Search Strategies

Informatics 2D: Reasoning and Agents **Lecture 3**



Search strategies

A search strategy is defined by picking the order of node expansion.

• Nodes are taken from the frontier.

Evaluating search strategies



completeness: does it always find a solution if one exists?



time complexity: number of nodes generated / expanded



space complexity: maximum number of nodes in memory



optimality: does it always find a least-cost solution?

Time and space complexity are measured in terms of:

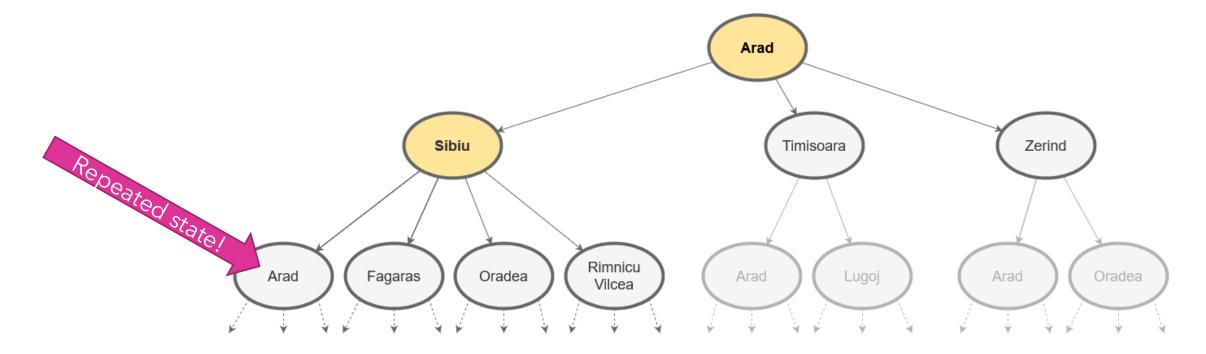
- **b**: maximum branching factor of the search tree
- **d**: depth of the least-cost solution
- m: maximum depth of the state space (may be ∞)

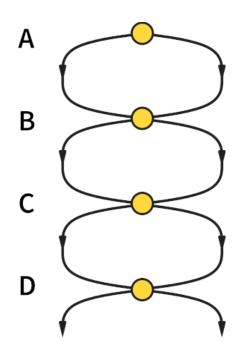
Recall: Tree Search

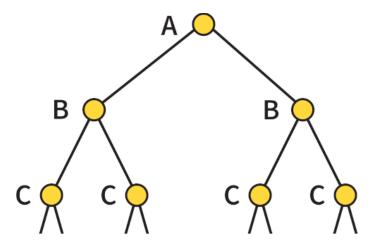
function TREE-SEARCH(*problem*) **returns** a solution, or failure initialize the frontier using the initial state of *problem* **loop do**

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







Repeated states

Failure to detect repeated states can turn a **linear** problem into an **exponential** one!

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

Graph search

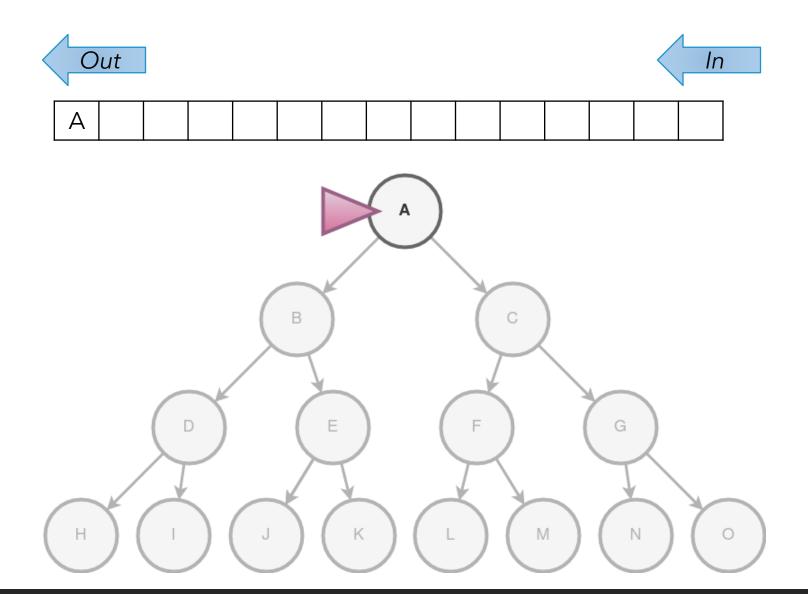
Augment TREE-SEARCH with a new data-structure:

- the **explored set** (closed list), which remembers every expanded node
- newly expanded nodes already in explored set are discarded

Expand **shallowest** unexpanded node

Implementation:

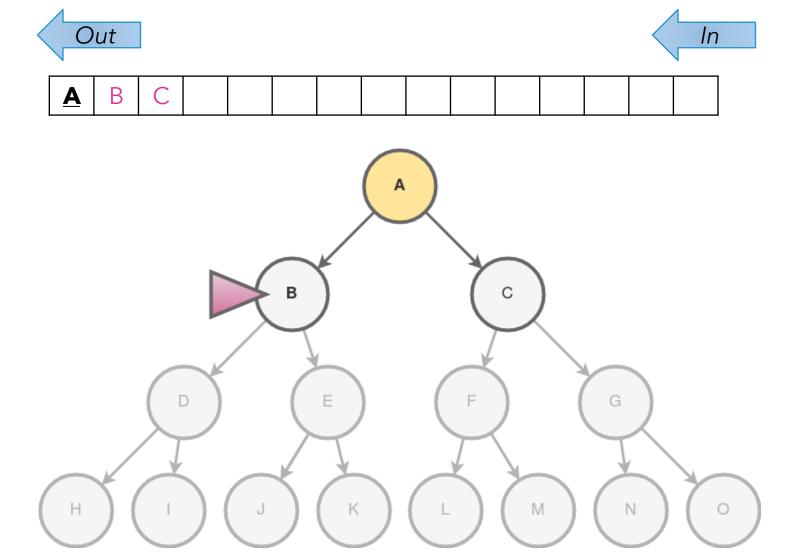
frontier is a FIFO queue,
 i.e., new successors go at end



Expand **shallowest** unexpanded node

Implementation:

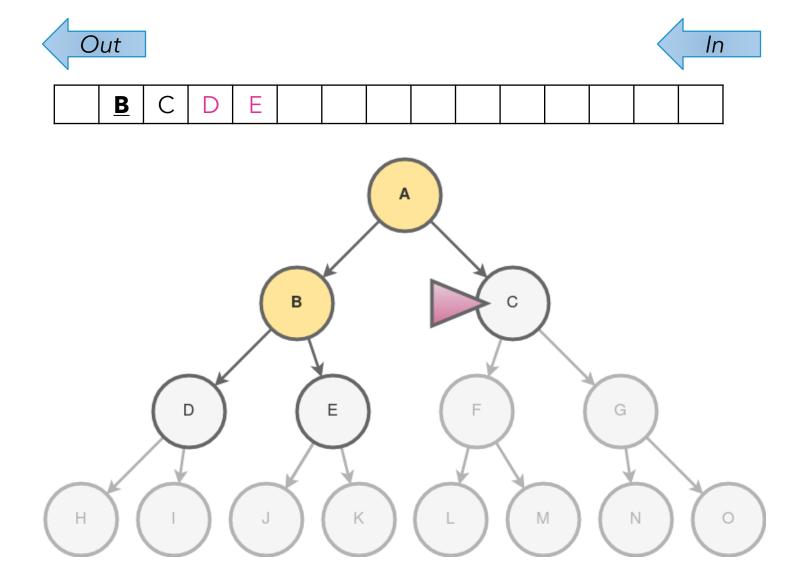
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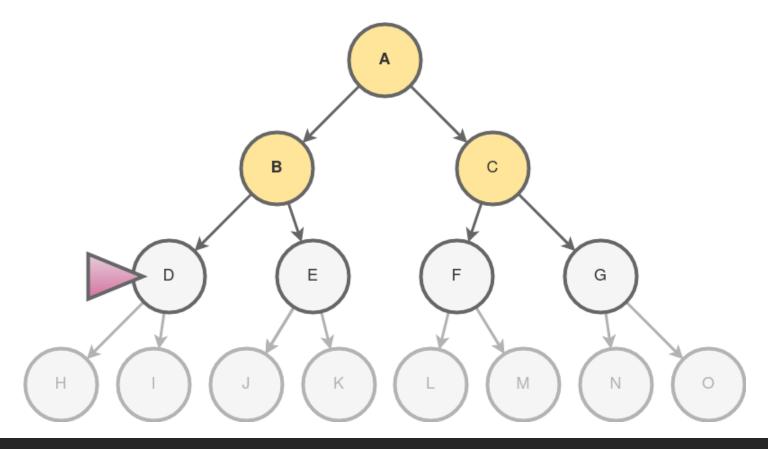


Expand **shallowest** unexpanded node

Implementation:

frontier is a FIFO queue,
 i.e., new successors go at end





```
function BREADTH-FIRST-SEARCH(problem) returns a solution, or failure

node ← a node with STATE = problem.INITIAL-STATE, PATH-COST = 0

if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)

frontier ← a FIFO queue with node as the only element

explored ← an empty set

loop do

if EMPTY?(frontier) then return failure

node ← POP(frontier) /* chooses the shallowest node in frontier */

add node.STATE to explored

for each action in problem.ACTIONS(node.STATE) do

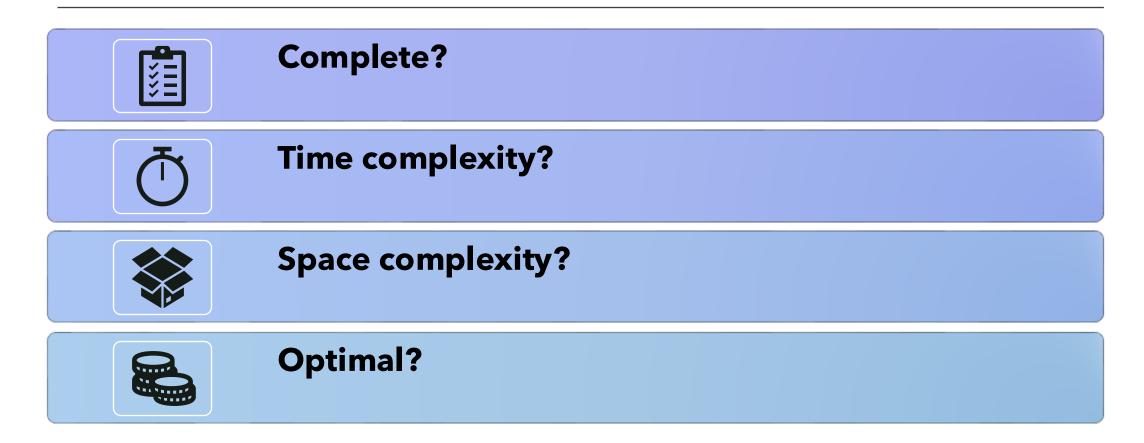
child ← CHILD-NODE(problem, node, action)

if child.STATE is not in explored or frontier then

if problem.GOAL-TEST(child.STATE) then return SOLUTION(child)

frontier ← INSERT(child, frontier)
```

