

CS540

Uninformed Search

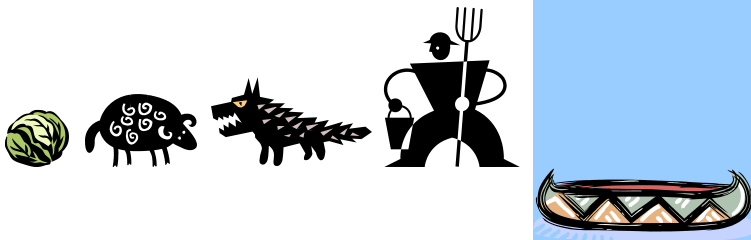
Yingyu Liang

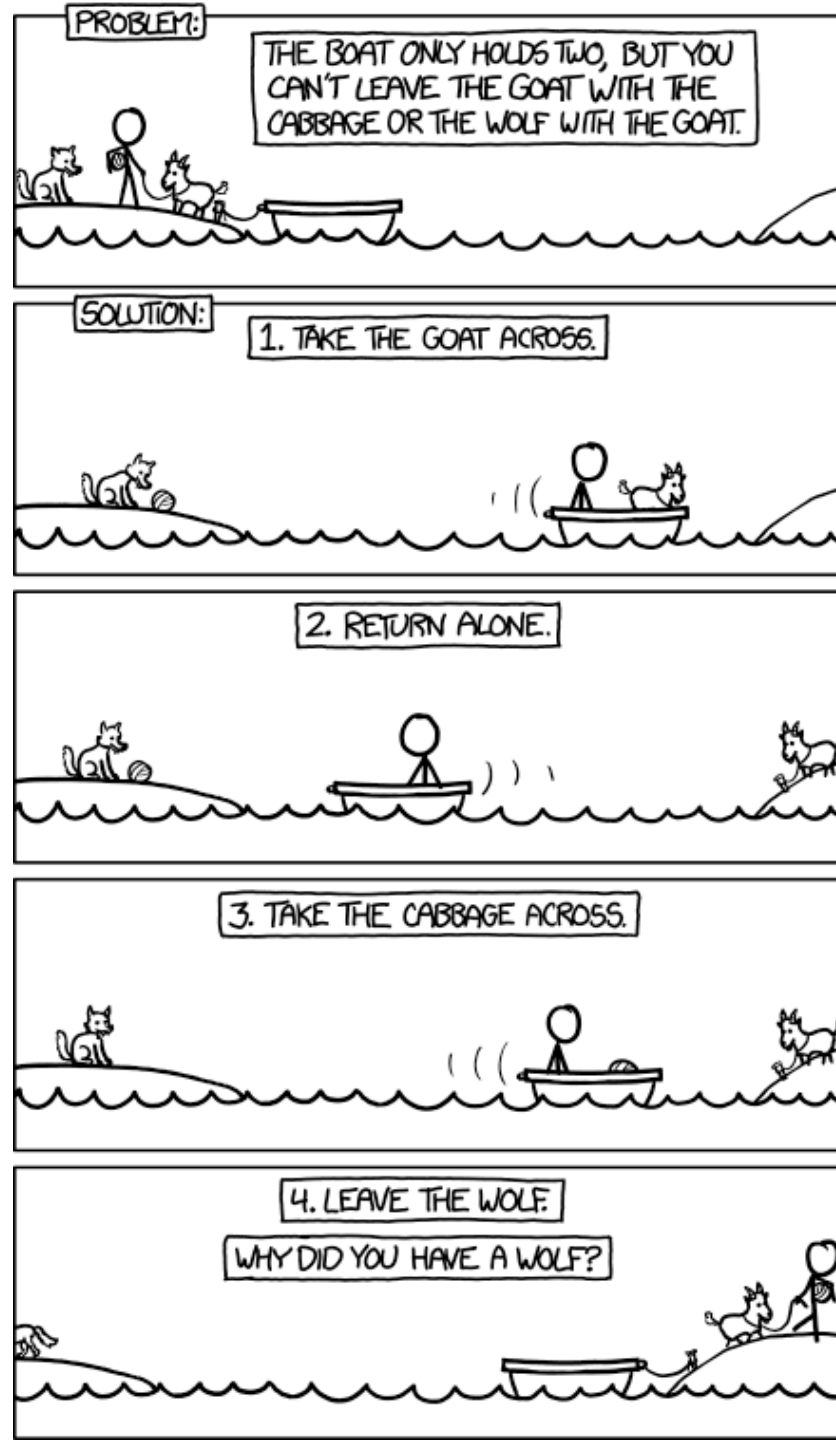
`yliang@cs.wisc.edu`

Computer Sciences Department
University of Wisconsin, Madison

Main messages

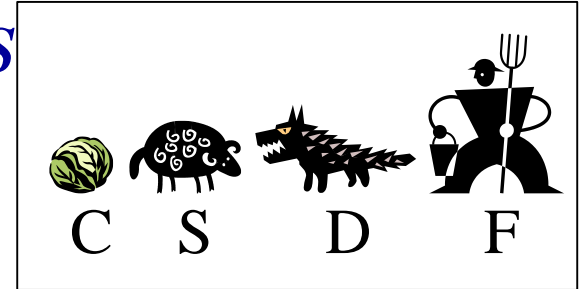
- Many AI problems can be formulated as search.
- Iterative deepening is good when you don't know much.





The search problem

- **State space** S : all valid configurations
- **Initial states (nodes)** $I = \{(\text{CSDF},)\} \subseteq S$
 - Where's the boat?
- **Goal states** $G = \{(\text{,CSDF})\} \subseteq S$
- **Successor function** $\text{succs}(s) \subseteq S$: states reachable in one step (one arc) from s
 - $\text{succs}((\text{CSDF},)) = \{(\text{CD}, \text{SF})\}$
 - $\text{succs}((\text{CDF}, \text{S})) = \{(\text{CD}, \text{FS}), (\text{D}, \text{CFS}), (\text{C}, \text{DFS})\}$
- **Cost** $\text{Cost}(s, s') = 1$ for all arcs. (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

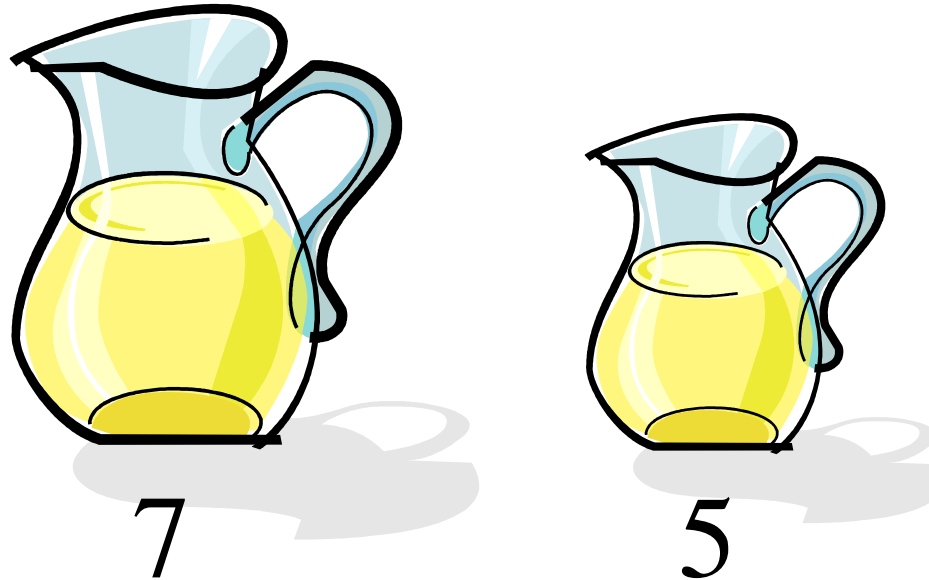
	1	2
3	4	5
6	7	8

Goal State

- States = configurations
- successor function = up to 4 kinds of movement
- Cost = 1 for each move

Search examples

- Water jugs: how to get 1?



