS Example

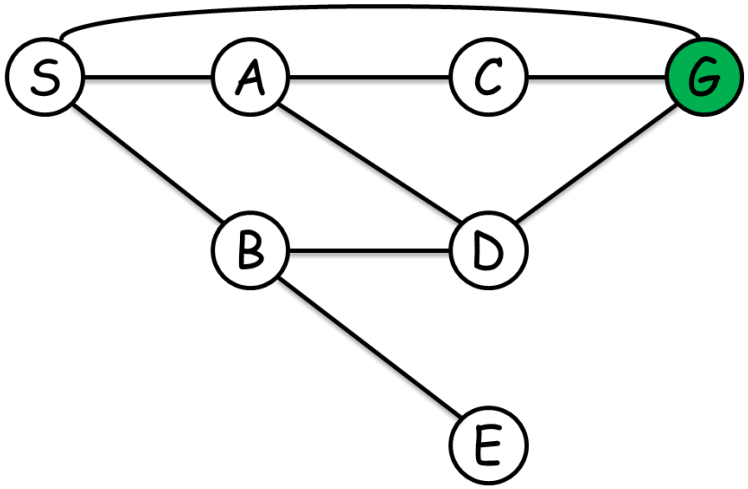
- Tree search
- Goal test when expanding a node
- Alphabetical order



