

# Uninformed search strategies (Section 3.4)



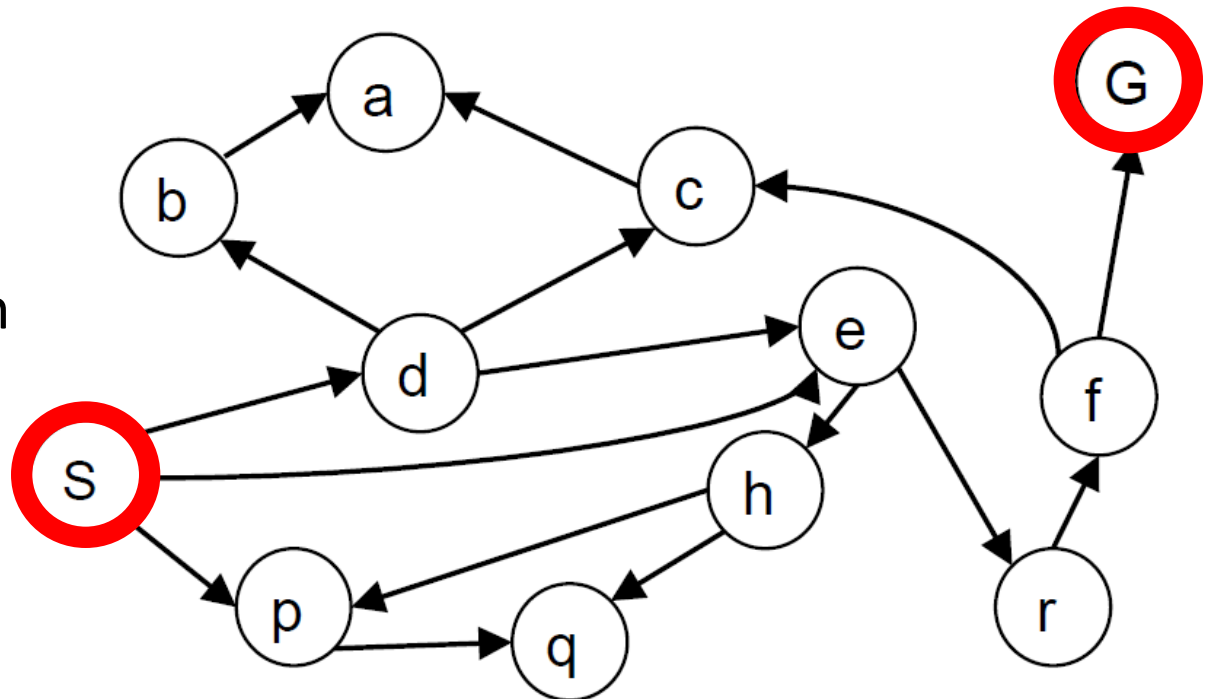
# Uninformed search strategies

- A **search strategy** is defined by picking the order of node expansion
- **Uninformed** search strategies use only the information available in the problem definition
  - Breadth-first search
  - Depth-first search
  - Iterative deepening search
  - Uniform-cost search

# Breadth-first search

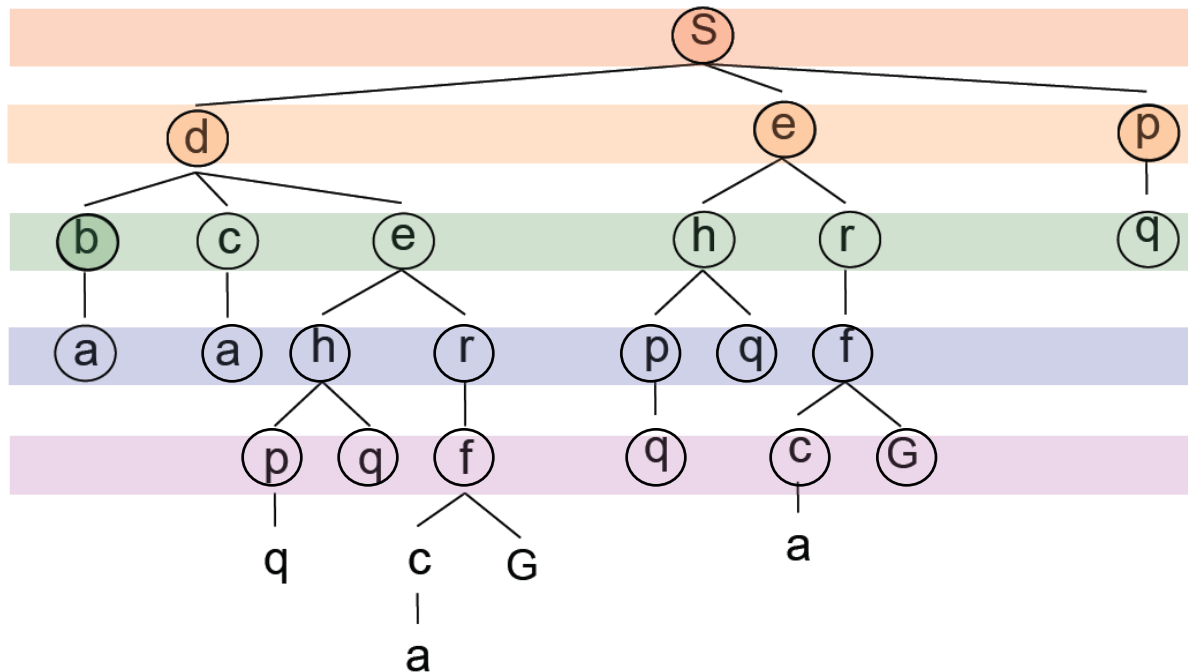
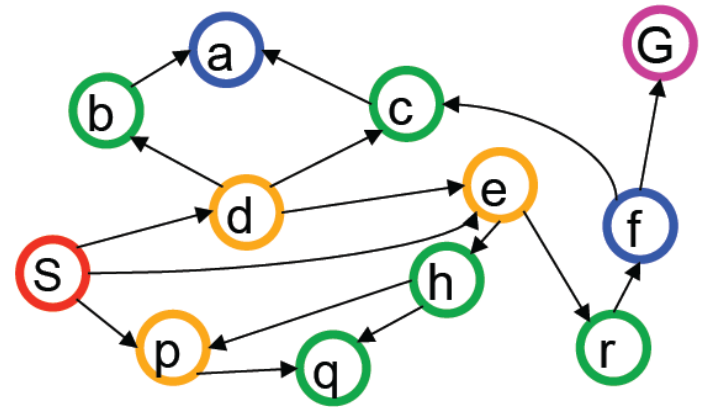
- Expand shallowest unexpanded node
- Implementation: *frontier* is a FIFO queue

Example state space graph for a tiny search problem



# Breadth-first search

- Expansion order:  
(S,d,e,p,b,c,e,h,r,q,a,a,  
h,r,p,q,f,p,q,f,q,c,G)

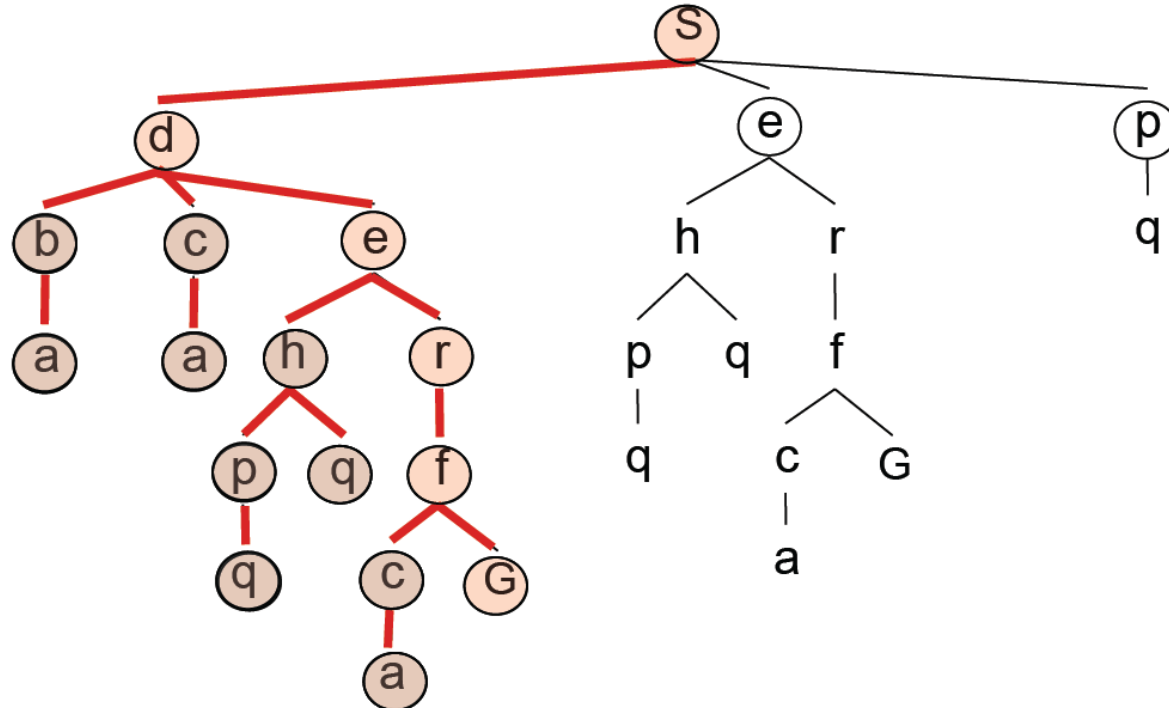
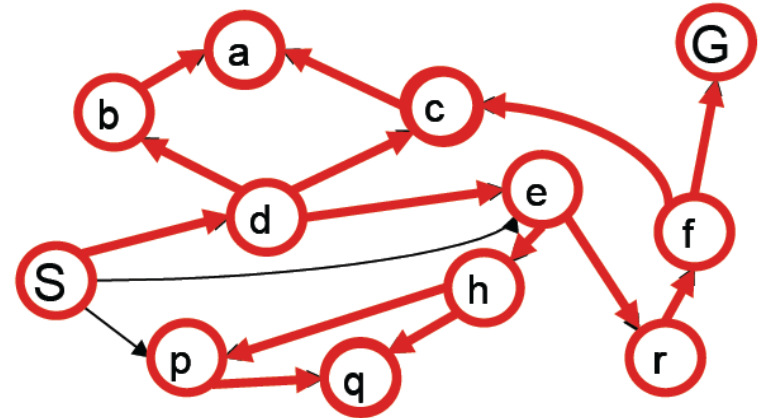


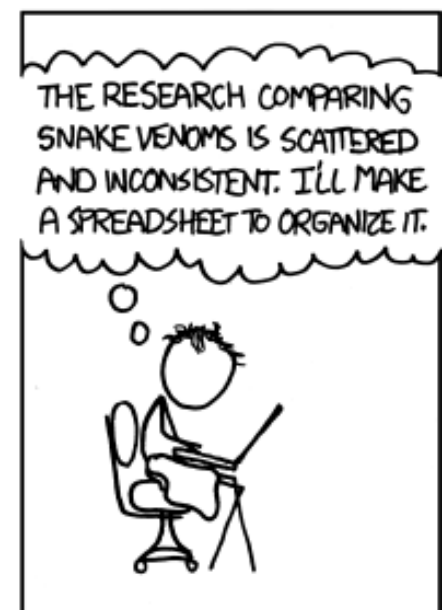
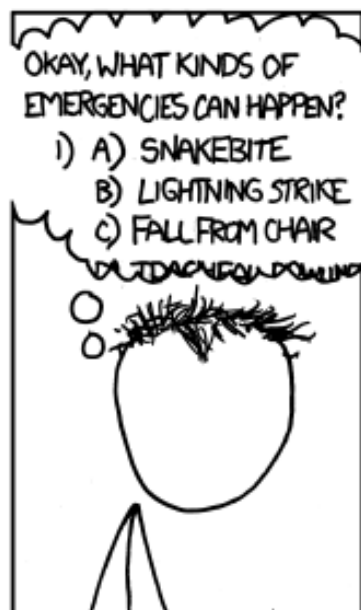
# Depth-first search