Breadthfirst search algorithm





Complete?

Yes (if b is finite)



Time complexity?



Space complexity?



Optimal?



Complete?

Yes (if b is finite)



Time complexity?

 $b+b^2+b^3+...+b^d=O(b^d)$ (worst-case)



Space complexity?



Optimal?



Complete?

Yes (if b is finite)



Time complexity?

 $b+b^2+b^3+...+b^d=O(b^d)$ (worst-case)



Space complexity?

 $O(b^d)$ (keeps every node in memory)



Optimal?



Complete?

Yes (if b is finite)



Time complexity?

 $b+b^2+b^3+...+b^d=O(b^d)$ (worst-case)



Space complexity?

 $O(b^d)$ (keeps every node in memory)



Optimal?

Yes (if cost = 1 per step)



Complete?

Yes (if b is finite)



Time complexity?

 $b+b^2+b^3+...+b^d=O(b^d)$ (worst-case)



Space complexity?

 $O(b^d)$ (keeps every node in memory)



Optimal?

Yes (if cost = 1 per step)

then optimal solution is closest to start!



Complete?

Yes (if b is finite)



Time complexity?

 $b+b^2+b^3+...+b^d=O(b^d)$ (worst-case)



Space complexity?

 $O(b^d)$ (keeps every node in memory)



Optimal?

Yes (if cost = 1 per step)

Space is the bigger problem (more than time)

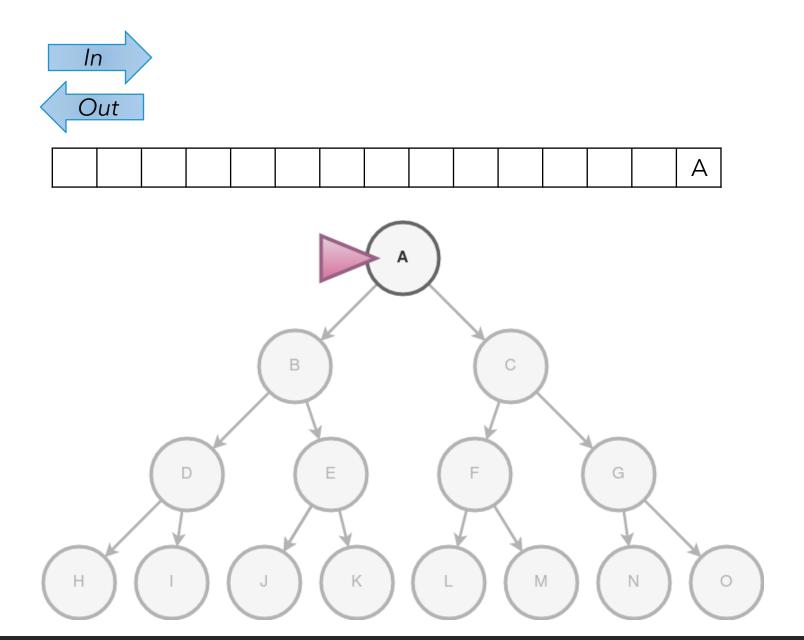
Depth	Nodes	Time	Memory
2	110	.11 milliseconds	107 kilobytes
4	11,110	11 milliseconds	10.6 megabytes
6	10^{6}	1.1 seconds	1 gigabyte
8	10^{8}	2 minutes	103 gigabytes
10	10^{10}	3 hours	10 terabytes
12	10^{12}	13 days	1 petabyte
14	10^{14}	3.5 years	99 petabytes
16	10^{16}	350 years	10 exabytes

Figure 3.13 Time and memory requirements for breadth-first search. The numbers shown assume branching factor b=10; 1 million nodes/second; 1000 bytes/node.