Frontier

DFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

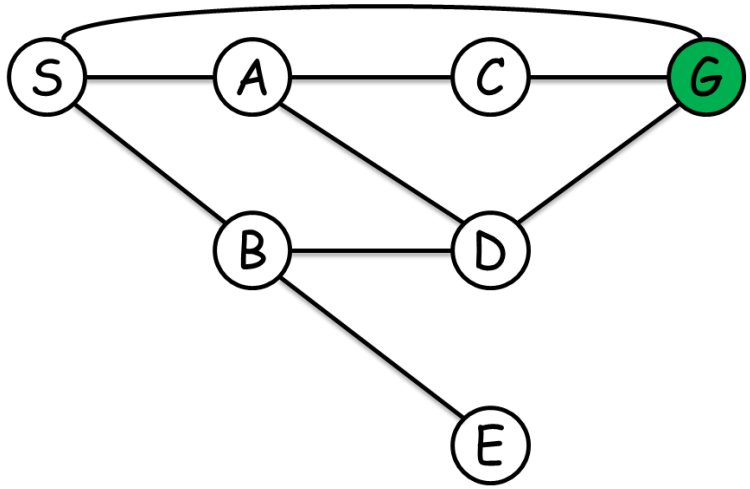


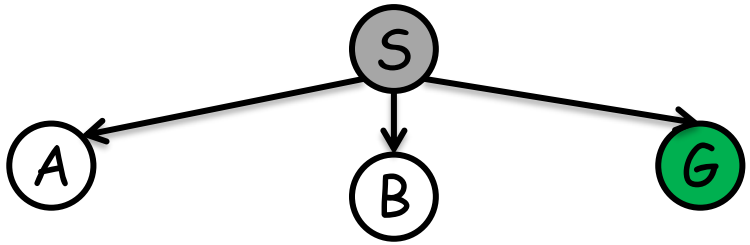
Frontier

S

DFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

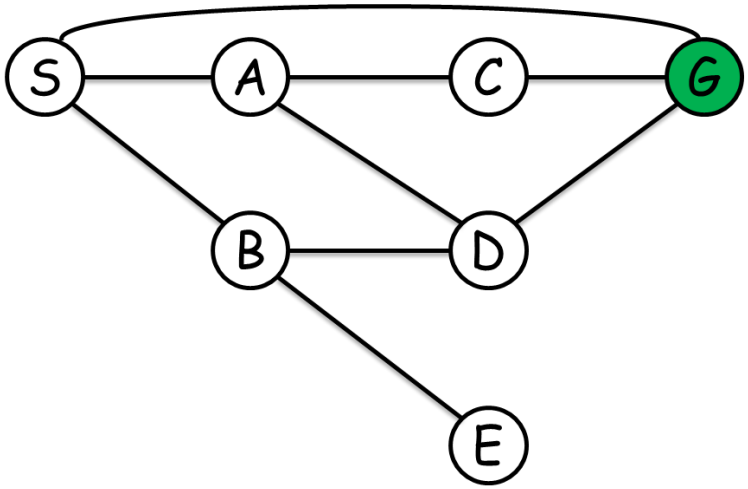


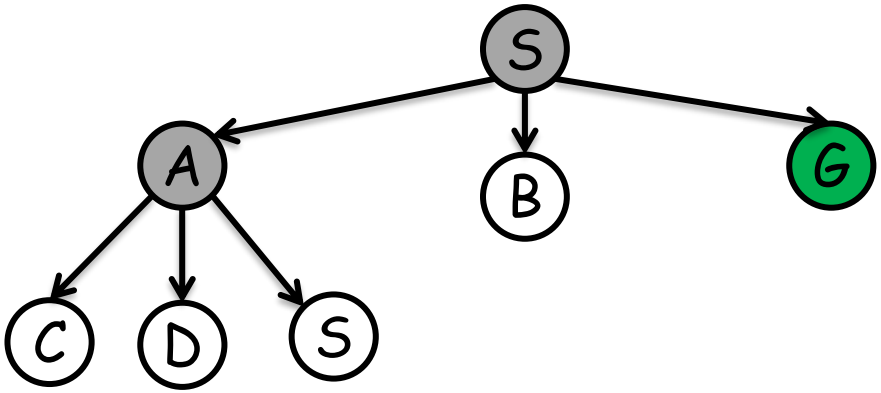


Frontier
S
A, B, G

DFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

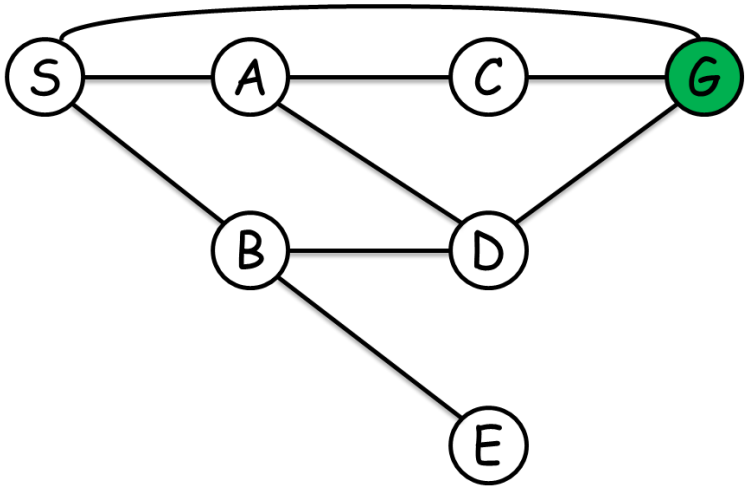




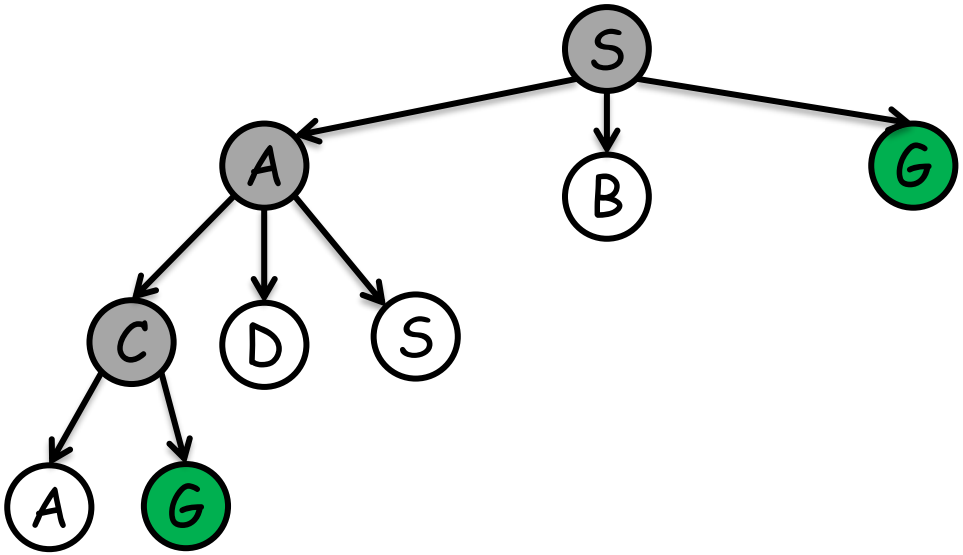
Frontier
S
A, B, G
C, D, S, B, G

DFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

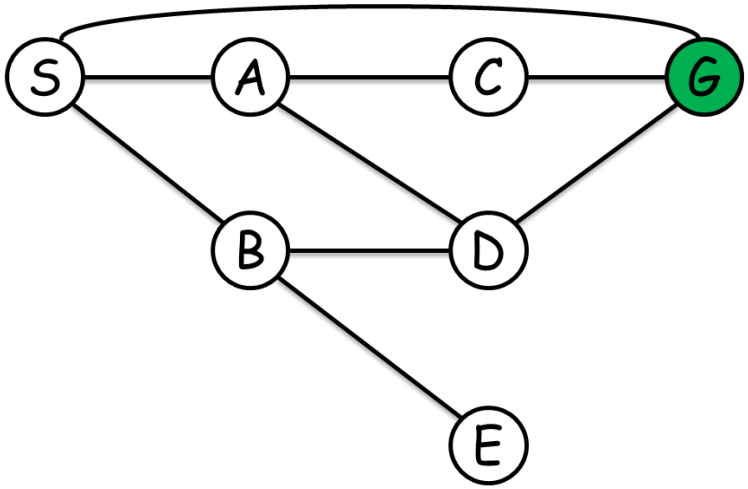


Frontier
S
A, B, G
C, D, S, B, G
A, G, D, S, B, G

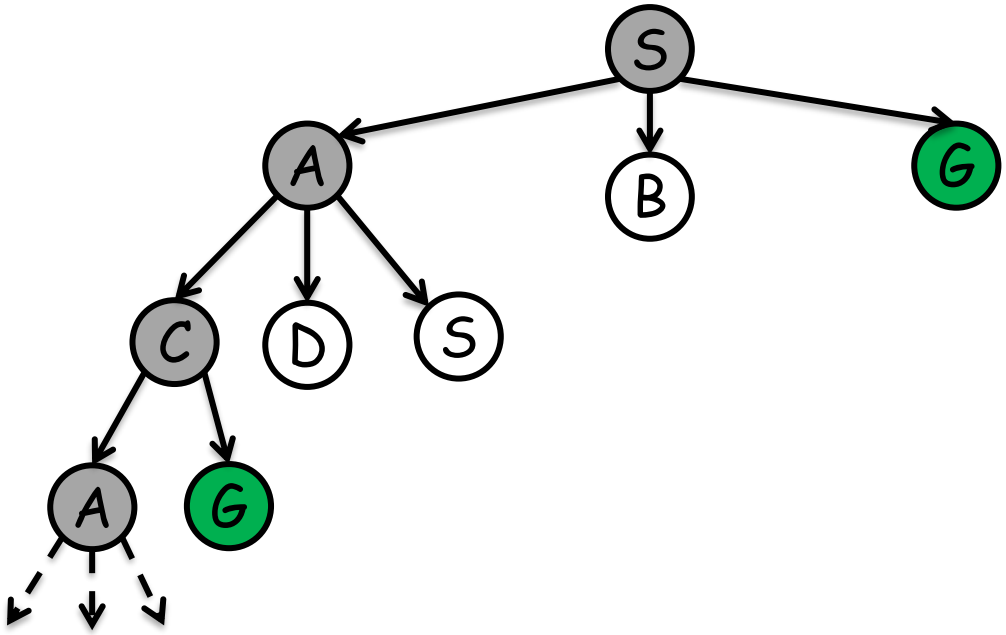


DFS Example

- Tree search
- Goal test when expanding a node
- Alphabetical order

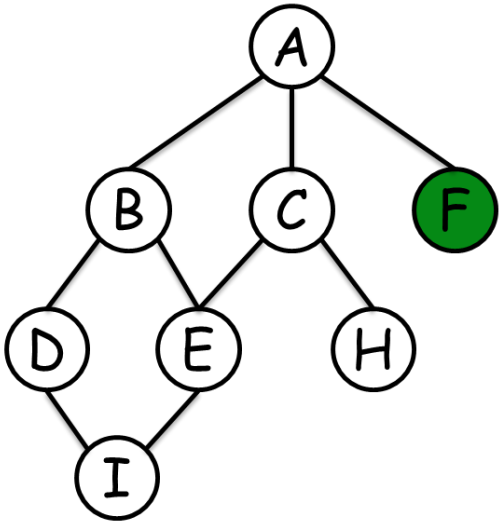


Frontier
S
A, B, G
C, D, S, B, G
A, G, D, S, B, G



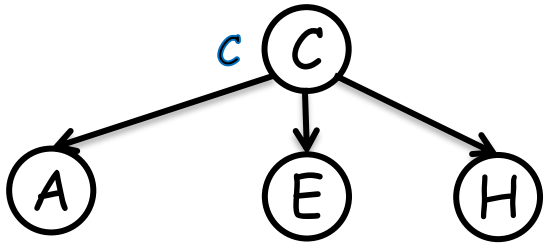
DFS Example

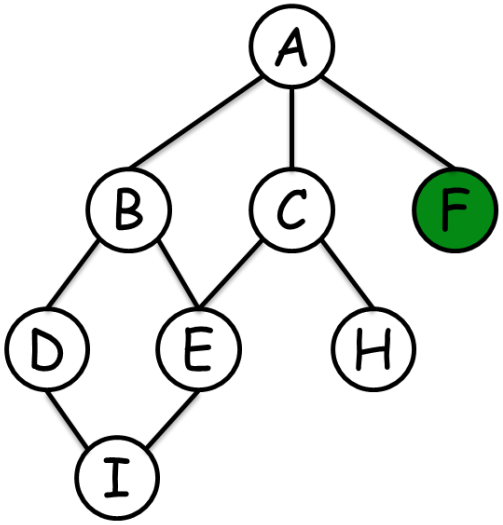
- Improved Tree search
- Goal test when expanding a node
- Alphabetical order



DFS Example

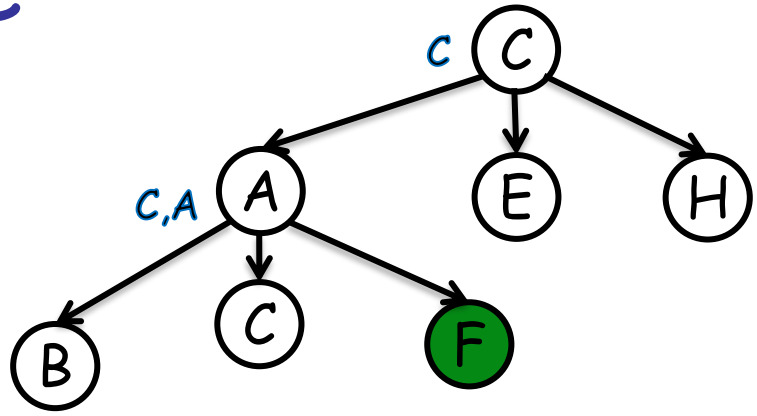
- Improved Tree search
- Goal test when expanding a node
- Alphabetical order

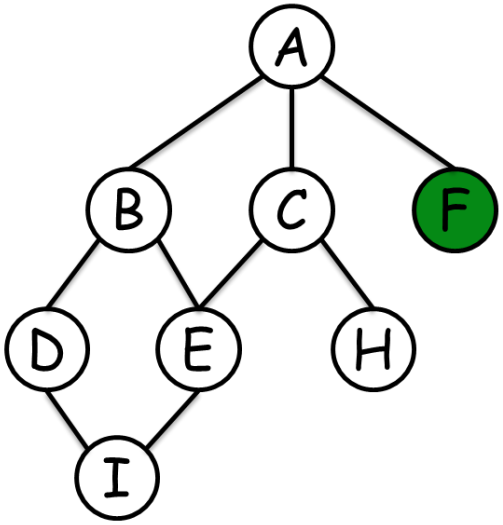




DFS Example

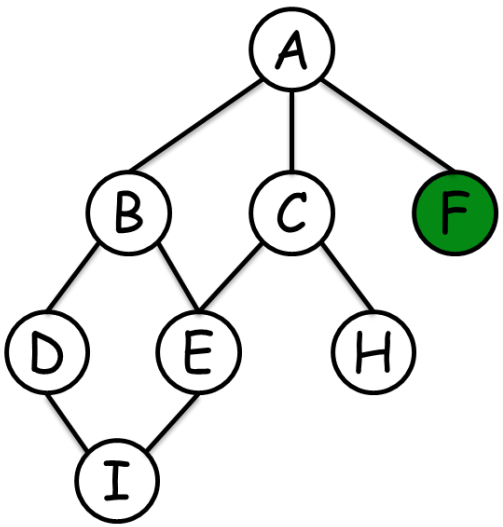
- Improved Tree search
- Goal test when expanding a node
- Alphabetical order