- Goal? (How many goal states?)
- Successor function: fill up (from tap or other jug), empty (to ground or other jug)

Search examples

- Route finding (state? Successors? Cost weighted)

The screenshot shows the Google Maps interface in a Mozilla Firefox browser window. The address bar indicates the route from 1210 W Dayton St, Madison, WI 53706 to State St, Madison, WI 53703. The Google Maps logo and navigation controls are visible. A satellite view of the city of Madison is shown, with a blue line indicating the route. The route starts at 1210 W Dayton St, goes east, then turns left at N Frances St, right at W Gilman St, right at N Henry St, and right at W Gorham St. The distance is 1.2 mi (about 2 mins).

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

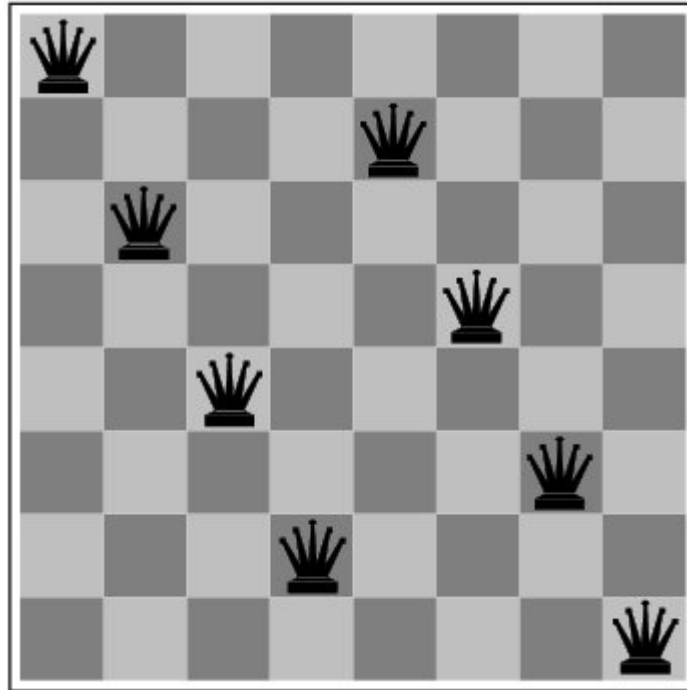
1. Head east from W Dayton St - go 0.5 mi
2. Turn left at N Frances St - go 0.2 mi
3. Turn right at W Gilman St - go 0.3 mi
4. Turn right at N Henry St - go 0.1 mi
5. Turn right at W Gorham St - go 0.1 mi

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 NAVTEQ™, Tele Atlas

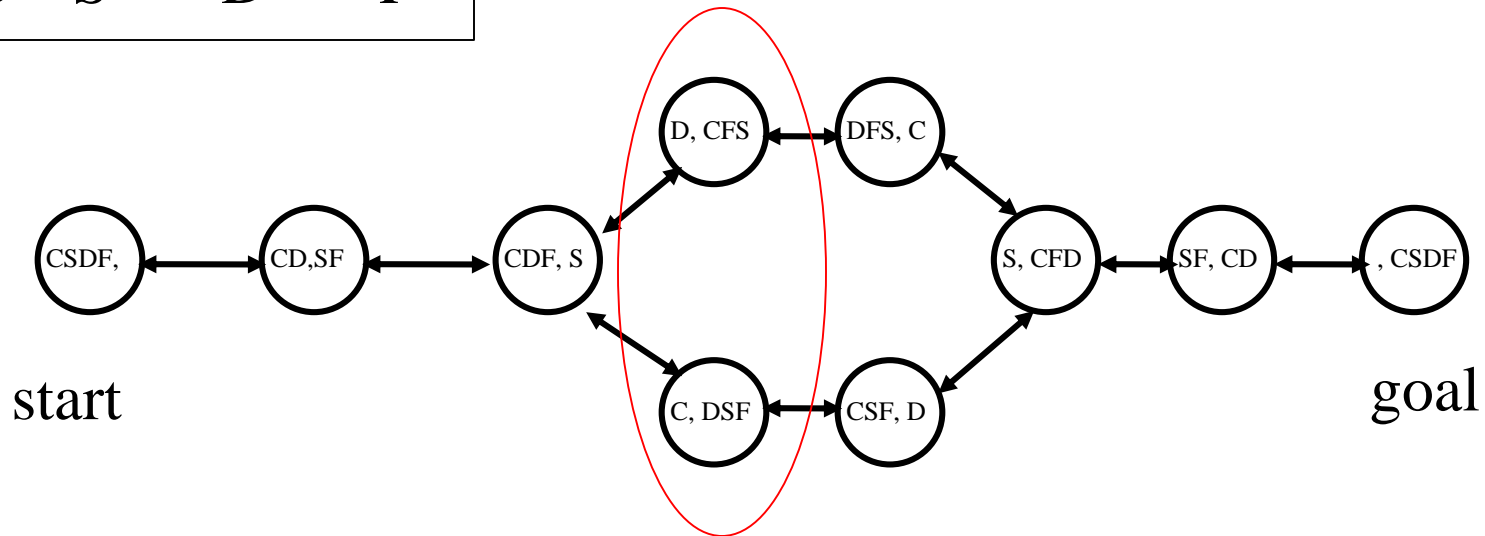
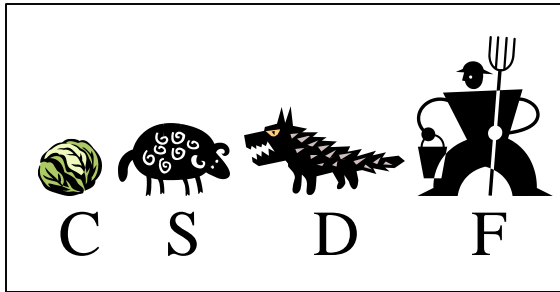
8-queens

