Artificial Intelligence

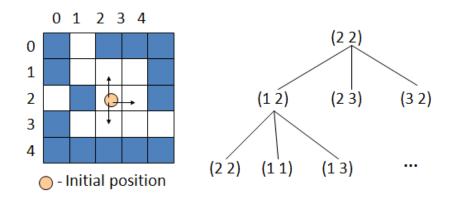
COURSE INSTRUCTOR: MUHAMMAD SAIF UL ISLAM

Outline

- Uninformed search
- Search strategies
- Problem formulation
- Uniform Search Strategies
 - Depth First Search
 - Breadth First Search
 - Uniform cost Search
 - Depth-Limit Search
 - Iterative Deepening Search
 - Bidirectional Search

Uninformed Search Strategies

Uninformed (blind) strategies use only the information available in the problem definition

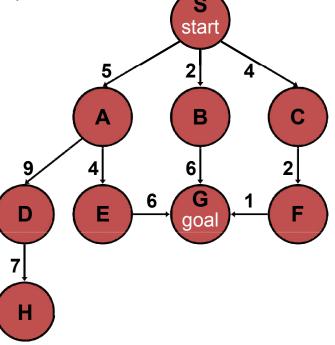


- Root node is expanded first -> successor of root node -> their successors
- All nodes are expanded at a given depth in the search tree before any nodes at the next level are expanded
- Easy to implement by using FIFOqueue

generalSearch(problem, queue)

of nodes tested: 0, expanded: 0

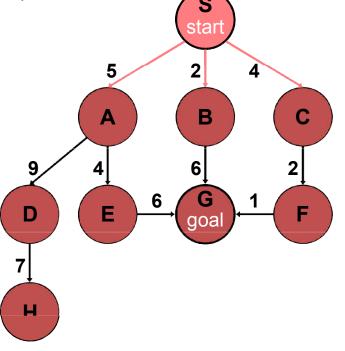
expnd. node	Frontier list
	{S}



generalSearch(problem, queue)

of nodes tested: 1, expanded: 1

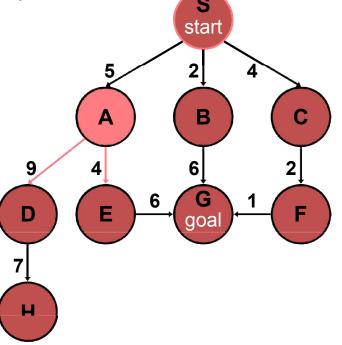
expnd. node	Frontier list
	{S}
S not goal	{A,B,C}



generalSearch(problem, queue)

of nodes tested: 2, expanded: 2

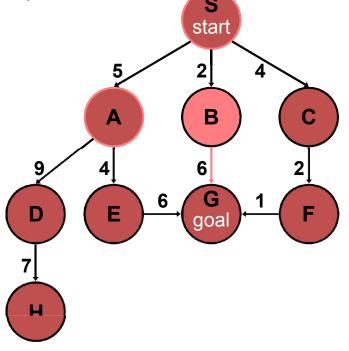
expnd. node	Frontier list
	{S}
S	{A,B,C}
A not goal	$\{B,C,D,E\}$



generalSearch(problem, queue)

of nodes tested: 3, expanded: 3

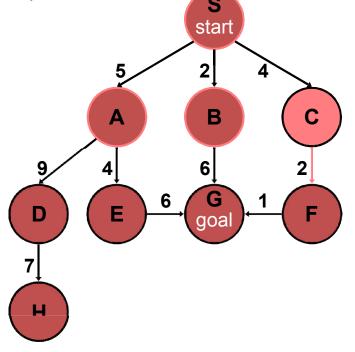
expnd. node	Frontier list	
	{S}	
S	{A,B,C}	
А	$\{B,C,D,E\}$	
B not goal	{C,D,E,G}	



generalSearch(problem, queue)

of nodes tested: 4, expanded: 4

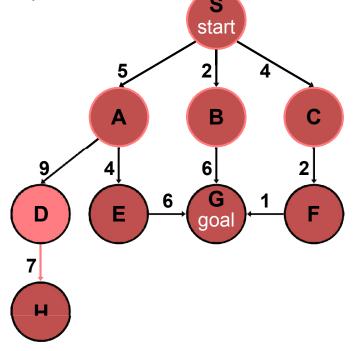
expnd. node Frontier list	
	{S}
S	{A,B,C}
A	{B,C,D,E}
В	{C,D,E,G}
C not goal	{D,E,G,F}



generalSearch(problem, queue)

of nodes tested: 5, expanded: 5

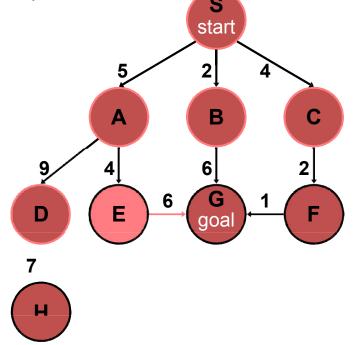
expnd. node	Frontier list
	{S}
S	{A,B,C}
A	{B,C,D,E}
В	{C,D,E,G}
С	{D,E,G,F}
D not goal	{E,G,F,H}