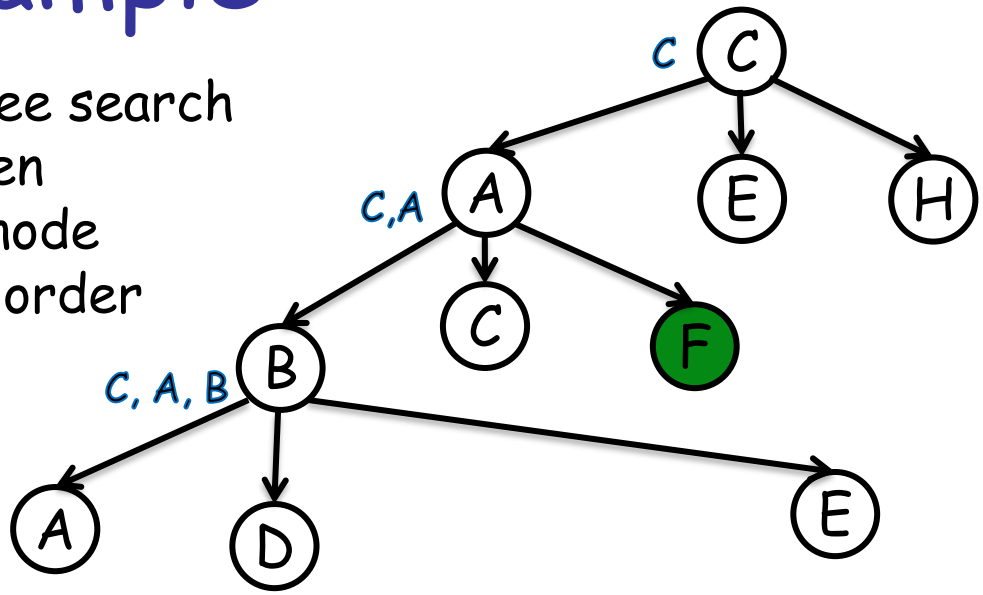




DFS Example

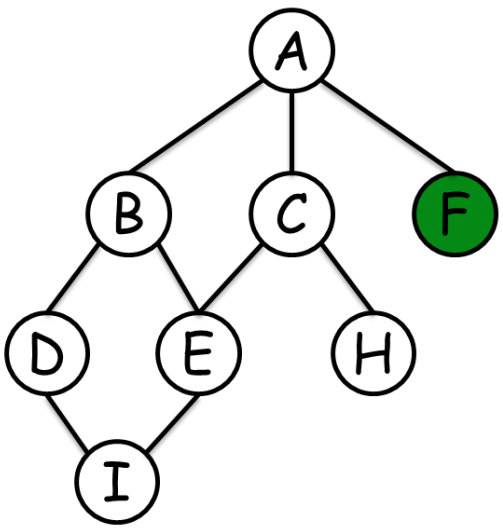
- Improved Tree search
- Goal test when expanding a node
- Alphabetical order

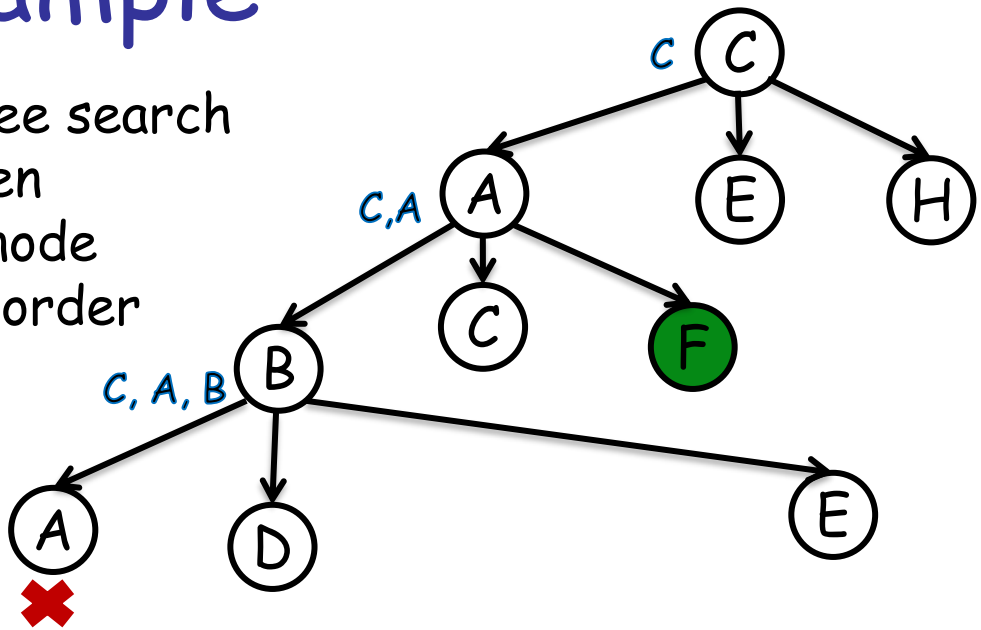




DFS Example

- Improved Tree search
- Goal test when expanding a node
- Alphabetical order





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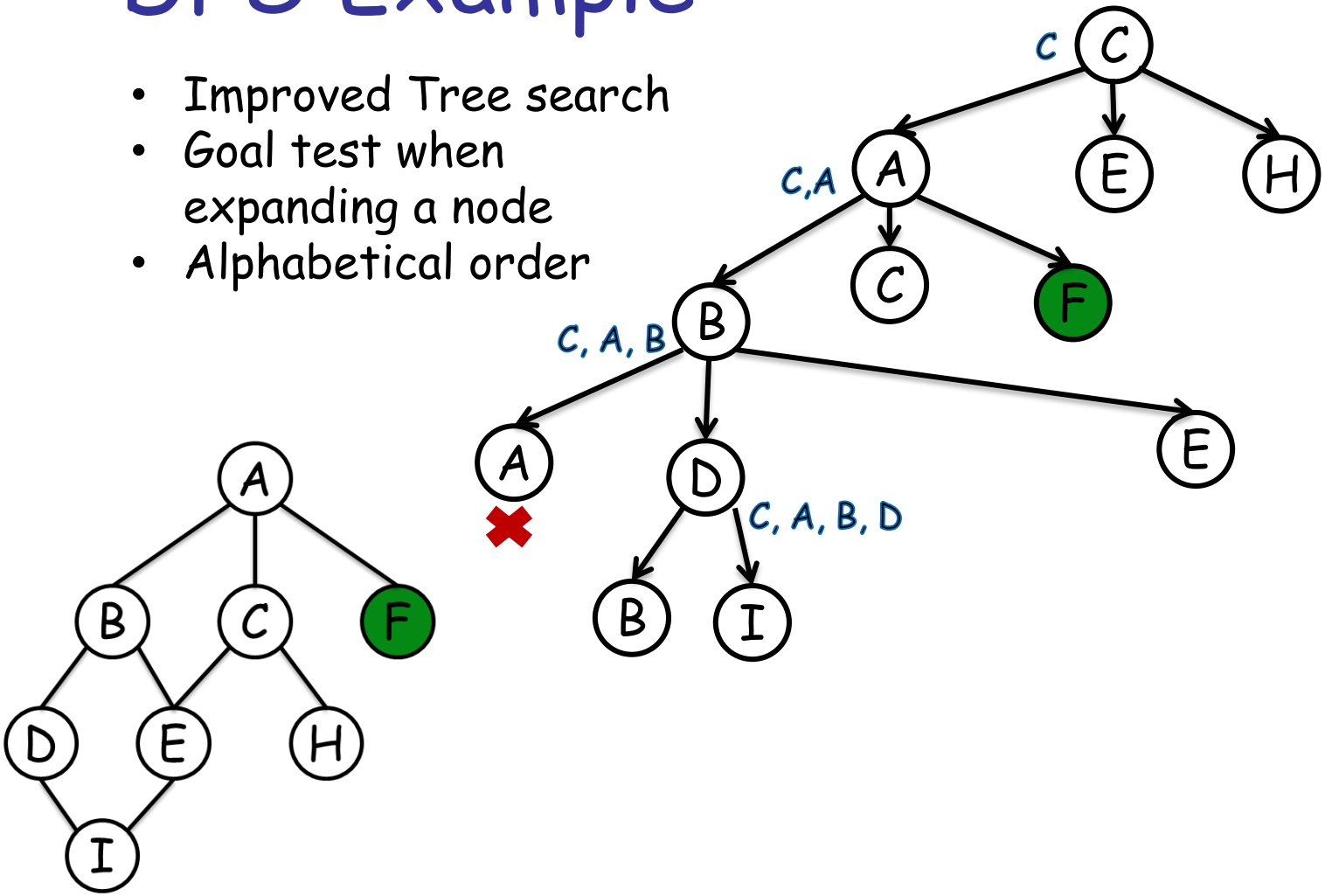

