



# **CS540 Intro to AI Uninformed Search**

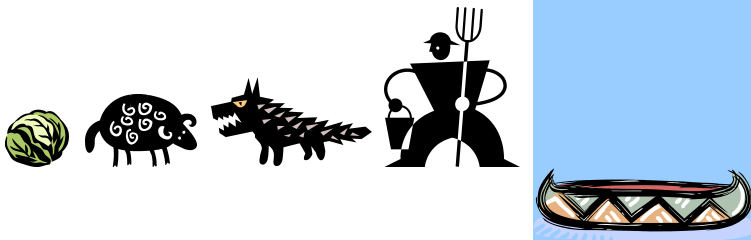
**Sharon Li**

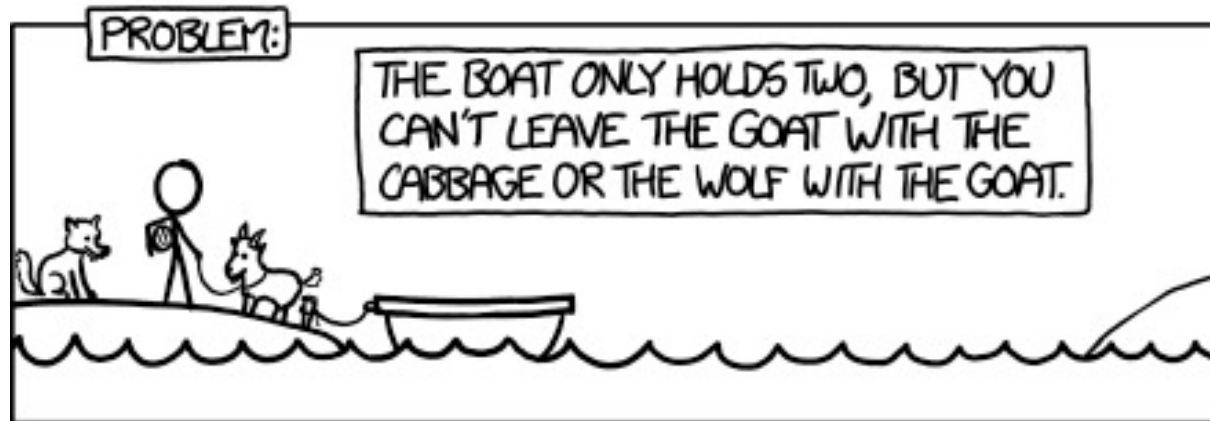
**University of Wisconsin-Madison**

Slides created by Xiaojin Zhu (UW-Madison),  
lightly edited by Anthony Gitter

slide 1

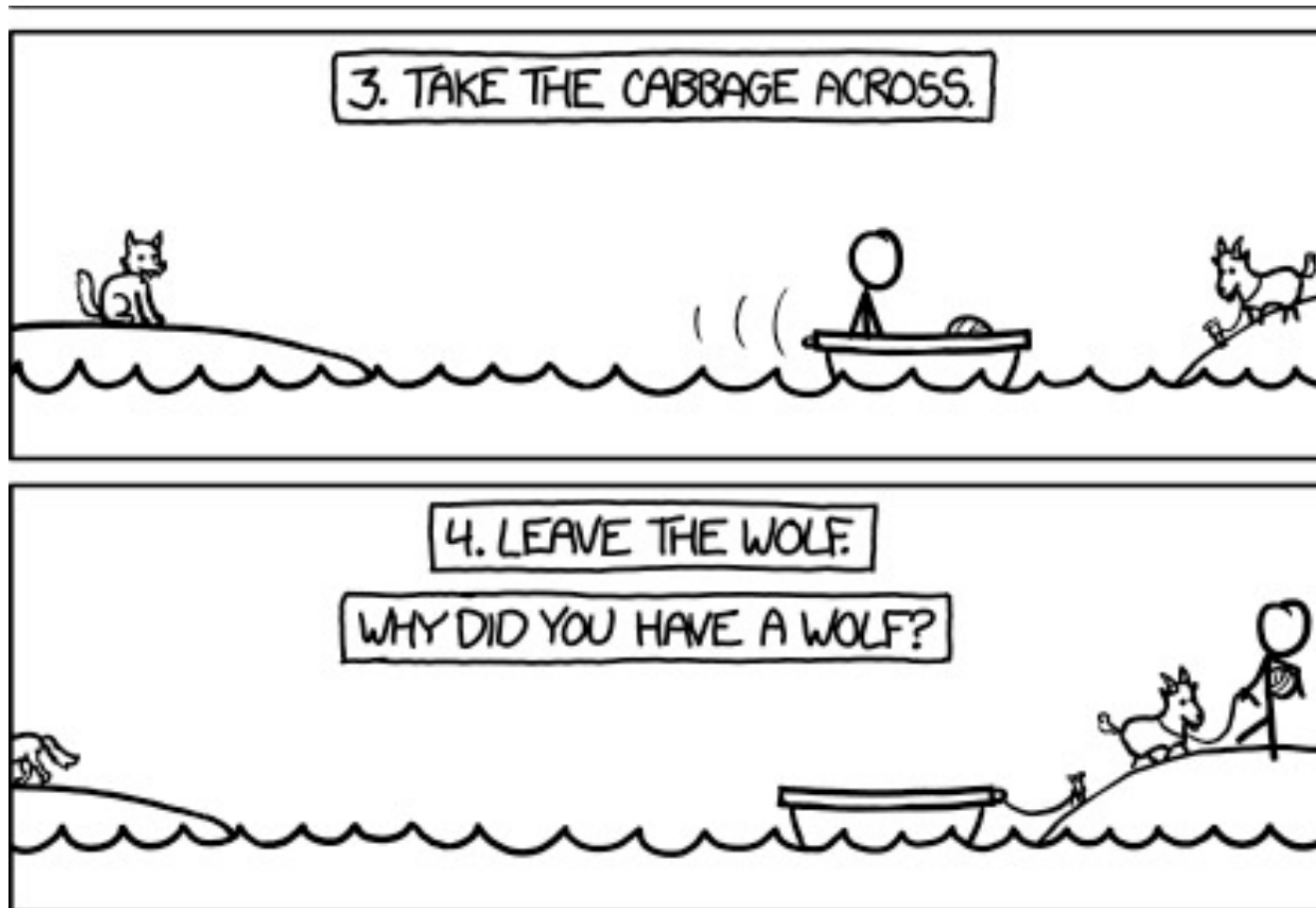
**Many AI problems can be formulated as search.**





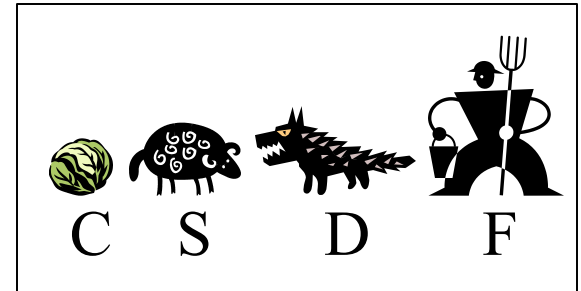
<http://xkcd.com/1134/>





# The search problem

- State space  $\mathcal{S}$  : all valid configurations
- Initial state  $I = \{(CSDF, )\} \subseteq \mathcal{S}$
- Goal state  $G = \{(:, CSDF)\} \subseteq \mathcal{S}$
- Successor function  $\text{succs}(s) \subseteq \mathcal{S}$  : states reachable in one step from state  $s$ 
  - $\text{succs}((CSDF, )) = \{(CD, SF)\}$
  - $\text{succs}((CDF, S)) = \{(CD, FS), (D, CFS), (C, DFS)\}$
- Cost( $s, s'$ )=1 for all steps. (weighted later)
- The search problem: find a solution path from a state in  $I$  to a state in  $G$ .
  - Optionally minimize the cost of the solution.