generalSearch(problem, queue)

of nodes tested: 6, expanded: 6

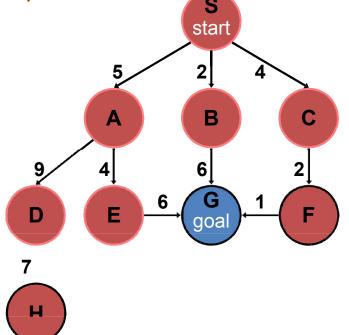
expnd. node	Frontier list
	{S}
S	{A,B,C}
A	$\{B,C,D,E\}$
В	{C,D,E,G}
С	{D,E,G,F}
D	{E,G,F,H}
E not goal	{G,F,H,G}



generalSearch(problem, queue)

of nodes tested: 7, expanded: 6

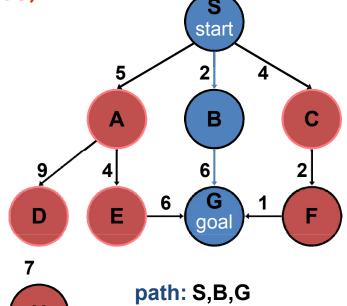
expnd. node	Frontier list
	{S}
S	{A,B,C}
Α	$\{B,C,D,E\}$
В	{C,D,E,G}
С	{D,E,G,F}
D	{E,G,F,H}
E	{G,F,H,G}
G goal	{F,H,G} no expand



generalSearch(problem, queue)

of nodes tested: 7, expanded: 6

expnd. node	Frontier list	
	{S}	
S	{A,B,C}	
A	$\{B,C,D,E\}$	
В	{C,D,E,G}	
С	{D,E,G,F}	
D	{E,G,F,H}	
E	{G,F,H,G}	
G	{F,H,G}	



cost: 8

Properties of Breadth-First Search

• Completeness: Yes, if b is finite

Optimality: Yes (assuming cost = 1 per step)

Time complexity: $1+b+b^2+...+b^d = O(b^d)$, i.e., exponential in d

• Space complexity: O(b d)

maximum branching factor of the search tree

d - depth of the least-cost solutionm - maximum depth of the state space (may be infinite)

Exponential Growth

Depth	Nodes		Time		Memory
2	110	0.11	millisecond	107	kbytes
4	11,110	11	millisecond	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

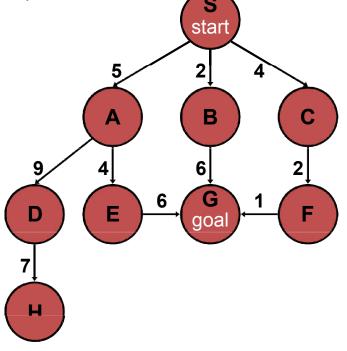
Time and memory requirements for breadth-first search, assuming a branching factor of 10, 100 bytes per node and searching 1000 nodes/second

- Expands the deepest node in the current frontier of the search tree until the nodes have no successors, then backs up to the next deepest node that still has unexplored successors.
- Use LIFO queue
 - The most recently generated node is chosen for expansion, the deepest unexpanded node
- Use recursive

generalSearch(problem, stack)

of nodes tested: 0, expanded: 0

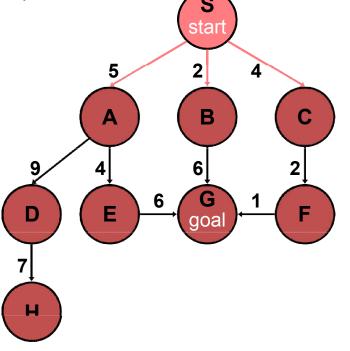
expnd. node	Frontier
	{S}



generalSearch(problem, stack)

of nodes tested: 1, expanded: 1

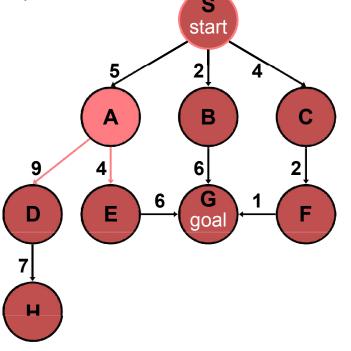
expnd. node	Frontier
	{S}
S not goal	{A,B,C}



generalSearch(problem, stack)

of nodes tested: 2, expanded: 2

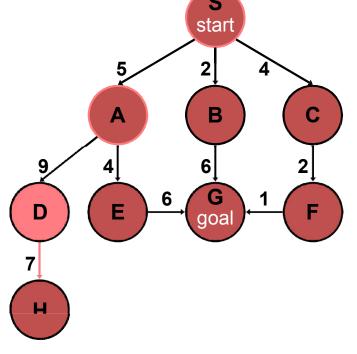
expnd. node	Frontier
	{S}
S	{A,B,C}
A not goal	$\{D,E,B,C\}$



