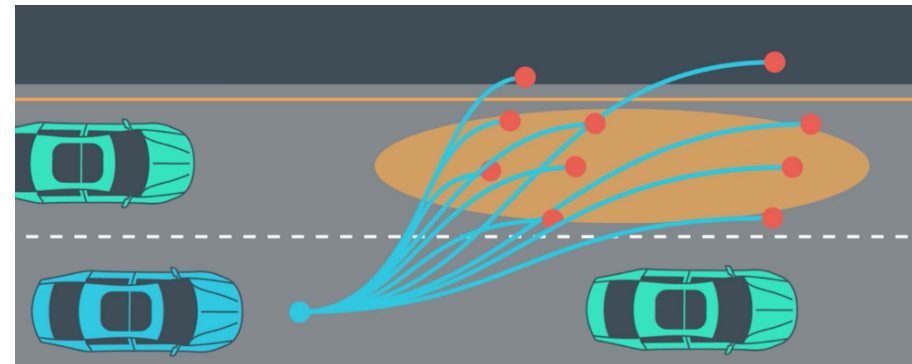
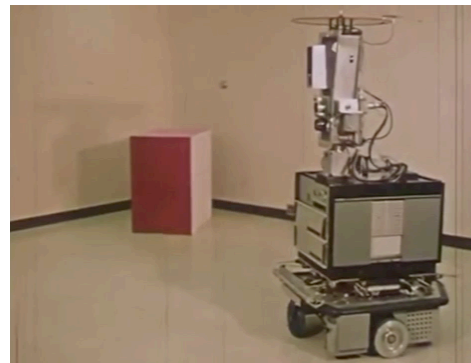
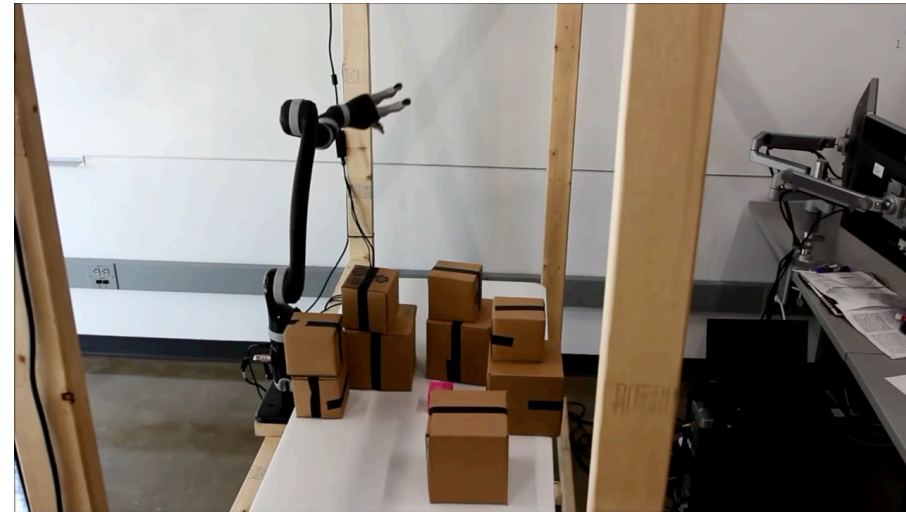


# Uninformed Search

COMP3411/9814: Artificial Intelligence

# When is Search Needed?

- Motion Planning
- Navigation
- Speech and Natural Language
- Task Planning
- Machine Learning
- Game Playing



# Search Methods

- Uninformed search
  - use no problem-specific information
  - Uninformed (or “blind”) search strategies use only the information available in the problem definition (can only distinguish a goal from a non-goal state)
- Informed search
  - use heuristics to improve efficiency
  - Informed (or “heuristic”) search strategies use task-specific knowledge.