# Search examples

- 8-puzzle

7	2	4
5		6
8	3	1

Start State

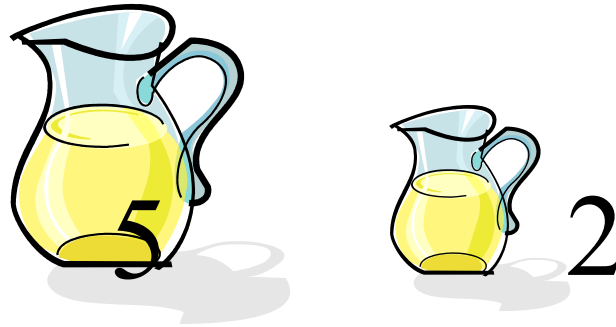
	1	2
3	4	5
6	7	8

Goal State

- States = 3x3 array configurations
- action = up to 4 kinds of movement
- Cost = 1 for each move

# Search examples

- Water jugs: how to get 1?



State =  $(x,y)$ , where  $x$  = number of gallons of water in the 5-gallon jug and  $y$  is gallons in the 2-gallon jug

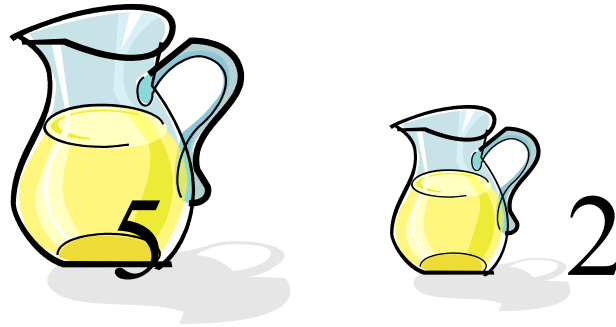
Initial State =  $(5,0)$

Goal State =  $(*,1)$ , where  $*$  means any amount



# Search examples

- Water jugs: how to get 1?



State =  $(x,y)$ , where  $x$  = number of gallons of water in the 5-gallon jug and  $y$  is gallons in the 2-gallon jug

Initial State =  $(5,0)$

Goal State =  $(*,1)$ , where  $*$  means any amount

Operators

$(x,y) \rightarrow (0,y)$  ; empty 5-gal jug

$(x,y) \rightarrow (x,0)$  ; empty 2-gal jug

$(x,2)$  and  $x \leq 3 \rightarrow (x+2,0)$  ; pour 2-gal into 5-gal

$(x,0)$  and  $x \geq 2 \rightarrow (x-2,2)$  ; pour 5-gal into 2-gal

$(1,0) \rightarrow (0,1)$  ; empty 5-gal into 2-gal

# Search examples

# Search examples

- Route finding (State? Successors? Cost weighted)

Google Maps - from: 1210 W Dayton St, Madison, WI 53706 to: State St, Madison, WI 53703 - Mozilla Firefox

File Edit View Go Bookmarks Tools Help

Google Maps Local Search Directions

1210 W Dayton St, Madison, WI 53706 Start address

State St, Madison, WI 53703 End address

Search Help

Maps

Map Satellite

Print Email Link to this page

Start address: 1210 W Dayton St  
Madison, WI 53706

End address: State St  
Madison, WI 53703

Distance: 1.2 mi (about 2 mins)

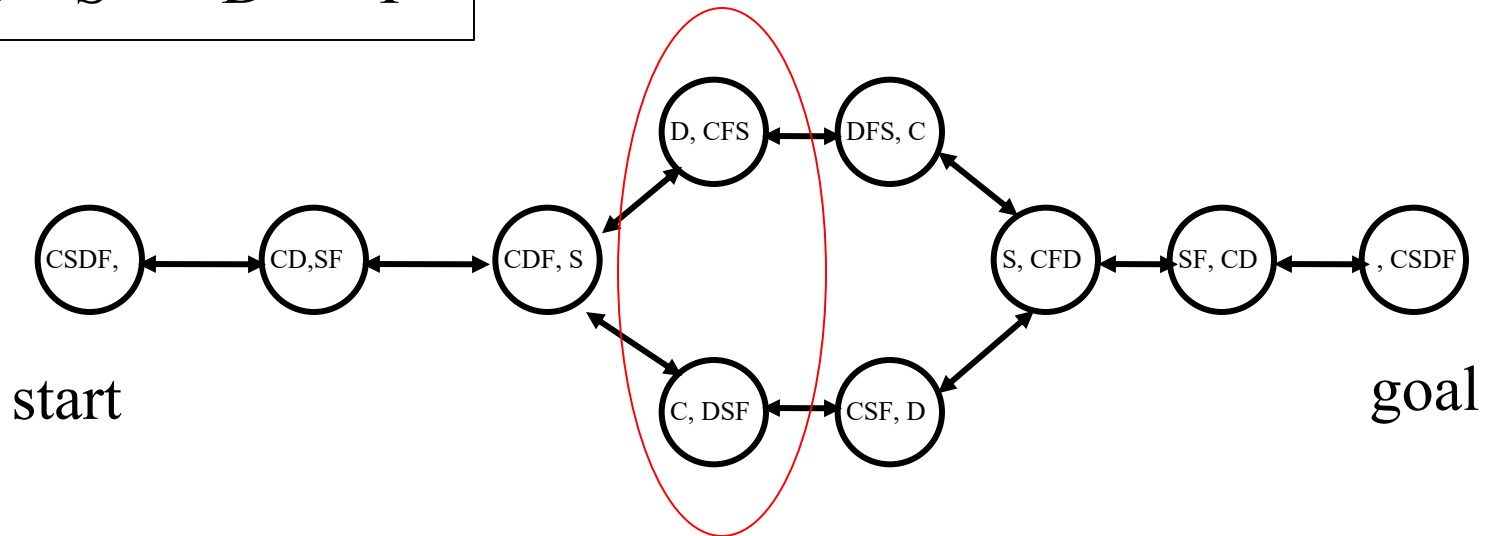
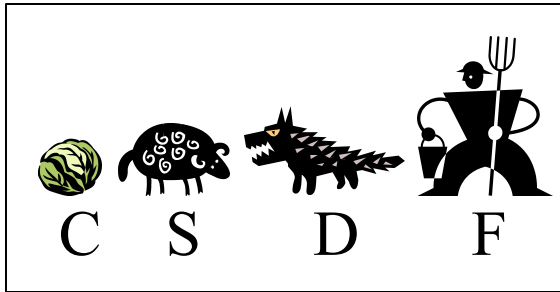
Reverse directions

1. Head east from W Dayton St - go 0.5 mi
2. Turn left at N Frances St - go 0.2 mi
3. Turn right at W Gilman St - go 0.3 mi
4. Turn right at N Henry St - go 0.1 mi
5. Turn right at W Gorham St - go 0.1 mi

These directions are for planning purposes only. You may find that construction projects, traffic, or other events may cause road conditions to differ from the map results.