generalSearch(problem, stack)

of nodes tested: 3, expanded: 3

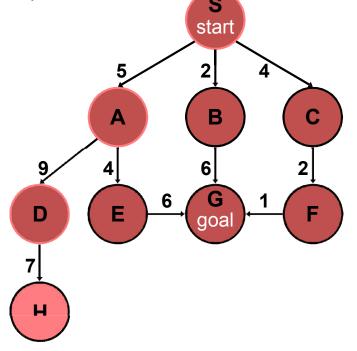
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{D,E,B,C}
D not goal	{H,E,B,C}



generalSearch(problem, stack)

of nodes tested: 4, expanded: 4

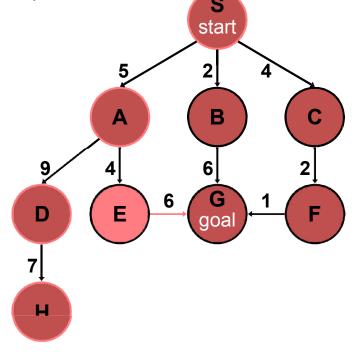
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{D,E,B,C}
D	{H,E,B,C}
H not goal	{E,B,C}



generalSearch(problem, stack)

of nodes tested: 5, expanded: 5

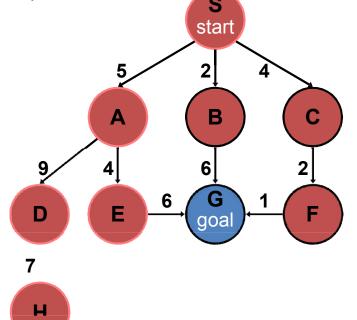
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{D,E,B,C}
D	{H,E,B,C}
Н	{E,B,C}
E not goal	{G,B,C}



generalSearch(problem, stack)

of nodes tested: 6, expanded: 5

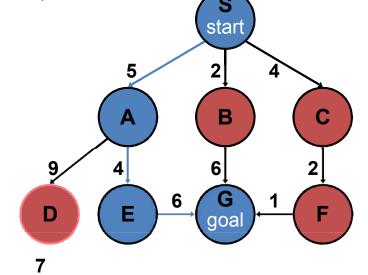
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{D,E,B,C}
D	{H,E,B,C}
Н	{E,B,C}
E	{G,B,C}
G goal	{B,C} no expand



generalSearch(problem, stack)

of nodes tested: 6, expanded: 5

expnd. node	Frontier
	{S}
S	{A,B,C}
A	{D,E,B,C}
D	{H,E,B,C}
Н	{E,B,C}
E	{G,B,C}
G	{B,C}



path: S,A,E,G

cost: 15

Properties of Depth-First Search

• Completeness: (yes, if finite state space)

• Time complexity: O(b m)

• Space complexity: O(bm)

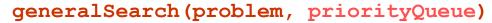
Optimality: No

Remember:

b = branching factor

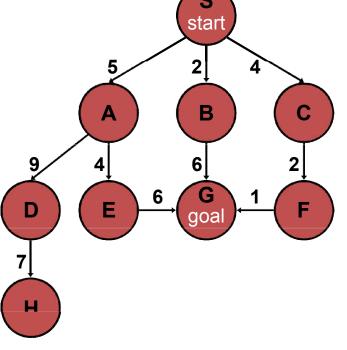
m = max depth of search tree

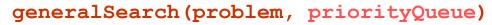
- Use a "Priority Queue" to order nodes on the Frontier list, sorted by path cost
- Let $g(n) = \cos t$ of the path from start node s to current node n
- Sort nodes by increasing the value of g



of nodes tested: 0, expanded: 0

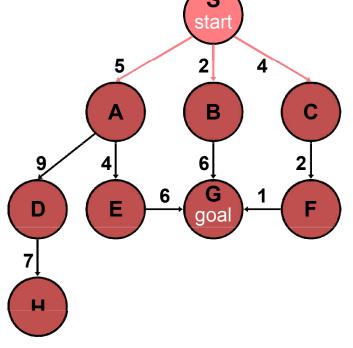
expnd. node	Frontier list
	{S}





of nodes tested: 1, expanded: 1

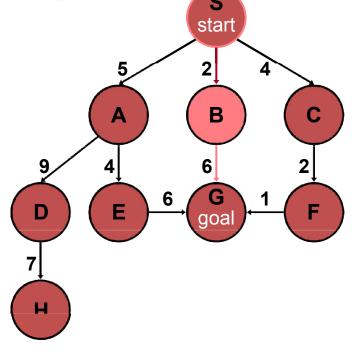
expnd. node	Frontier list
	{S:0}
S not goal	{B:2,C:4,A:5}



generalSearch(problem, priorityQueue)