- State: complete configuration vs. column-by-column



- Tree** instead of graph

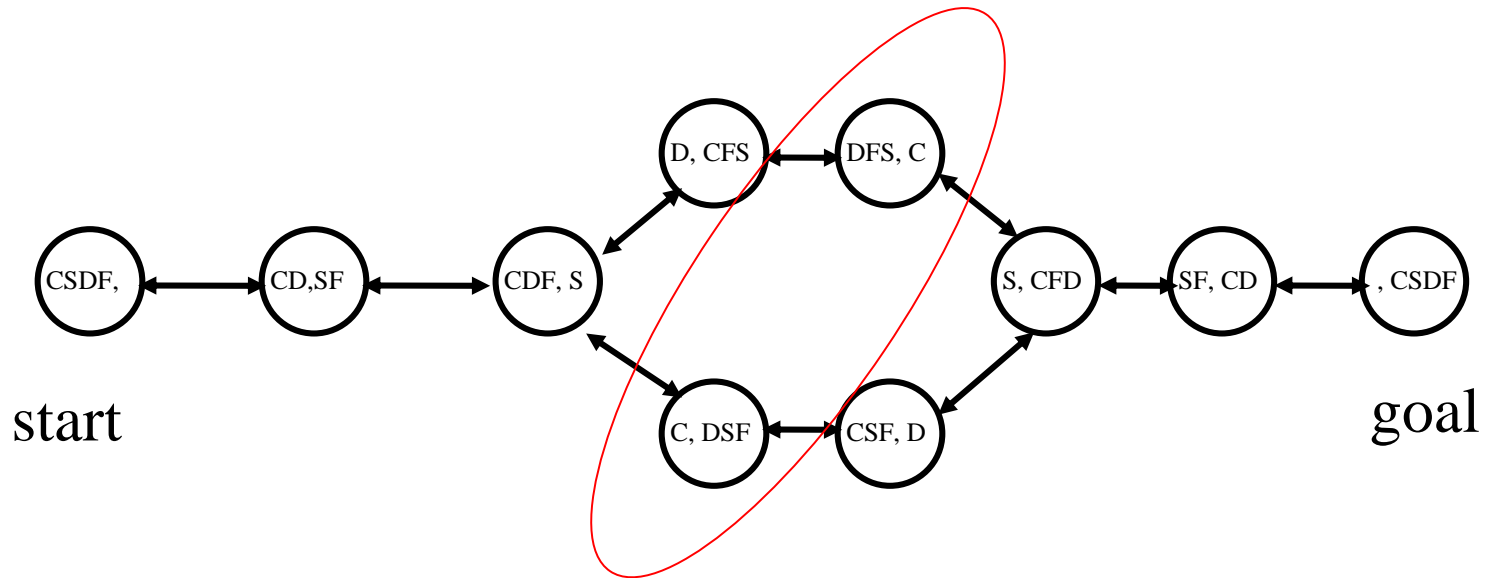
A directed graph in state space



- 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

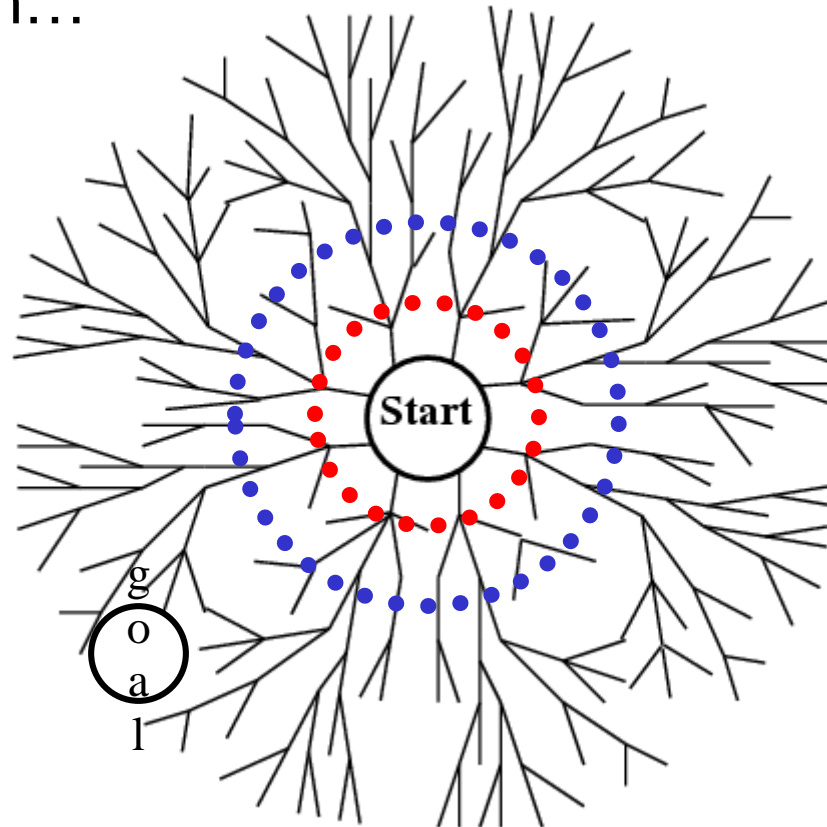
- **Uninformed** means we only know:
 - The goal test
 - The *succs()* function
- But **not** which non-goal states are better: that would be informed search (next lecture).
- For now, we also assume *succs()* graph is **a tree**.
 - Won't encounter repeated states.
 - We will relax it later.
- Search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

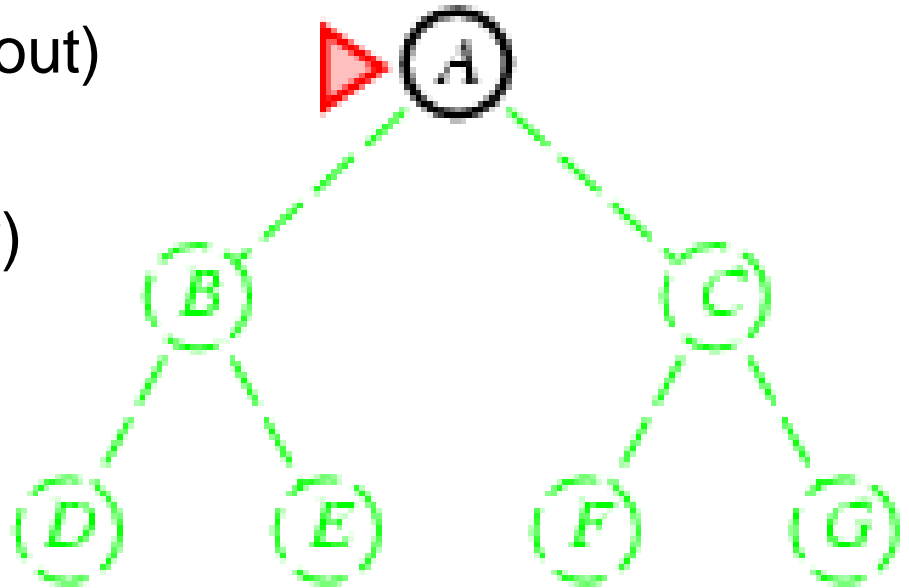
ripple



Breadth-first search (BFS)

Use a **queue** (First-in First-out)

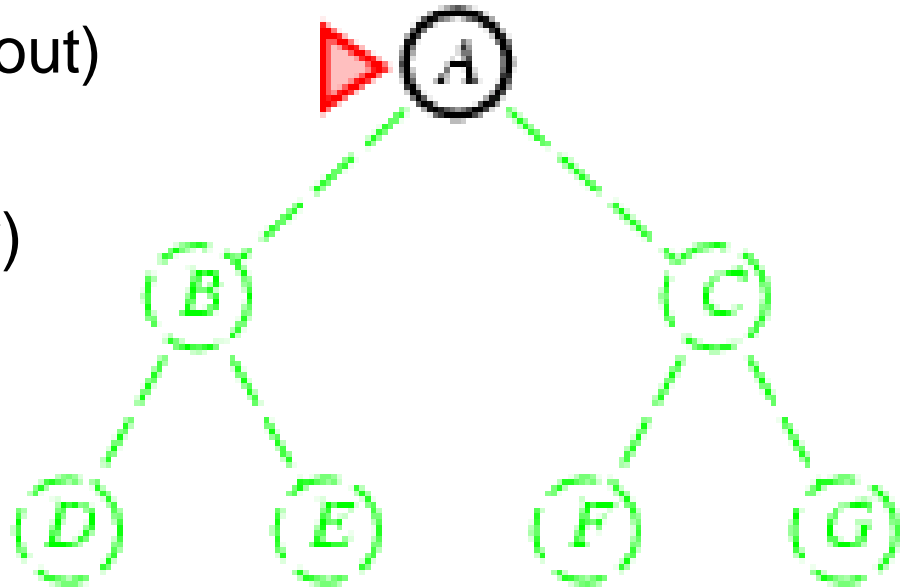
1. en_queue(Initial states)
2. While (queue not empty)
3. s = de_queue()
4. if (s==goal) success!
5. T = succs(s)
6. en_queue(T)
7. endwhile



Breadth-first search (BFS)

Use a **queue** (First-in First-out)

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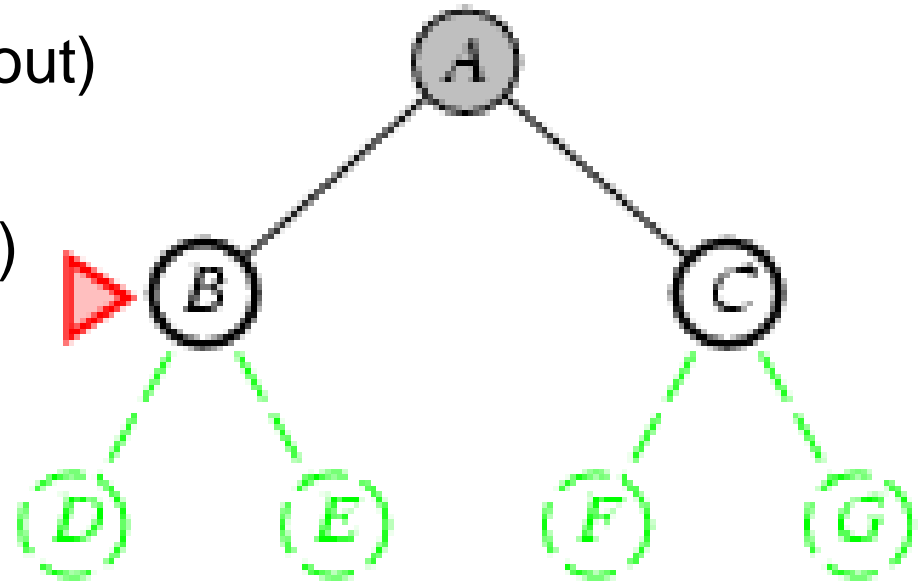


queue (**fringe**, **OPEN**)
→ [A] →

Breadth-first search (BFS)