Map data ©2005 NAVTEG™, Tele Atlas

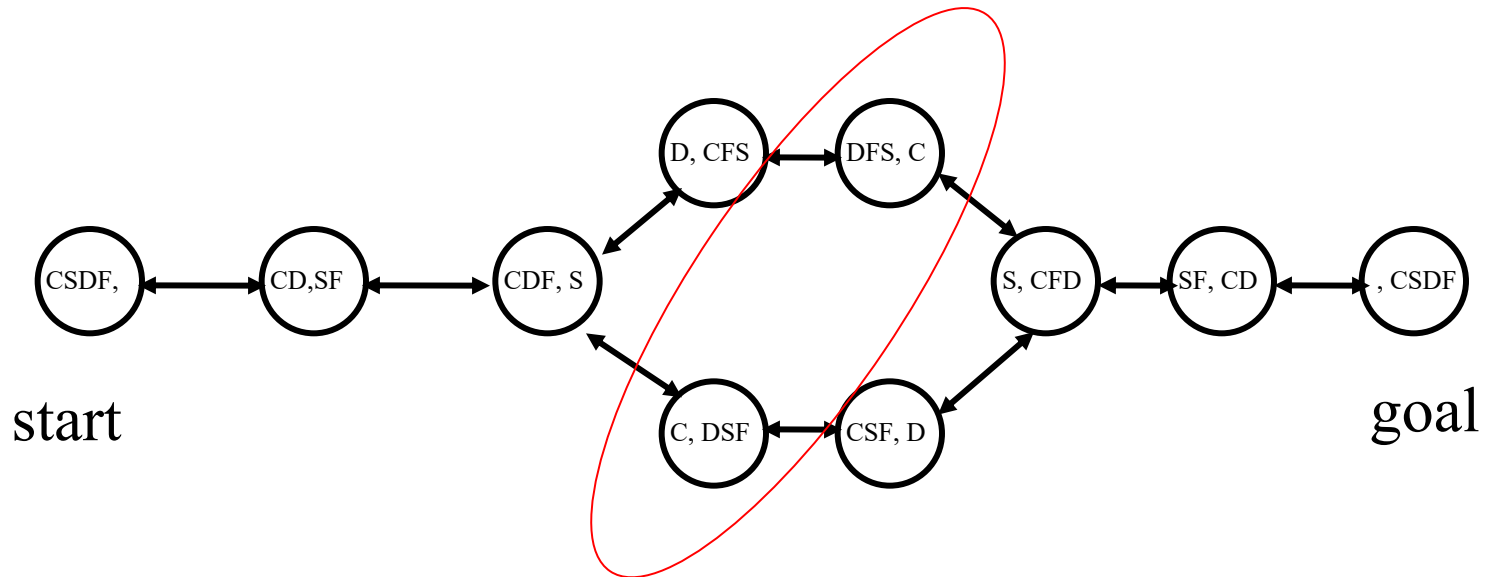
# A directed graph in state space



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

# Different search strategies

- The generated, but not yet expanded states form the **fringe** (**OPEN**).
- The essential difference is **which one to expand first**.
- Deep or shallow?



# Uninformed search on trees

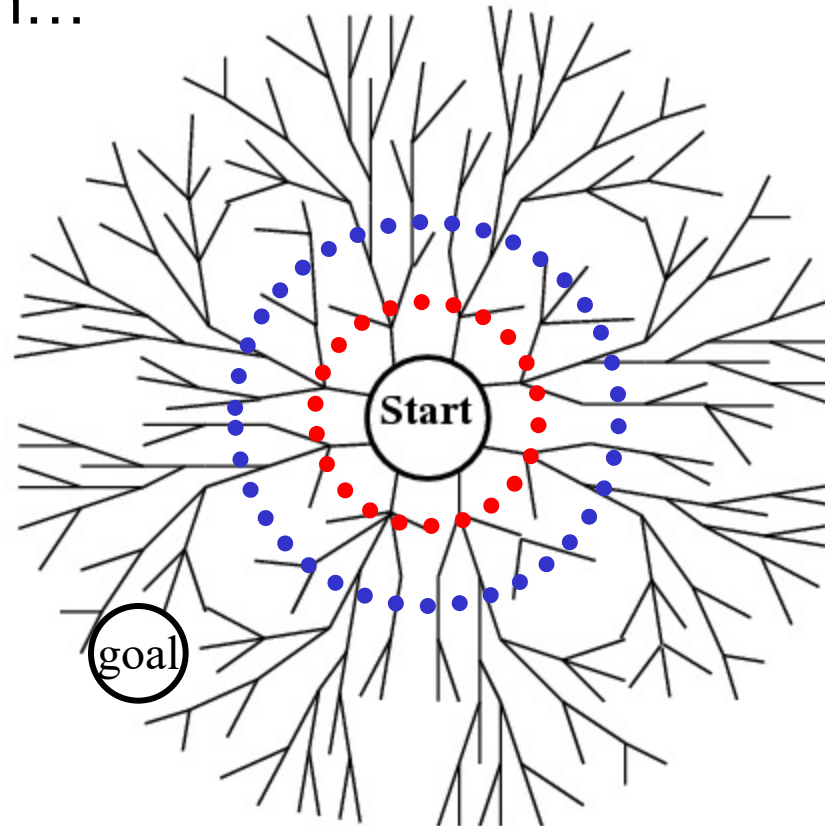
- **Uninformed** means we only know:
  - The goal test
  - The *succs()* function
- But **not** which non-goal states are better: that would be informed search (next topic).
- For now, we also assume *succs()* graph is **a tree**.
  - Won't encounter repeated states.
  - We will relax it later.
- Search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

# Breadth-first search (BFS)

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



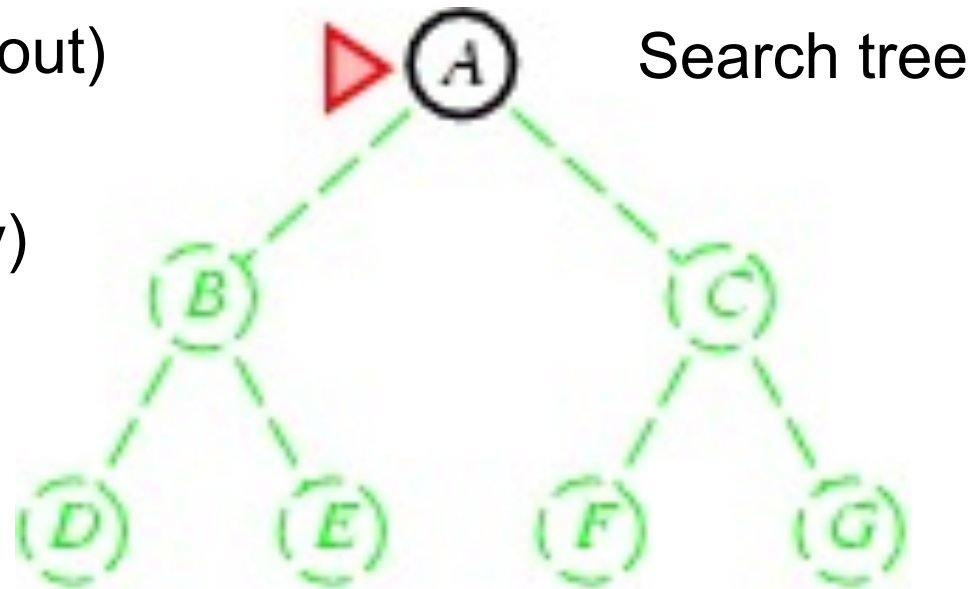
# Breadth-first search (BFS)

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

Initial state: **A**

Goal state: **G**





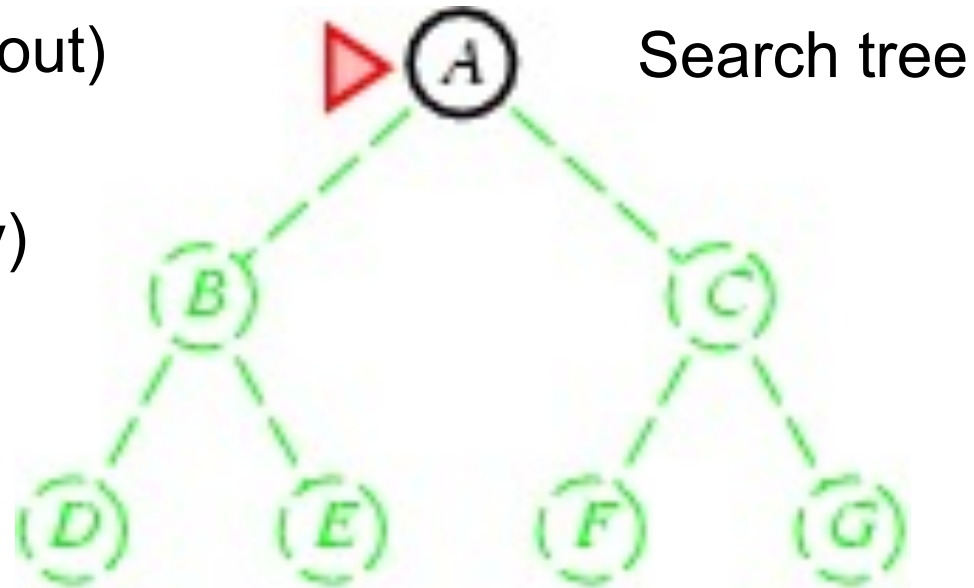
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Initial state: **A**

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queue (**fringe**, **OPEN**)  
→ [A] →