- Expand deepest unexpanded node
- Implementation: *frontier* is a LIFO queue

# Depth-first search

- Expansion order:  
(d,b,a,c,a,e,h,p,q,q,  
r,f,c,a,G)

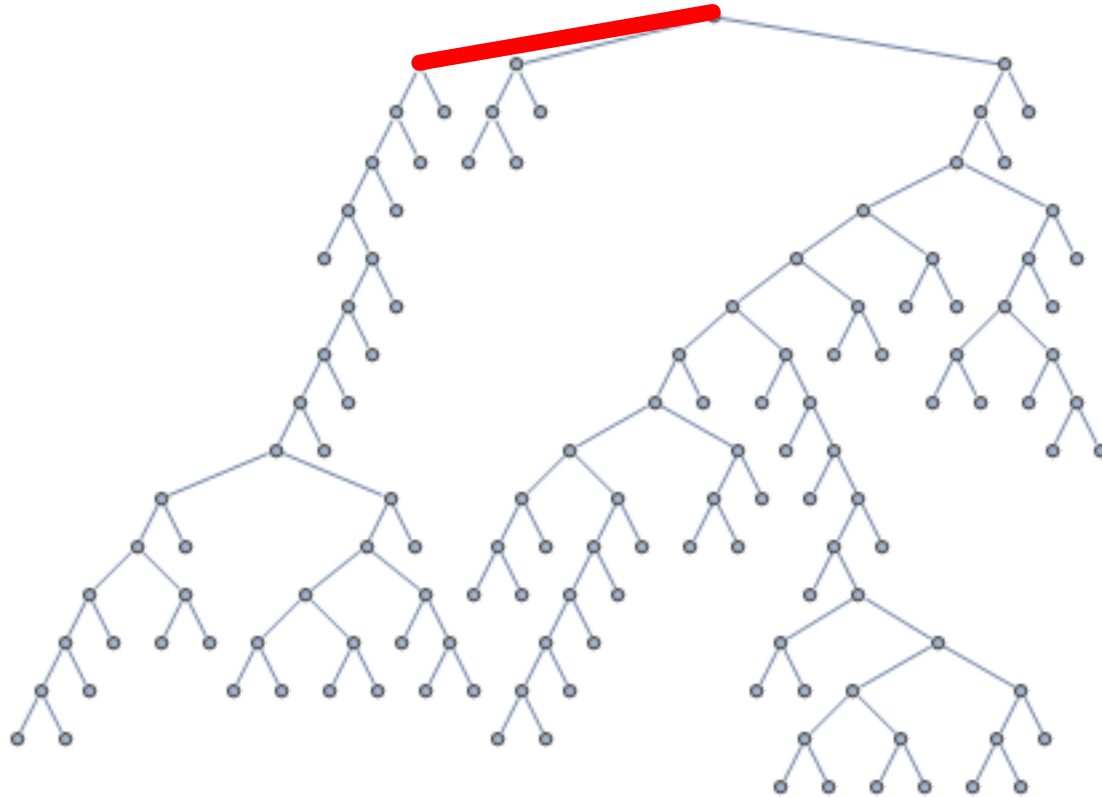




I REALLY NEED TO STOP USING DEPTH-FIRST SEARCHES.

