Al 2002 Artificial Intelligence

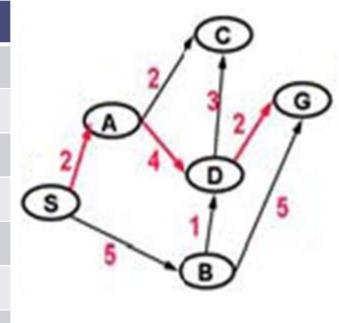
The uniform-cost search expands the node n with the lowest path cost g(n).

Breadth First Search vs Uniform Cost Search:

- Uniform cost search has two significant differences from breadth-first search.
 - 1. The goal test is applied to a node when it is selected for expansion rather than when it is first generated.
 - The goal test is added in case if a better path is found to a node currently on the frontier

- ☐ Pick best (by **path length**) element of Q
- ☐ Add path extensions to Q

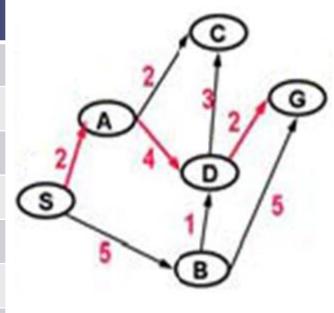
	Q
1	<u>(0 S)</u>
2	(2 A S) (5 B S)
3	(4 C A S) (6 D A S) (5 B S)
4	(6 D A S) <u>(5 B S)</u>



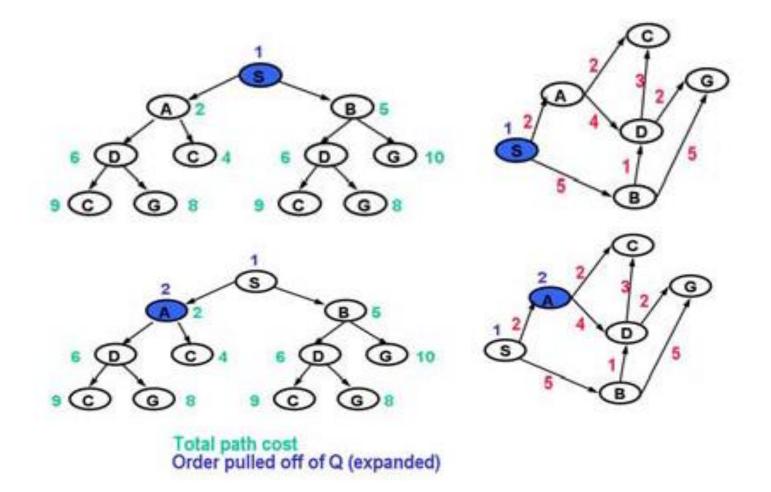
- **☐** Blue Color represents added paths
- Underline paths are chosen for extension.

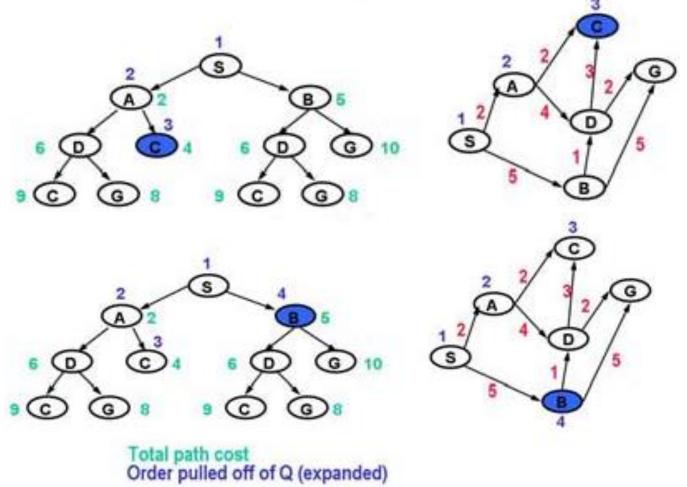
- ☐ Pick best (by **path length**) element of Q
- Add path extensions to Q