Use a **queue** (First-in First-out)

1. en_queue(Initial states)
2. While (queue not empty)
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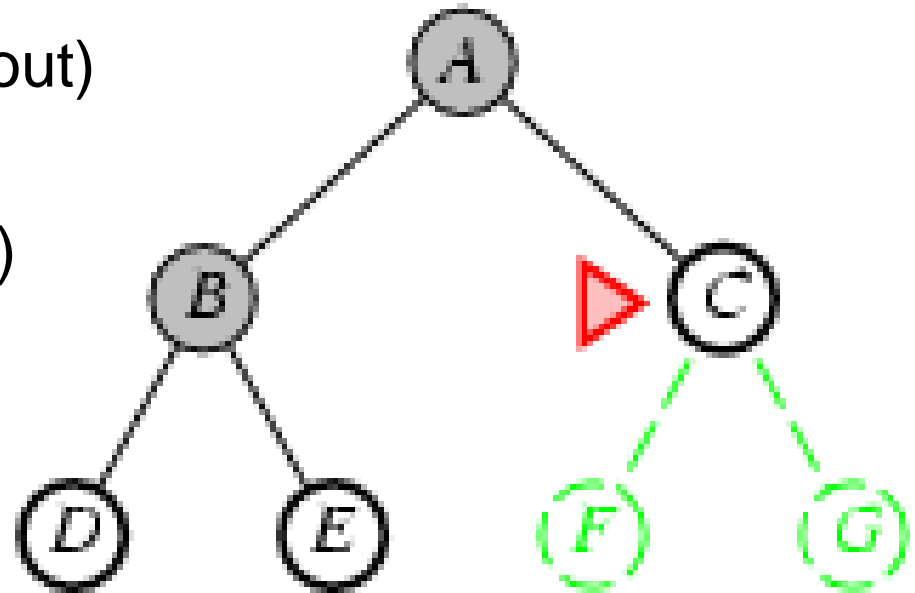


queue (**fringe**, **OPEN**)
→ [CB] → A

Breadth-first search (BFS)

Use a **queue** (First-in First-out)

1. en_queue(Initial states)
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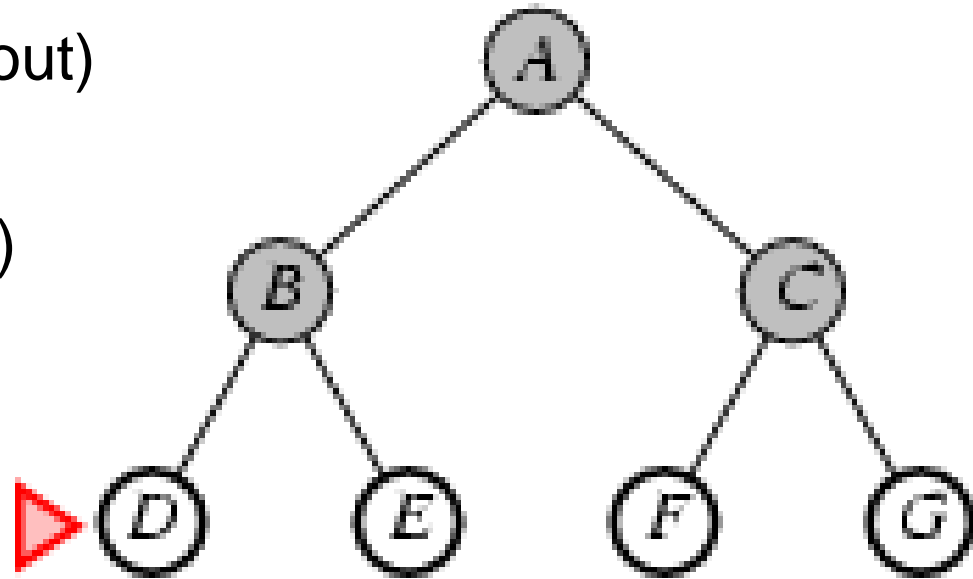


queue (**fringe**, **OPEN**)
→ [EDC] → B

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



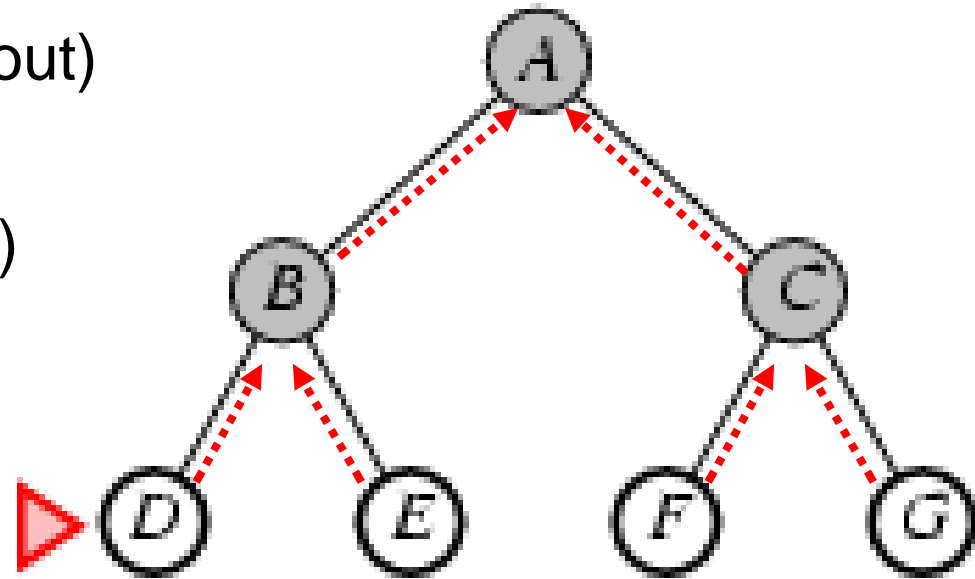
queue (**fringe** , **OPEN**)
→[GFED] → C

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)

- en_queue(Initial states)
- While (queue not empty)
- s = de_queue()
- if (s==goal) success!
- T = succs(s)
- **for t in T: t.prev=s**
- en_queue(T)
- endwhile



queue
→ [] → G

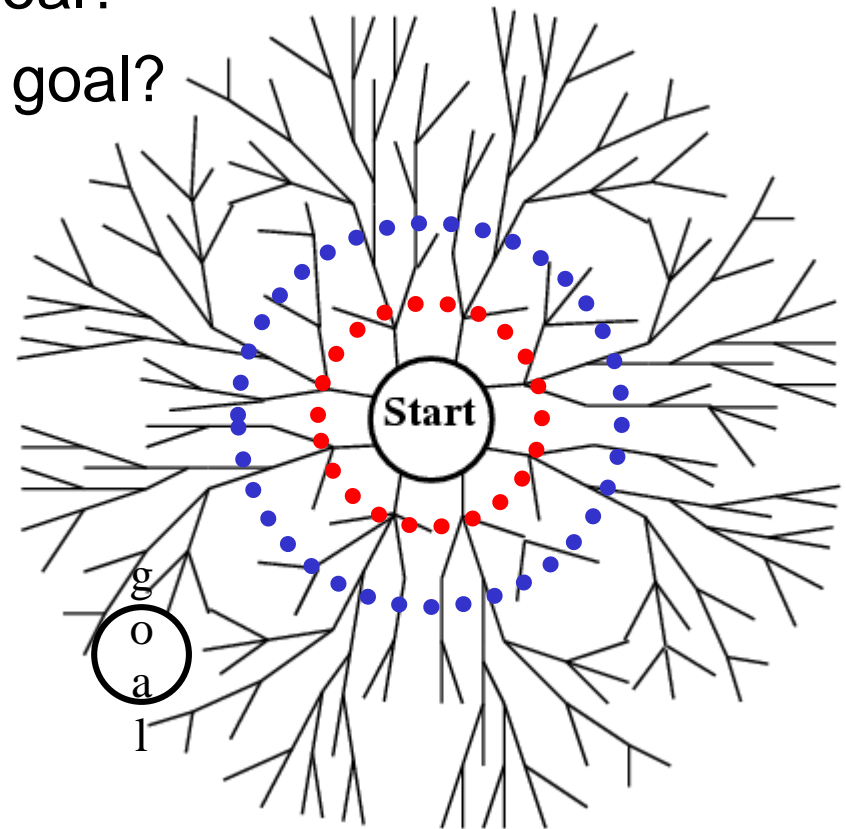
... until much later we pop G.

