

# CS540 Intro to Al Uninformed Search II

University of Wisconsin–Madison Fall 2025, Section 3 November 5, 2025

## Many Al problems can be formulated as search.

How to make a sequence of decisions to reach a desired goal.

Leverage computation and a known model of world dynamics to make decisions.

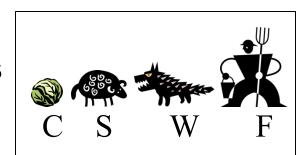
"How the world changes in response to agent actions"





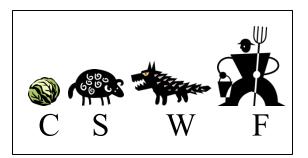
## The search problem

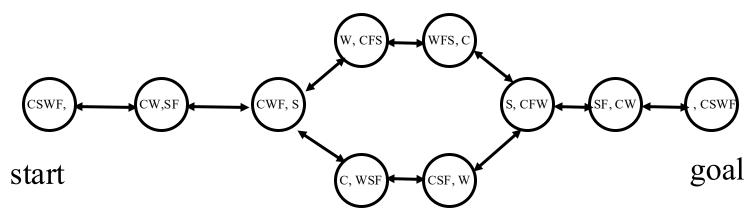
- State space S: all valid configurations
- Initial state *I* = {(CSWF,)} ⊆ *S*
- Goal state **G** = {(,CSWF)} ⊆ **S**



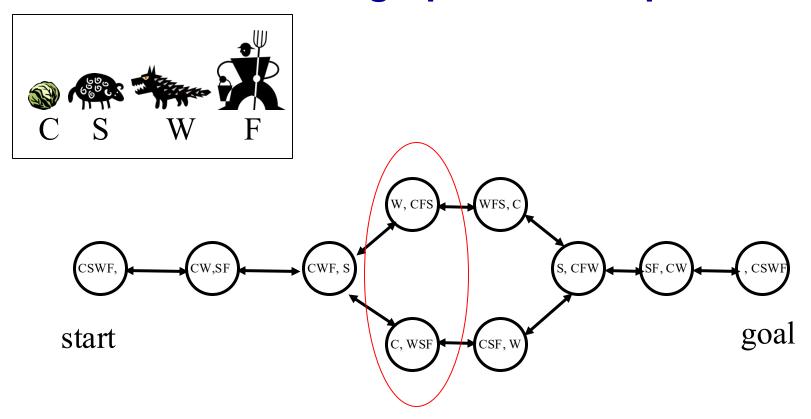
- Successor function succs(s) ⊆ S: states reachable in one step from state s
  - **succs**((CSWF,)) = {(CW, SF)}
  - succs((CWF,S)) = {(CW,FS), (W,CFS), (C, WFS)}
- Cost(s,s')=1 for all steps. (weighted later)
- The search problem: find a solution path from a state in / to a state in G.
  - Optionally minimize the cost of the solution.

## A directed graph in state space





## A directed graph in state space



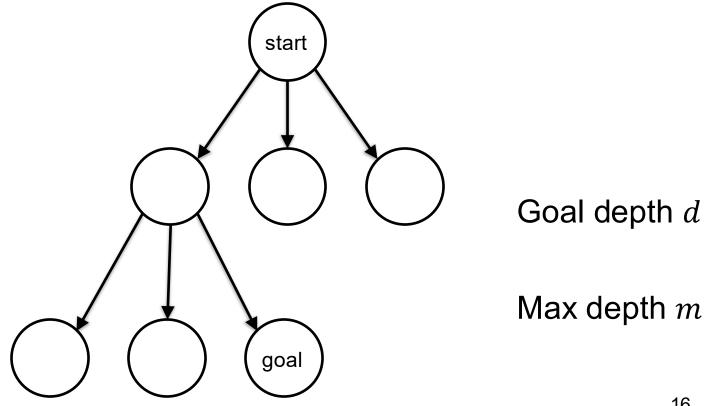
- In general there will be many generated, but un-expanded states at any given time
- One has to choose which one to expand next