# Overview

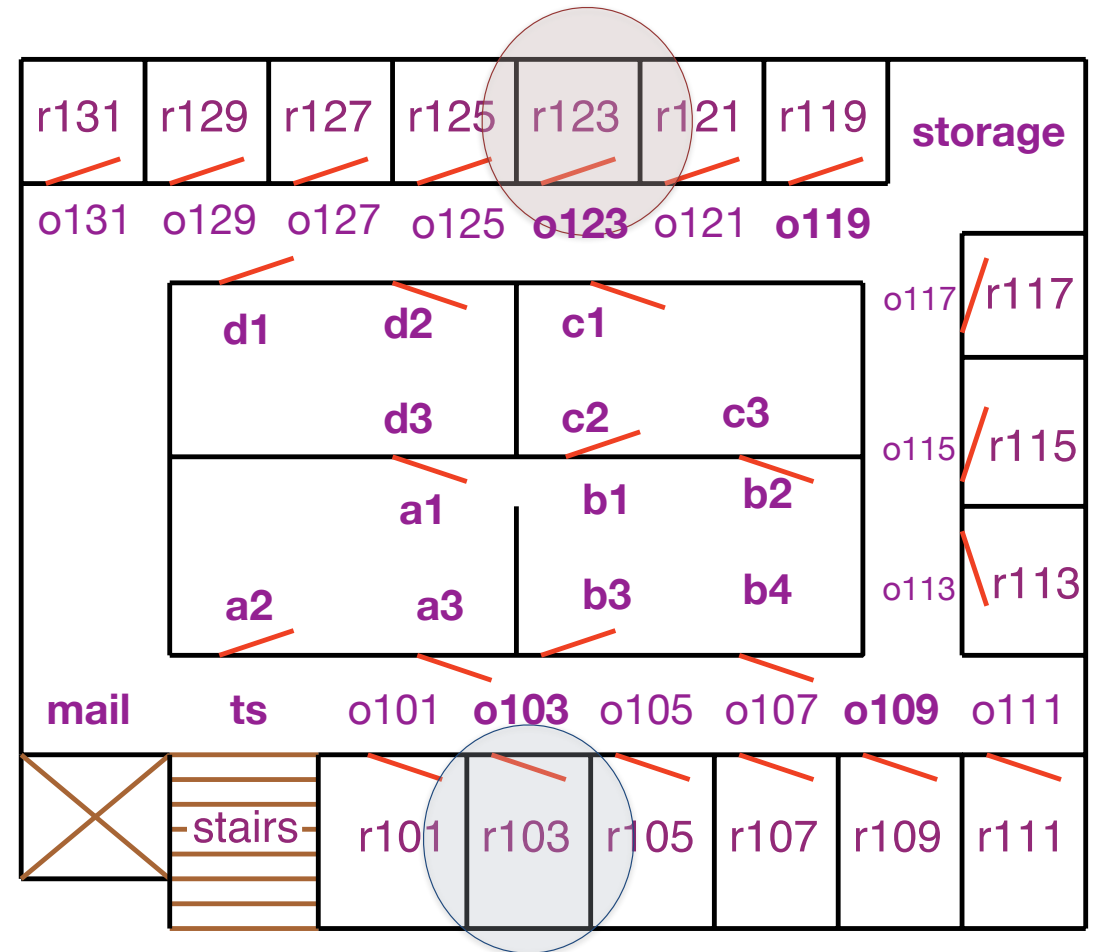
- Basic search algorithms
  - Breadth First Search
  - Depth First Search
  - Uniform Cost Search
  - Depth Limited Search
  - Iterative Deepening Search
  - Bidirectional Search

# State Space Search Problems

- **State space** — set of all states reachable from initial state(s) by any action sequence
- **Initial state(s)** — element(s) of the state space
- Transitions
  - **Operators** — set of possible actions at agent's disposal; describe state reached after performing action in current state, or
  - **Successor function** —  $s(x)$ = set of states reachable from state  $x$  by performing a single action
- **Goal state(s)** — element(s) of the state space
- **Path cost** — cost of a sequence of transitions used to evaluate solutions (applies to optimisation problems)