of nodes tested: 2, expanded: 2

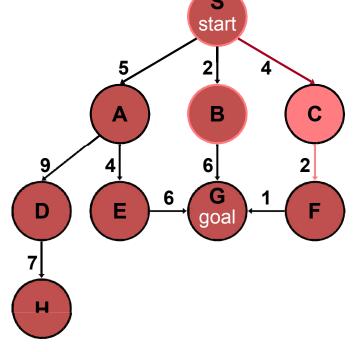
expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
B not goal	{C:4,A:5,G:2+6}



generalSearch(problem, priorityQueue)

of nodes tested: 3, expanded: 3

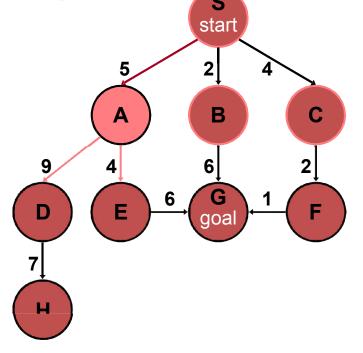
expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
В	{C:4,A:5,G:8}
C not goal	{A:5,F:4+2,G:8}



generalSearch(problem, priorityQueue)

of nodes tested: 4, expanded: 4

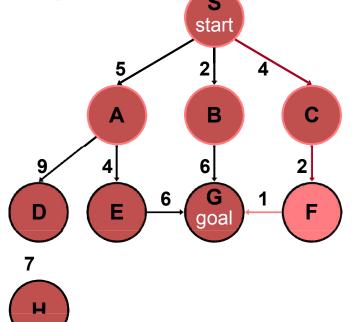
expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
В	{C:4,A:5,G:8}
С	{A:5,F:6,G:8}
A not goal	{F:6,G:8, E:5+4
	, D:5+9}



generalSearch(problem, priorityQueue)

of nodes tested: 5, expanded: 5

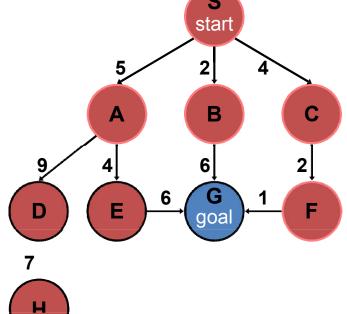
expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
В	{C:4,A:5,G:8}
С	{A:5,F:6,G:8}
A	{F:6,G:8,E:9,D:14}
F not goal	{G:4+2+1,G:8,E:9
	, D:14}



generalSearch(problem, priorityQueue)

of nodes tested: 6, expanded: 5

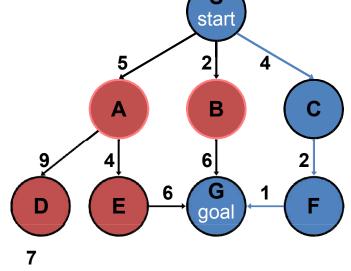
expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
В	{C:4,A:5,G:8}
С	{A:5,F:6,G:8}
Α	{F:6,G:8,E:9,D:14}
F	{G:7,G:8,E:9,D:14}
G goal	{G:8,E:9,D:14}
	no expand



generalSearch(problem, priorityQueue)

of nodes tested: 6, expanded: 5

expnd. node	Frontier list
	{S}
S	{B:2,C:4,A:5}
В	{C:4,A:5,G:8}
С	{A:5,F:6,G:8}
A	{F:6,G:8,E:9,D:14}
F	{G:7,G:8,E:9,D:14}
G	{G:8,E:9,D:14}



path: S,C,F,G cost: 7

Properties of Uniform-Cost Search

- Complete: Yes if arc costs > 0.
- Optimal: Yes
- Time and space complexity: $O(b^d)$ (i.e., exponential)
 - -d is the depth of the solution
 - is the branching factor at each non-leafnode
- More precisely, time and space complexity is $O(b^{C^*/\epsilon})$ where all edge costs $\epsilon \sum > 0$, and is the best goalpath cost

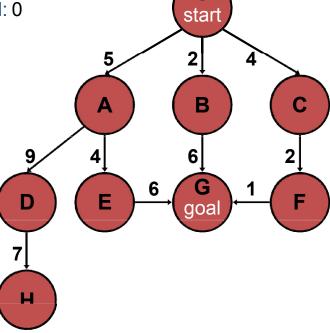
Iterative-Deepening Search (IDS)