We need **back pointers** to recover the solution path.

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

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



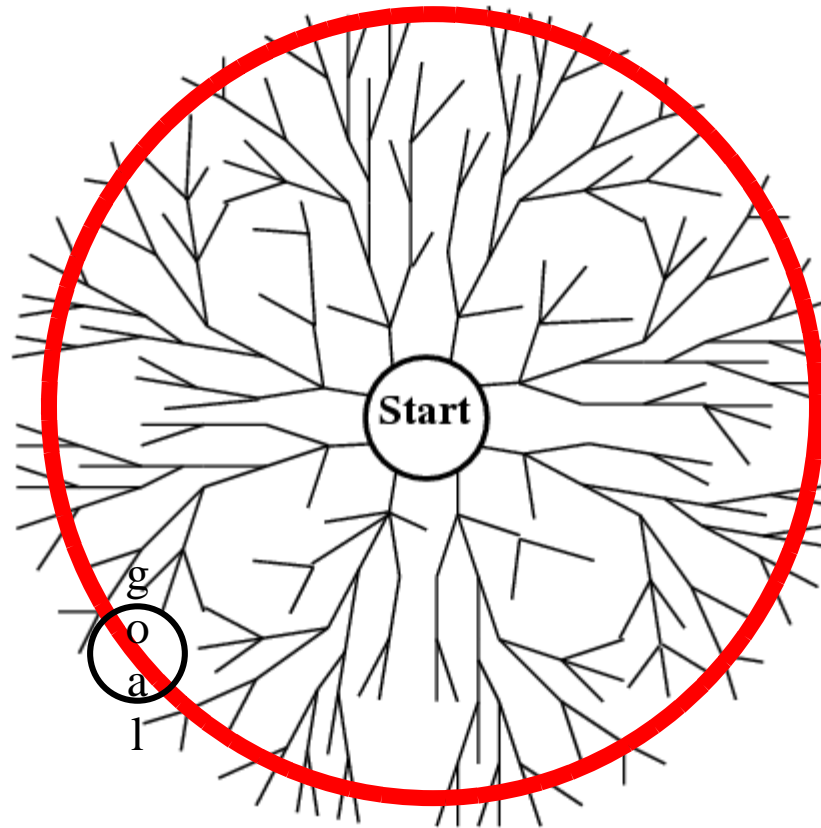
Performance of BFS

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 ¹	$O(b^d)$	$O(b^d)$

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.
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 - 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

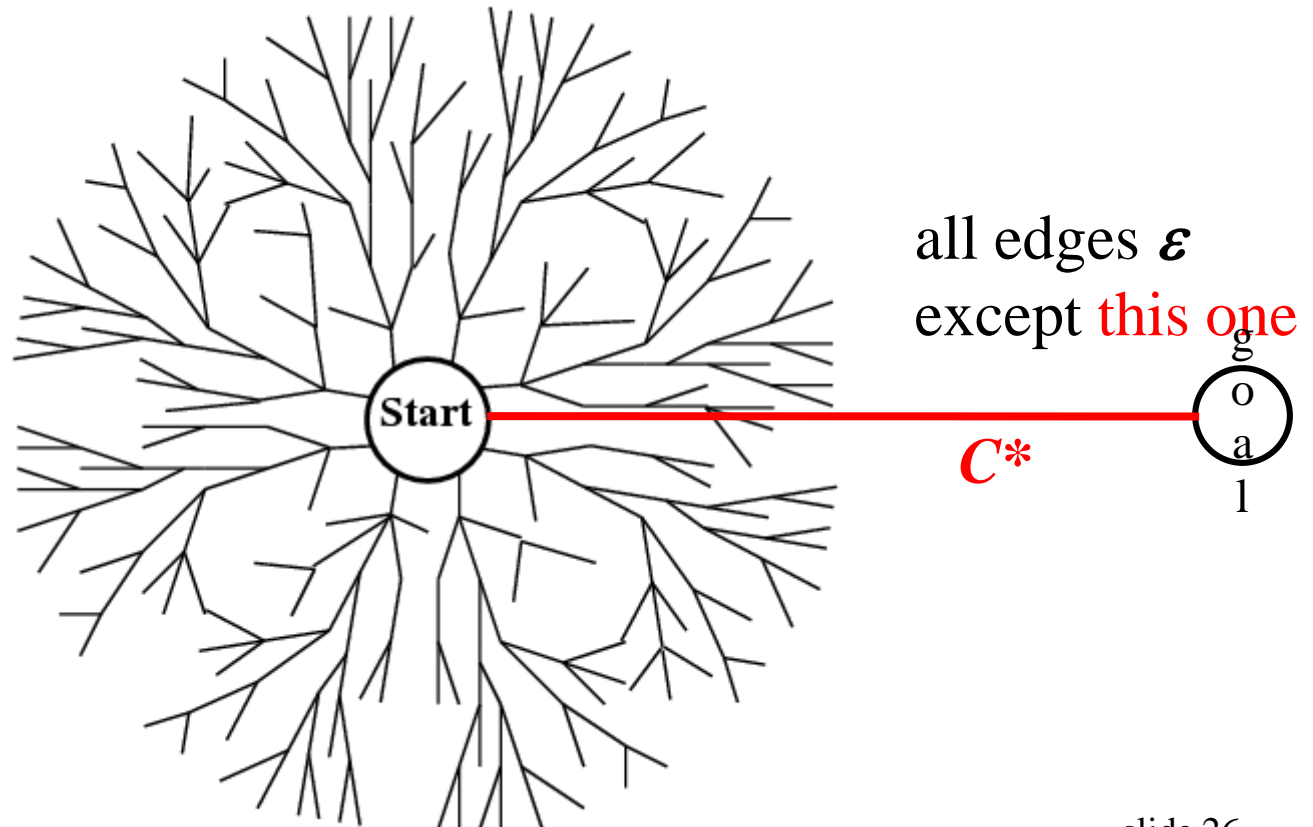
- Find the least-cost goal
- Each node has a path cost from start (= sum of edge costs along the path). Expand the least cost node first.
- Use a **priority queue** instead of a normal queue
 - Always take out the least cost item
 - Remember *heap*? time $O(\log(\text{\#items in heap}))$

That's it*

* Complications on graphs (instead of trees). Later.