# Delivery Robot

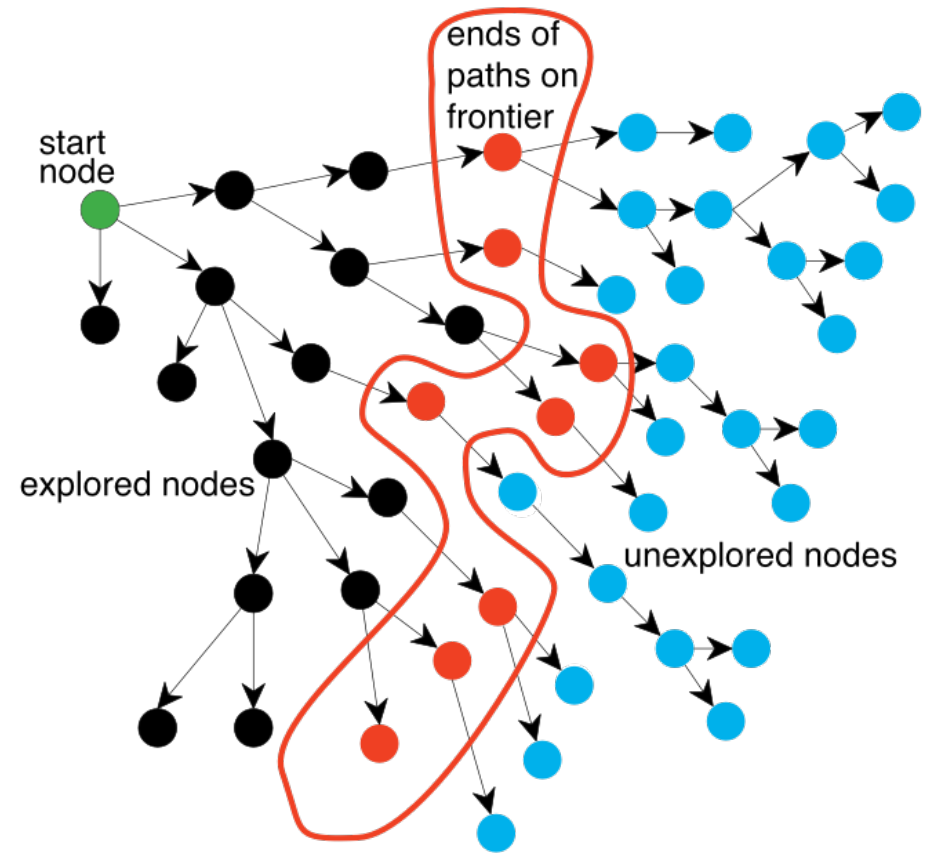
- The robot wants to get from outside room 103 to the inside of room 123.
- The only way a robot can get through a doorway is to push the door open in the direction shown.
- The task is to find a path from o103 to to r123





# Problem Solving by Graph Searching

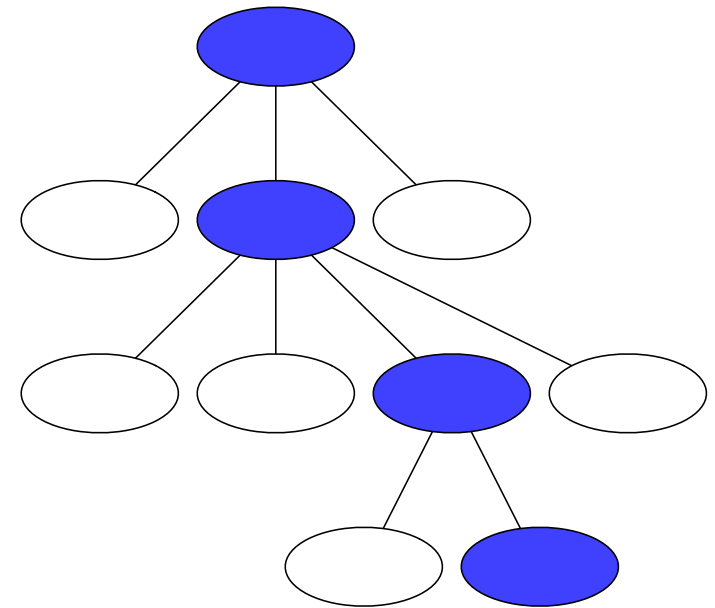
**Search strategy differ in the way they expand the frontier**