DFS Example

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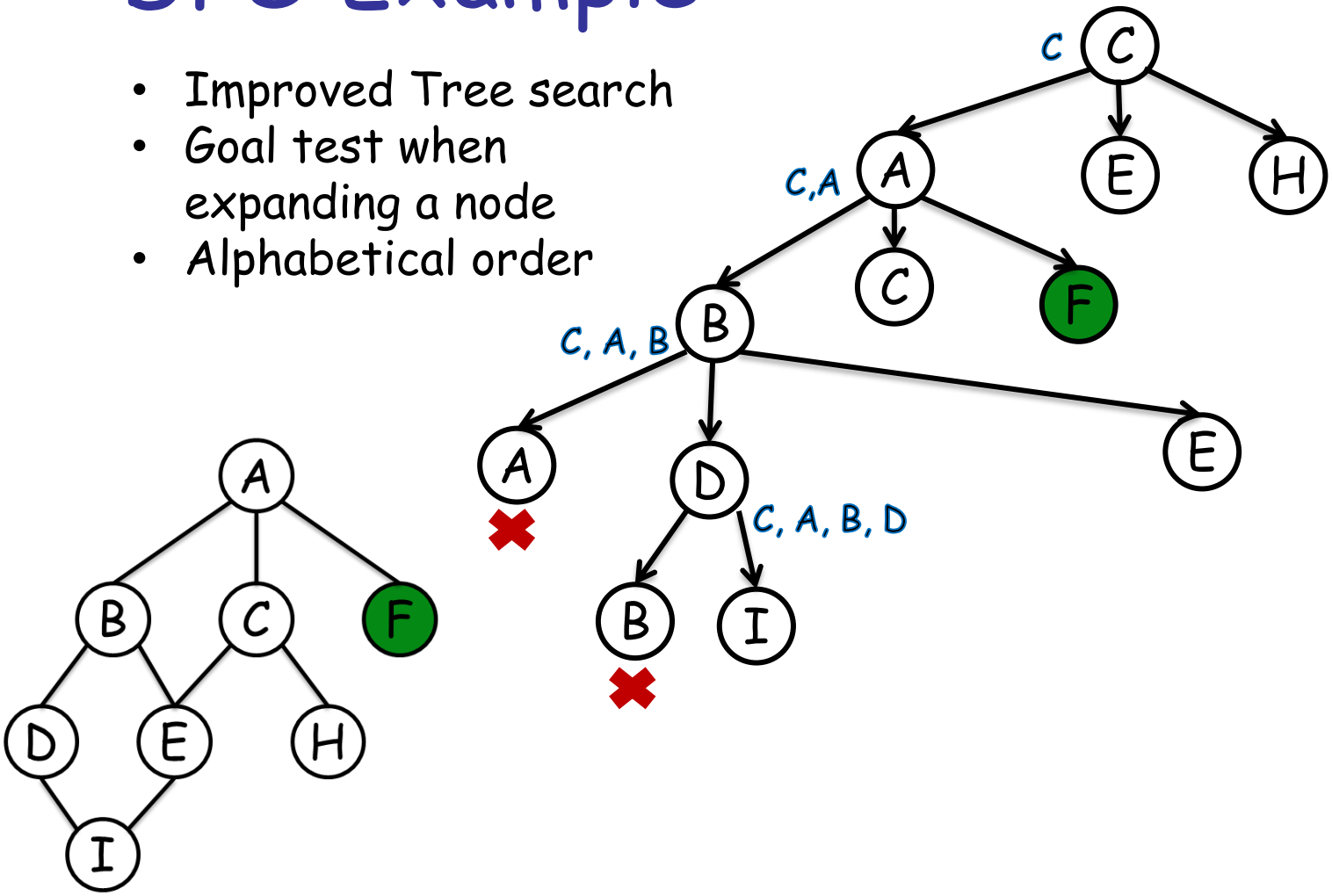

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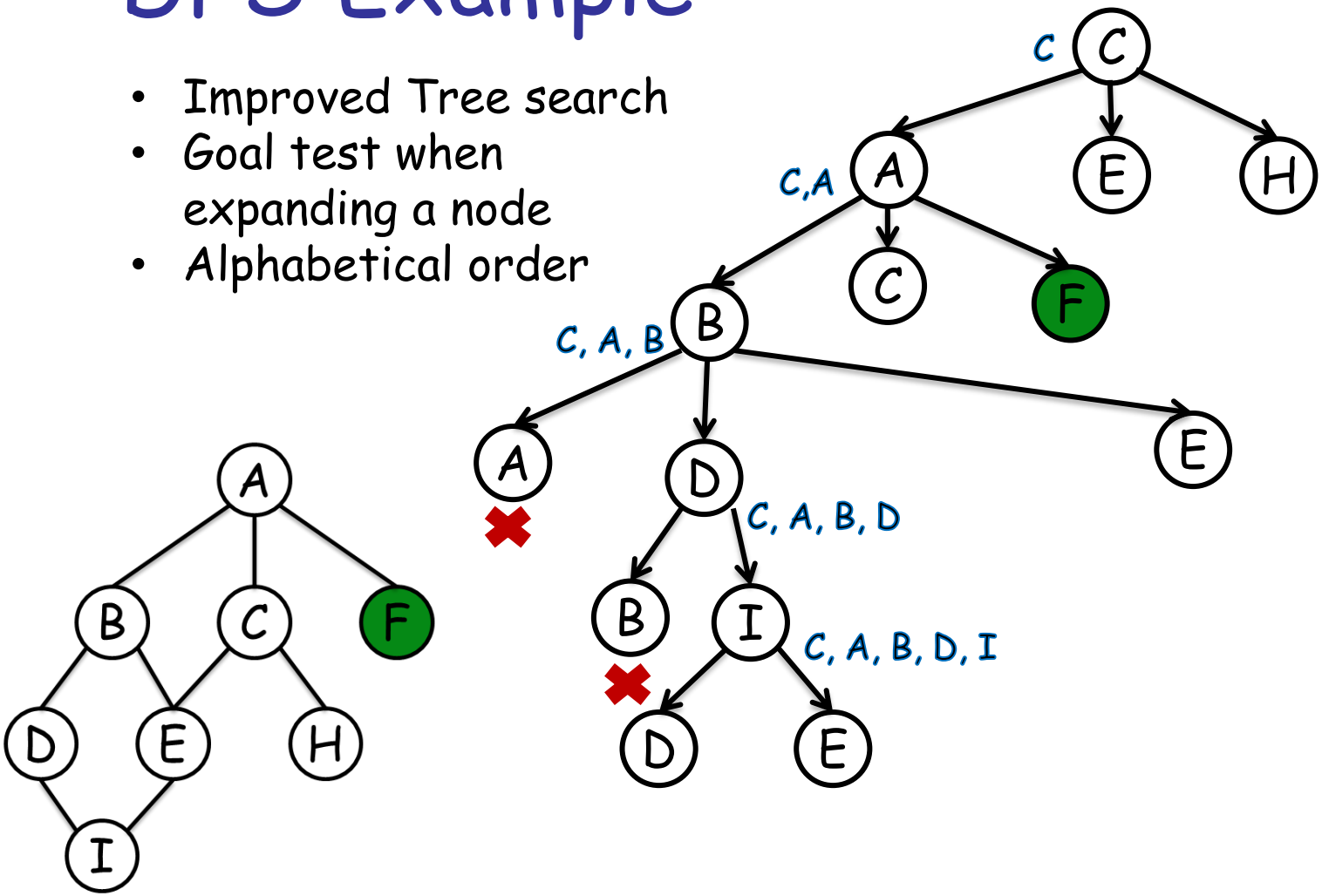

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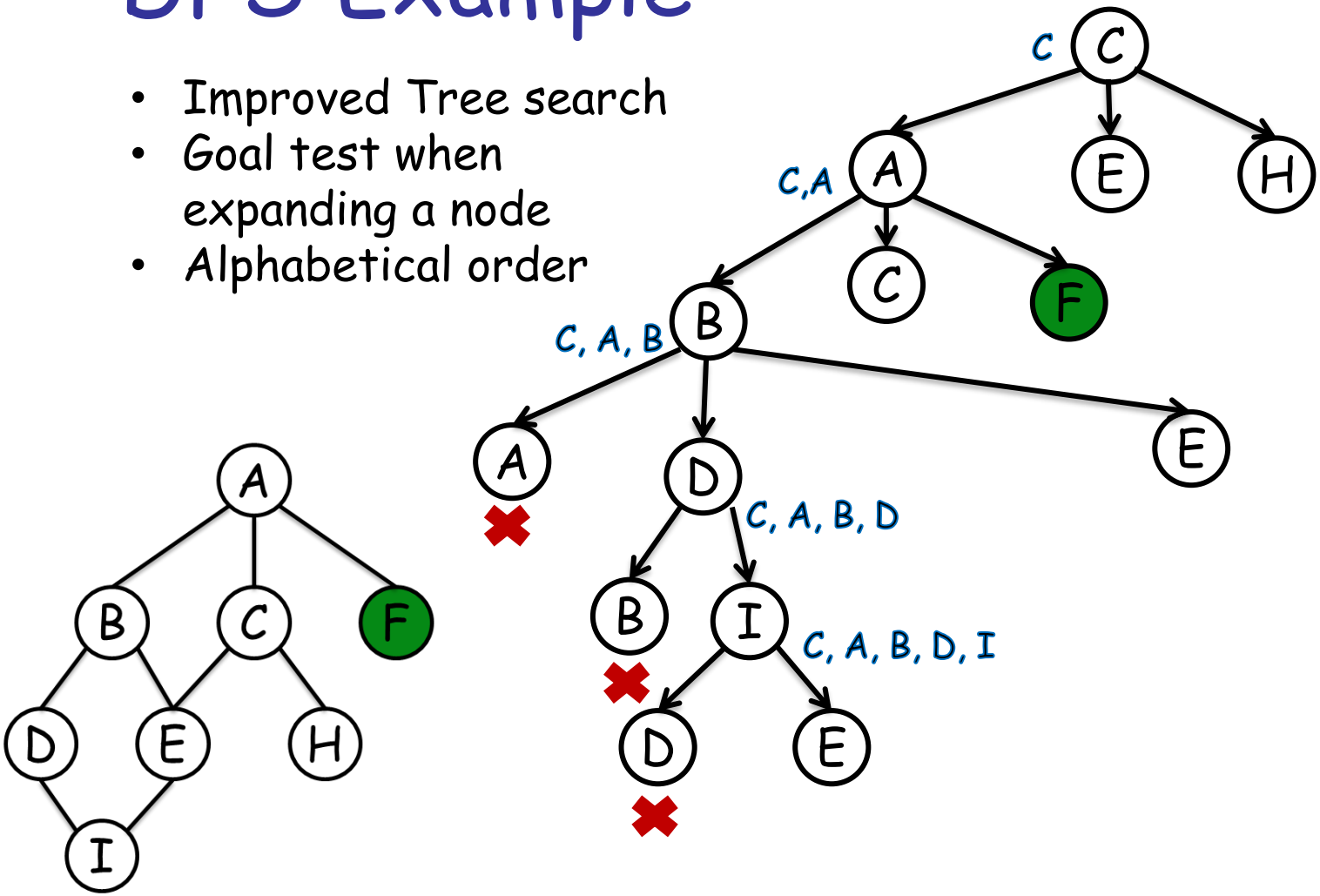