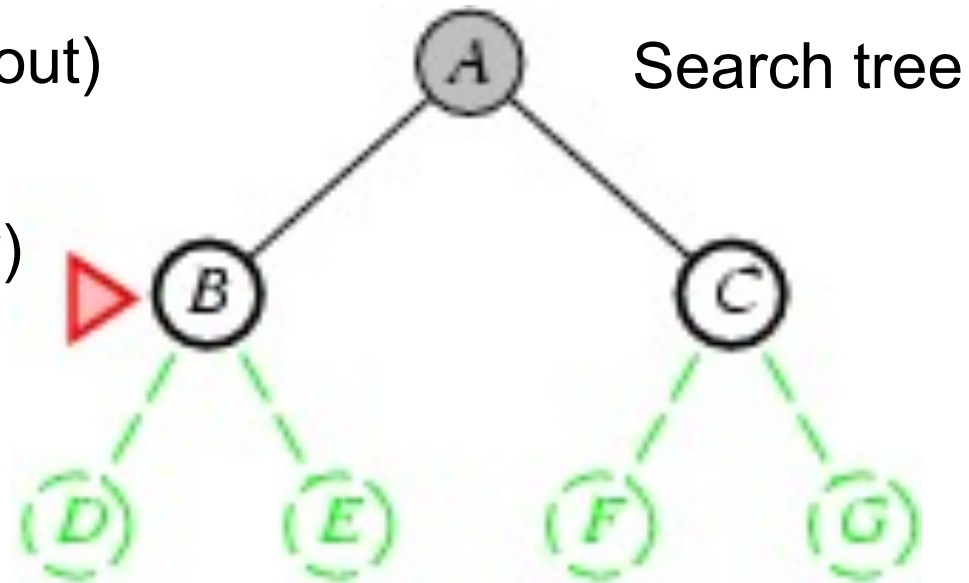
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Initial state: **A**

Goal state: **G**



queue (**fringe**, **OPEN**)  
→ [CB] → A

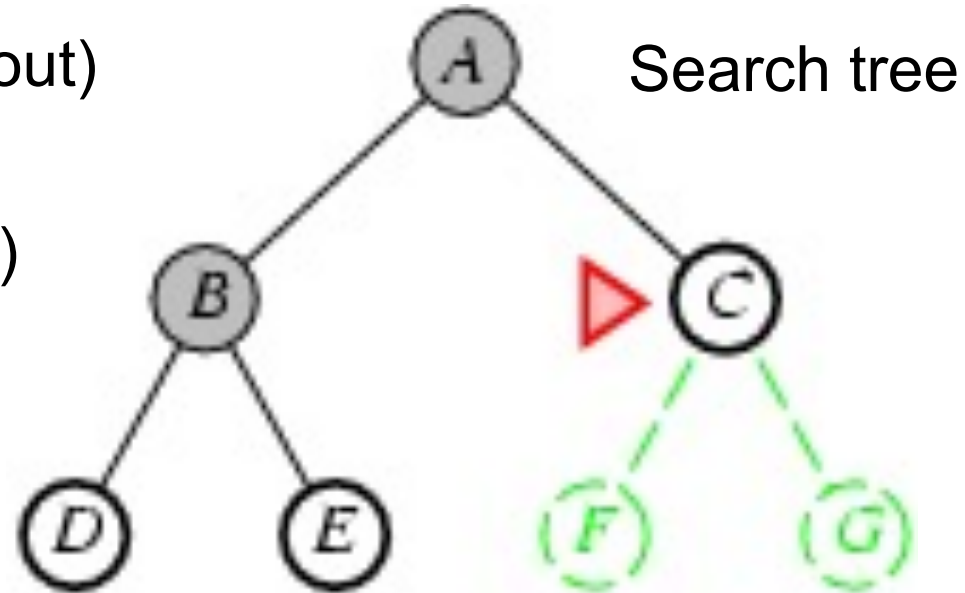
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6.   en\_queue(T)
7. endwhile

Initial state: **A**

Goal state: **G**



queue (**fringe**, **OPEN**)  
→ [EDC] → B

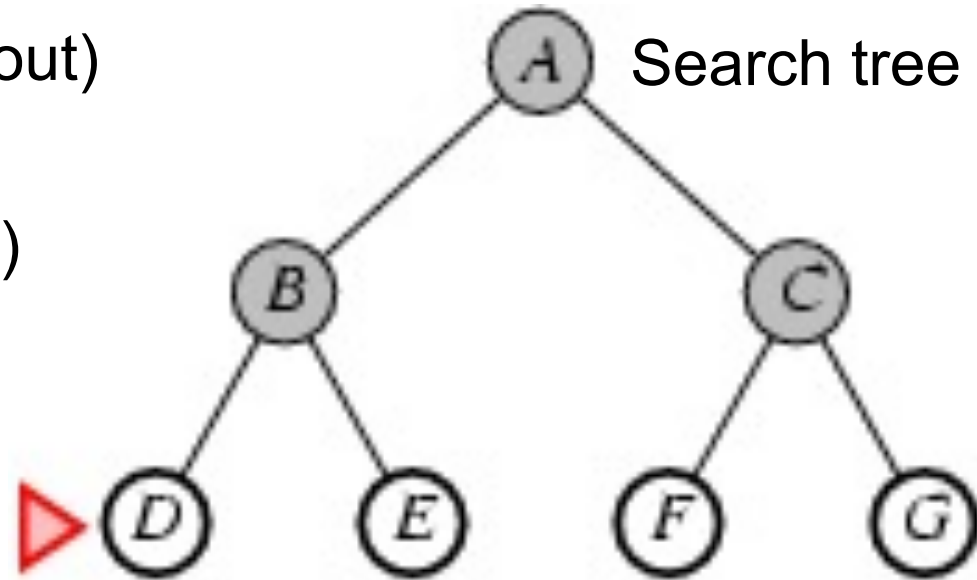
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Initial state: **A**