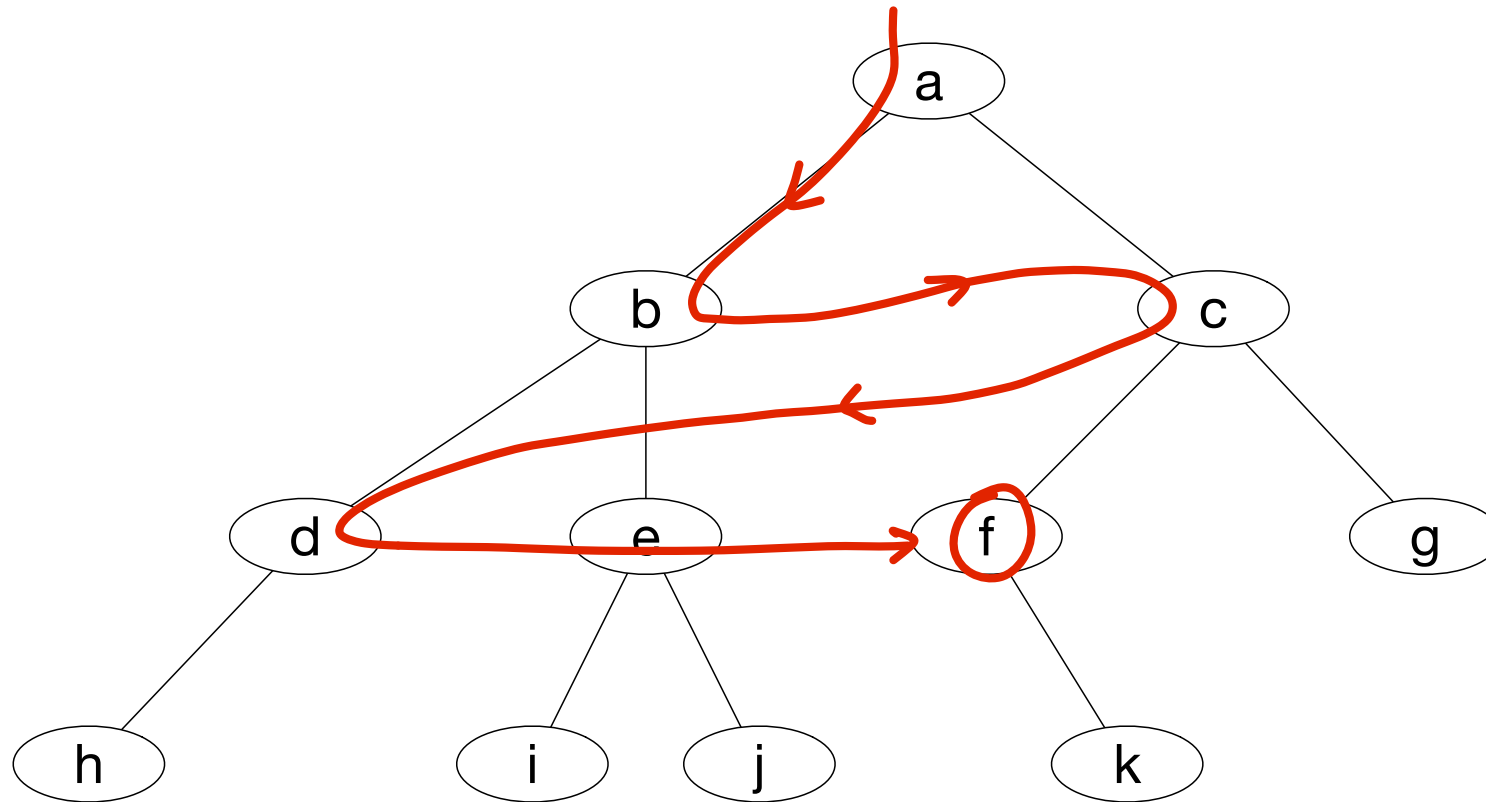


# Search Tree

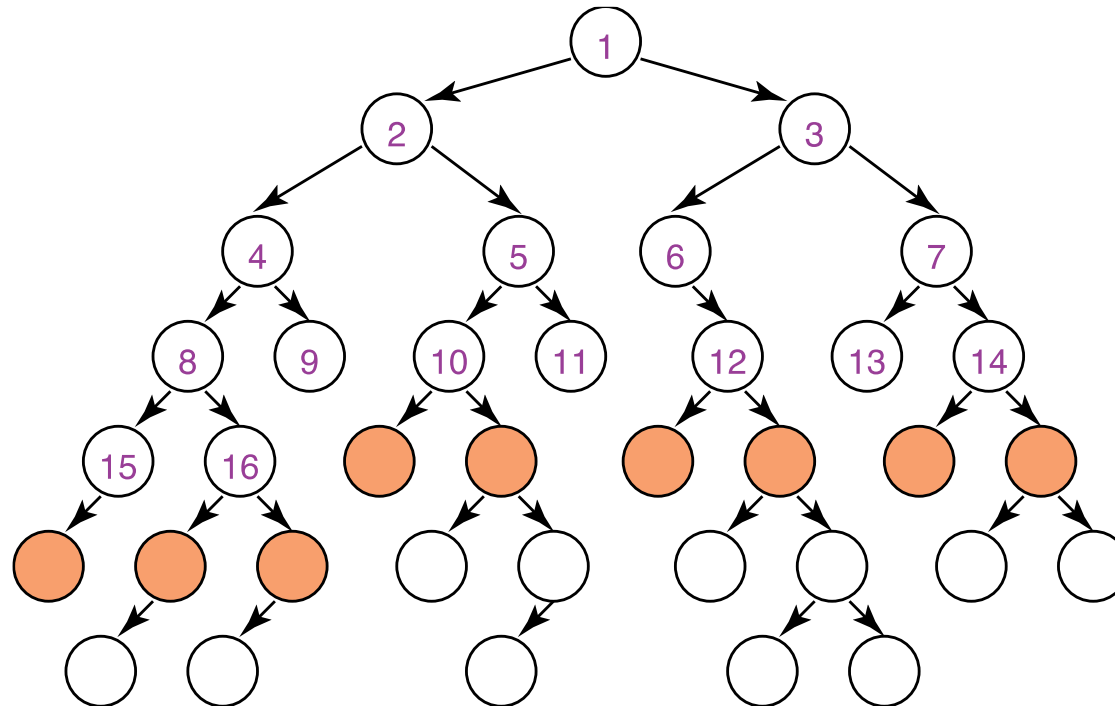
- **Search tree:** superimposed over the state space.
- **Root:** search node corresponding to the initial state.
- **Leaf nodes:** correspond to states that have no successors in the tree because they were not expanded or generated no new nodes.