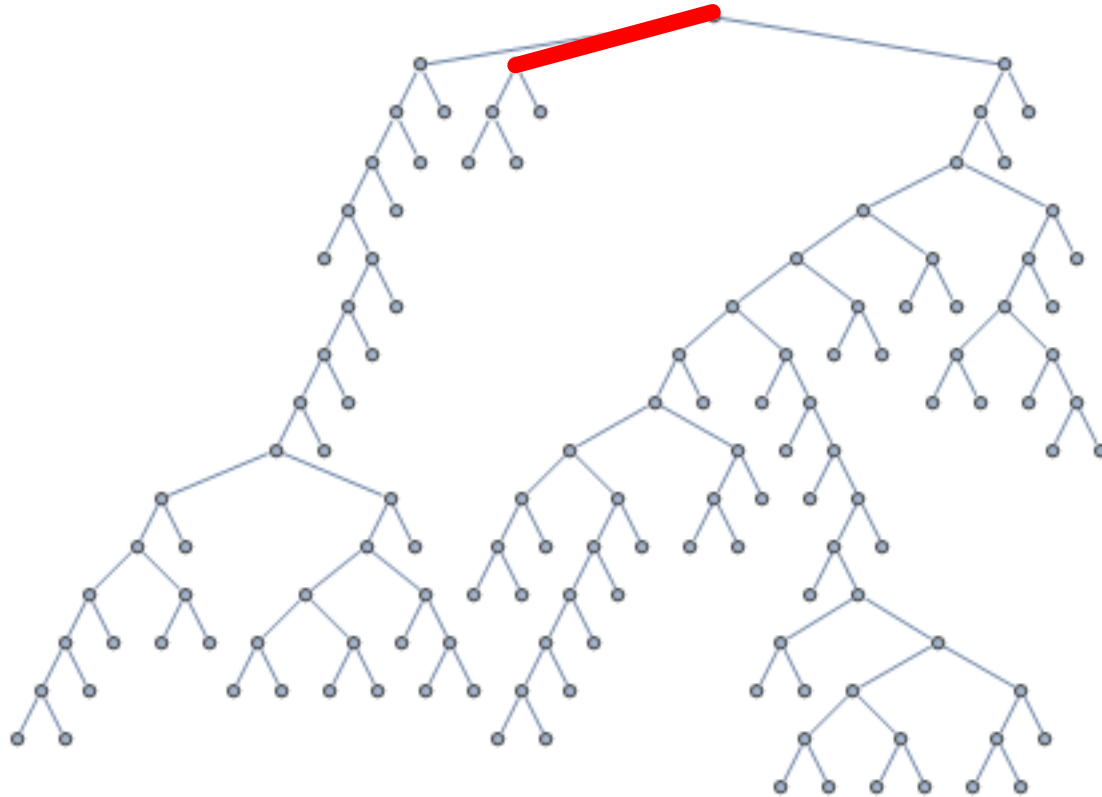
<http://xkcd.com/761/>

# BFS vs. DFS

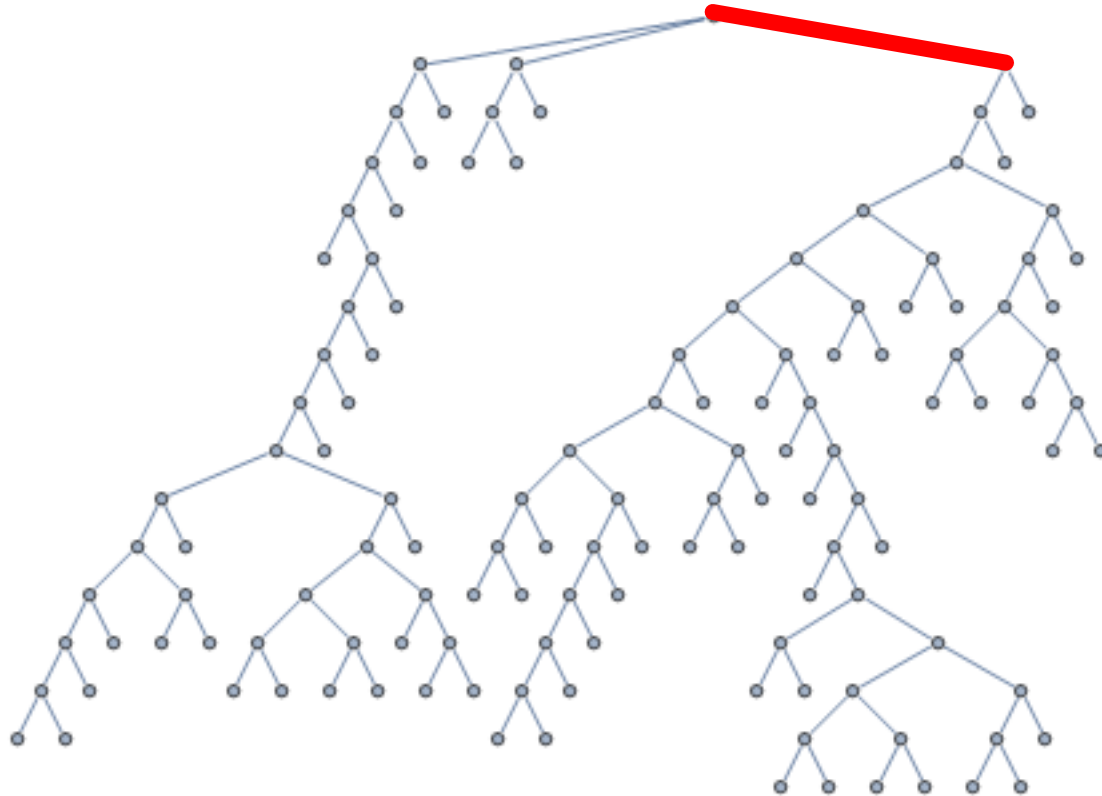




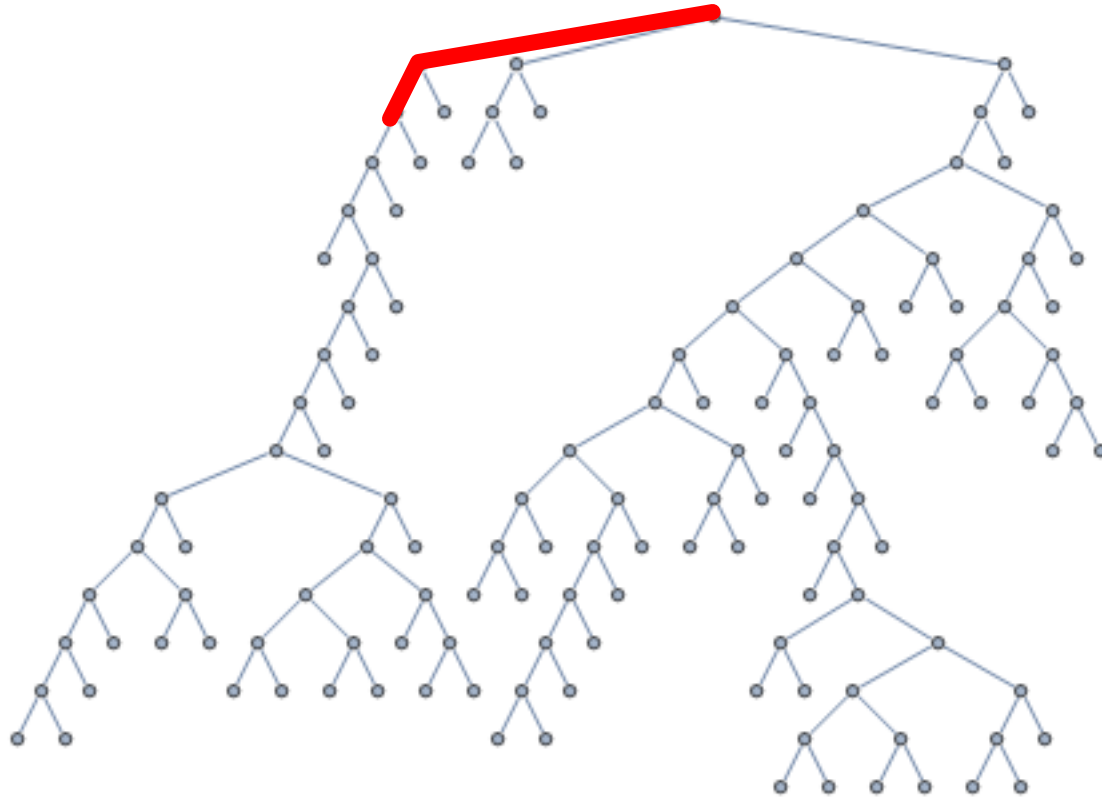
# BFS vs. DFS



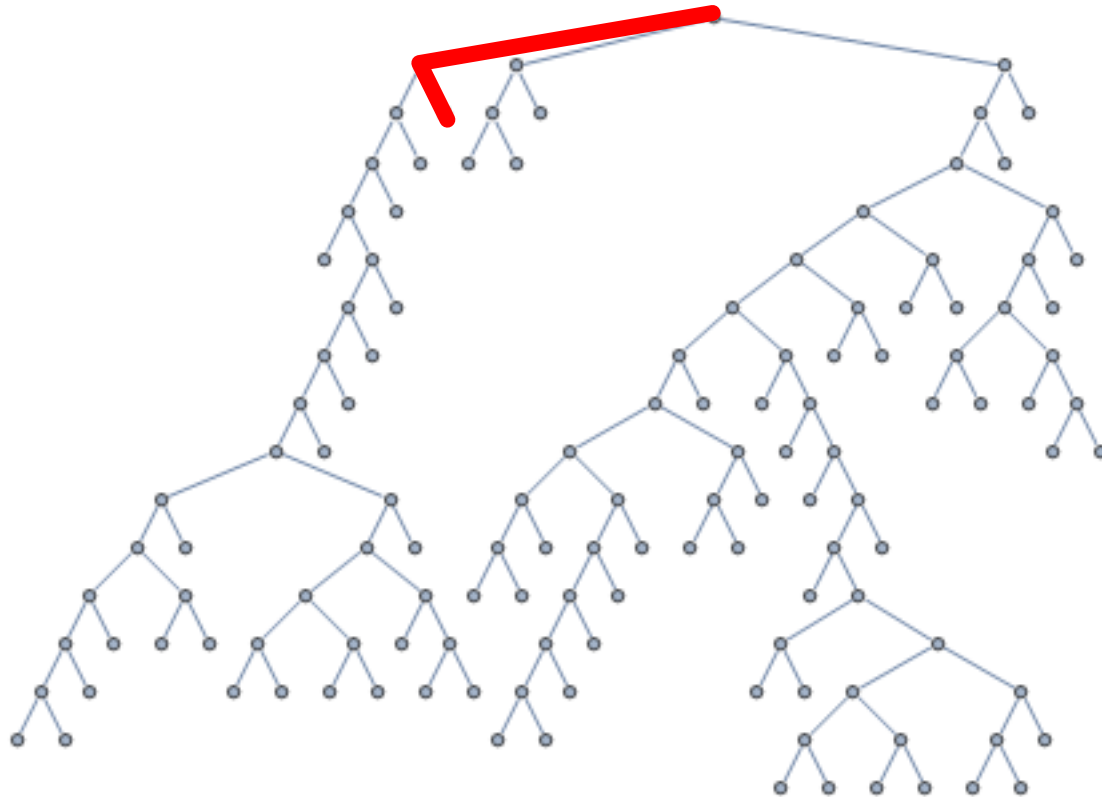
# BFS vs. DFS



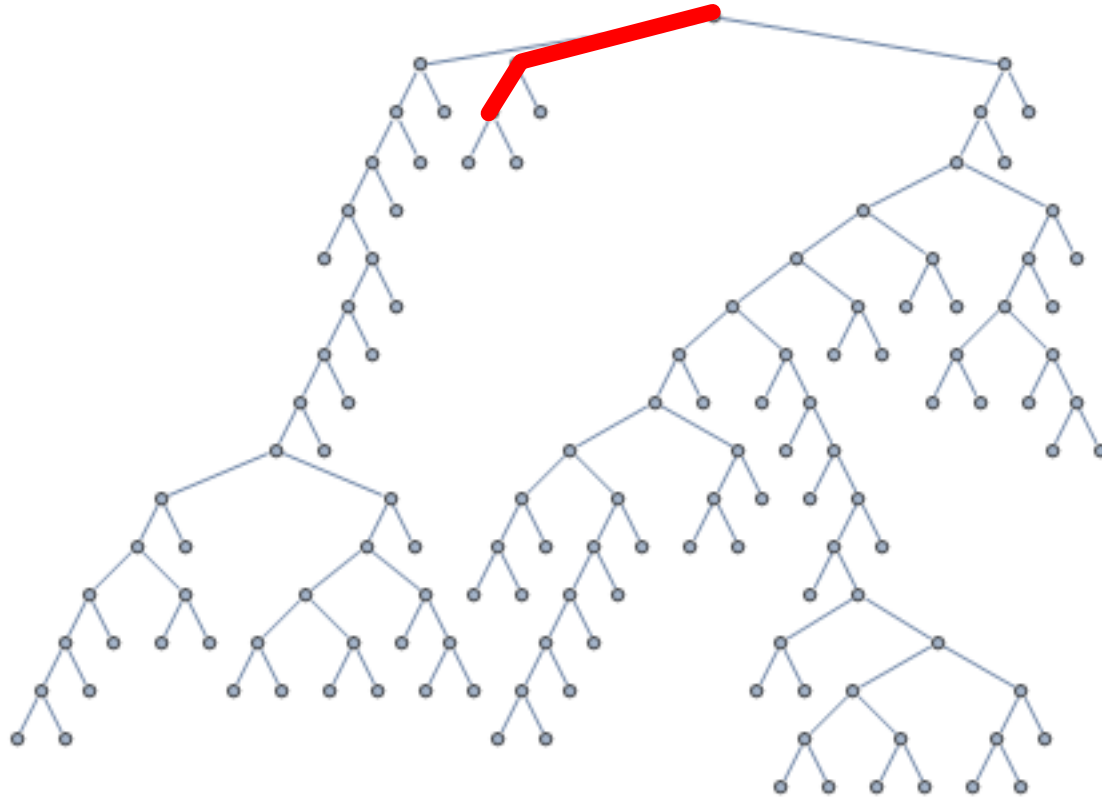
# BFS vs. DFS



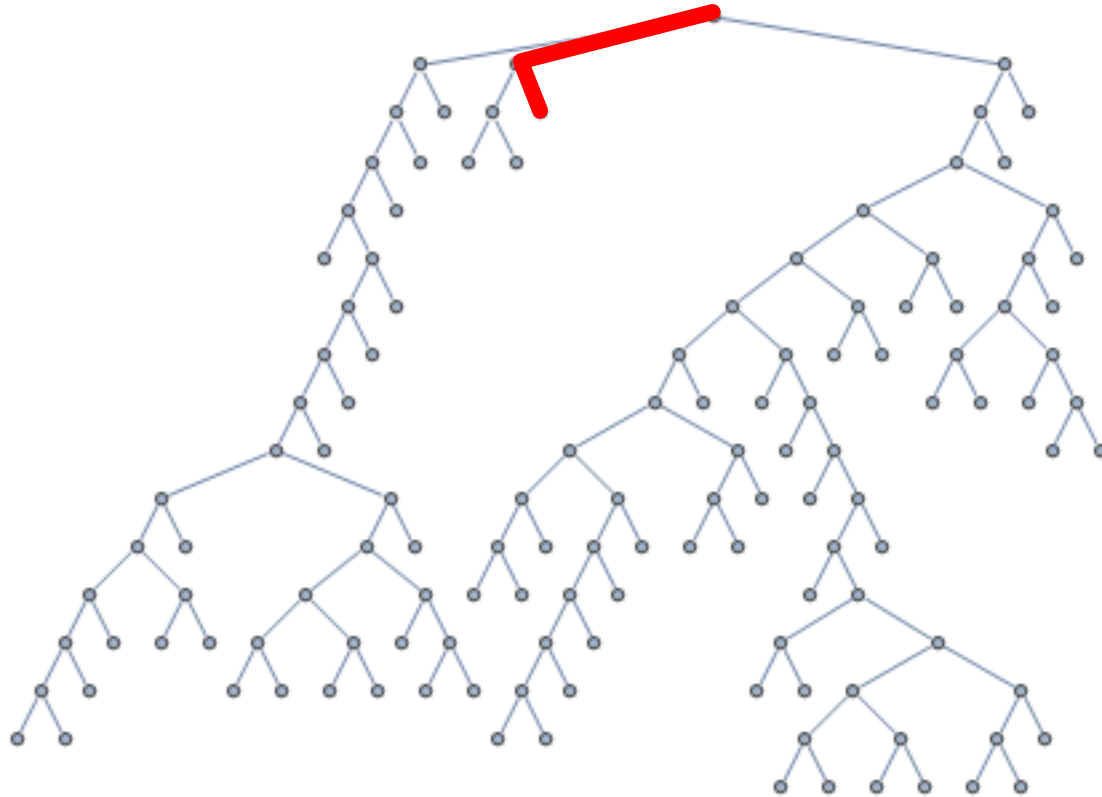
# BFS vs. DFS



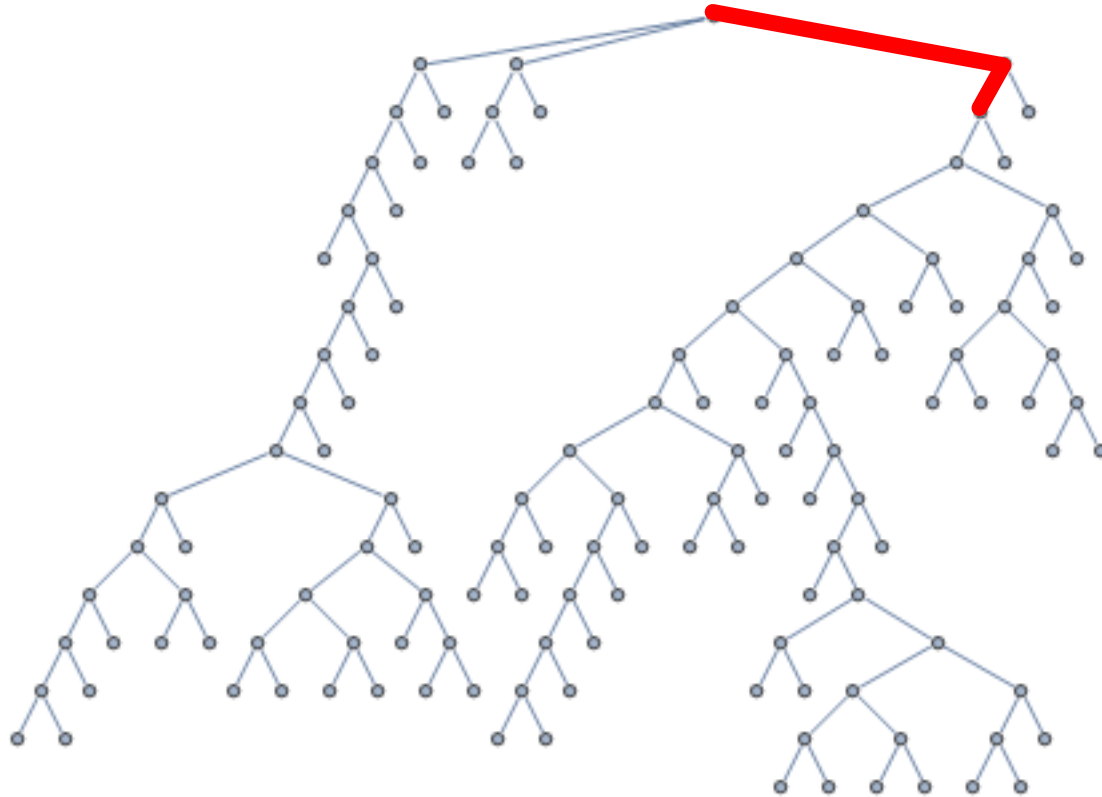
# BFS vs. DFS



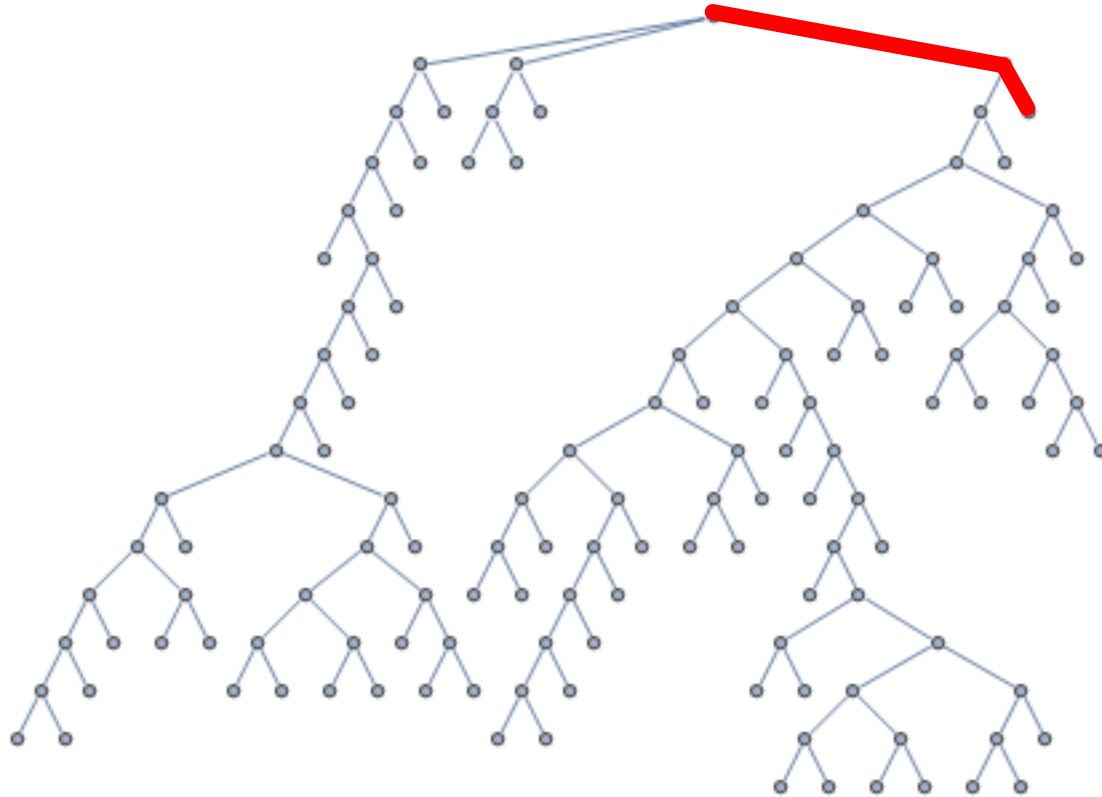
# BFS vs. DFS



# BFS vs. DFS

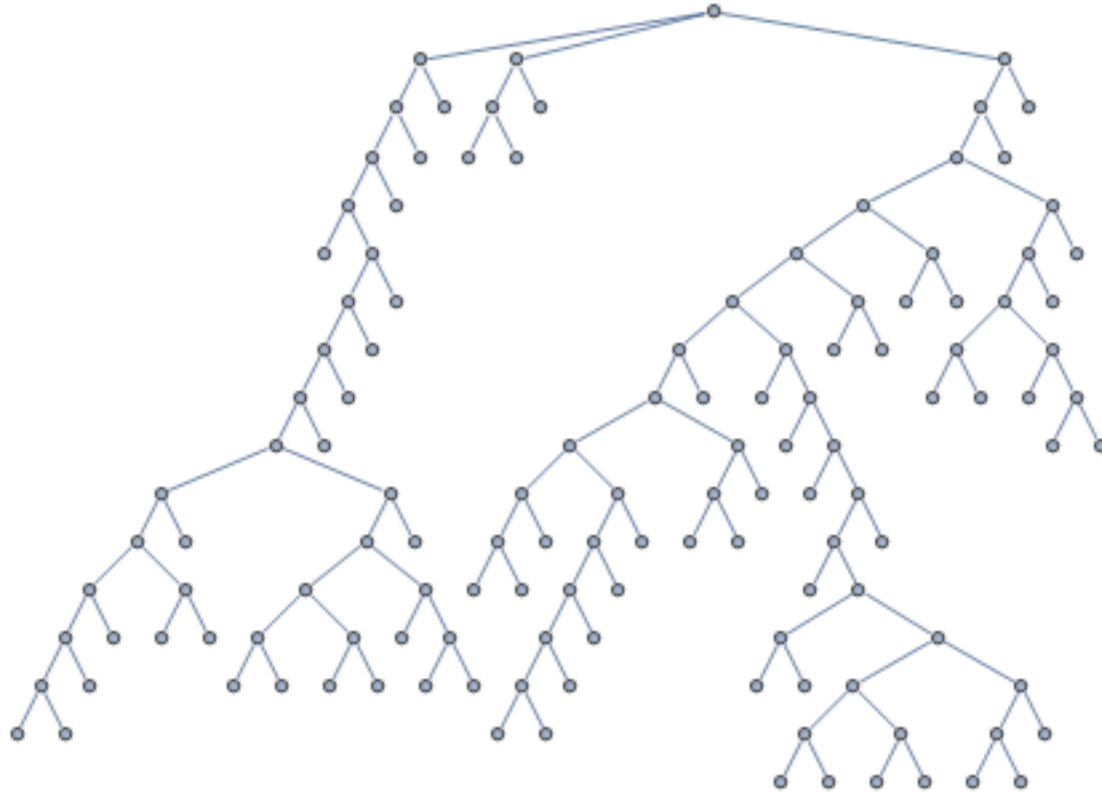


# BFS vs. DFS

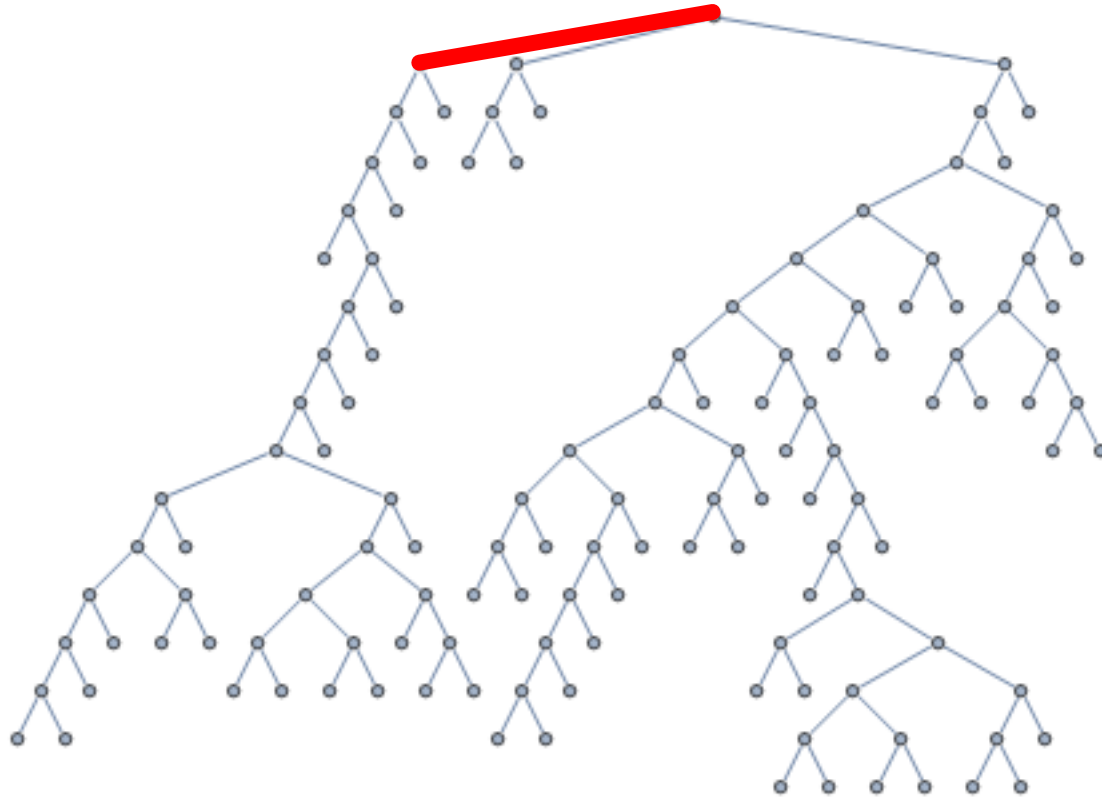