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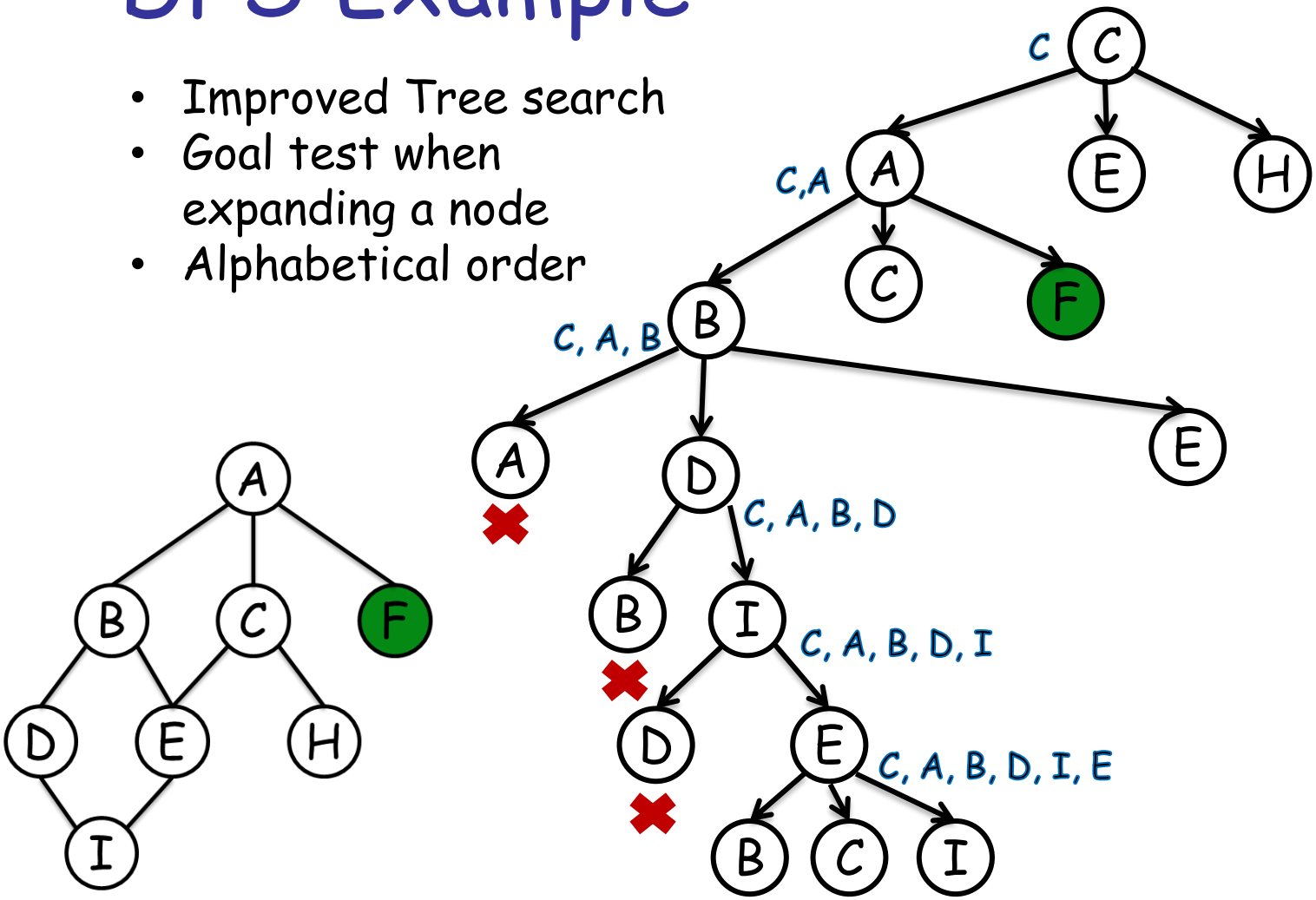

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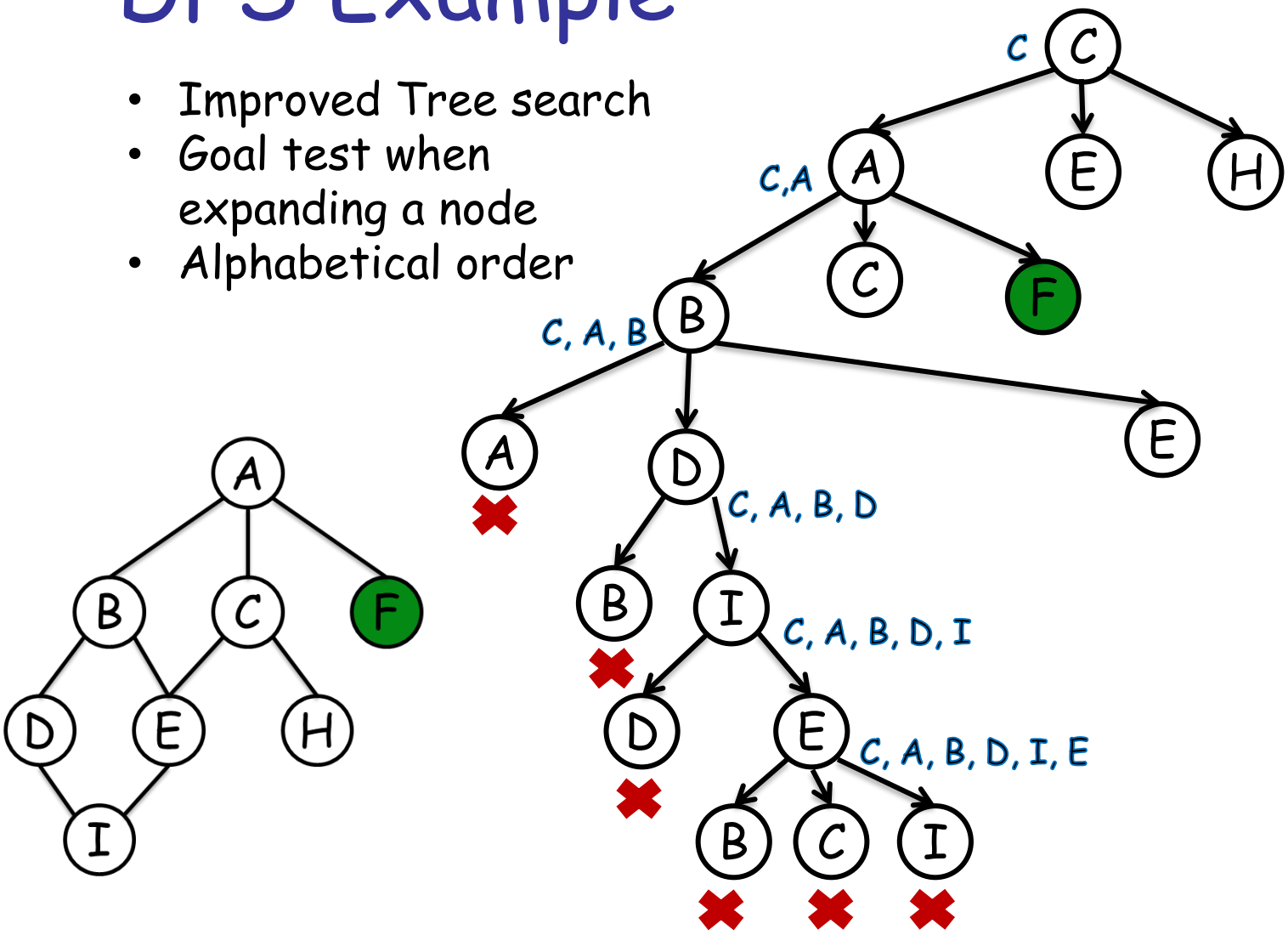