- requires modification to DFS search algorithm:
 - do DFS to depth 1
 and treat all children of the start node as leaves
 - if no solution is found, do DFS to depth 2
 - repeat by increasing "depth bound" until a solution found
- Start node is at depth 0

deepeningSearch(problem)

depth: 1, # of nodes expanded: 0, tested: 0

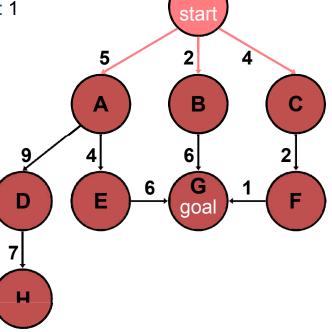
expnd. node	Frontier
	{S}



deepeningSearch(problem)

depth: 1, # of nodes tested: 1, expanded: 1

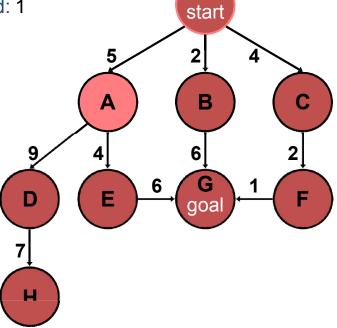
expnd. node	Frontier
	{S}
S not goal	{A,B,C}



deepeningSearch(problem)

depth: 1, # of nodes tested: 2, expanded: 1

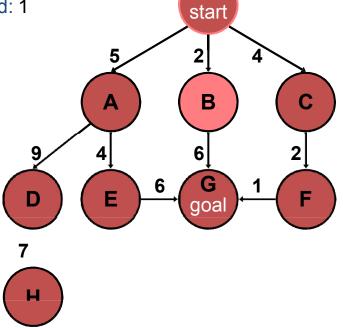
expnd. node	Frontier
	{S}
S	{A,B,C}
A not goal	{B,C} no expand



deepeningSearch(problem)

depth: 1, # of nodes tested: 3, expanded: 1

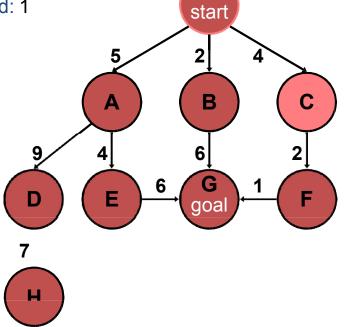
expnd. node	Frontier
	{S}
S	{A,B,C}
Α	{B,C}
B not goal	{C} no expand



deepeningSearch(problem)

depth: 1, # of nodes tested: 4, expanded: 1

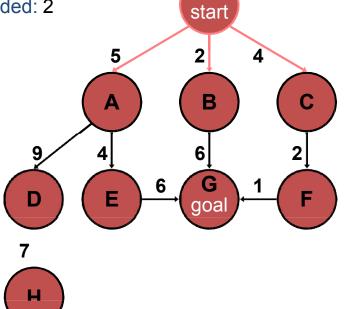
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{B,C}
В	{C}
C not goal	{ } no expand-FAIL



deepeningSearch(problem)

depth: 2, # of nodes tested: 4(1), expanded: 2

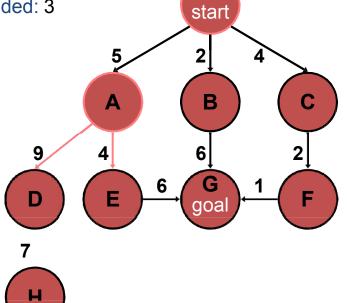
expnd. node	Frontier
	{S}
S	{A,B,C}
A	{B,C}
В	{C}
С	{}
S no test	{A,B,C}



deepeningSearch(problem)

depth: 2, # of nodes tested: 4(2), expanded: 3

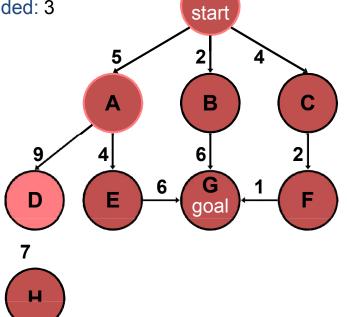
expnd. node	Frontier
	{S}
S	{A,B,C}
Α	{B,C}
В	{C}
С	{}
S	{A,B,C}
A no test	{D,E,B,C}



deepeningSearch(problem)

depth: 2, # of nodes tested: 5(2), expanded: 3

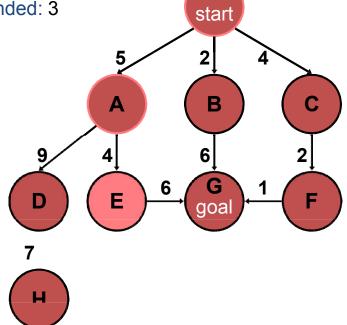
expnd. node	Frontier
	{S}
S	{A,B,C}
Α	{B,C}
В	{C}
С	{}
S	{A,B,C}
Α	{D,E,B,C}
D not goal	{E,B,C} no expand



deepeningSearch(problem)

depth: 2, # of nodes tested: 6(2), expanded: 3

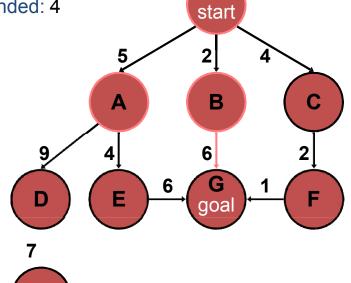
expnd. node	Frontier
	{S}
S	{A,B,C}
Α	{B,C}
В	{C}
С	{}
S	{A,B,C}
Α	{D,E,B,C}
D	{E,B,C}
E not goal	{B,C} no expand