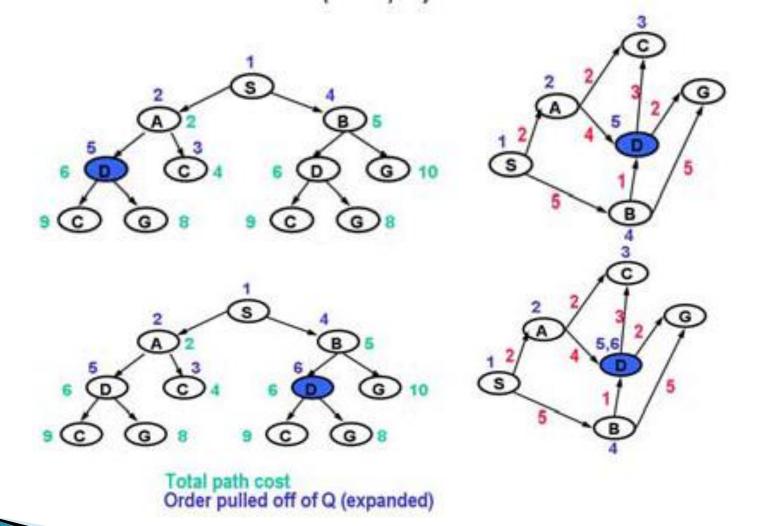
- Add path extensions to Q			
	Q		
1	<u>(0 S)</u>		
2	(2 A S) (5 B S)		
3	(4 C A S) (6 D A S) (5 B S)		
4	(6 D A S) <u>(5 B S)</u>		
5	(6 D B S) (10 G B S) (6 D A S)		
6	(8 G D B S) (9 C D B S) (10 G B S) (6 D A S)		
7	(8 G D A S) (9 C D A S) (8 G D B S) (9 C D B S) (10 G B S)		

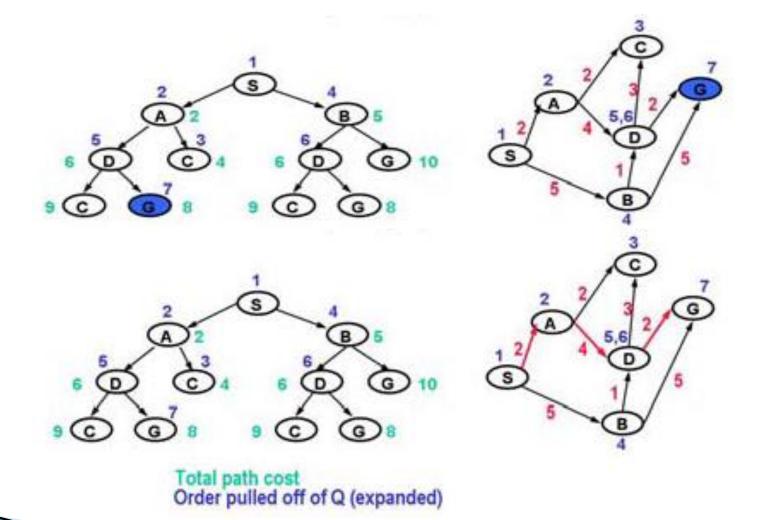


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- Almost Equivalent to breadth-first
 - if step costs are all equal.
 - except that the Breadth-first search stops as soon as it generates a goal, whereas uniform-cost search examines all the nodes at the goal's depth to see if one has a lower cost;
 - Thus uniform-cost search does strictly more work by expanding nodes at depth d unnecessarily.
- It is implemented by the queue ordered by path cost rather than depths,
- Its complexity cannot easily be characterized in terms of b and d.

Completeness

- Uniform-Cost Search is considered complete
 - when the cost of every step is greater than or equal to some small positive constant " ϵ ".
- It will get stuck in an infinite loop if there is a path with an infinite sequence of zero-cost actions

Optimality

- The above condition is also sufficient to ensure optimality.
 - It means that the cost of a path always increases as we go along the path.