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
 - $O(b^{C^*/\epsilon})$, possibly $C^*/\epsilon \gg d$



Performance of search algorithms on trees

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

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

1. edge cost constant, or positive non-decreasing in depth
 - 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 (goal is the last node at radius d):
 - Have to generate all nodes at radius d .
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Solution:
Uniform-cost
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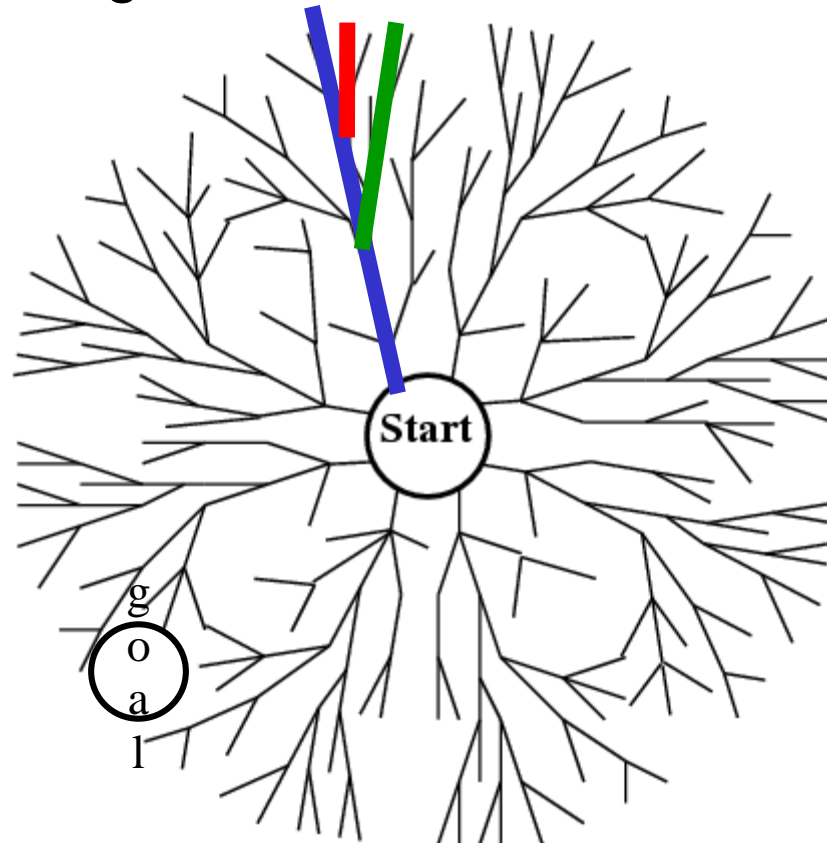
Solution:
Depth-first
search

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... ————

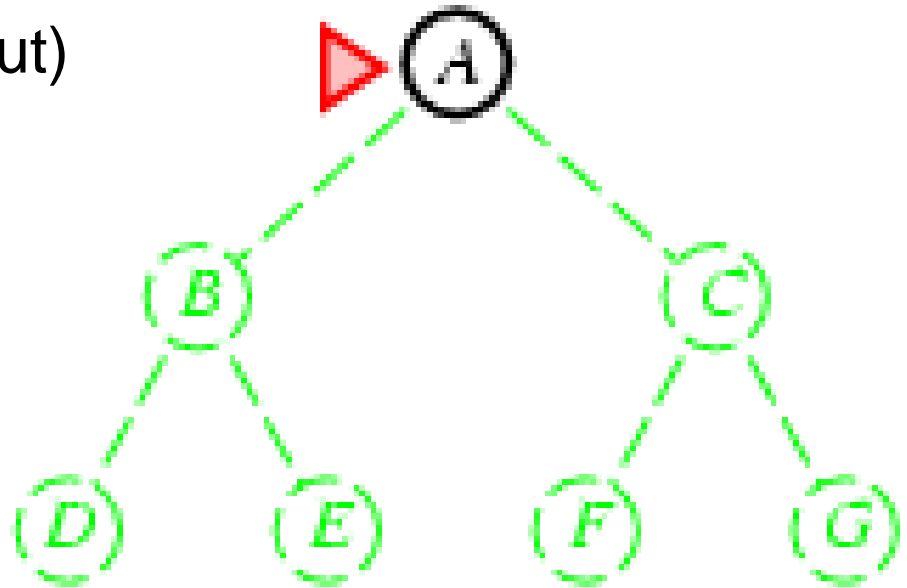
fan



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

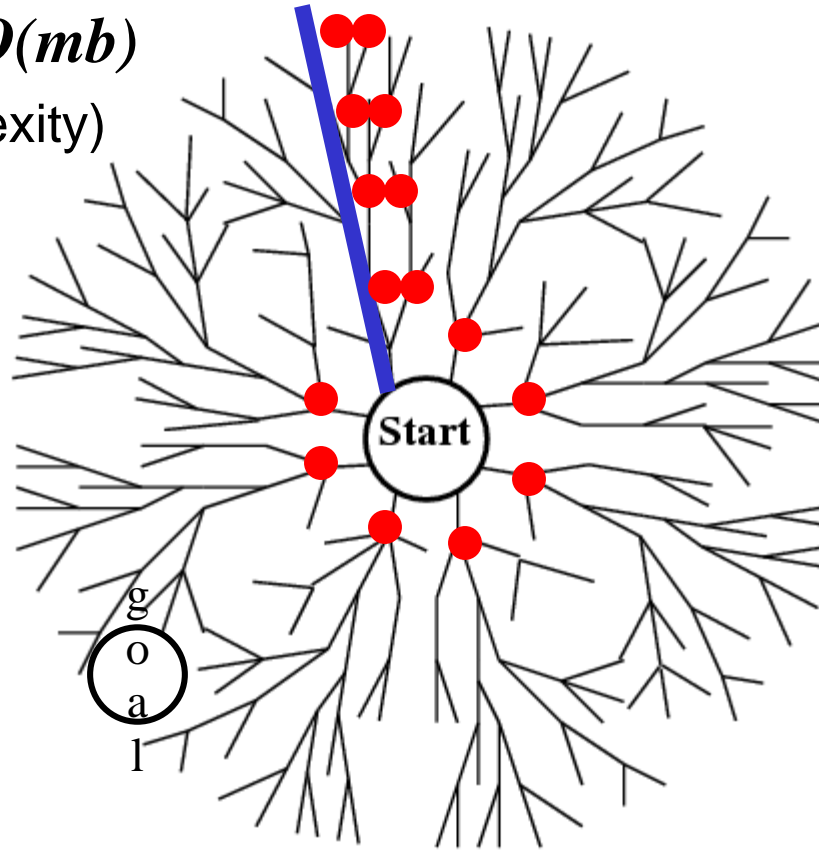


stack (**fringe**)

[] ⇔

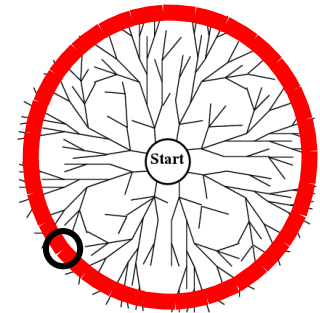
What's in the fringe for DFS?

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



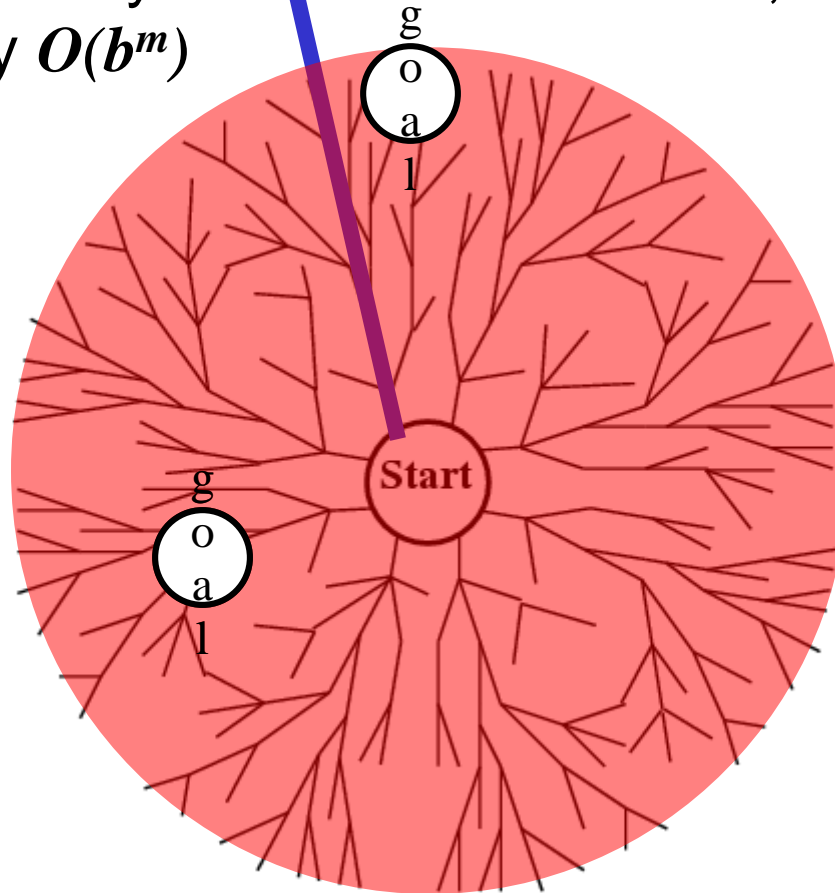
- “backtracking search” even less space
 - generate siblings (if applicable)

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

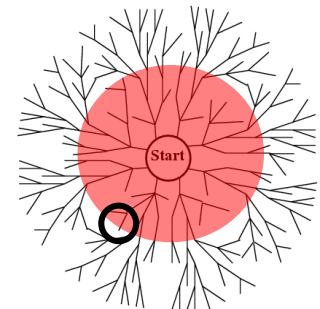


What's wrong with DFS?