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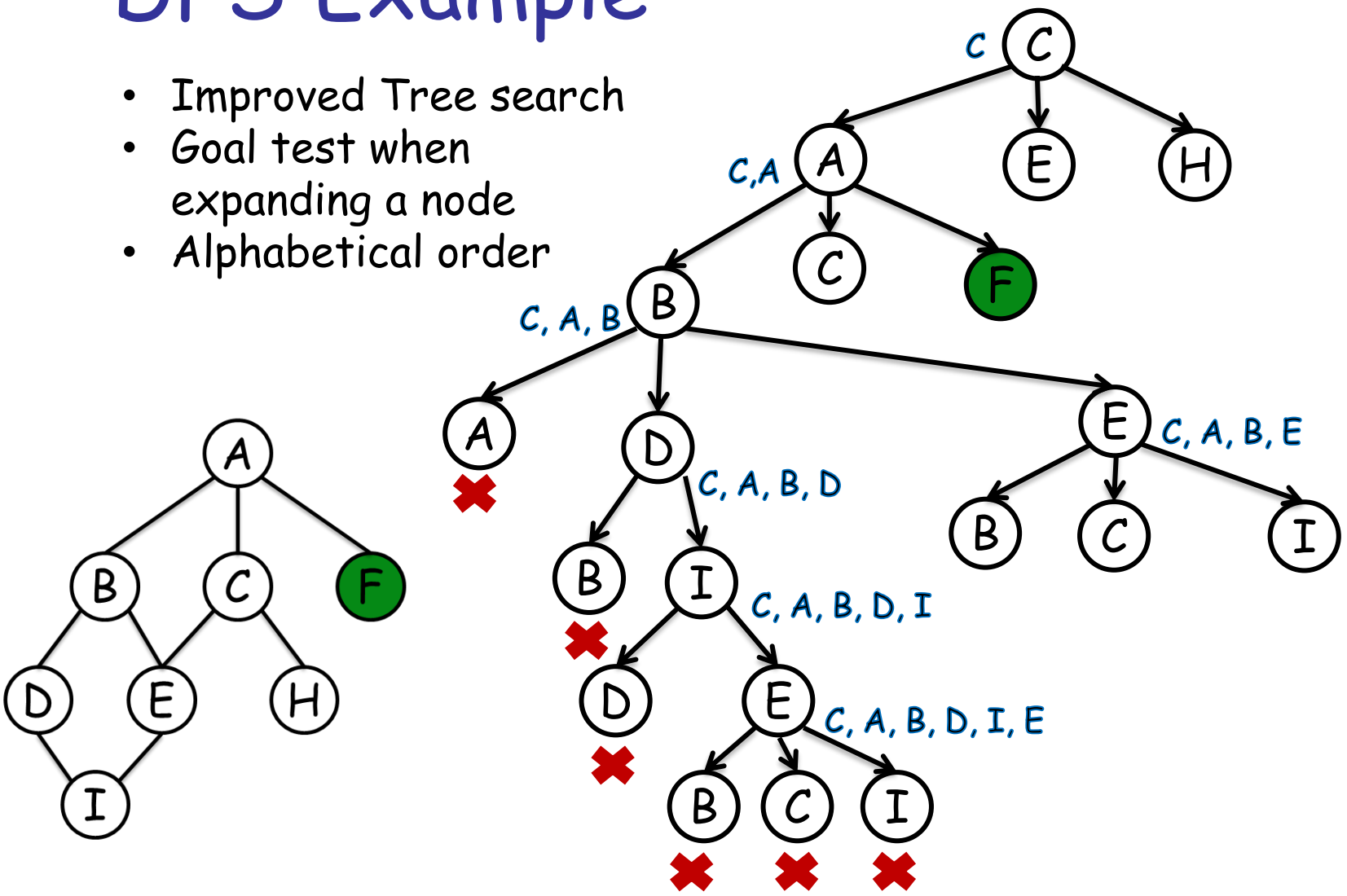

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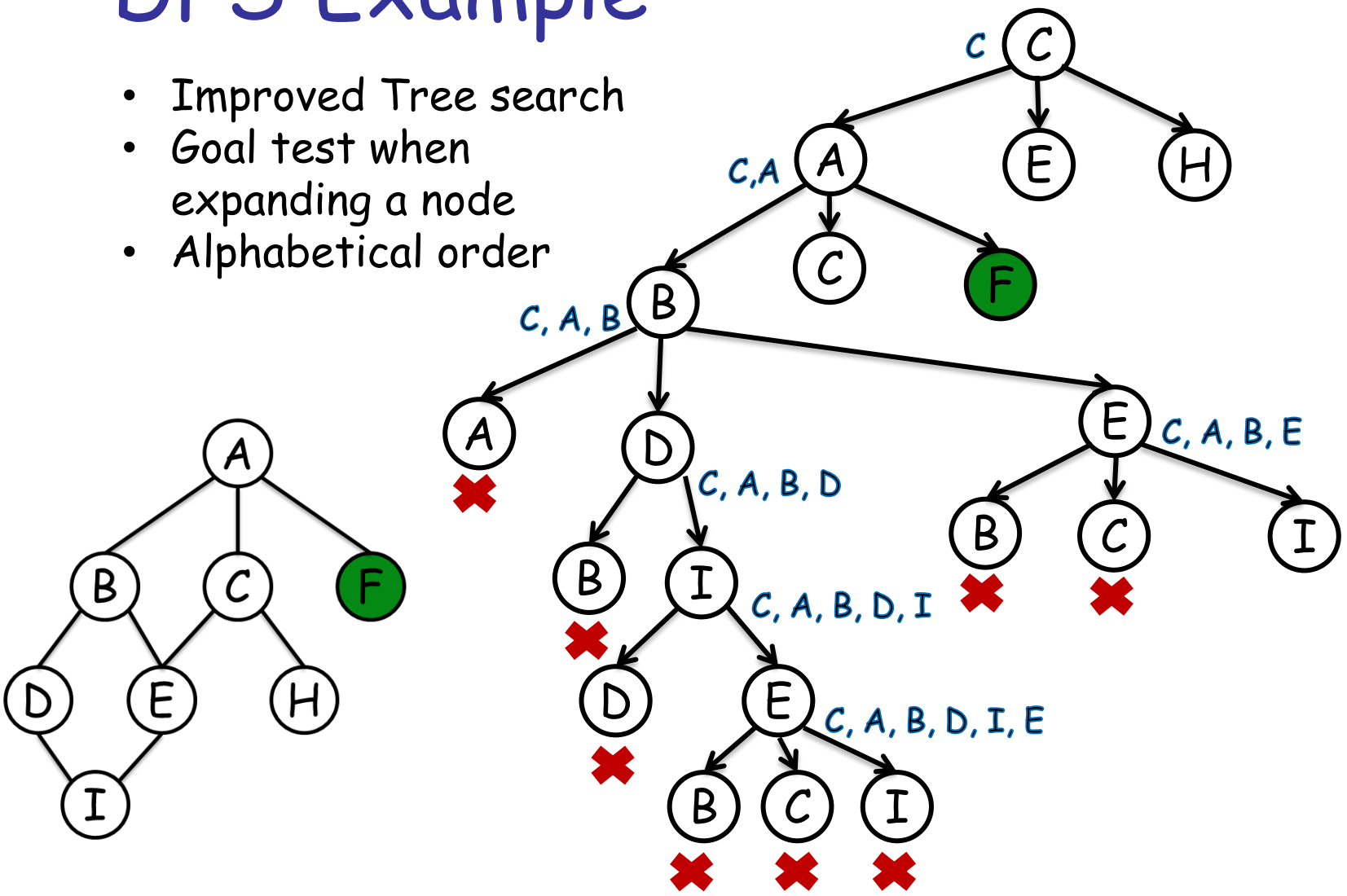