Expand **deepest** unexpanded node

Implementation:

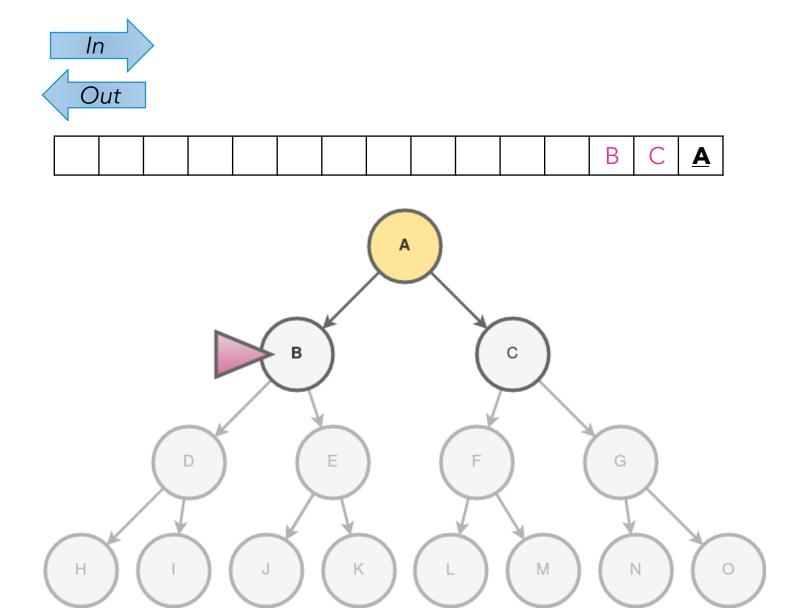
frontier is a LIFO queue,
 i.e., new successors go at front



Expand **deepest** unexpanded node

Implementation:

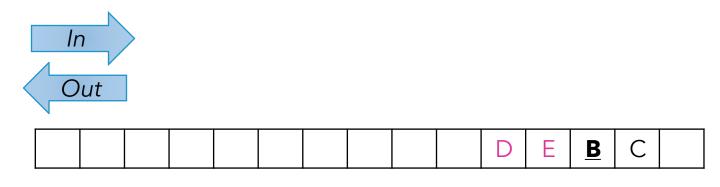
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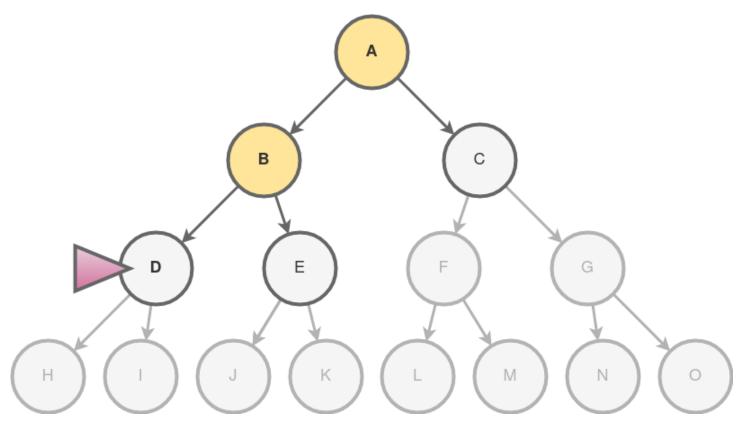


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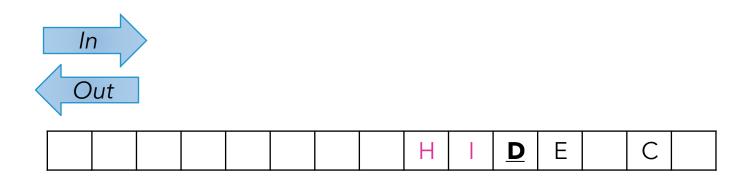


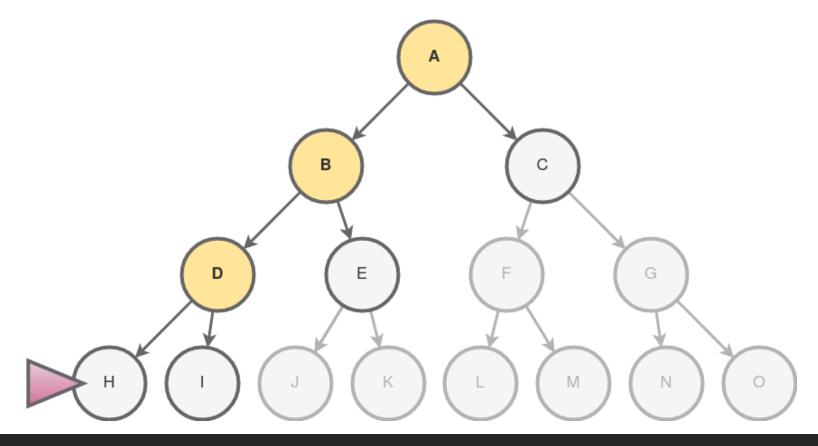


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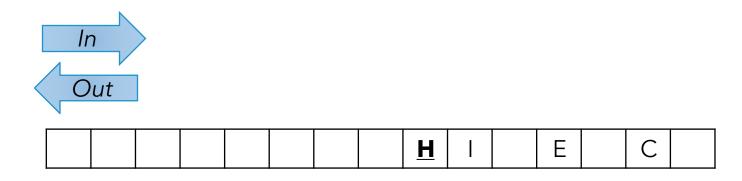