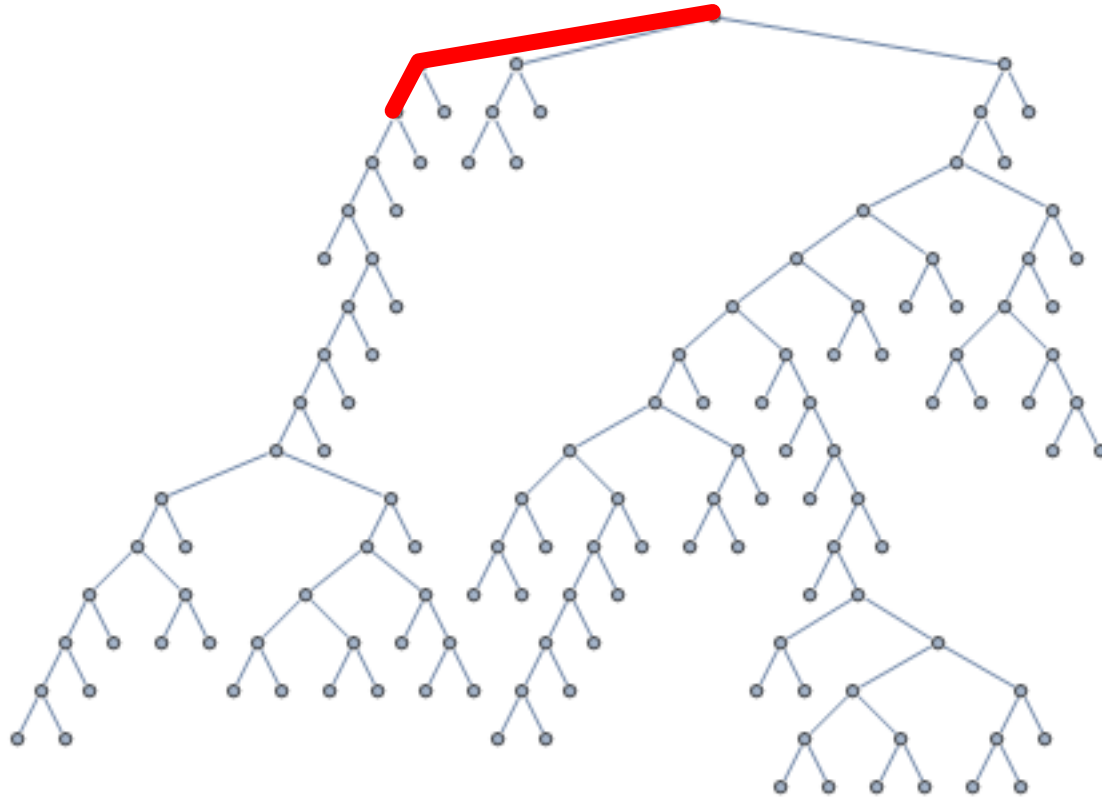
# BFS vs. DFS



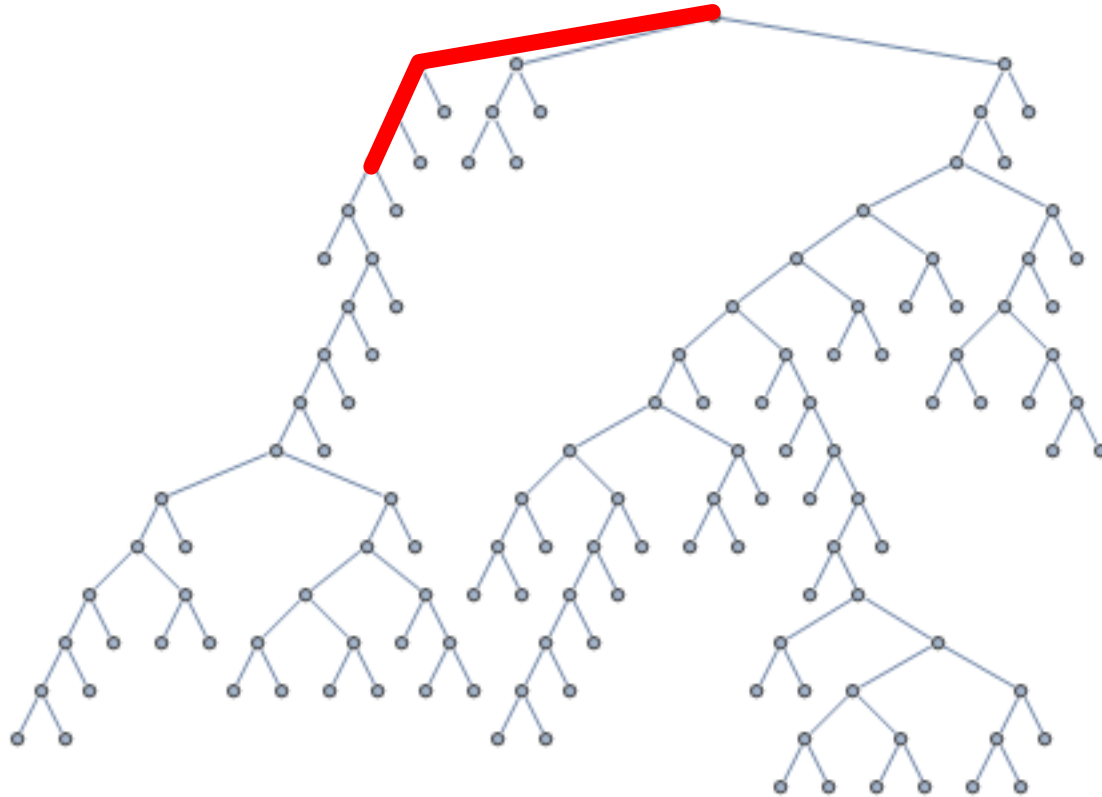
# BFS vs. DFS



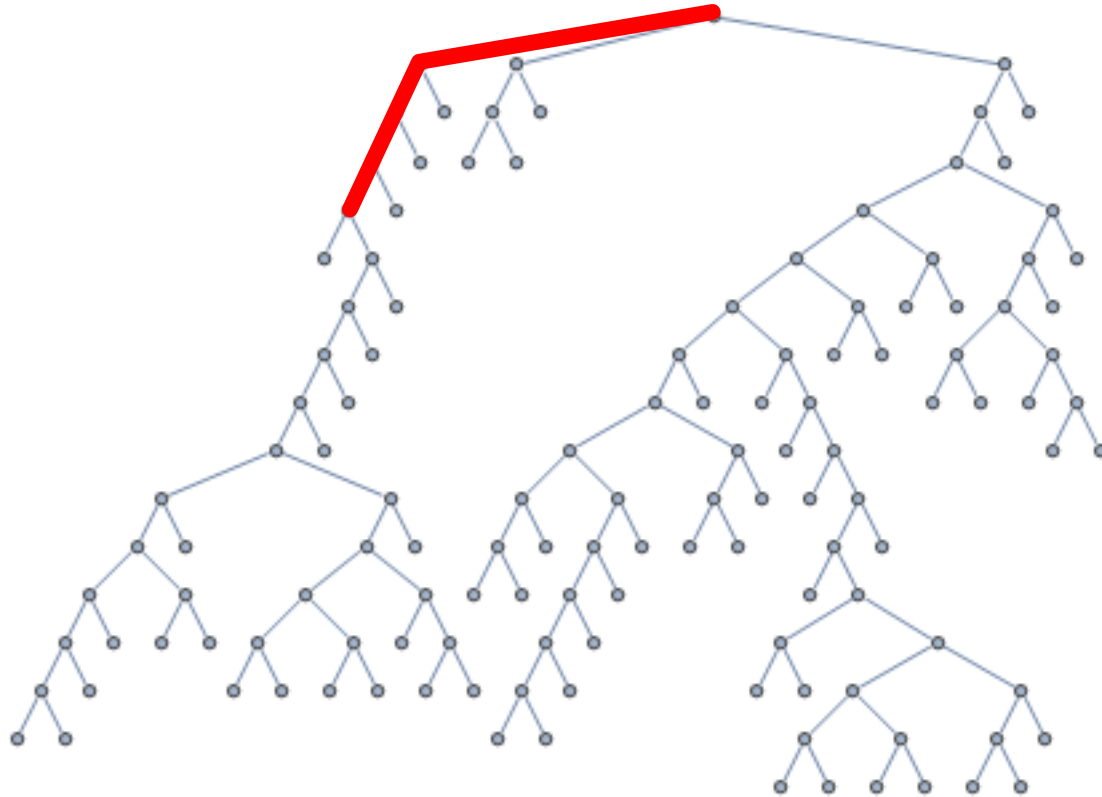
# BFS vs. DFS



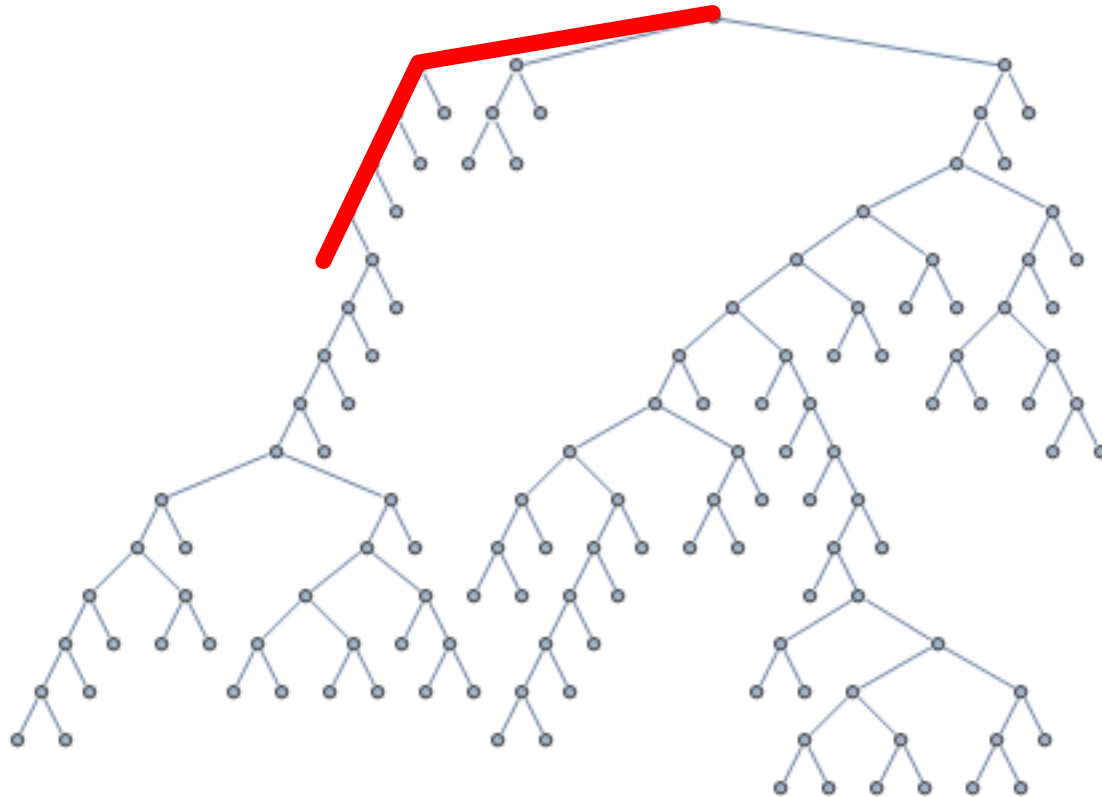
# BFS vs. DFS



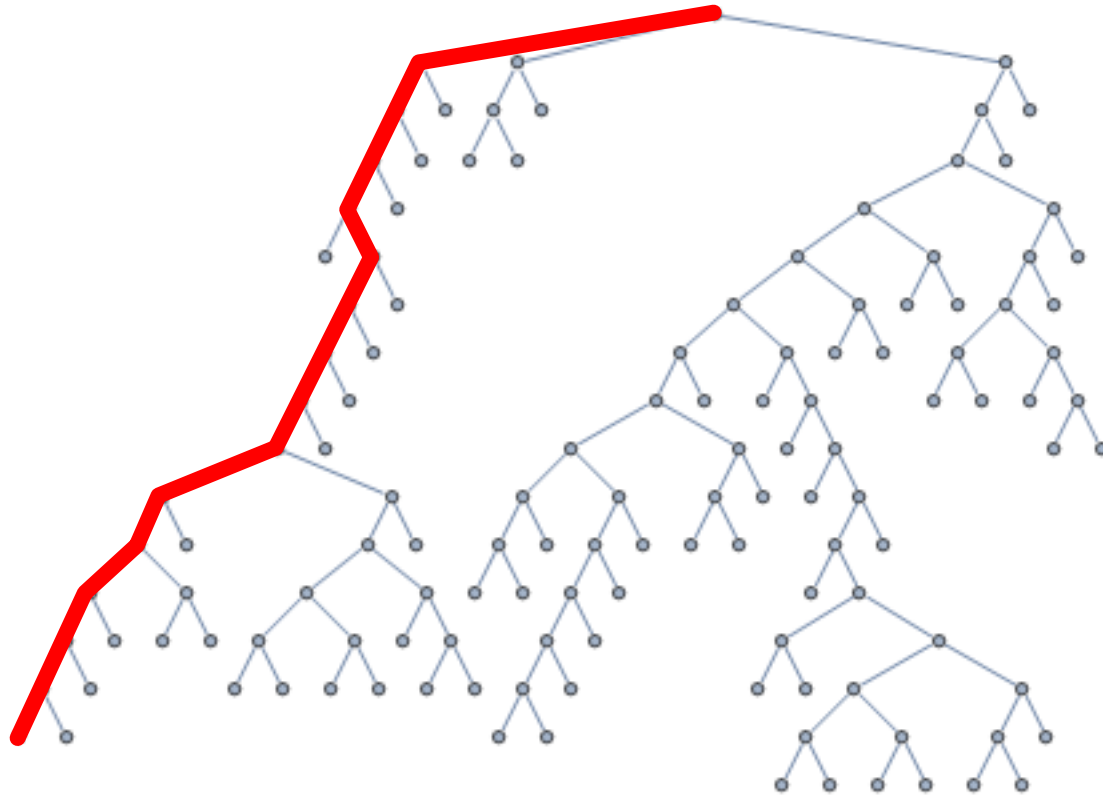
# BFS vs. DFS



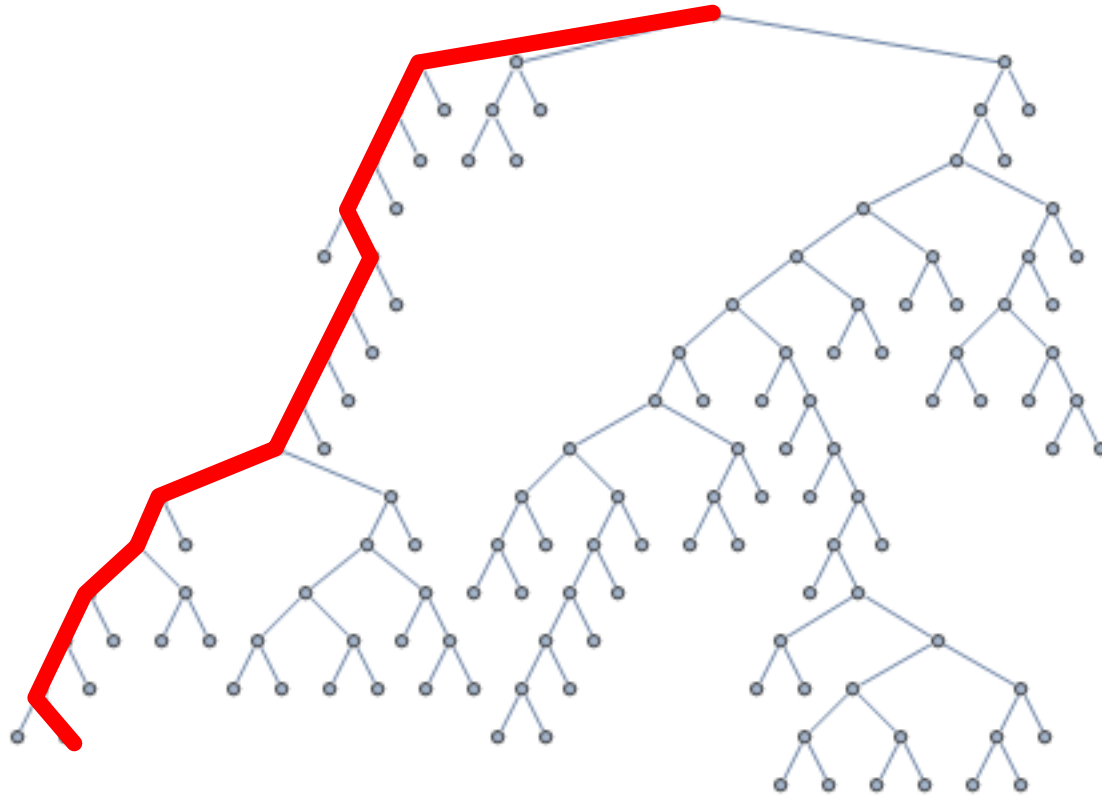
# BFS vs. DFS



# BFS vs. DFS

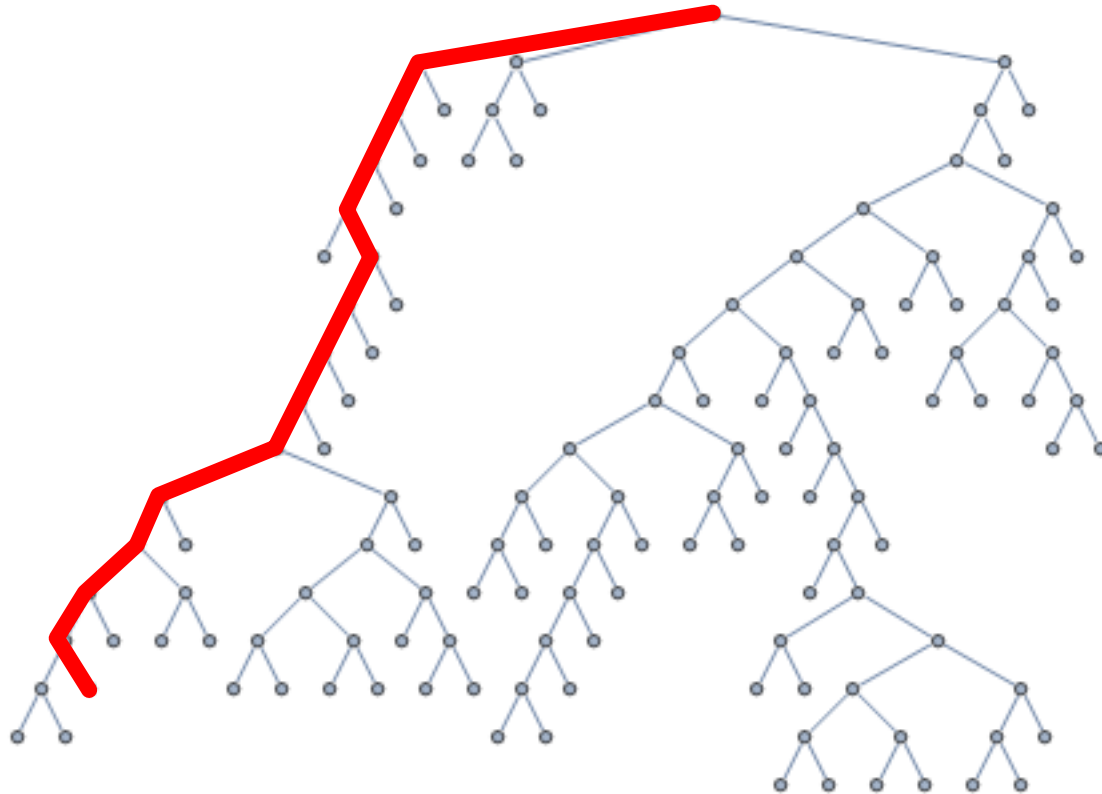


# BFS vs. DFS

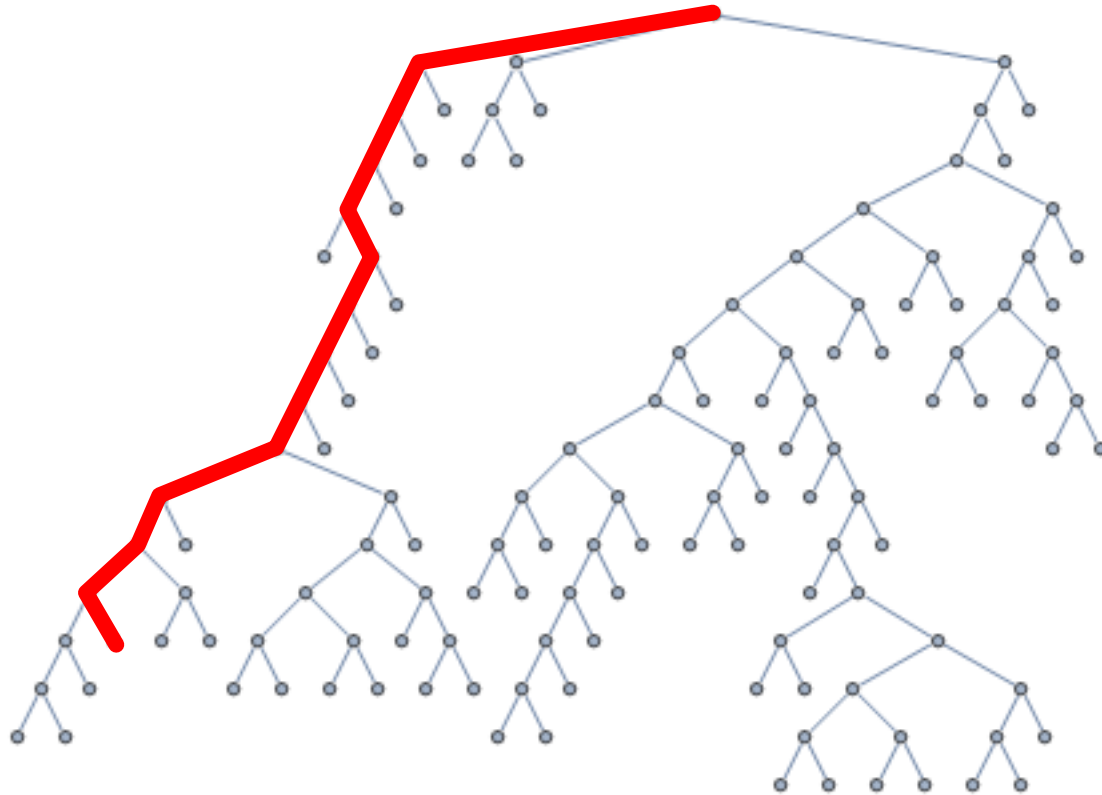




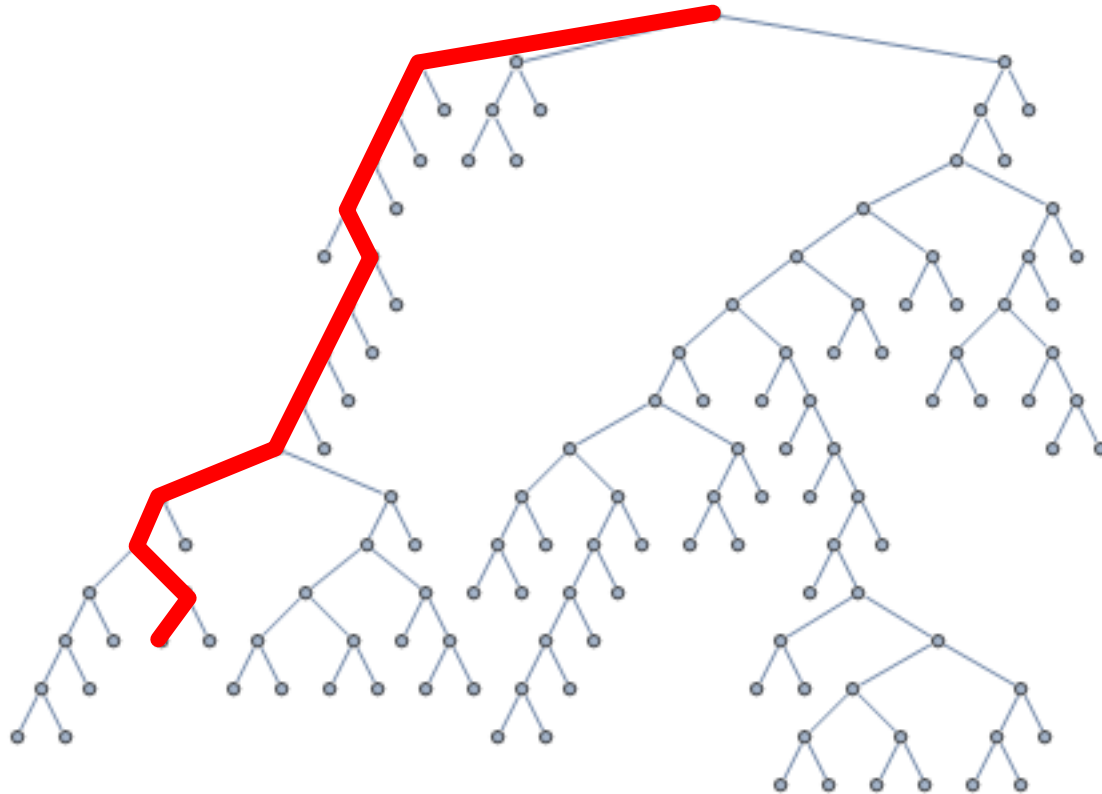
# BFS vs. DFS



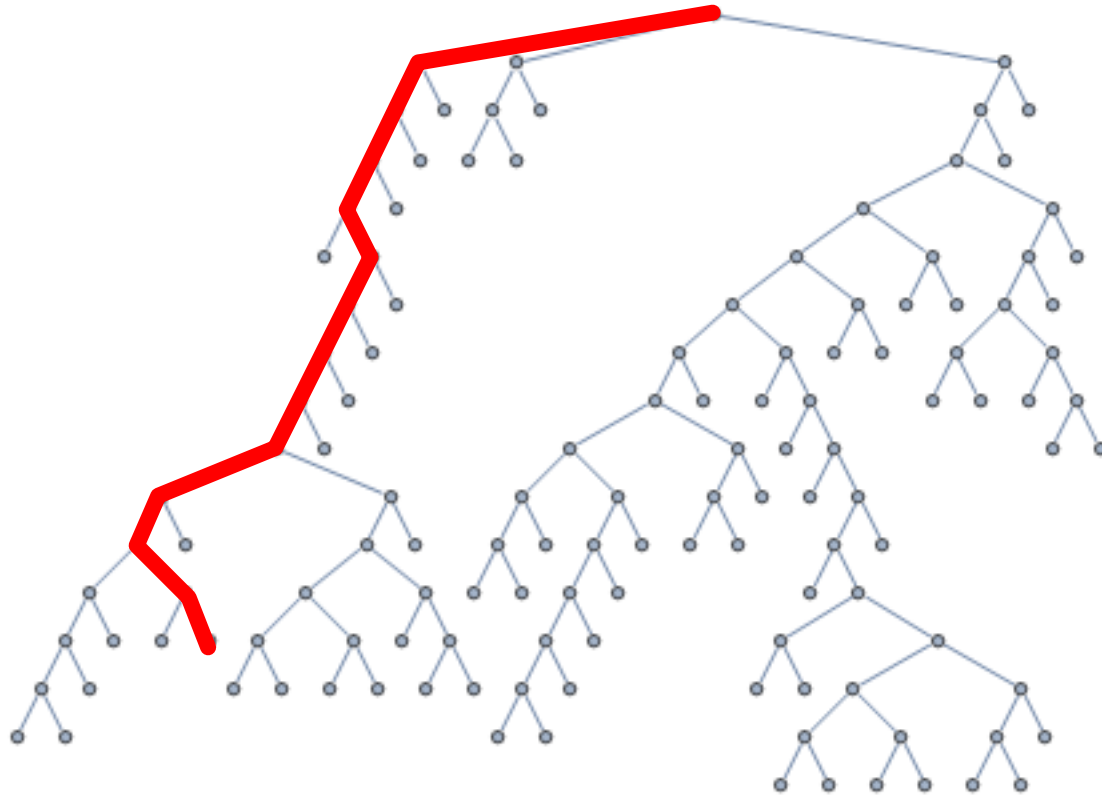