#### Uninformed search on trees

- Uninformed means we only know:
  - The goal test
  - The *succs*() function
- But not which non-goal states are better: that would be informed search (next topic).
- For now, we also assume succs() graph is a tree.
  - Won't encounter repeated states.
  - We will relax it later.
- Many search strategies:
  - We will see BFS, UCS, DFS, IDS
- Differ by what un-expanded nodes to expand

#### Uninformed search on trees

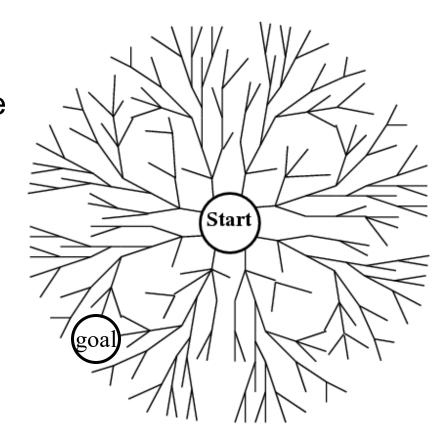
- Assume graph of state space is a tree
  - Only one path from start to each node
  - Could be infinite
- Assume every state has exactly b successor states



#### Uninformed search on trees

- Assume graph of state space is a tree
  - Only one path from start to each node
  - Could be infinite
- Assume every state has exactly b successor states

Alternative Picture:

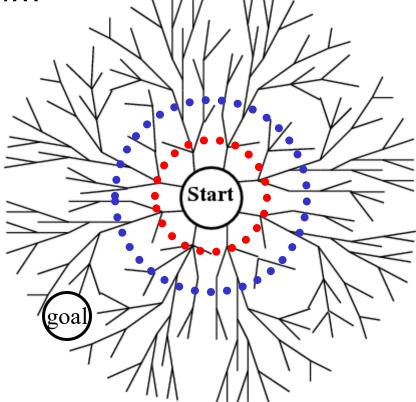


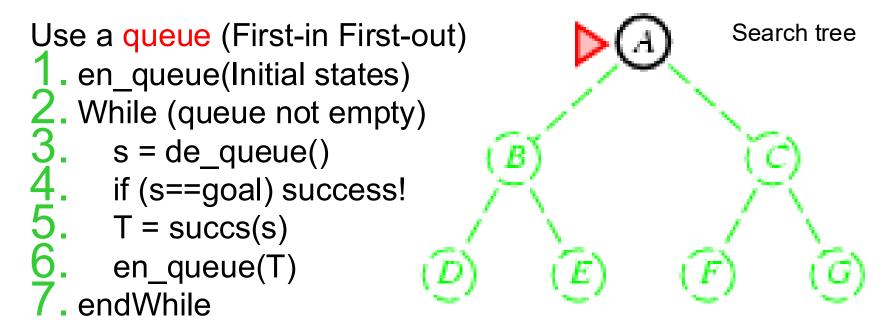
#### Expand the shallowest node first

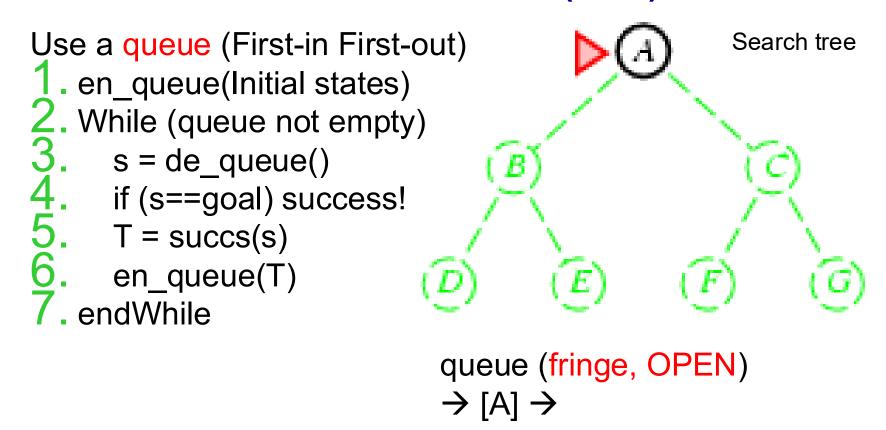
- Examine states one step away from the initial states
- Examine states two steps away from the initial states