deepeningSearch(problem)

depth: 2, # of nodes tested: 6(3), expanded: 4

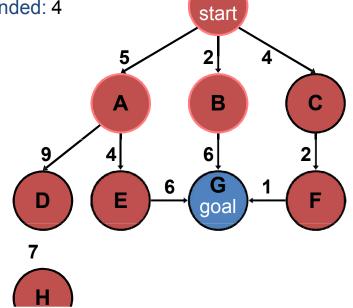
	\ / /
expnd. node	Frontier
	{S}
S	{A,B,C}
A.	{B,C}
B.	{C}
C.	{}
S	{A,B,C}
A	{D,E,B,C}
D.	$\{E,B,C\}$
E.	{B,C}
B no test	{G,C}



deepeningSearch(problem)

depth: 2, # of nodes tested: 7(3), expanded: 4

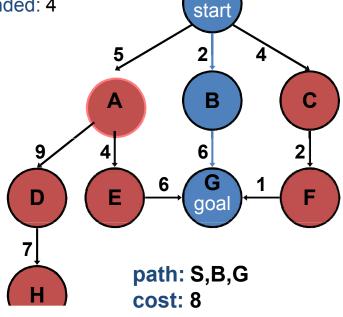
	(//
expnd. node	Frontier
	{S}
S	{A,B,C}
Α	{B,C}
B.	{C}
C.	{}
S	{A,B,C}
A	{D,E,B,C}
D.	$\{E,B,C\}$
E.	{B,C}
В	{G,C}
G goal	(C) no expand



deepeningSearch(problem)

depth: 2, # of nodes tested: 7(3), expanded: 4

expnd. node	Frontier
	{S}
S	{A,B,C}
A.	{B,C}
B.	{C}
C.	{}
S	{A,B,C}
A	{D,E,B,C}
D.	{E,B,C}
E.	{B,C}
В	{G,C}
G	{C}



- Has advantages of BFS
 - completeness
 - optimality as stated for BFS
- Has advantages of DFS
 - limited space
 - in practice, even with the redundant effort it still finds longer paths more quickly than BFS

- Space complexity: O(bd) (i.e., linear like DFS)
- Time complexity is a little worse than BFS or DFS
 - because nodes near the top of the search tree are generated multiple times (redundant effort)
- Worst case time complexity: $O(b^d)$ exponential
 - because most nodes are near the bottom of tree

Howmuch redundant effort is done?

• The number of times the nodes are generated:

$$1b^{d} + 2b^{(d-1)} + \dots + db \le b^{d} / (1 - 1/b)^{2}$$

= $O(b^{d})$

- -d: the solution's depth
- the branching factor at each non-leaf node
- For example: b = 4

$$4^{d} / (1 - \frac{1}{4})^{2} = 4^{d} / (.75)^{2} = 1.78 \times 4^{d}$$

- in the worst case, 78% more nodes are
 searched (redundant effort) than exist at depth d
- as increases, this % decreases

Iterative-Deepening Search

- Trades a little time for a huge reduction in space
 - lets you do breadth-first search with (more space efficient) depth-first search

applications, e.g., games

Anytime algorithm: good for response-time critical

An "anytime" algorithm is an <u>algorithm</u> that can return a valid solution to a <u>problem</u> even if
it's interrupted at any time before it ends. The algorithm is expected to find better and better
solutions the more time it keepsrunning.

Iterative Deepening Search

- It's a Depth First Search, but it does it one level at a time, gradually increasing the limit, until a goal is found.
- Combine the benefits of depth-first and breadth-first search
- like DFS, modest memory requirements O(bd)
- like BFS, it is complete when branching factor is finite, and optimal when the path cost is a nondecreasing function of the dept of the node

Iterative Deepening Search

- May seem wasteful because states are generated multiple times
- but actually not very costly, because nodes at the bottom level are generated only once
- In practice, however, the overhead of these multiple expansions is small, because most of the nodes are towards leaves (bottom) of the search tree:
 - thus, the nodes that are evaluated several times (towards top of tree) are in relatively small number.
- Iterative depending is the preferred uninformed search method when the search space is large and the depth of the solution is unknown

Properties of Iterative DeepeningSearch

Combines the best of breadth-first and depth-first search strategies.