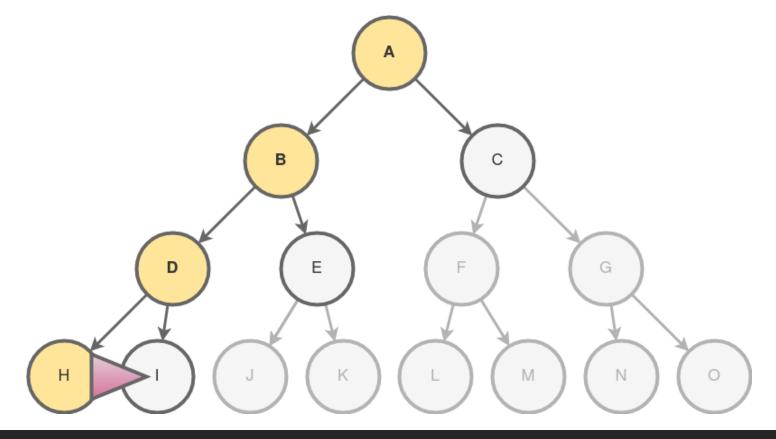


Expand **deepest** unexpanded node

Implementation:

frontier is a LIFO queue,
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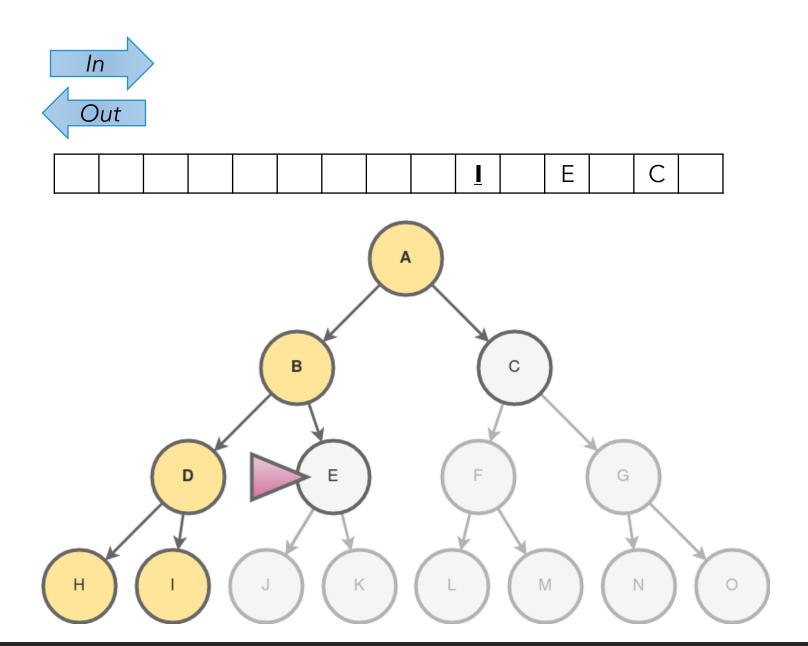




Expand **deepest** unexpanded node

Implementation:

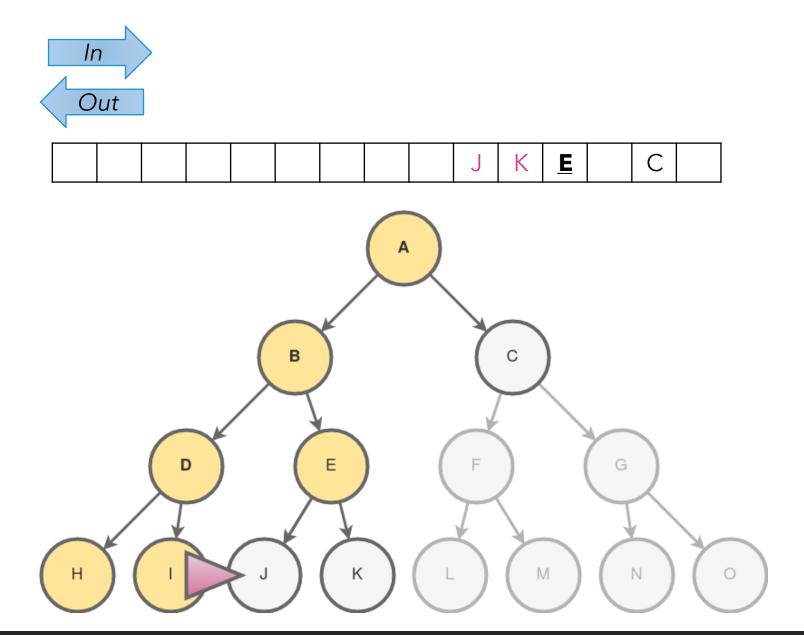
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Expand **deepest** unexpanded node

Implementation:

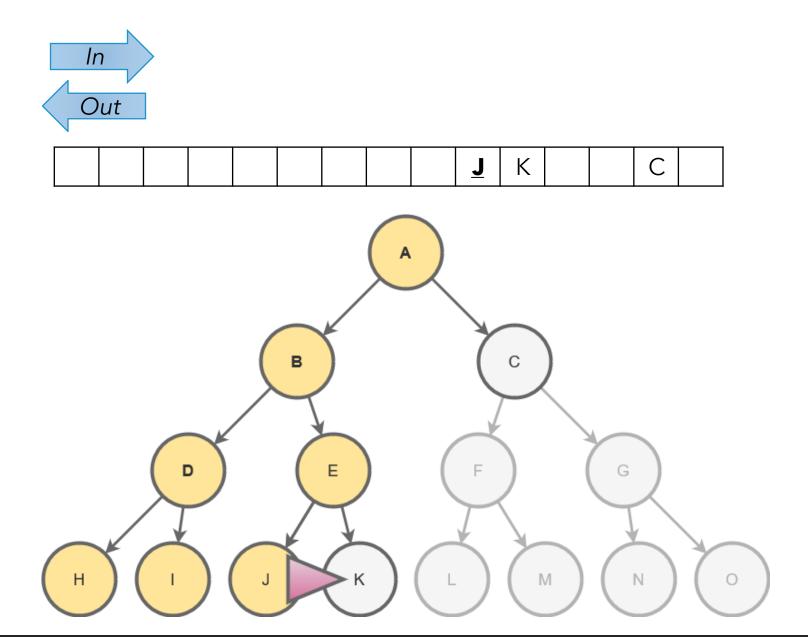
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Expand **deepest** unexpanded node

Implementation:

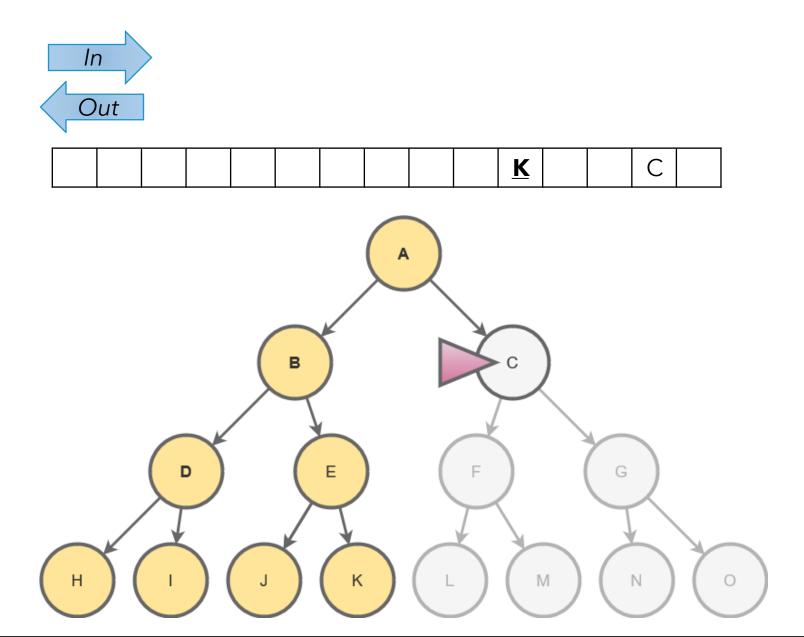
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Expand **deepest** unexpanded node

Implementation:

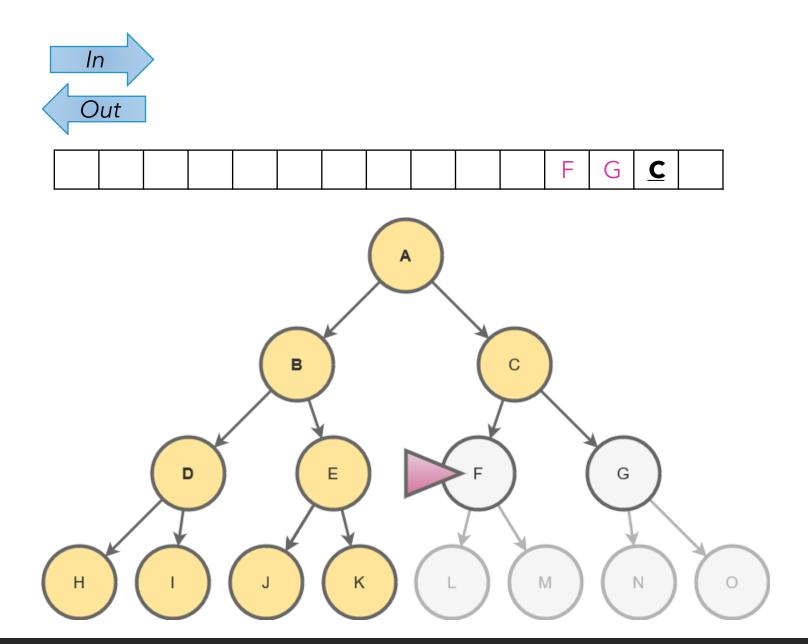
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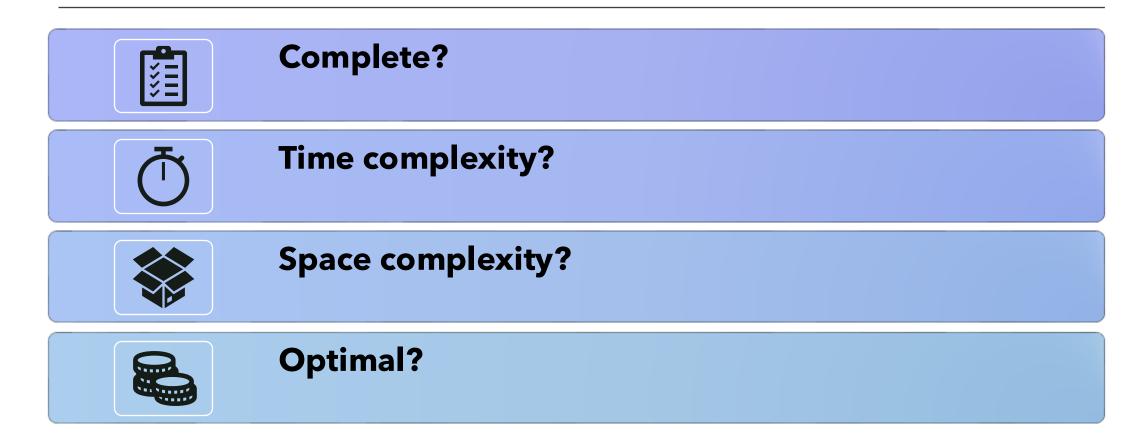