and so on...

ripple







Use a queue (First-in First-out)

1. en\_queue(Initial states)

2. While (queue not empty)

3. s = de\_queue()

4. if (s==goal) success!

5. T = succs(s)

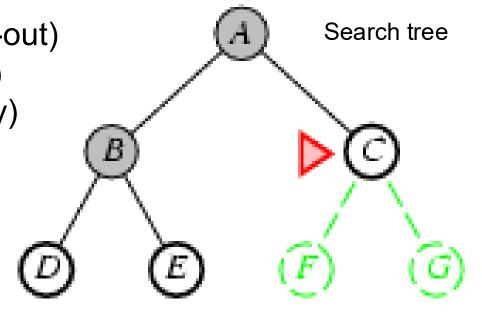
6. en\_queue(T)

7. endWhile

queue (fringe, OPEN)

→ [CB] → A

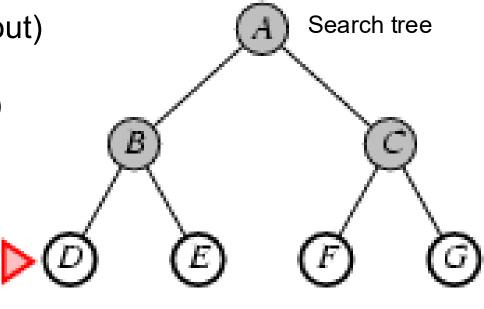
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queue (fringe, OPEN)

→ [EDC] → B

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queue (fringe, OPEN)

 $\Box$ [GFED]  $\rightarrow$  C

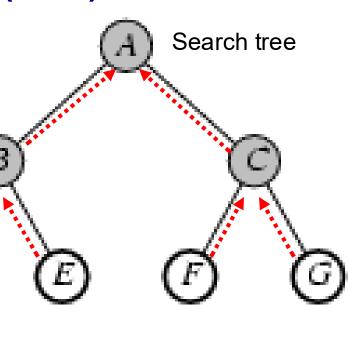
Initial state: **A** Goal state: **G** 

If G is a goal, we've seen it, but we don't stop!

Use a queue (First-in First-out)

- 1. en\_queue(Initial states)
- 2. While (queue not empty)
- 3.  $s = de_queue()$
- 4. if (s==goal) success!
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□[] <del>→</del>G

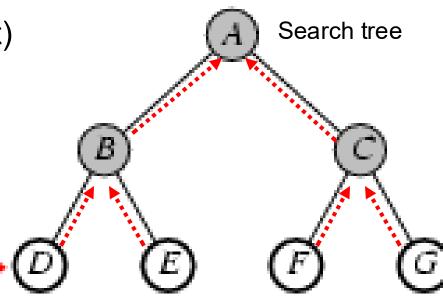
Looking foolish? Indeed. But let's be consistent...

... until much later we pop G.

Use a queue (First-in First-out)

- 1. en\_queue(Initial states)
- 2. While (queue not empty)
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queue

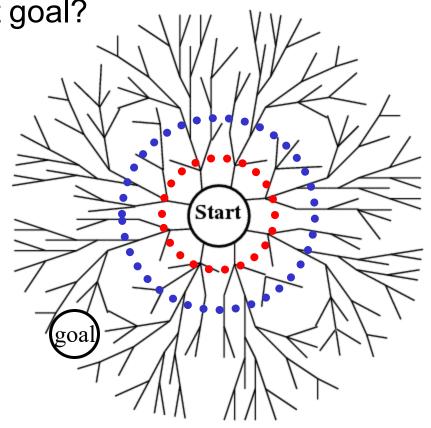
□[] **→**G

... until much later we pop G.