# BFS vs. DFS



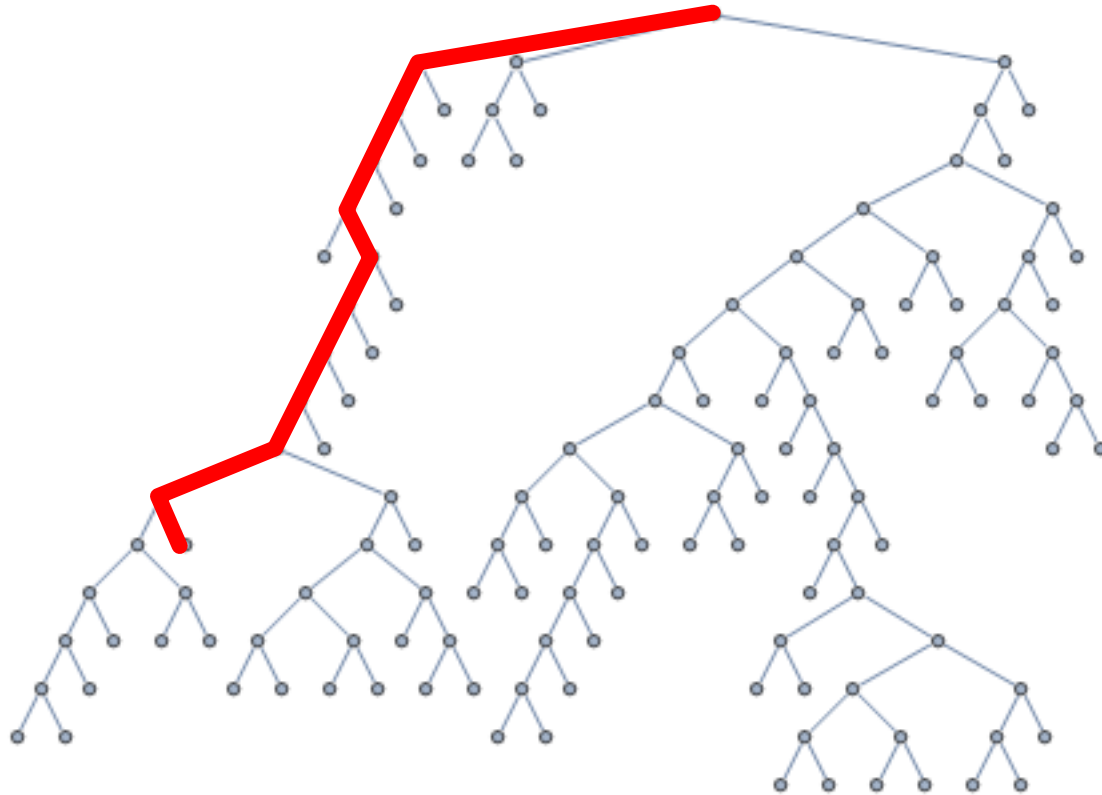
# BFS vs. DFS



# BFS vs. DFS



# BFS vs. DFS



# Analysis of search strategies

- Strategies are evaluated along the following criteria:
  - **Completeness:** does it always find a solution if one exists?
  - **Optimality:** does it always find a least-cost solution?
  - **Time complexity:** number of nodes generated
  - **Space complexity:** maximum number of nodes in memory
- Time and space complexity are measured in terms of
  - **$b$ :** maximum branching factor of the search tree
  - **$d$ :** depth of the optimal solution
  - **$m$ :** maximum length of any path in the state space (may be infinite)

# Properties of breadth-first search

- **Complete?**

Yes (if branching factor  $b$  is finite)

- **Optimal?**

Yes – if cost = 1 per step

- **Time?**

Number of nodes in a  $b$ -ary tree of depth  $d$ :  $O(b^d)$   
( $d$  is the depth of the optimal solution)

- **Space?**

$O(b^d)$

- Space is the bigger problem (more than time)

# Properties of depth-first search

- **Complete?**

Fails in infinite-depth spaces, spaces with loops

Modify to avoid repeated states along path

→ complete in finite spaces

- **Optimal?**

No – returns the first solution it finds

- **Time?**

Could be the time to reach a solution at maximum depth  $m$ :  $O(b^m)$

Terrible if  $m$  is much larger than  $d$

But if there are lots of solutions, may be much faster than BFS

- **Space?**

$O(bm)$ , i.e., linear space!



# Iterative deepening search

- Use DFS as a subroutine
  1. Check the root
  2. Do a DFS searching for a path of length 1
  3. If there is no path of length 1, do a DFS searching for a path of length 2
  4. If there is no path of length 2, do a DFS searching for a path of length 3...