• Completeness: Yes,

• Time complexity: O(b d

• Space complexity: *O(bd)*

• Optimality: Yes, if step cost = 1

Is a depth-first search with a predetermined depth $\,$ limit $\,$ $\,$

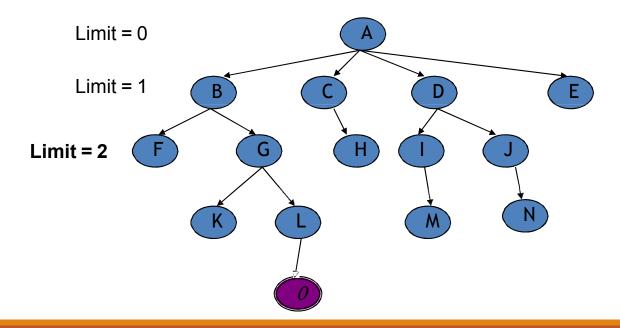
Implementation:

Nodes at depth I are treated as if they have no successors.

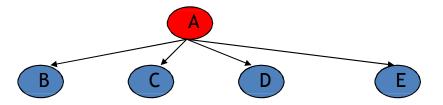
Complete: if cutoff chosen appropriately then it is guaranteed to find a solution.

Optimal: it does not guarantee to find the least-cost solution (If choosing I > d)

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

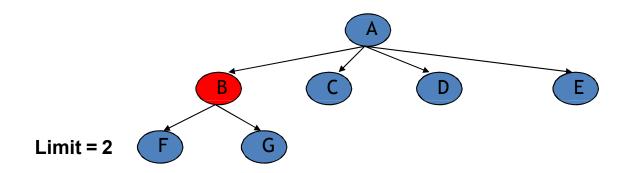


A,

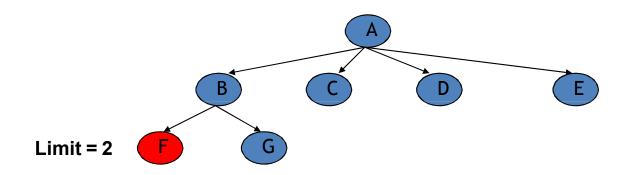


Limit = 2

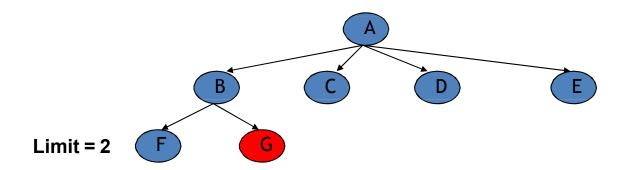
A,B,



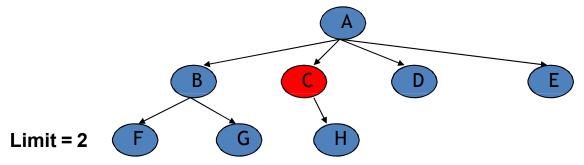
A,B,F,



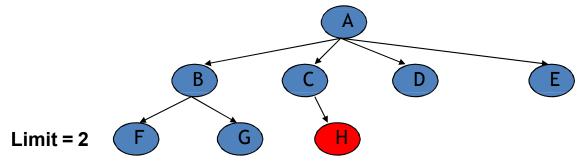
- A,B,F,
- G,

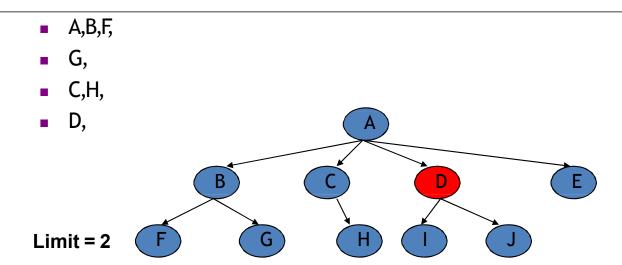


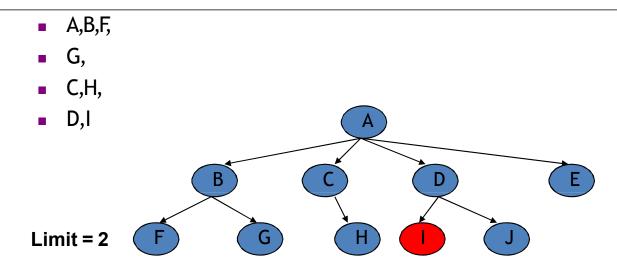
- A,B,F,
- G,
- C.