Let C* be the cost of the optimal solution, and assume that every action costs at least ε. Then the algorithm's worst-case time and space is

$$o(b^{1+\left\lfloor c^*/_{\varepsilon}\right\rfloor})$$

Which can be much greater than $m{b}^{m{d}}$

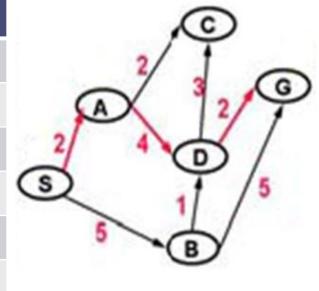
When all step costs are equal, of course,

$$o(b^{1+|c^*/\varepsilon|}) = o(b^{d+1})$$

(with strict Expanded List)

- ☐ Pick best (by **path length**) element of Q
- ☐ Add path extensions to Q

	Q	Expanded		
1	(<u>0 S)</u>	S		
2	(2 A S) (5 B S)	S, A		
3	(4 C A S) (6 D A S) (5 B S)	S, A, C		

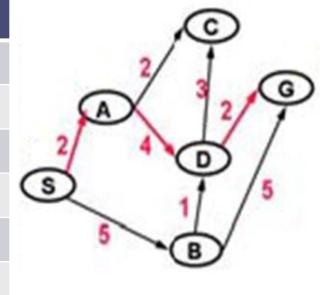


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(with strict Expanded List)

- ☐ Pick best (by **path length**) element of Q
- ☐ Add path extensions to Q

	Q	Expanded		
1	(<u>0 S)</u>	S		
2	(2 A S) (5 B S)	S, A		
3	(4 C A S) (6 D A S) (5 B S)	S, A, C		
4	(6 D A S) <u>(5 B S</u>)	S, A, C, B		
5	(6 D B S) (10 G B S) (<u>6 D A S</u>)	S, A, C, B, D		

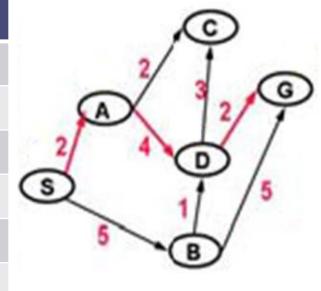


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(with strict Expanded List)

- ☐ Pick best (by **path length**) element of Q
- ☐ Add path extensions to Q

	Q	Expanded	
1	(<u>0 S)</u>	S	
2	(2 A S) (5 B S)	S, A	
3	(4 C A S) (6 D A S) (5 B S)	S, A, C	
4	(6 D A S) <u>(5 B S</u>)	S, A, C, B	
5	(6 D B S) (10 G B S) (<u>6 D A S</u>)	S, A, C, B, D	
6	(8 G D A S) (9 C D A S) (10 G B S)	S, A, C, B, D, G	