# Iterative deepening search

Limit = 0



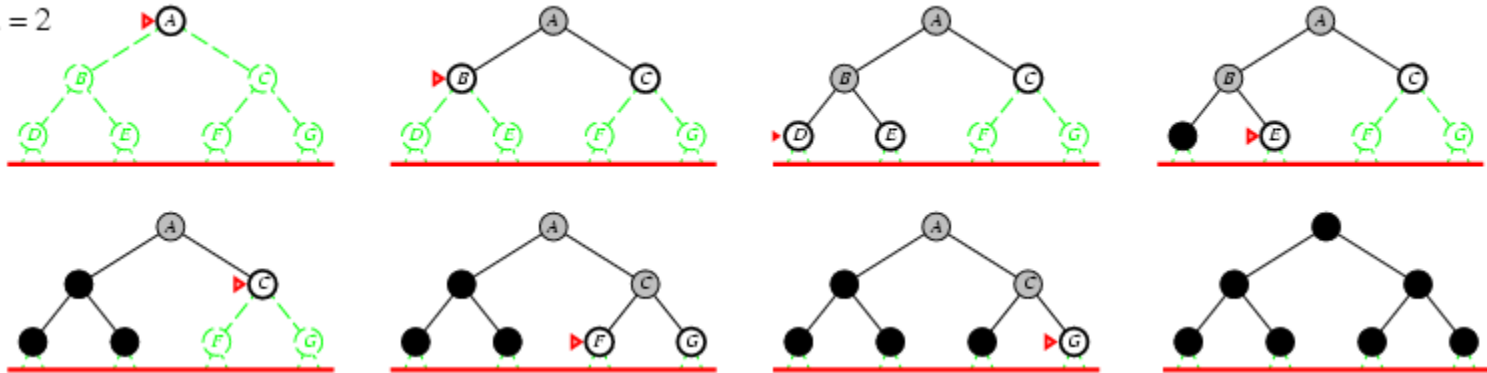
# Iterative deepening search

Limit = 1



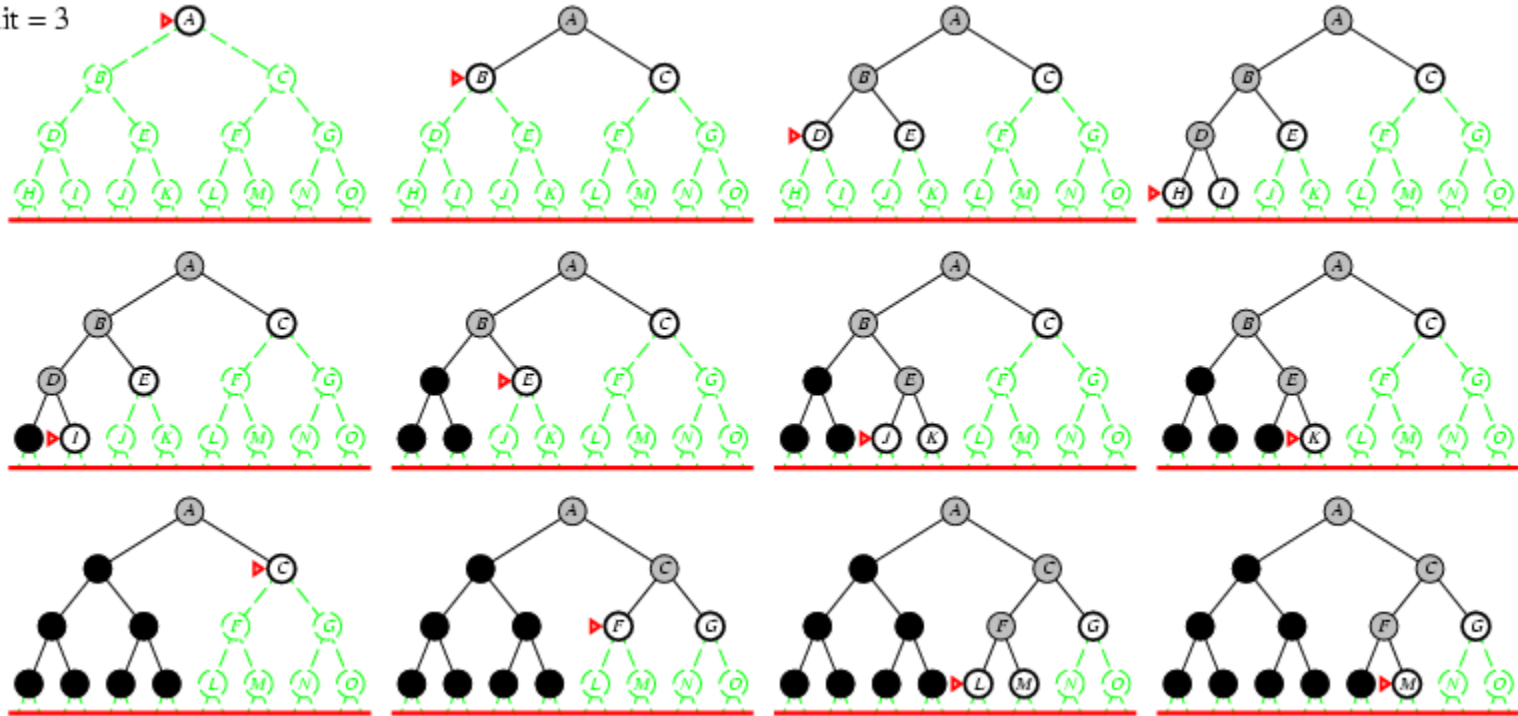
# Iterative deepening search

Limit = 2



# Iterative deepening search

Limit = 3



# Properties of iterative deepening search

- **Complete?**

Yes

- **Optimal?**

Yes, if step cost = 1

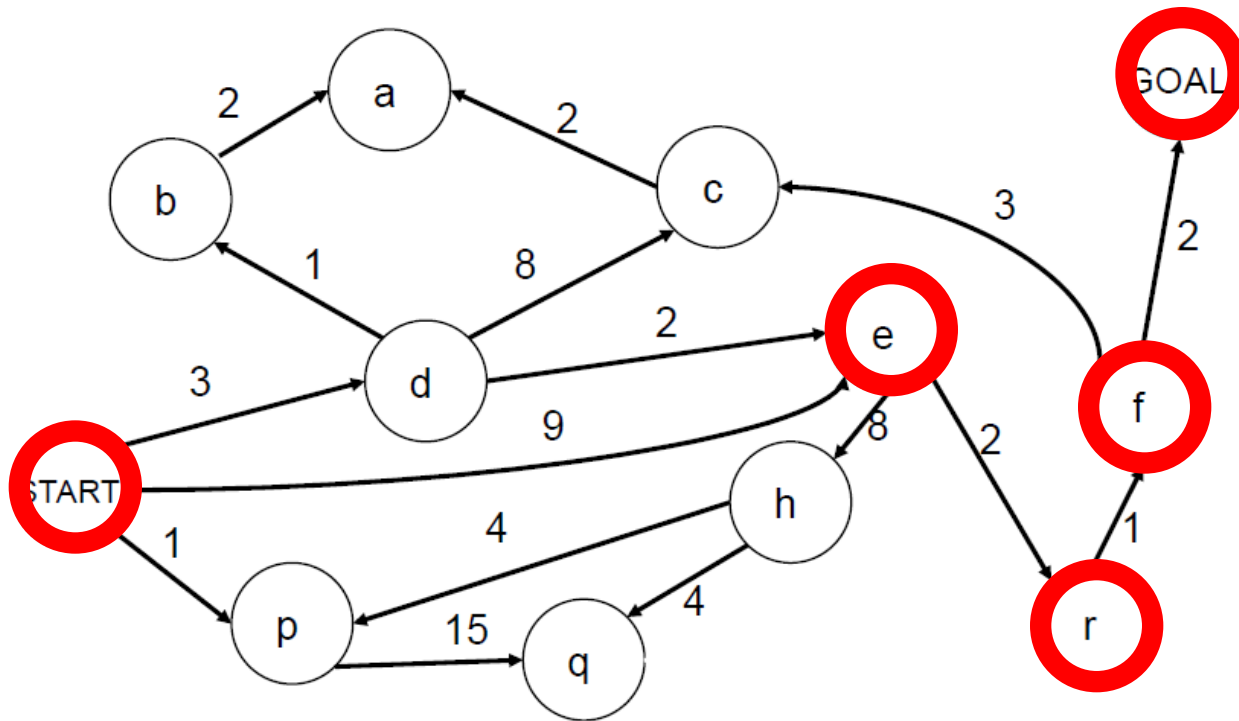
- **Time?**

$$(d+1)b^0 + d b^1 + (d-1)b^2 + \dots + b^d$$

- **Space?**

$$O(bd)$$

# Search with varying step costs



- BFS finds the path with the fewest steps, but does not always find the cheapest path

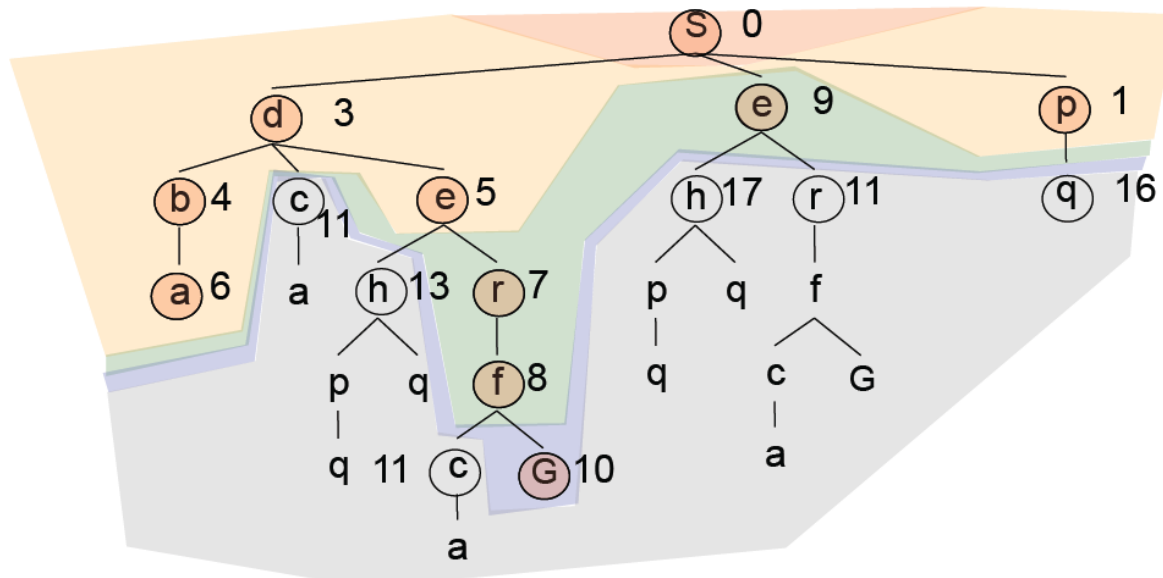
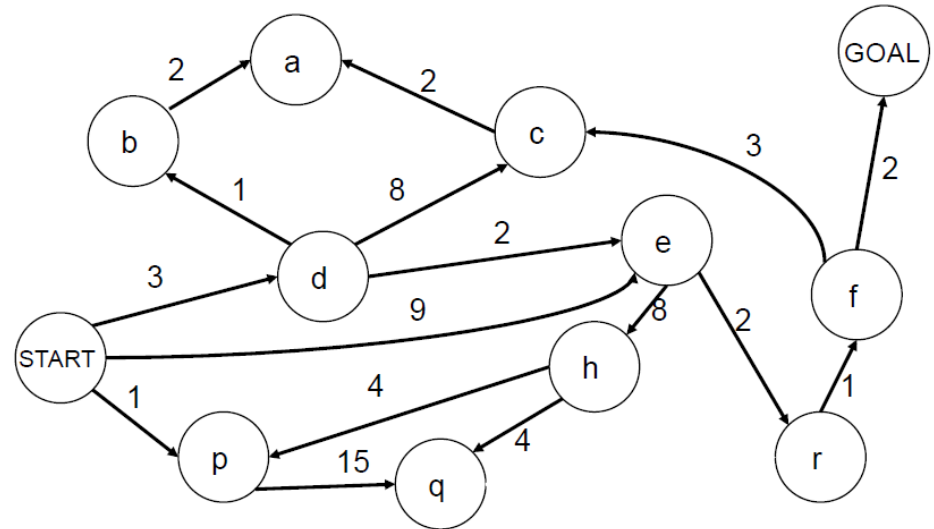
# Uniform-cost search

- For each frontier node, save the total cost of the path from the initial state to that node
- Expand the frontier node with the lowest path cost
- Implementation: *frontier* is a priority queue ordered by path cost
- Equivalent to BFS if step costs all equal
- Equivalent to Dijkstra's algorithm in general



# Uniform-cost search example

- Expansion order:  
(S,p,d,b,e,a,r,f,e,G)



# Properties of uniform-cost search

- **Complete?**

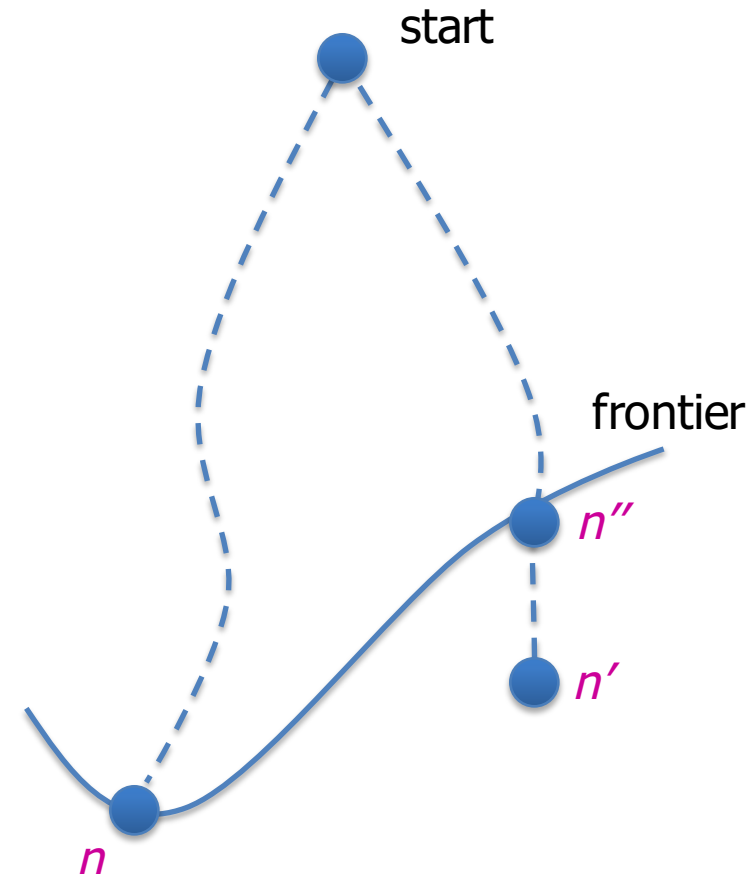
Yes, if step cost is greater than some positive constant  $\epsilon$   
(we don't want infinite sequences of steps that have a finite total cost)

- **Optimal?**

Yes

# Optimality of uniform-cost search

- **Graph separation property**: every path from the initial state to an unexplored state has to pass through a state on the frontier
  - Proved inductively
- Optimality of UCS: proof by contradiction
  - Suppose UCS terminates at goal state  $n$  with path cost  $g(n)$  but there exists another goal state  $n'$  with  $g(n') < g(n)$
  - By the graph separation property, there must exist a node  $n''$  on the frontier that is on the optimal path to  $n'$
  - But because  $g(n'') \leq g(n') < g(n)$ ,  $n''$  should have been expanded first!



# Properties of uniform-cost search

- **Complete?**

Yes, if step cost is greater than some positive constant  $\epsilon$  (we don't want infinite sequences of steps that have a finite total cost)

- **Optimal?**

Yes – nodes expanded in increasing order of path cost

- **Time?**

Number of nodes with path cost  $\leq$  cost of optimal solution ( $C^*$ ),  
 $O(b^{C^*/\epsilon})$

This can be greater than  $O(b^d)$ : the search can explore long paths consisting of small steps before exploring shorter paths consisting of larger steps

- **Space?**

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

# Review: Uninformed search strategies

Algorithm	Complete?	Optimal?	Time complexity	Space complexity
<b>BFS</b>	Yes	If all step costs are equal	$O(b^d)$	$O(b^d)$
<b>DFS</b>	No	No	$O(b^m)$	$O(bm)$
<b>IDS</b>	Yes	If all step costs are equal	$O(b^d)$	$O(bd)$
<b>UCS</b>	Yes	Yes	Number of nodes with $g(n) \leq C^*$	

b: maximum branching factor of the search tree

d: depth of the optimal solution

m: maximum length of any path in the state space

$C^*$ : cost of optimal solution

$g(n)$ : cost of path from start state to node  $n$

# Attribution

Slides developed by Svetlana Lazebnik based on content from Stuart Russell and Peter Norvig, [Artificial Intelligence: A Modern Approach](#), 3rd edition