- 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

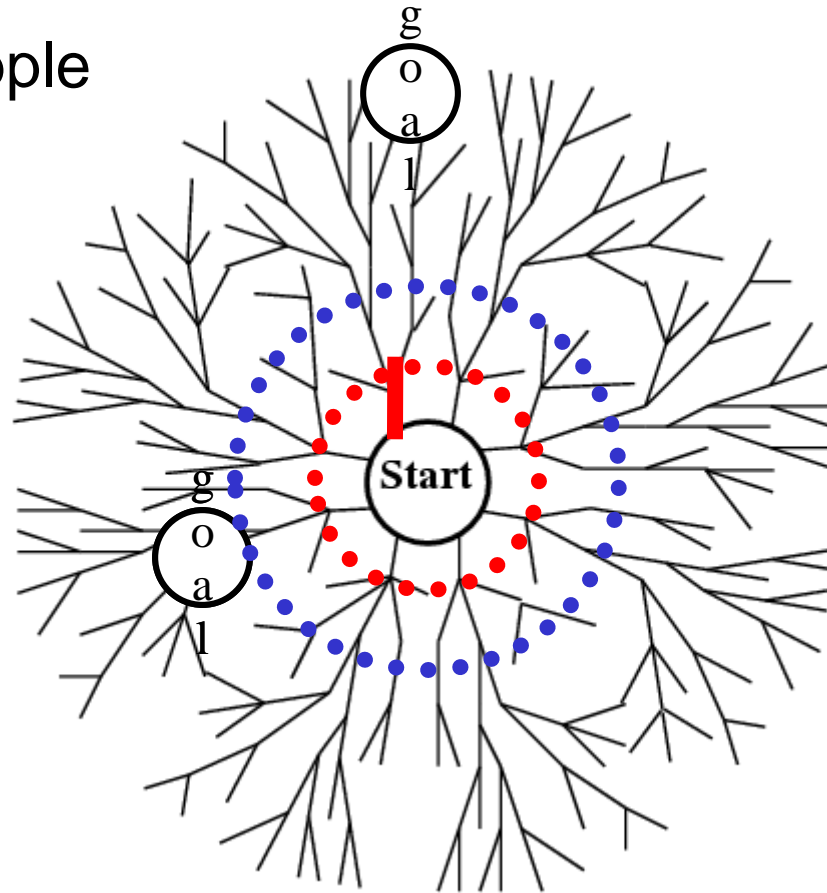
	Complete	optimal	time	space
Breadth-first search	Y	Y, if ¹	$O(b^d)$	$O(b^d)$
Uniform-cost search ²	Y	Y	$O(b^{C^*/\epsilon})$	$O(b^{C^*/\epsilon})$
Depth-first search	N	N	$O(b^m)$	$O(bm)$

1. edge cost constant, or positive non-decreasing in depth
 - 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 ¹	$O(b^d)$	$O(b^d)$
Uniform-cost search ²	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 ¹	$O(b^d)$	$O(bd)$

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Performance of search algorithms on trees

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

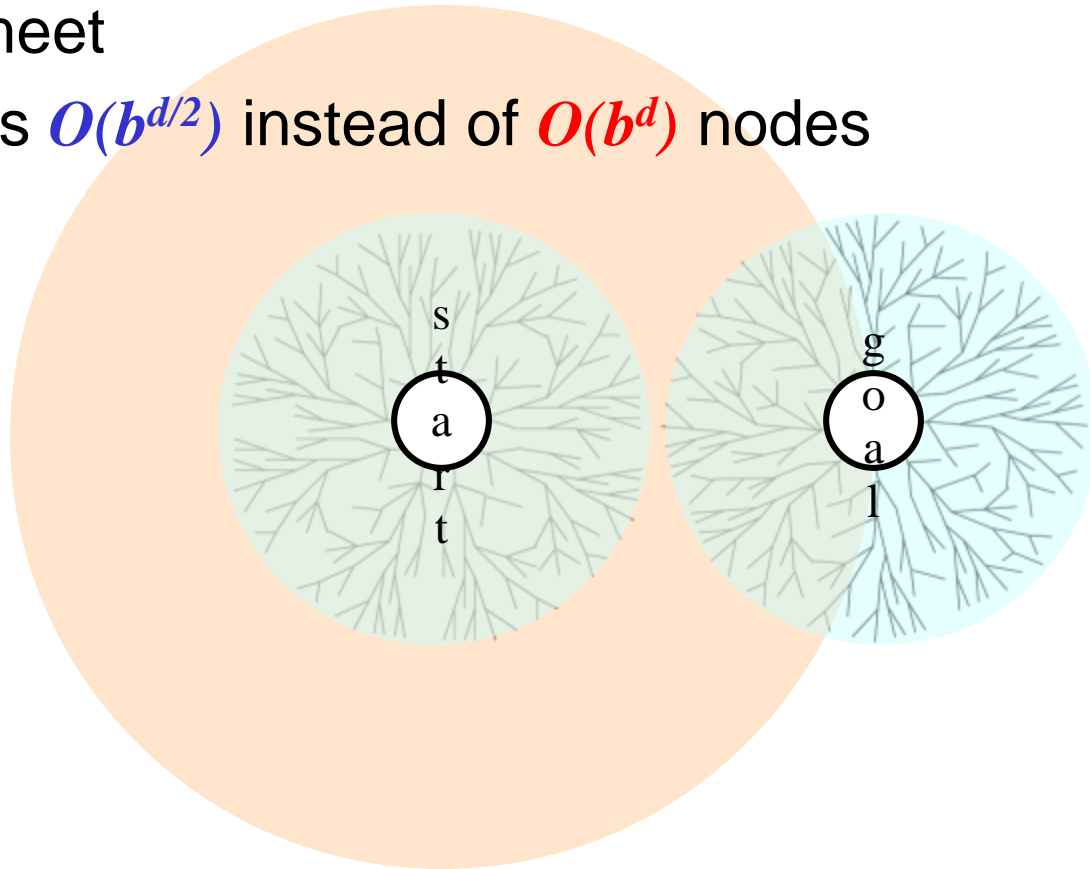
	Complete	optimal	time	space
Breadth-first			$O(b^d)$	$O(b^d)$
			$O(b^{C^*/\epsilon})$	$O(b^{C^*/\epsilon})$
			$O(b^m)$	$O(bm)$
			$O(b^d)$	$O(bd)$

How to reduce the number of states we have to generate?