Expand **deepest** unexpanded node

Implementation:

frontier is a LIFO queue,
 i.e., new successors go at front







Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?



Space complexity?



Optimal?



Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?

avoid repeated states along path; complete in finite spaces



Space complexity?



Optimal?



Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?

 $O(b^m)$: terrible if m is much larger than d



Space complexity?



Optimal?



Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?

 $O(b^m)$: terrible if m is much larger than d



Space complexity?

If solutions are dense, depth-first may be much faster than breadth-first!



Optimal?

Properties of depth-first search



Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?

 $O(b^m)$: terrible if m is much larger than d



Space complexity?

O(bm), i.e., linear space!



Optimal?

Properties of depth-first search



Complete?

No: fails in infinite-depth spaces, spaces with loops



Time complexity?

 $O(b^m)$: terrible if m is much larger than d



Space complexity?

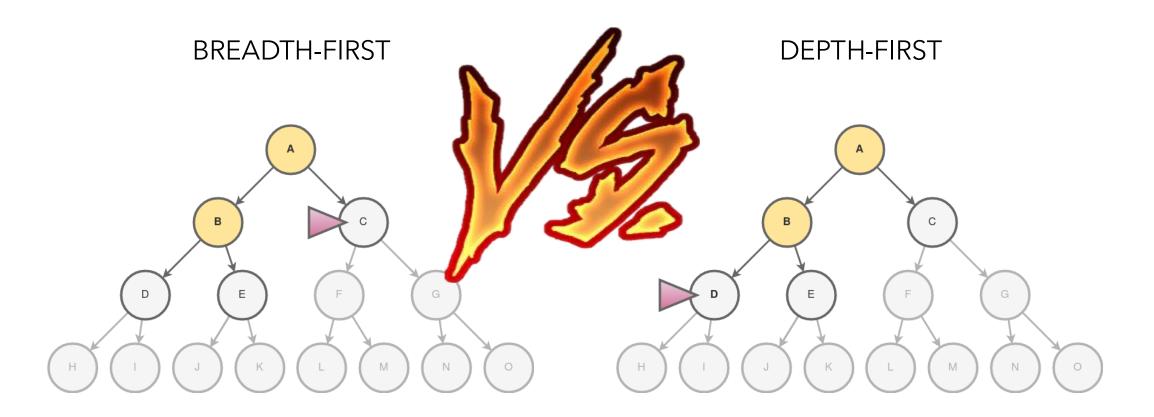
O(bm), i.e., linear space!



Optimal?

No

Mid-Lecture Exercise



Mid-Lecture Exercise

BREADTH-FIRST

DEPTH-FIRST

- When completeness is important.
- When optimal solutions are important.

 When solutions are dense and low-cost is important, especially space costs.

```
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
```

Depth-limited search

This is depth-first search with depth limit *l*, i.e., nodes at depth *l* have no successors

Properties of depth-limited tree search



Complete?

No



Time complexity?

 $O(b^l)$



Space complexity?

O(bl), i.e., linear space!



Optimal?

No

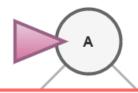
Iterative deepening search

... or how to improve depth-first search

Iterative deepening search

```
\begin{aligned} &\textbf{function ITERATIVE-DEEPENING-SEARCH}(\textit{problem}) \textbf{ returns} \text{ a solution, or failure} \\ &\textbf{for } \textit{depth} = 0 \textbf{ to } \infty \textbf{ do} \\ &\textit{result} \leftarrow \text{DEPTH-LIMITED-SEARCH}(\textit{problem}, \textit{depth}) \\ &\textbf{if } \textit{result} \neq \text{cutoff } \textbf{then } \textbf{return } \textit{result} \end{aligned}
```