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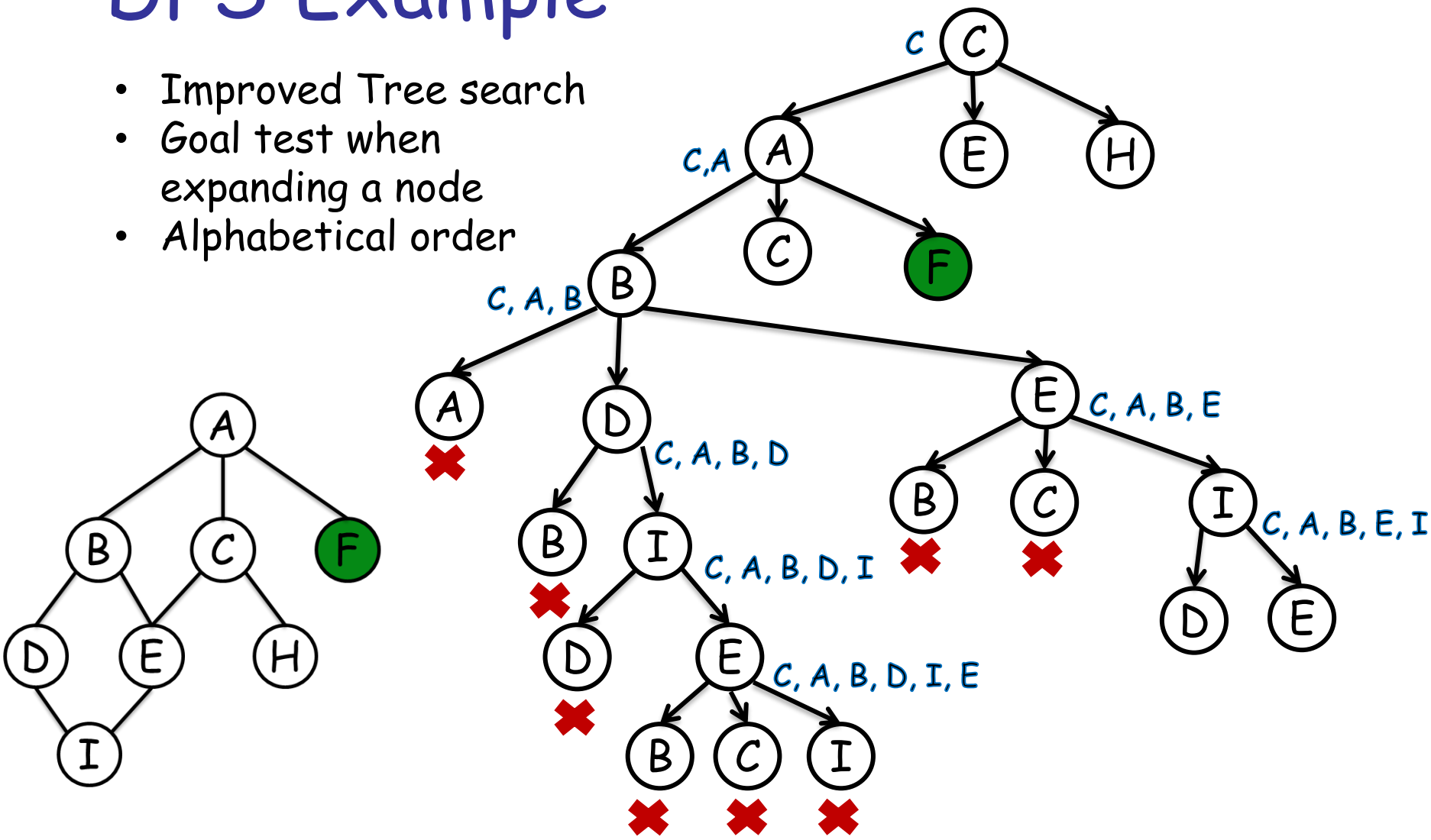

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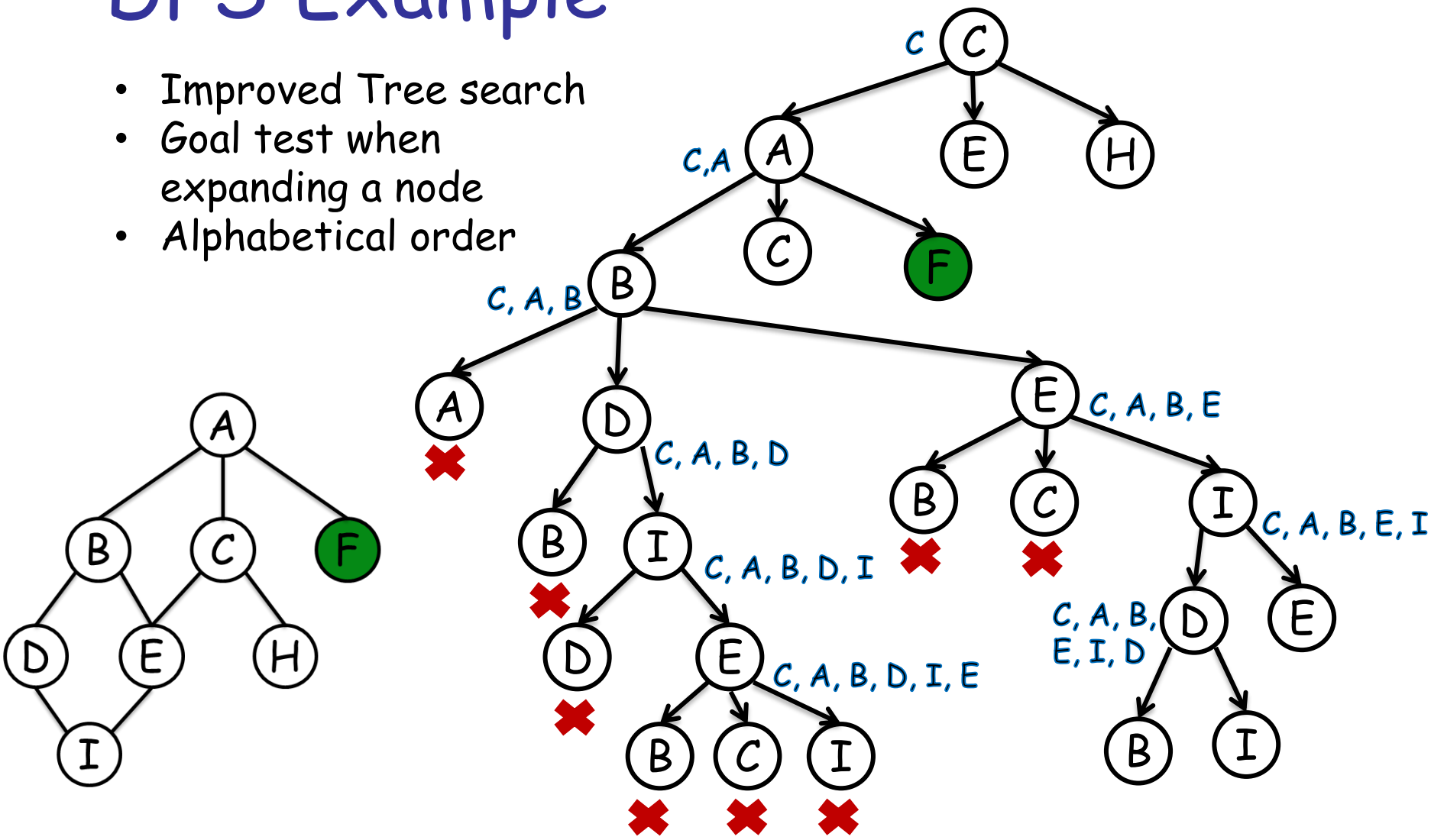

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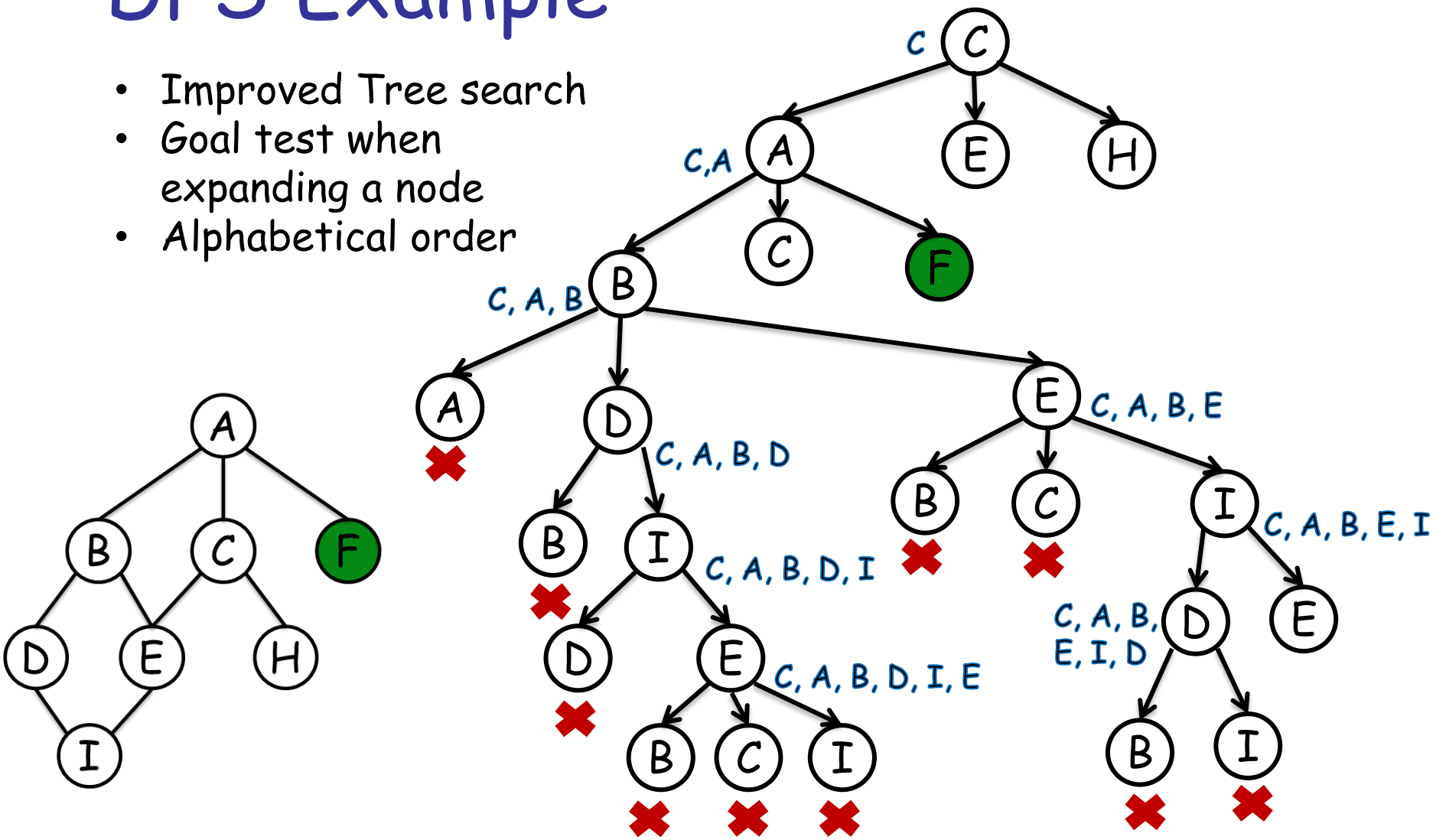