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

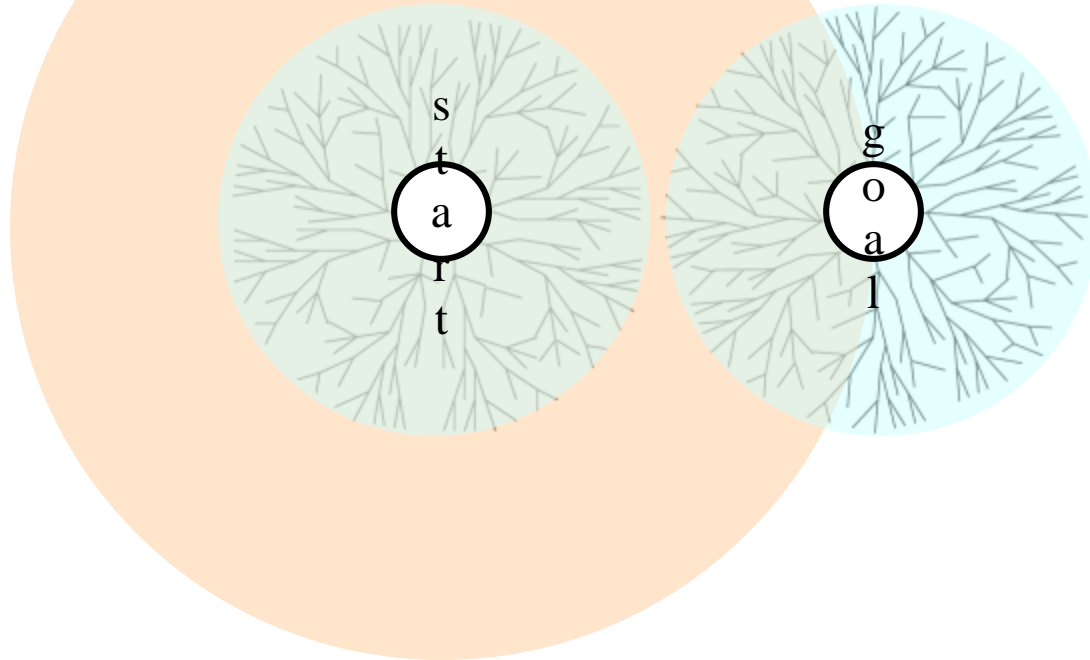
Bidirectional search

- Breadth-first search from both start and goal
- Fringes meet
- Generates $O(b^{d/2})$ instead of $O(b^d)$ nodes



Bidirectional search

- But
 - The fringes are $O(b^{d/2})$
 - How do you start from the 8-queens goals?



Performance of search algorithms on trees

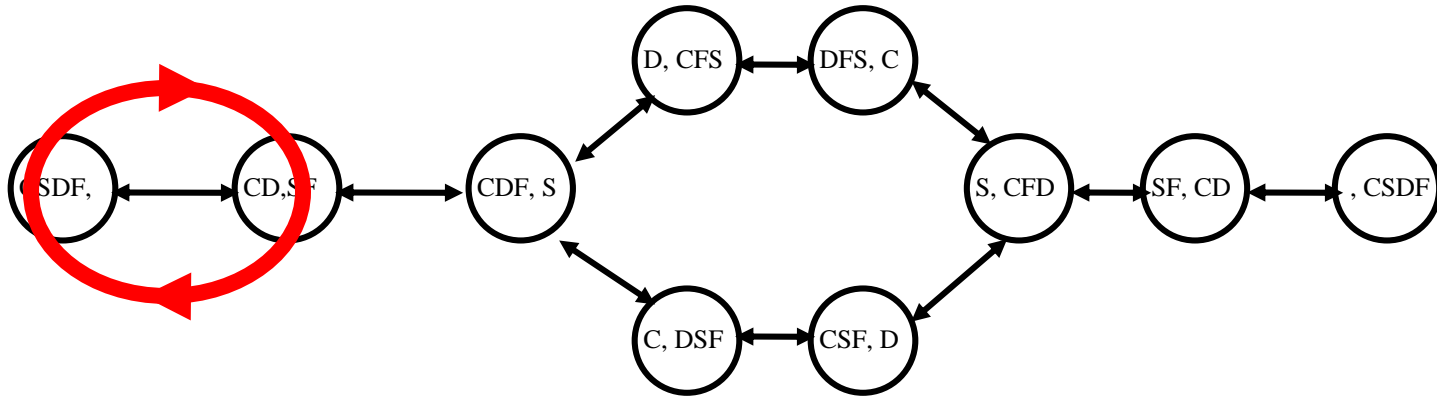
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	Complete	optimal	time	space
Breadth-first search	Y	Y, if ¹	$O(b^d)$	$O(b^d)$
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Depth-first search	N	N	$O(b^m)$	$O(bm)$
Iterative deepening	Y	Y, if ¹	$O(b^d)$	$O(bd)$
Bidirectional search ³	Y	Y, if ¹	$O(b^{d/2})$	$O(b^{d/2})$

1. edge cost constant, or positive non-decreasing in depth
 - edge costs $\geq \epsilon > 0$. C^* is the best goal path cost.
 - both directions BFS; not always feasible.

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.

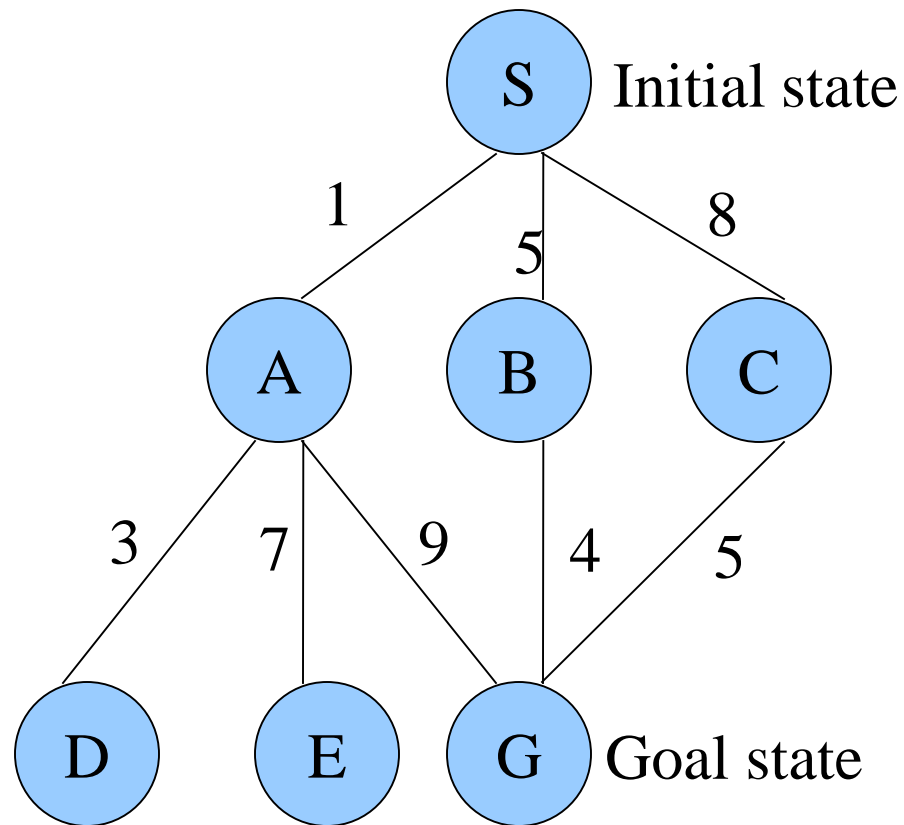
If state space graph is not a tree

- BFS:
 - Still $O(b^d)$ space complexity, not worse
- DFS:
 - Known as Memorizing DFS (MEMDFS)
 - Space and time now $O(\min(N, b^M))$ – much worse!
 - N: number of states in problem
 - M: length of longest cycle-free path from start to anywhere
 - Alternative: Path Check DFS (PCDFS): remember only expanded states on current path (from start to the current node)
 - Space $O(M)$
 - Time $O(b^M)$

Path Checking DFS

1. Maintain a “prefix” path from root to current node, initially empty.
2. Pop a state **s**. If **s** in prefix, skip to next pop
3. Goal-checking **s**.
4. **s** comes with a backpointer to its parent **p**. The prefix should contain **p** somewhere as in **initial, ..., p, ...**
5. Remove everything after **p** and put **s** there, so prefix is now **initial, ..., p, s**.
6. When you generate a successor **t** of **s**, check if **t** is in **prefix or stack**. If no, push **t** to the stack; if yes, do not push it.

Example



(All edges are directed, pointing downwards)

Nodes expanded by:

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

Breadth-First Search

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

Solution path found is S A G <-- this G also has cost 10
Number of nodes expanded (including goal node) = 7

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 ★
 - Bidirectional search
- Can you unify them (except bidirectional) using the same algorithm, with different priority functions?
- Performance measures
 - Completeness, optimality, time complexity, space complexity