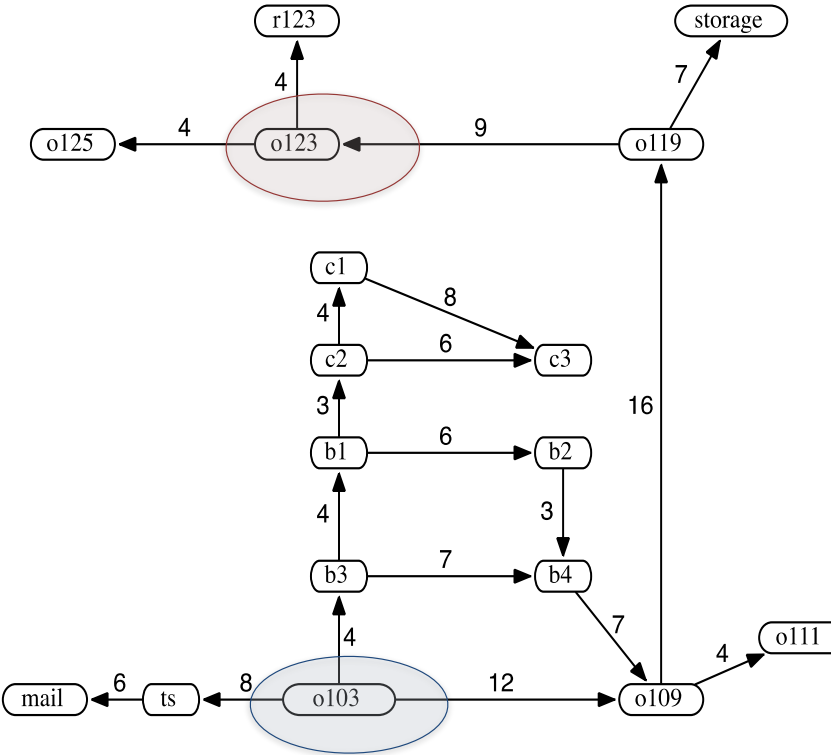
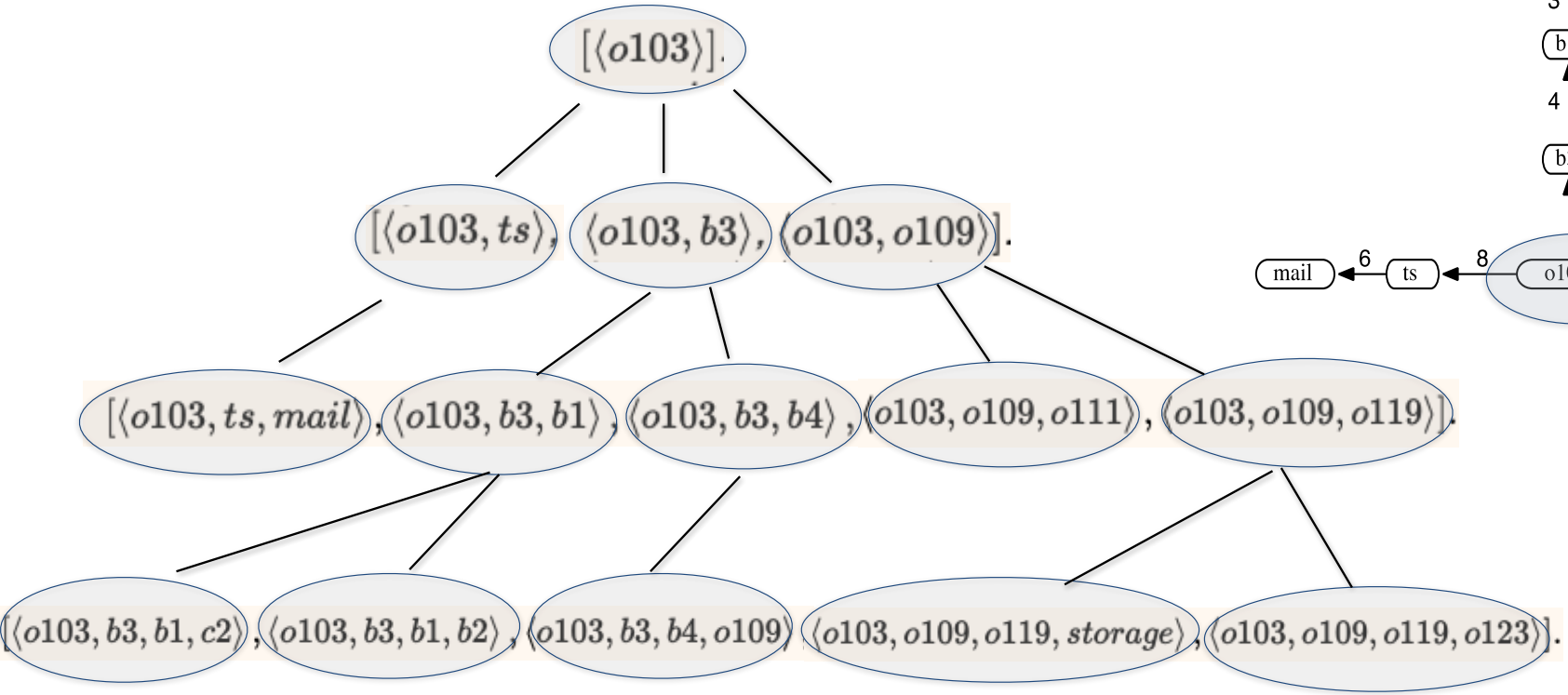
# Breadth-First Search



# Breadth-first Search Frontier



# Breadth-First Search



After each iteration, each path on the frontier has either the same number of arcs

# Breadth-first Search

- Breadth-first search treats the frontier as a queue
- It selects the first element in the queue to explore next
- If the list of paths on the frontier is  $[p_1, p_2, \dots, p_r]$ :
  - $p_1$  is selected. Its neighbours are added to the end of the queue, after  $p_r$ .
  - $p_2$  is selected next.

# Breadth-First Search

- All nodes are expanded at a same depth in the tree before any nodes at the next level are expanded
- Can be implemented by using a queue to store frontier nodes
  - put newly generated successors at end of queue
- Stop when node with goal state is reached
- Include check that state has not already been explored
  - Needs a new data structure for set of explored states
- Finds the shallowest goal first

# Complexity of Breadth-first Search

- Does breadth-first search guarantee finding the shortest path?
- What happens on infinite graphs or on graphs with cycles if there is a solution?
- What is the time complexity as a function of the length of the path selected?
- What is the space complexity as a function of the length of the path selected?
- How does the goal affect the search?