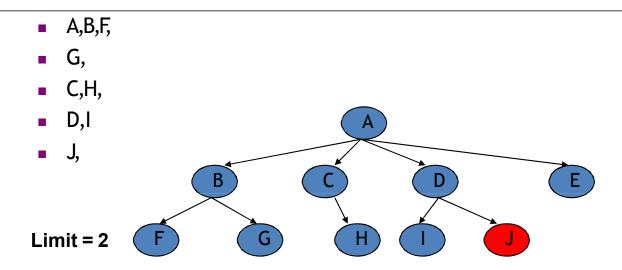


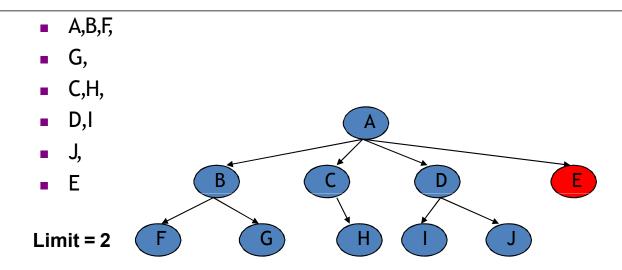
- A,B,F,
- G,
- C,H,

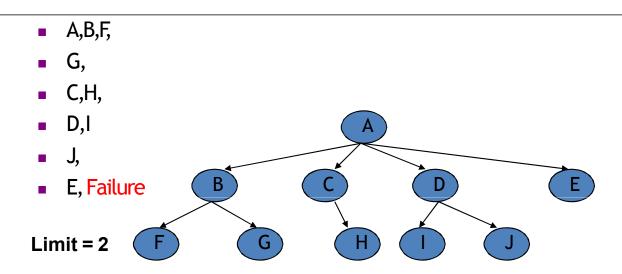




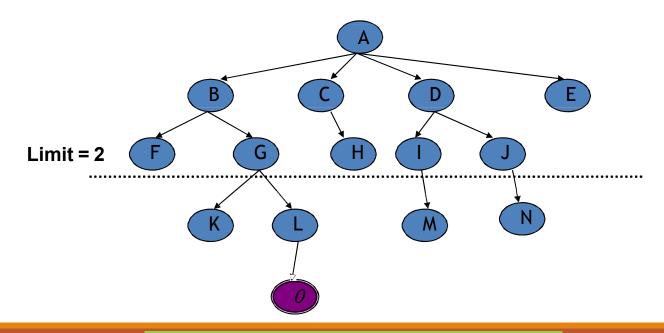








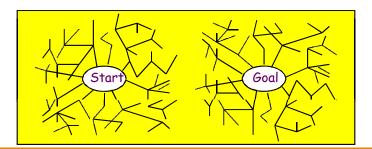
- DLS algorithm returns Failure (no solution)
- The reason is that the goal is beyond the limit (Limit =2): the goal depth is (d=4)

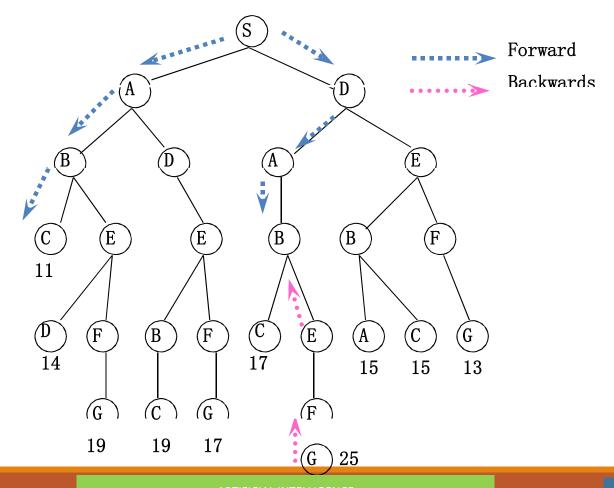


How to choose depthlimit I

- E.g. Romania case, 20 cities
 - The longest is 19, so I can be 19
 - From the study, cities can be reached by other cities by 9 steps (diameter)
 - If the diameter is known, more efficient depth-limited search
 - Depth limit search can be terminated by

- Both search forward from the initial state, and backward from the goal.
- Stop when the two searches meet in the middle.
- Motivation: bd/2 + bd/2 is much less than bd
- Implementation
 - Replace the goal test with a check to see whether the frontiers of the two searches intersect, if yes
 - solution is found





- Check when each node is expanded or selected for expansion
- Can be implemented using BFS or iterative deepening (but at least one frontier needs to be kept in memory)
- Significant weakness
 - Space requirement
- Time Complexity is good

- Problem: how do we search backward from goal??
 - predecessor of node n = all nodes that have n as successor
 - this may not always be easy to compute!
 - if several goal states, apply predecessor function to them just as we applied successor (only works well if goals are explicitly known; may be difficult if goals only characterized implicitly).
 - for bidirectional search to work well, there must be an efficient way to check whether a given node belongs to the other search tree.
 - select a given search algorithm for each half.

- Completeness: Yes,
- Time complexity: $2*O(b^{d/2}) = O(b^{d/2})$
- Space complexity: *O*(*b* ^{*m*/2})
- Optimality: Yes
- To avoid one by one comparison, we need a hash table of size $O(b^{m/2})$
- If hash table is used, the cost of comparison is O(1)

Comparison Uninformed Search Strategies

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

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 b for positive b optimal if step costs are all identical; b if both directions use breadth-first search.

ARTIFICIAL INTELLIGENCE 7

Conclusions

- Interesting problems can be cast as search problems, but you've got to set up state space
- Problem formulation usually requires abstracting away real-world details to define a state space that can feasibly be explored