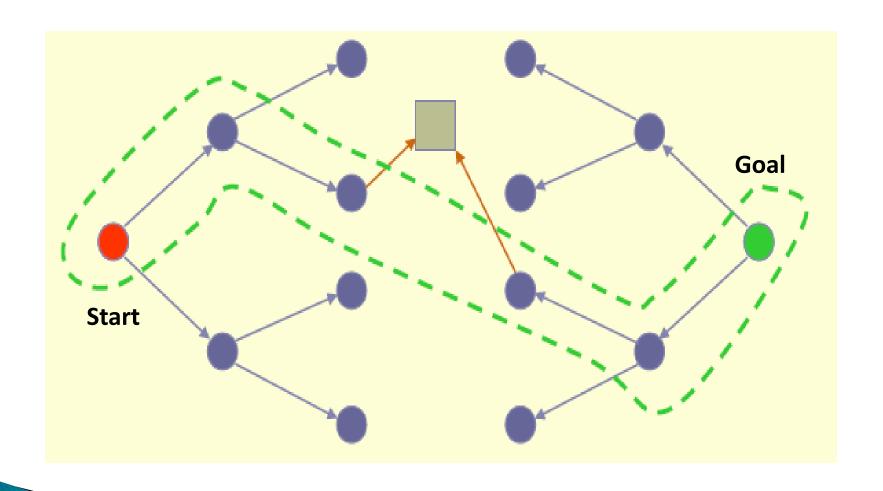


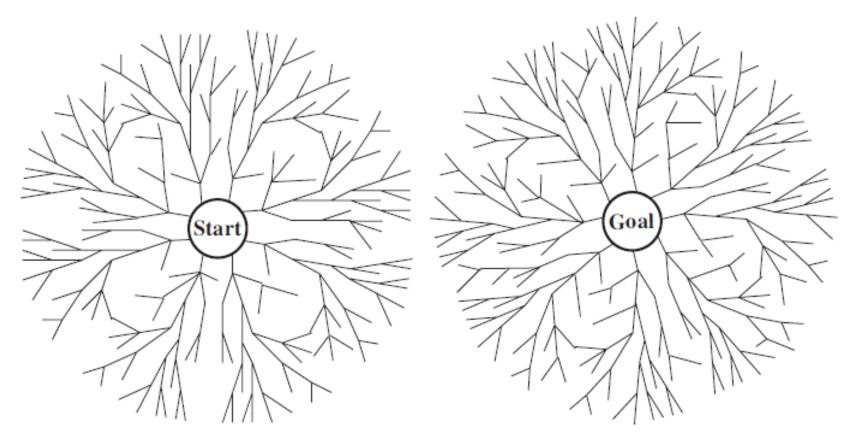
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- Uniform-cost search is concerned only with expanding shortest paths.
- It pays no particular attention to the goal (since it has no any way of knowing where it is).

Uniform-cost search is an algorithm for finding the shortest paths to all states in a graph rather than being focused in reaching a particular goal.

- Run two simultaneous searches
 - forward from the initial state
 - backward from the goal
 - stop when the two searches meet
- However, computing backward is difficult
 - A huge amount of goal states
 - At the goal state, which actions are used to compute it?
 - Can the actions be reversible to compute its predecessors?





A schematic view of a bidirectional search that is about to succeed when a branch from the start node **meets** a branch from the goal node.

Time and space complexity

$$O(b^{d/2}) << O(b^d)$$

For example,

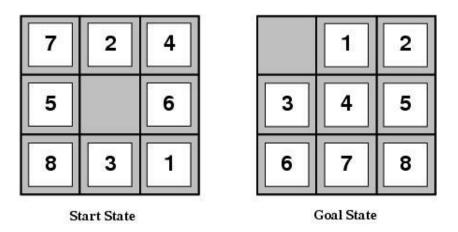
- If a problem has solution depth d=6, and each direction runs breadth-first search one node at a time, then in the worst case the two searches meet when they have generated all of the nodes at depth 3.
- For b = 10, this means a total of 2,220 node generations, compared with 1,111,110 for a standard breadth-first search.

$$(10 + 100 + 1000 + 10,000 + 100,000 + 1,000,000) = 1,111,110$$

 $2 \times (10 + 100 + 1000) = 2220$

"Backward from the goal." How?

For example, the 8-puzzle problem or for finding a route in Romania, there is just one goal state, so the backward search is very much like the forward search,



Evaluation of tree-search strategies

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

- •b: branching factor
- d: depth of the shallowest solution;
- •m: maximum depth of the search tree;
- *l*: is the depth limit.

- a) complete if **b** is finite
- b) complete if step costs $\geq \varepsilon$ for positive ε ;
- c) optimal if step costs are all identical;
- d) if both directions use breadth-first search

Avoiding Repeated States

For all search strategies:

- There is possibility of expanding states that have already been encountered and expanded before, on some other path
 - It may cause the path to be infinite
 - loop forever
- Algorithms that forget their history are doomed to repeat it.

Avoiding Repeated States

Three ways to deal with this possibility:

- Do not return to the state it just came from
 - Refuse generation of any successor same as its parent state
- Do not create paths with cycles
 - Refuse generation of any successor same as its ancestor states
- Do not generate any generated state
 - Not only its ancestor states, but also all other expanded states have to be checked against

Reading Material

- Artificial Intelligence, A Modern Approach Stuart J. Russell and Peter Norvig
 - Chapter 3.