Goal state: **G**



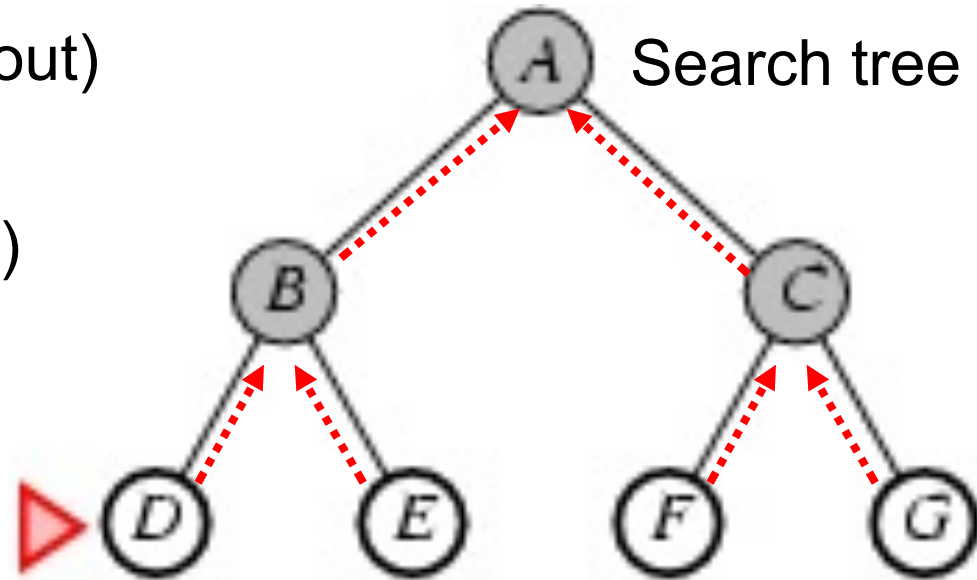
queue (**fringe, OPEN**)  
→[GFED] → C

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

# Breadth-first search (BFS)

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  
→ [] → G

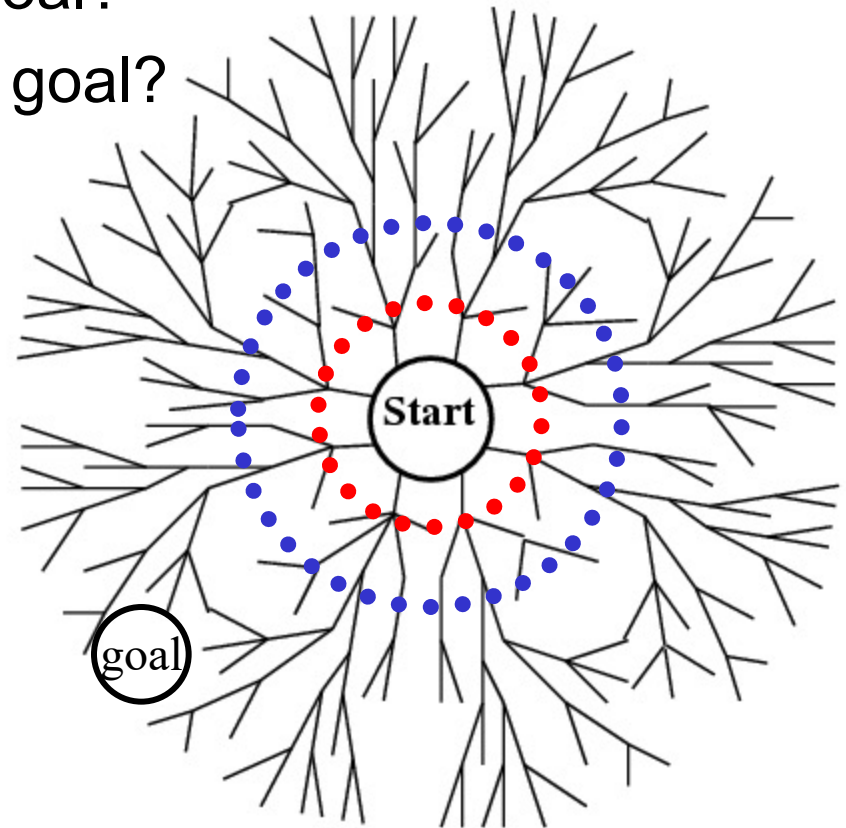
... until much later we pop G.

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

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



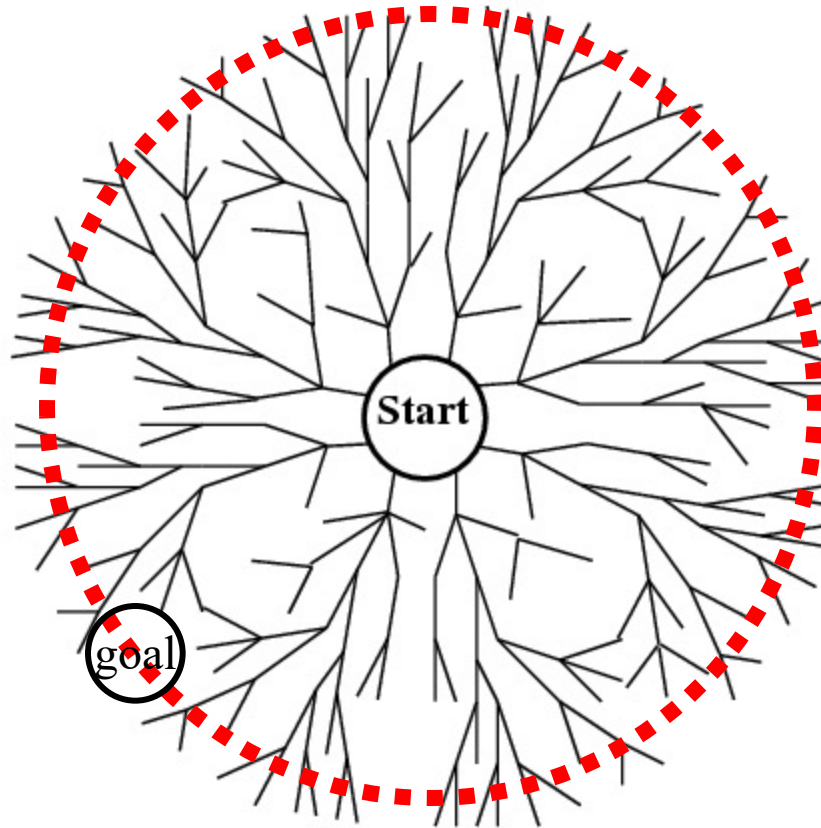
# Performance of BFS

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

# 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	Y	Y, if <sup>1</sup>	$O(b^d)$	$O(b^d)$

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

# Performance of BFS

Four measures of search algorithms:

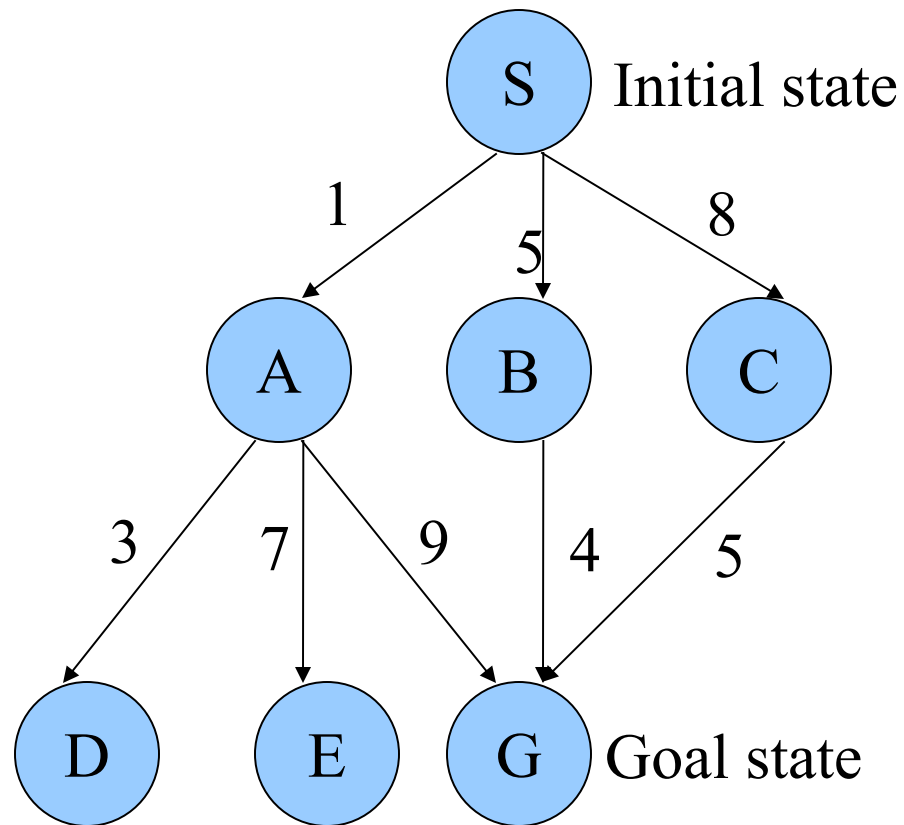
- **Completeness** (not finding all goals): find a goal.
- **Optimality**: yes if edges cost 1 (more generally positive non-decreasing with depth), **no otherwise**.
- **Time** complexity (worst case): goal is the last node at radius  $d$ .
  - Have to generate all nodes at radius  $d$ .
  - $b + b^2 + \dots + b^d \sim O(b^d)$
- **Space** complexity (bad, Figure 3.11)
  - Back points for all generated nodes  $O(b^d)$
  - The queue (smaller, but still  $O(b^d)$ )