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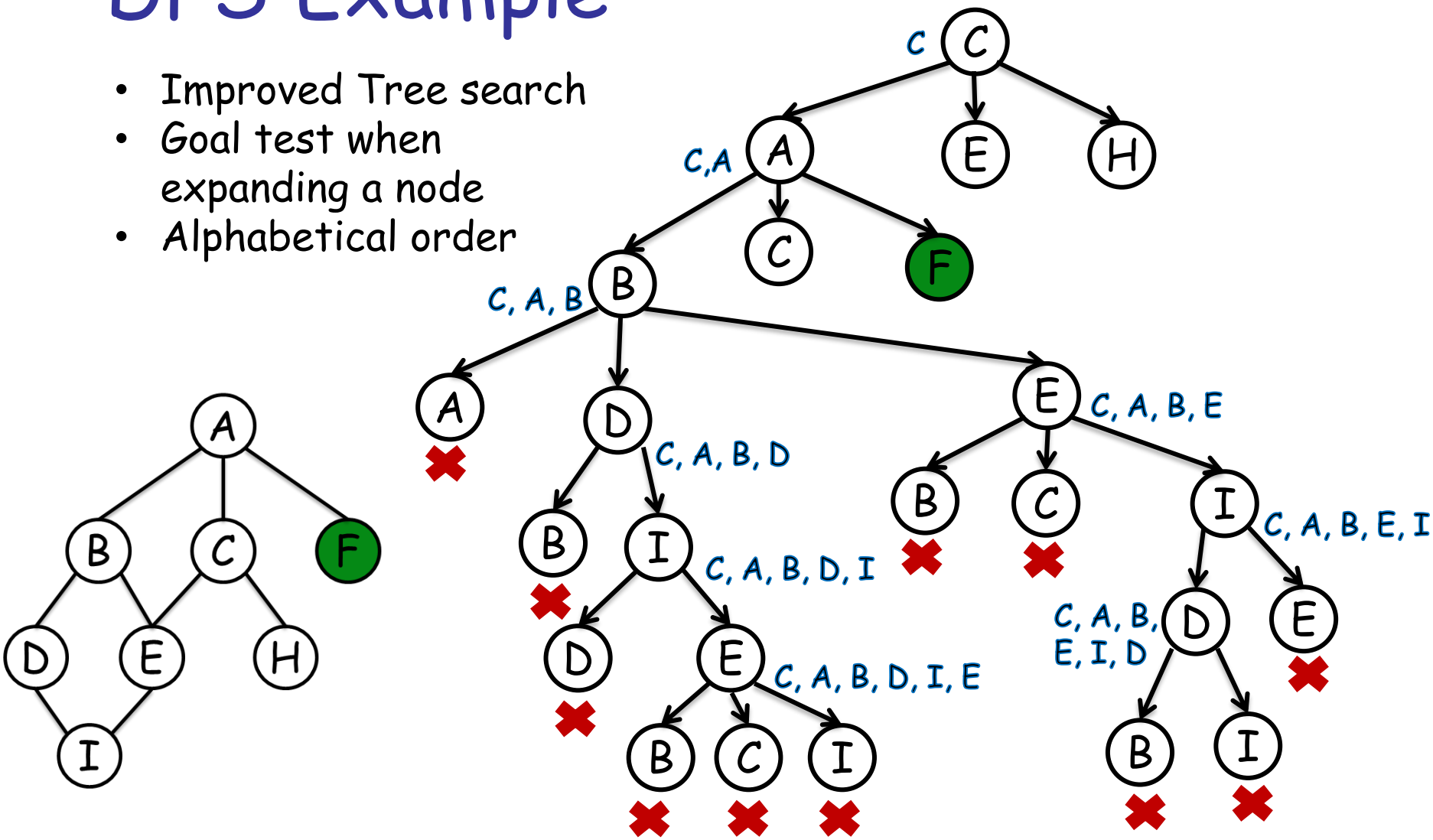

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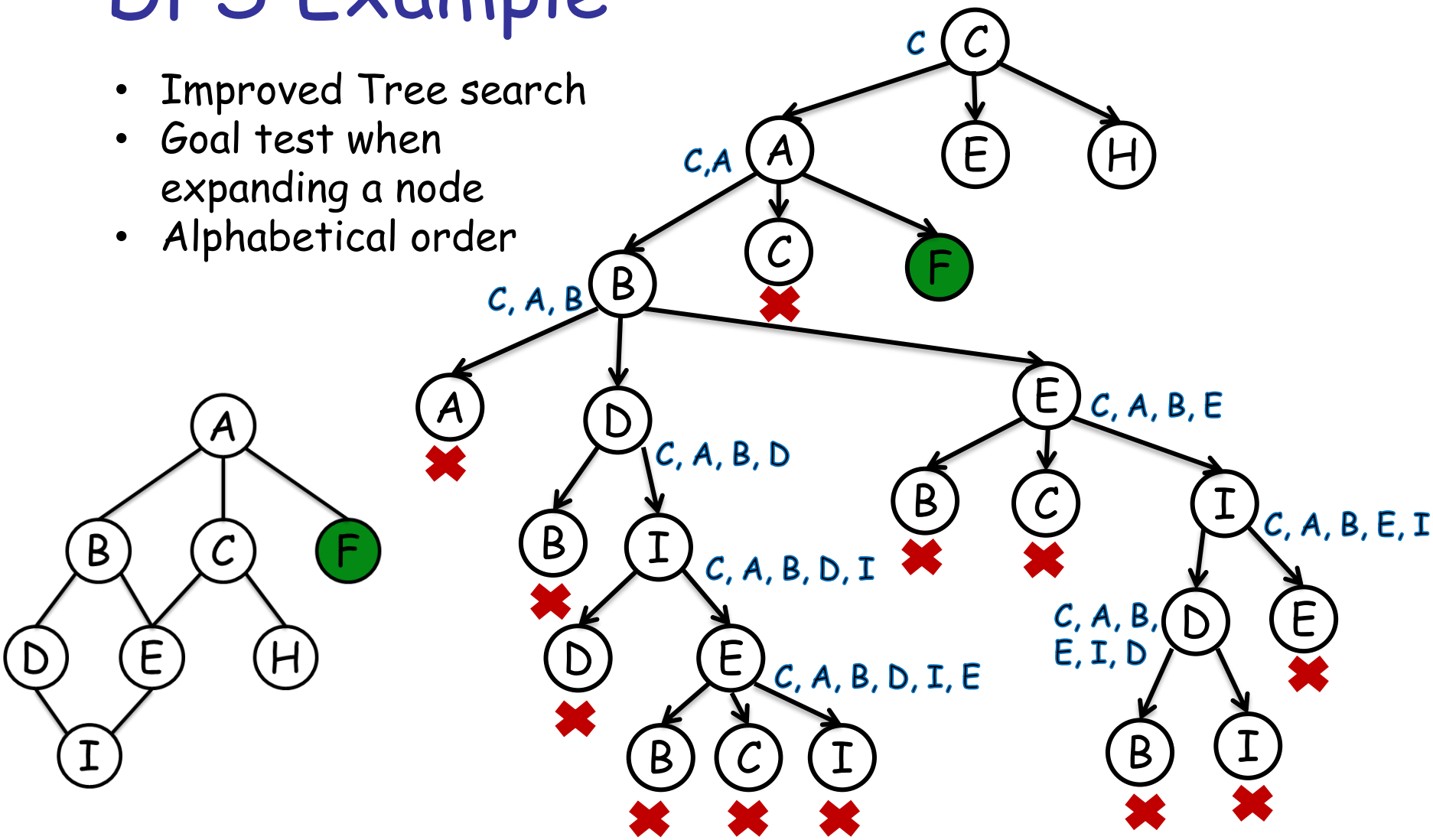

Summary of algorithms

Criterion	Breadth-First	Uniform-Cost	Depth-Frist	Depth-Limited	Iterative-Deepening	Bidirectional (If applicable)
Complete?	Yes[a]	Yes[a,b]	No	No	Yes[a]	Yes[a,d]
Time	$O(b^d)$	$O(b^{1+C^*/\varepsilon})$	$O(b^m)$	$O(b^l)$	$O(b^d)$	$O(b^{d/2})$
Space	$O(b^d)$	$O(b^{1+C^*/\varepsilon})$	$O(bm)$	$O(bl)$	$O(bd)$	$O(b^{d/2})$
Optimal?	Yes[c]	Yes	No	No	Yes[c]	Yes[c,d]

Superscript caveats are as follows:
[a] complete if b is finite
[b] complete if step costs ≥ ε for positive ε
[c] optimal if step costs are all identical
[d] if both directions use breadth-first search