Iterative deepening search I = 0

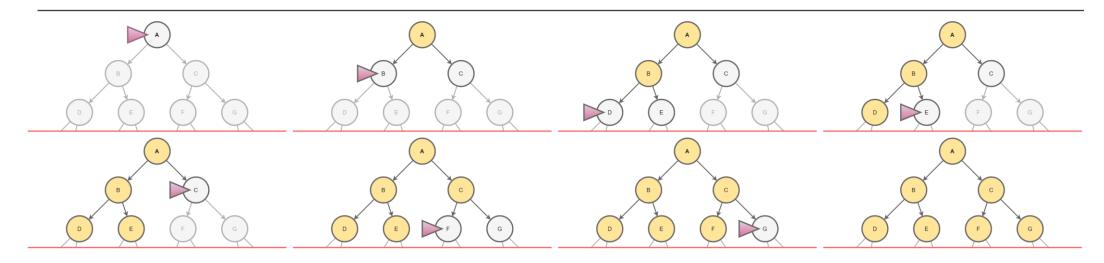




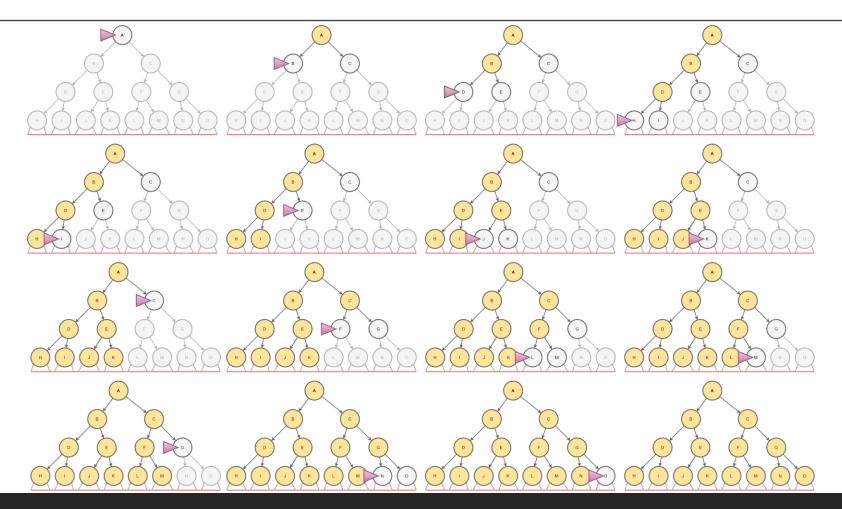
Iterative deepening search / = 1



Iterative deepening search l=2



Iterative deepening search l = 3



Iterative deepening search

Number of nodes generated in an iterative deepening search to depth d with branching factor b:

$$N_{IDS} = (d)b + (d-1)b^2 + ... + (2)b^{d-1} + (1)b^d$$

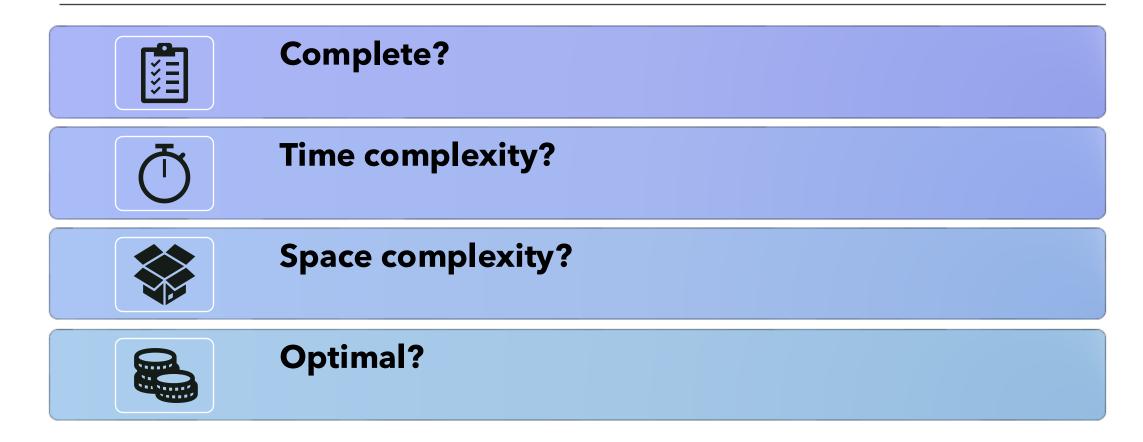
Some cost associated with generating upper levels multiple times

Example: For b = 10, d = 5,

$$\circ$$
 N_{BFS} = 10 + 100 + 1,000 + 10,000 + 100,000 = 111,110

$$\circ$$
 N_{IDS} = 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450

Overhead = (123,450 - 111,110)/111,110 = 11%







Complete?

Yes



Time complexity?

$$(d)b + (d-1)b^2 + ... + (1)b^d = O(b^d)$$



Space complexity?



Optimal?



Complete?

Yes



Time complexity?

$$(d)b + (d-1)b^2 + ... + (1)b^d = O(b^d)$$



Space complexity?

O(bd)



Optimal?



Complete?

Yes



Time complexity?

$$(d)b + (d-1)b^2 + ... + (1)b^d = O(b^d)$$



Space complexity?

O(bd)



Optimal?

Yes, if step cost = 1

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

Summary of algorithms

Summary

Variety of uninformed search strategies:

o breadth-first, depth-first, depth-limited, iterative deepening

Iterative deepening search uses only linear space and not much more time than other uninformed algorithms

Why?

- Very common algorithms.
- Used whenever we are looking for a path in a tree or graph.
 - Anywhere from games to programming languages.
- Properties matter!
 - time and/or space complexity.
- Understanding which algorithm to use in what circumstances.