We need back pointers to recover the solution path.

#### **Performance of BFS**

- Assume:
  - the graph may be infinite.
  - Goal(s) exists and is only finite steps away.
- Will BFS find at least one goal?
- Will BFS find the least cost goal?
- Time complexity?
  - # states generated
  - Goal d edges away
  - Branching factor b
- Space complexity?
  - # states stored



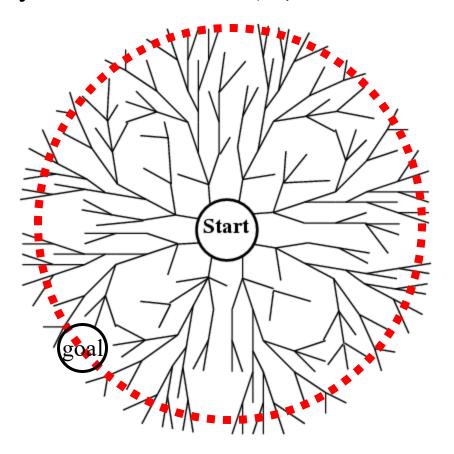
### **Performance of BFS**

Four measures of search algorithms:

- Completeness (not finding all goals): yes, BFS will find a goal.
- Optimality: yes if edges cost 1 (more generally positive non-decreasing in depth), no otherwise.
- Time complexity (worst case): goal is the last node at radius d.
  - Have to generate all nodes at radius d.
  - $b + b^2 + ... + b^d \sim O(b^d)$
- Space complexity (bad)
  - Back pointers for all generated nodes O(b<sup>d</sup>)
  - The queue / fringe (smaller, but still O(b<sup>d</sup>))

## What's in the fringe (queue) for BFS?

• Convince yourself this is  $O(b^d)$ 



## Performance of search algorithms on trees

b: branching factor (assume finite) d: goal depth

	Complete	optimal	time	space
Breadth-first search	Υ	Y, if <sup>1</sup>	O(b <sup>d</sup> )	O(b <sup>d</sup> )

1. Edge cost constant, or positive non-decreasing in depth

#### **Performance of BFS**

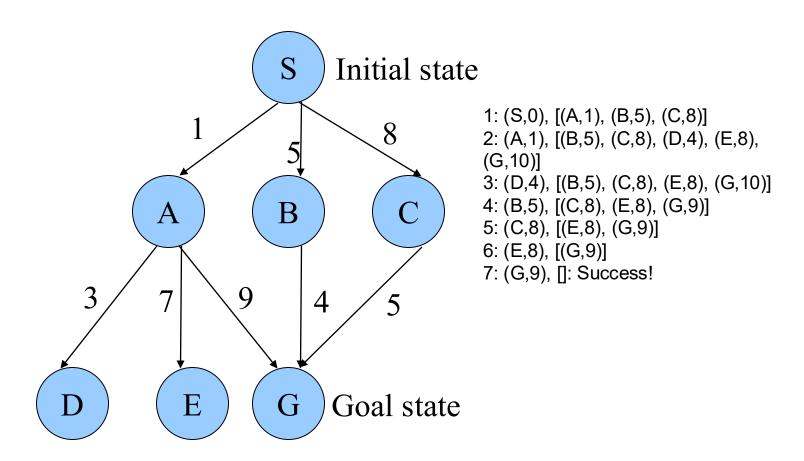
Four measures of search algorith

- Solution:
  Uniform-cost
  search
- Completeness (not finding all gind a goal.
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#### **Uniform-cost search**