**Solution:**  
**Uniform-cost**  
**search**

# 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

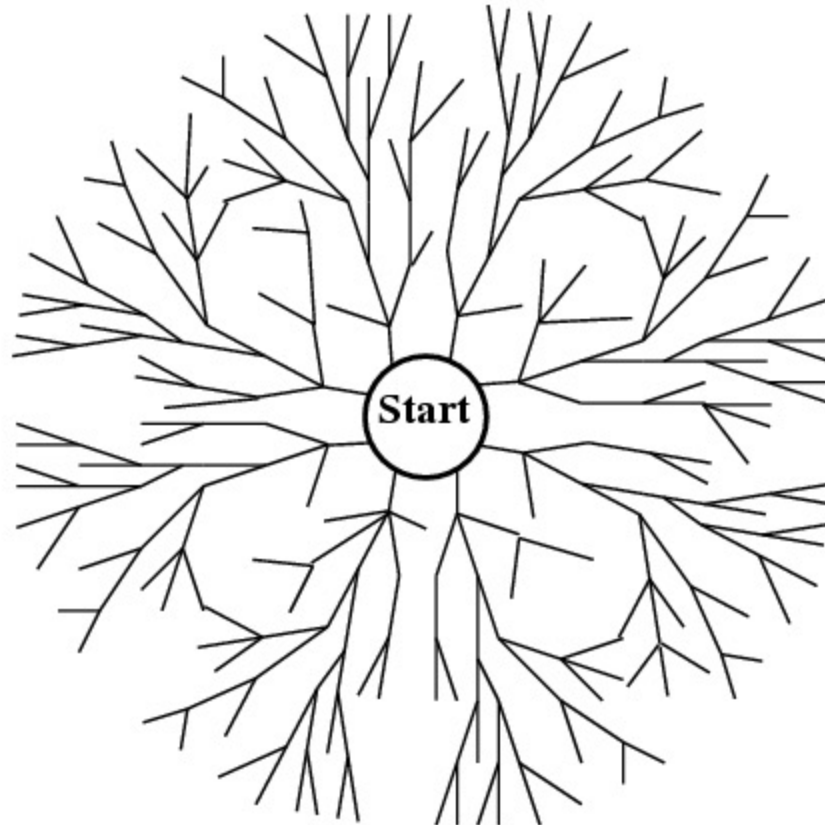
# Example



(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
  - $O(b^{C^*/\epsilon})$



goal

# 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>	$O(b^d)$	$O(b^d)$
Uniform-cost search <sup>2</sup>	Y	Y	$O(b^{C^*/\epsilon})$	$O(b^{C^*/\epsilon})$

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

# General State-Space Search Algorithm

```
function general-search(problem, QUEUEING-FUNCTION)
;; problem describes the start state, operators, goal test, and
;; operator costs
;; queueing-function is a comparator function that ranks two states
;; general-search returns either a goal node or "failure"

nodes = MAKE-QUEUE(MAKE-NODE(problem.INITIAL-STATE))
loop
  if EMPTY(nodes) then return "failure"
  node = REMOVE-FRONT(nodes)
  if problem.GOAL-TEST(node.STATE) succeeds then return node
  nodes = QUEUEING-FUNCTION(nodes, EXPAND(node,
                                         problem.OPERATORS))
;; succ(s)=EXPAND(s, OPERATORS)
;; Note: The goal test is NOT done when nodes are generated
;; Note: This algorithm does not detect loops
end
```

# Recall the bad space complexity of BFS

Four measures of search algorithms:

- **Completeness** (not finding all goals): find a goal.
- **Optimality**: yes if edges cost 1 (more generally positive non-decreasing with depth), **no otherwise**.
- **Time** complexity (bad, Figure 3.10): goal is the last node at radius  $d$ .
  - Have to generate all nodes at radius  $d$ .
  - $b + b^2 + \dots + b^d \sim O(b^d)$
- **Space** complexity (bad, Figure 3.11)
  - Back points for all generated nodes  $O(b^d)$
  - The queue (smaller, but still  $O(b^d)$ )

**Solution:**  
**Uniform-cost**  
**search**

**Solution:**  
**Depth-first**  
**search**

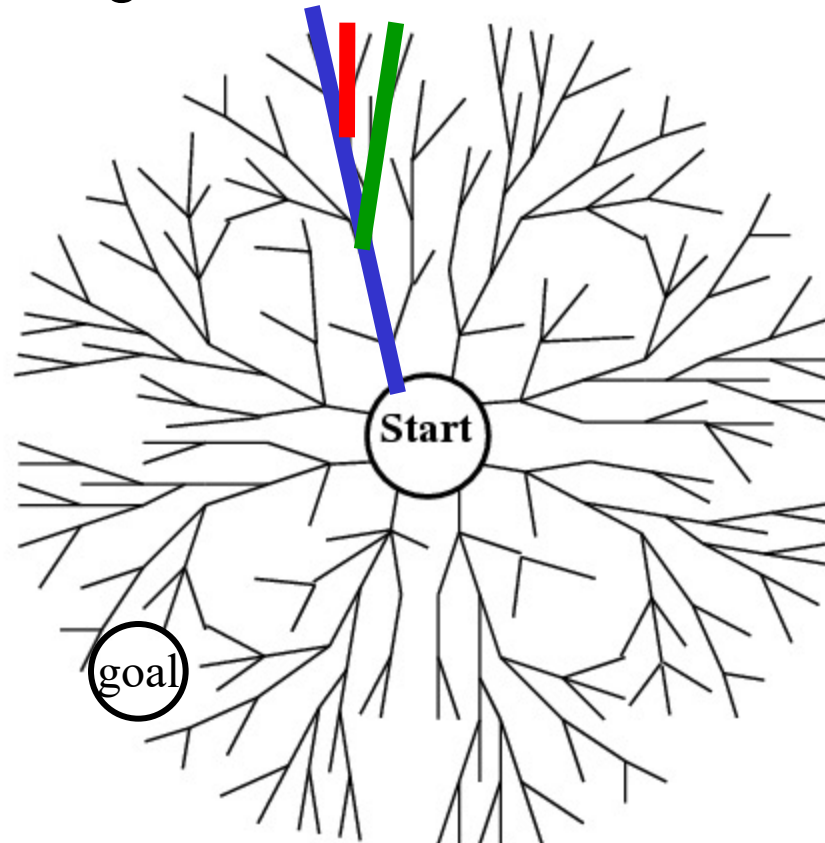


# Depth-first search

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... —————

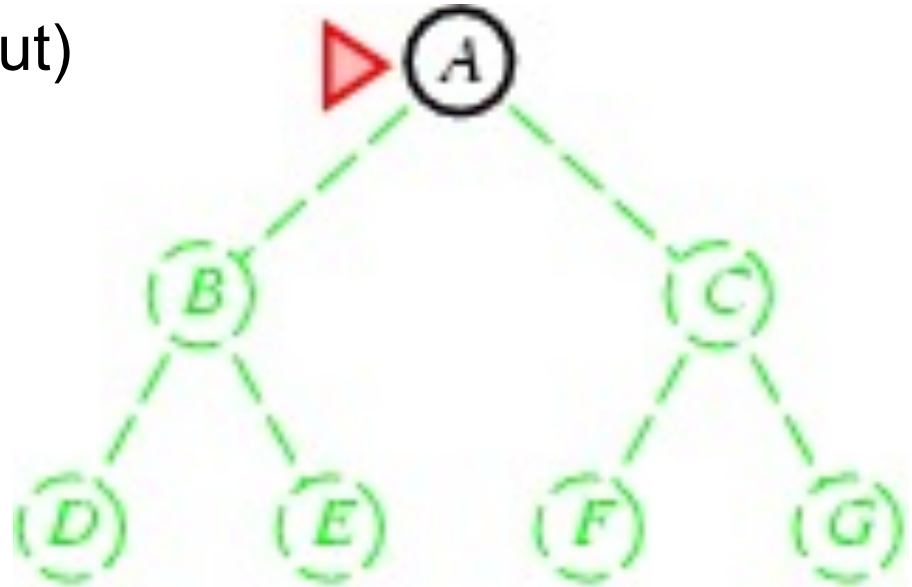
fan



# Depth-first search (DFS)

Use a **stack** (First-in Last-out)

1. push(Initial states)
2. While (stack not empty)
3.   s = pop()
4.   if (s==goal) success!
5.   T = succs(s)
6.   push(T)
7. endwhile