- Find the least-cost goal
- Each node has a path cost from start (= sum of edge costs along the path).
- Expand the least cost node first.
- Use a priority queue instead of a normal queue
  - Always take out the least cost item

## **Example**

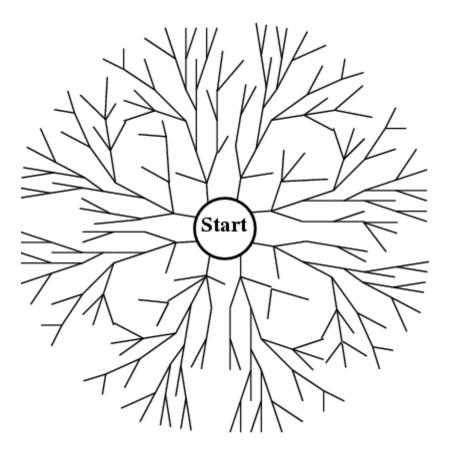


(All edges are directed, pointing downwards)

## **Uniform-cost search (UCS)**

- Complete and optimal (if edge costs  $\geq \epsilon > 0$ )
- Time and space: can be much worse than BFS
  - Let C\* be the cost of the least-cost goal

•  $O(b^{C*/\varepsilon})$ 





## Performance of search algorithms on trees

b: branching factor (assume finite) d: goal depth

	Complete	optimal	time	space
Breadth-first search	Y	Y, if <sup>1</sup>	O(b <sup>d</sup> )	O(b <sup>d</sup> )
Uniform-cost search <sup>2</sup>	Y	Y	$O(b^{C^*/\epsilon})$	O(b <sup>C*/ε</sup> )

- 1. edge cost constant, or positive non-decreasing in depth
- 2. edge costs  $\geq \epsilon > 0$ . C\* is the best goal path cost.

#### **Performance of BFS**

Four measures of search algorith

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  Uniform-cost Swill search
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- Time complexit at radius **d**. Solution: Depth-first search

  - $b + b^2 + ... + b^d \sim O(b^d)$
- Space complexity (bad)
  - Back pointers for all generated nodes O(b<sup>d</sup>)
  - The queue / fringe (smaller, but still O(b<sup>d</sup>))

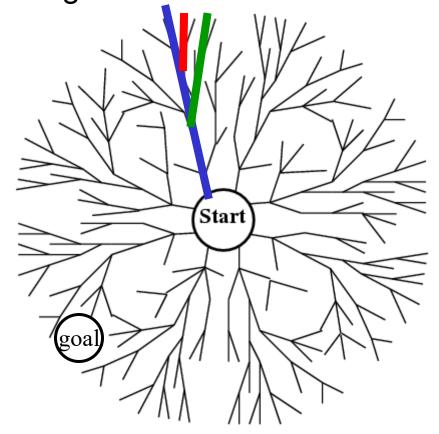
## **Depth-first search**

Expand the deepest node first

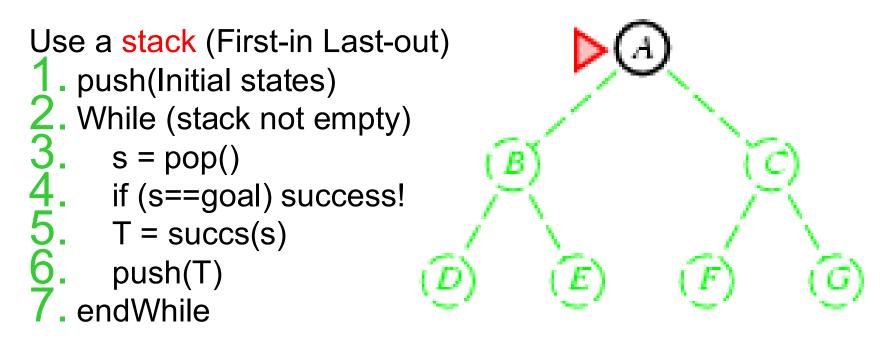
- 1. Select a direction, go deep to the end
- 2. Slightly change the end

3. Slightly change the end some more...

fan



## **Depth-first search (DFS)**

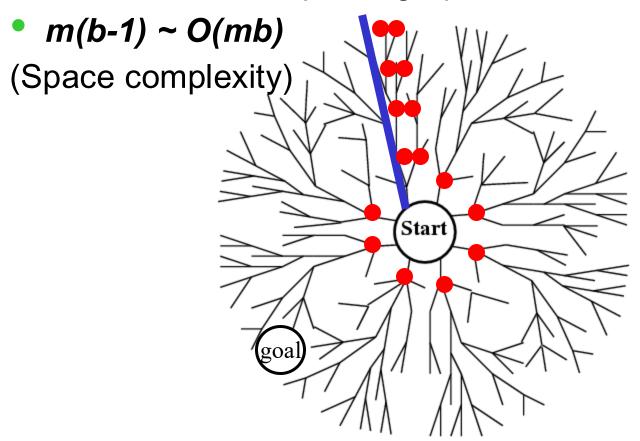


stack (fringe)

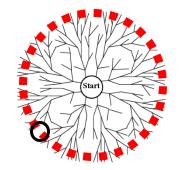
- 1. A, [B, C] 2. B, [D, E, C]
- 3. D, [E, C]
- 4. E, [C]
- 5. C, [F, G]
- 6. F, [G]
- 7. G

## What's in the fringe for DFS?

m = maximum depth of graph from start



c.f. BFS  $O(b^d)$ 



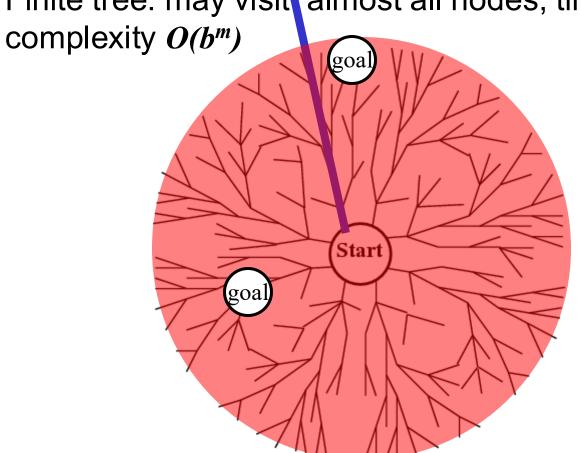
- "backtracking search" even less space
  - generate siblings (if applicable)

## What's wrong with DFS?

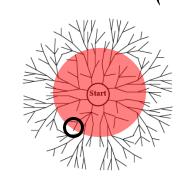
Infinite tree: may not find goal (incomplete)

May not be optimal

Finite tree: may visit almost all nodes, time



c.f. BFS  $O(b^d)$ 



## Performance of search algorithms on trees

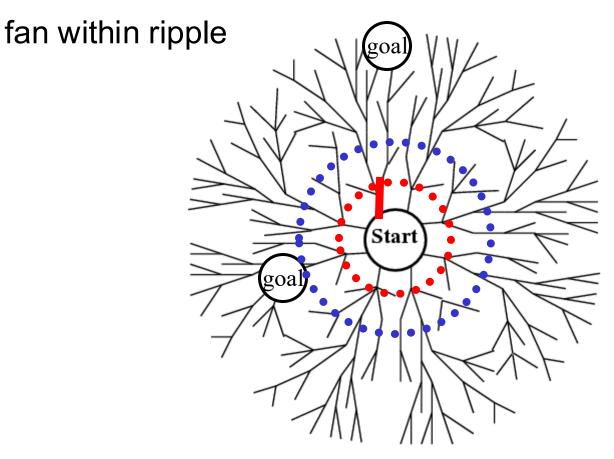
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Uniform-cost search <sup>2</sup>	Υ	Y	$O(b^{C^*/\epsilon})$	O(b <sup>C*/ε</sup> )
Depth-first search	N	N	O(b <sup>m</sup> )	O(bm)

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#### How about this?