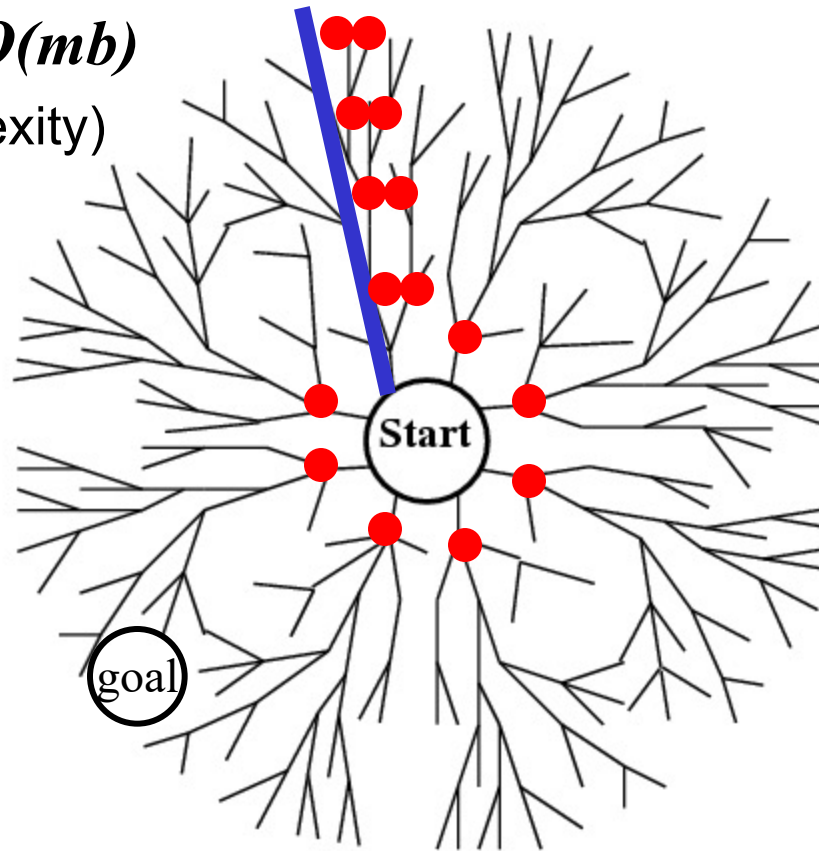


stack (**fringe**)

[] ⇔

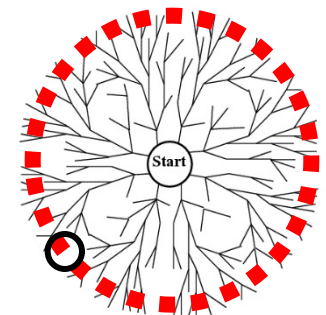
# What's in the fringe for DFS?

- $m$  = maximum depth of graph from start
- $m(b-1) \sim O(mb)$   
(Space complexity)



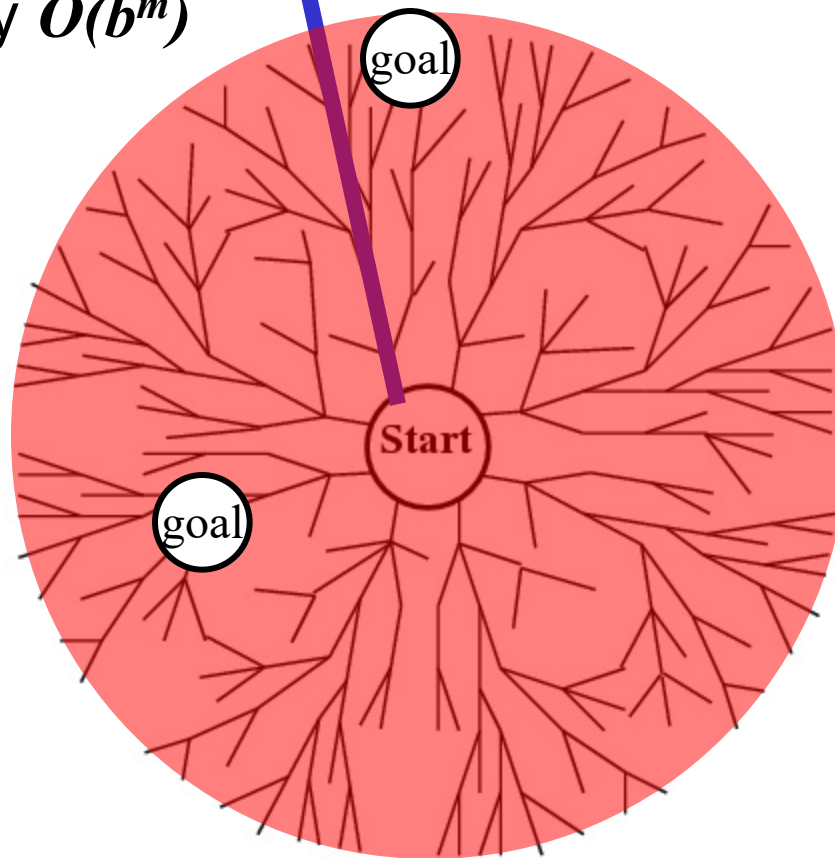
- “backtracking search” even less space
  - generate siblings (if applicable)

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

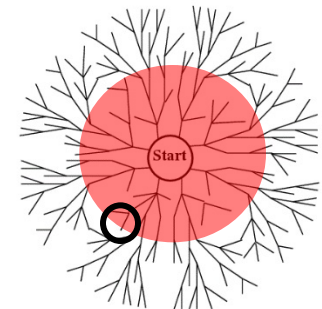


# What's wrong with DFS?

- Infinite tree: may not find goal (incomplete)
- May not be optimal
- Finite tree: may visit almost all nodes, time complexity  $O(b^m)$



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



# Performance of search algorithms on trees

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

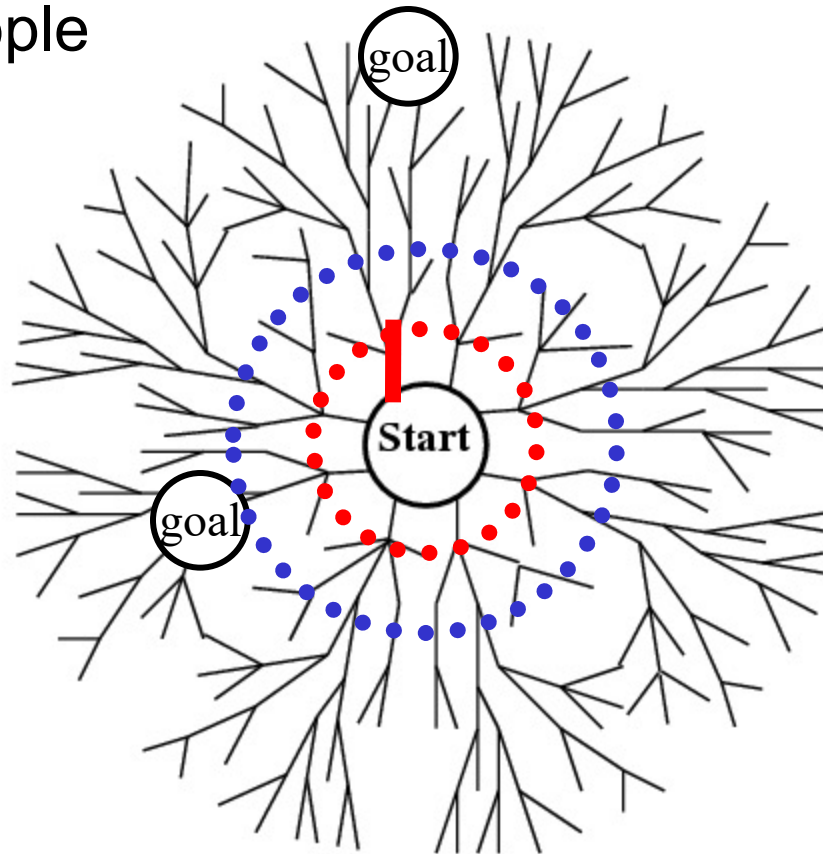
	Complete	optimal	time	space
Breadth-first search	Y	Y, if <sup>1</sup>	$O(b^d)$	$O(b^d)$
Uniform-cost search <sup>2</sup>	Y	Y	$O(b^{C^*/\epsilon})$	$O(b^{C^*/\epsilon})$
Depth-first search	N	N	$O(b^m)$	$O(bm)$