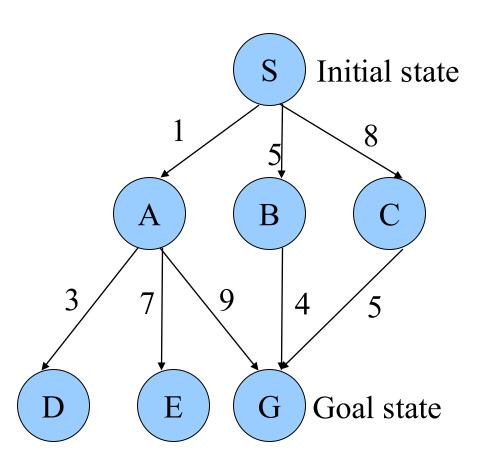
- 1. DFS, but stop if path length > 1.
- 2. If goal not found, repeat DFS, stop if path length > 2.
- 3. And so on...



## Iterative deepening

- Search proceeds like BFS, but fringe is like DFS
  - Complete, optimal like BFS
  - Small space complexity like DFS
  - Time complexity like BFS
- Preferred uninformed search method

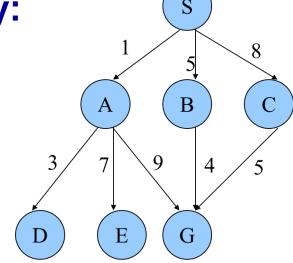
## **Example**



(All edges are directed, pointing downwards)

Nodes expanded by:

Breadth-First Search: S A B C D E G
 Solution found: S A G



Uniform-Cost Search: S A D B C E G
 Solution found: S B G (This is the only uninformed search that worries about costs.)

Depth-First Search: S A D E G
 Solution found: S A G

Iterative-Deepening Search: S A B C S A D E G
 Solution found: S A G

## Performance of search algorithms on trees

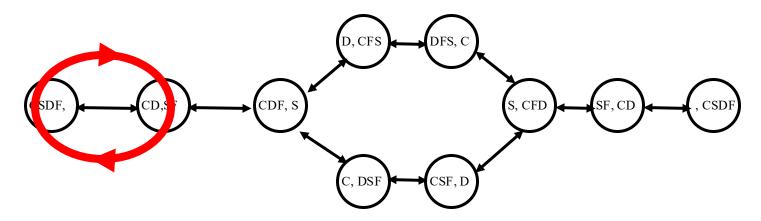
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Uniform-cost search <sup>2</sup>	Y	Y	O(b <sup>C*/ε</sup> )	O(b <sup>C*/ε</sup> )
Depth-first search	N	N	O(b <sup>m</sup> )	O(bm)
Iterative deepening	Υ	Y, if <sup>1</sup>	O(b <sup>d</sup> )	O(bd)

- 1. edge cost constant, or positive non-decreasing in depth
- 2. edge costs  $\geq \varepsilon > 0$ . C\* is the best goal path cost.

## If state space graph is not a tree

• The problem: repeated states



- Ignore the danger of repeated states: wasteful (BFS) or impossible (DFS). Can you see why?
- How to prevent it?

## If state space graph is not a tree

- We have to remember already-expanded states (CLOSED).
- When we take out a state from the fringe (OPEN), check whether it is in CLOSED (already expanded).
  - If yes, throw it away.
  - If no, expand it (add successors to OPEN), and move it to CLOSED.

## What you should know

- Problem solving as search: state, successors, goal test
- Uninformed search
  - Breadth-first search
    - Uniform-cost search
  - Depth-first search
  - Iterative deepening







- Can you unify them using the same algorithm, with different priority functions?
- Performance measures
  - Completeness, optimality, time complexity, space complexity