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

# 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...

fan within ripple



# Iterative deepening

- Search proceeds like BFS, but fringe is like DFS
  - Complete, optimal like BFS
  - Small space complexity like DFS
- A huge waste?
  - Each deepening repeats DFS from the beginning
  - No!  $db + (d-1)b^2 + (d-2)b^3 + \dots + b^d \sim O(b^d)$
  - Time complexity like BFS
- Preferred uninformed search method

# Performance of search algorithms on trees

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

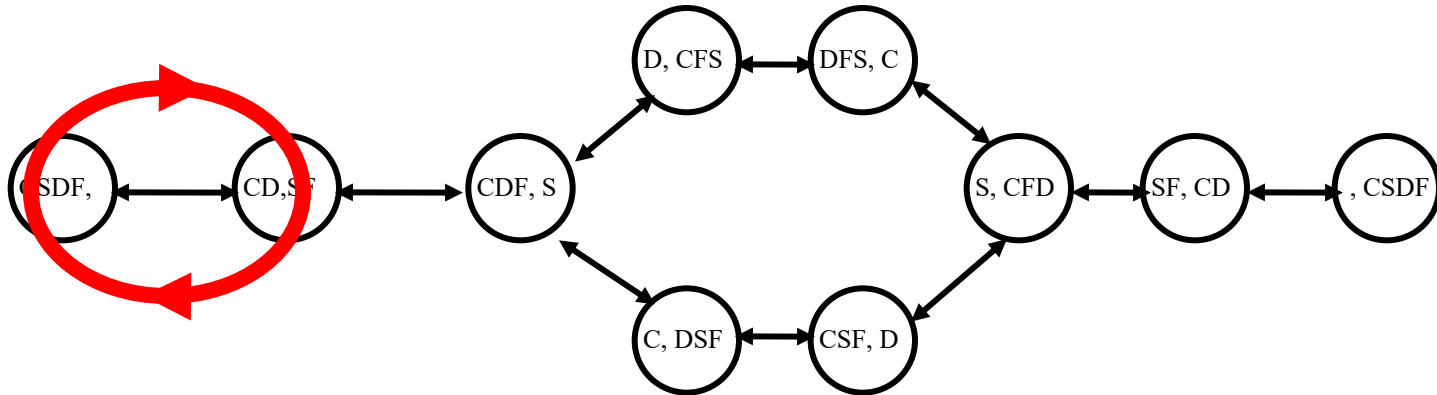
	Complete	optimal	time	space
Breadth-first search	Y	Y, if <sup>1</sup>	$O(b^d)$	$O(b^d)$
Uniform-cost search <sup>2</sup>	Y	Y	$O(b^{C^*/\epsilon})$	$O(b^{C^*/\epsilon})$
Depth-first search	N	N	$O(b^m)$	$O(bm)$
Iterative deepening	Y	Y, if <sup>1</sup>	$O(b^d)$	$O(bd)$

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



# If state space graph is not a tree

- The problem: repeated states

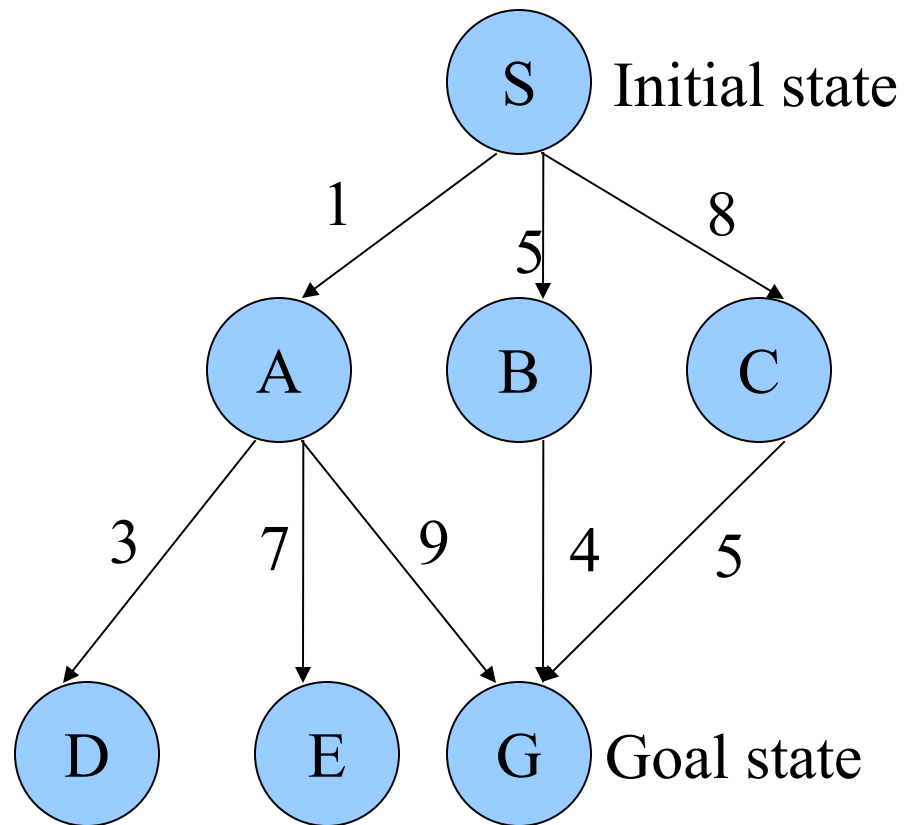


- Ignore the danger of repeated states: wasteful (BFS) or impossible (DFS). Can you see why?
- How to prevent it?

## 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.

# 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

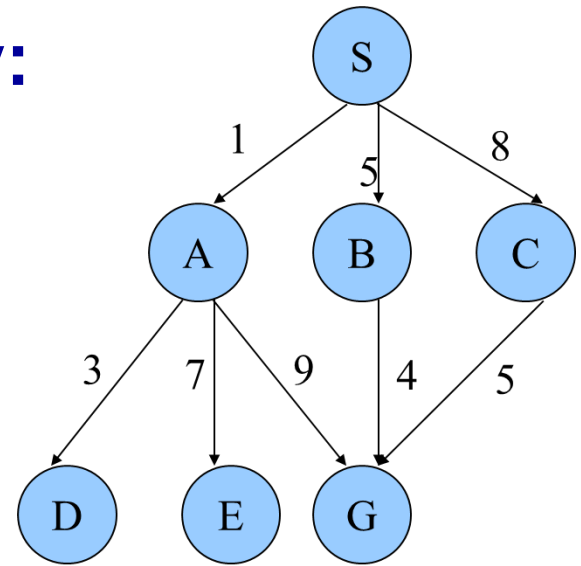
Solution found: S B G (This is the only uninformed search that worries about costs.)

- 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



# Depth-First Search

expanded node	nodes list
----	-----
	{ S }
S	{ A B C }
A	{ D E G B C }
D	{ E G B C }
E	{ G B C }
G	{ B C }


Solution path found is S A G <-- this G has cost 10  
Number of nodes expanded (including goal node) = 5

# Uniform-Cost Search

expanded node ----	nodes list -----
	{ S }
S	{ A(1) B(5) C(8) }
A	{ D(4) B(5) C(8) E(8) G(10) } (note, we don't return G)
D	{ B(5) C(8) E(8) G(10) }
B	{ C(8) E(8) G(9) G(10) }
C	{ E(8) G(9) G(10) G(13) }
E	{ G(9) G(10) G(13) }
G	{ }

Solution path found is S B G <-- this G has cost 9, not 10  
Number of nodes expanded (including goal node) = 7

# 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 ★
- 
- The icons are arranged horizontally. From left to right: a red five-pointed star, a small wooden canoe with orange and green patterns, a green cabbage, a black pig with white spots, a black wolf with orange eyes, and a black knight in armor holding a silver fork.
- Can you unify them using the same algorithm, with different priority functions?
  - Performance measures
    - Completeness, optimality, time complexity, space complexity