# Properties of breadth-first search

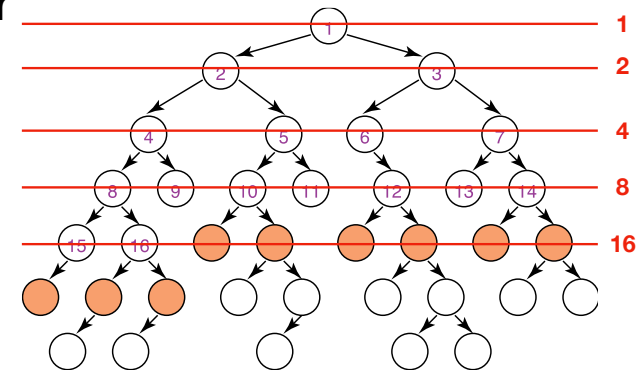
**Complete?** Yes (if breadth,  $b$ , is finite, the shallowest goal is at a fixed depth,  $d$ , and will be found before any deeper nodes are generated)

**Time?**  $1 + b^2 + b^3 + \dots + b^d = \frac{b^{d+1} - 1}{b - 1} = O(b^d)$

**Space?**  $O(b^d)$  (keeps every node in memory; generate all nodes up to level  $d$ )

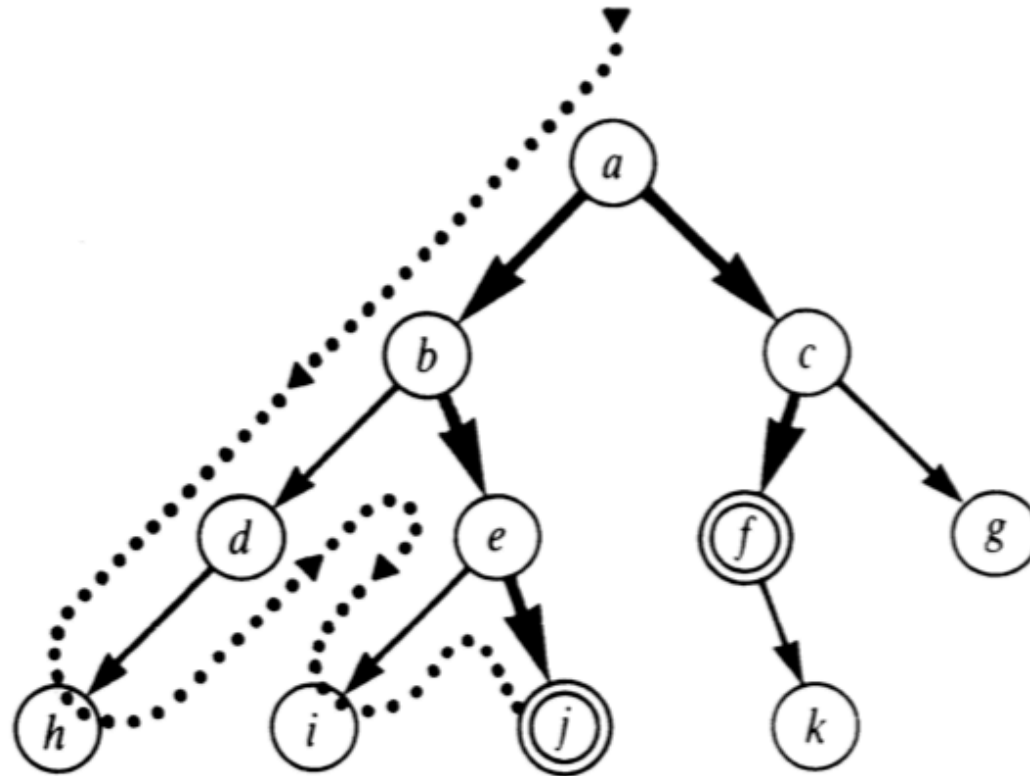
**Optimal?** Yes, but only if all actions have the same cost

**Space** is the big problem for BFS. It grows **exponentially** with depth





# Depth-first Search - DFS

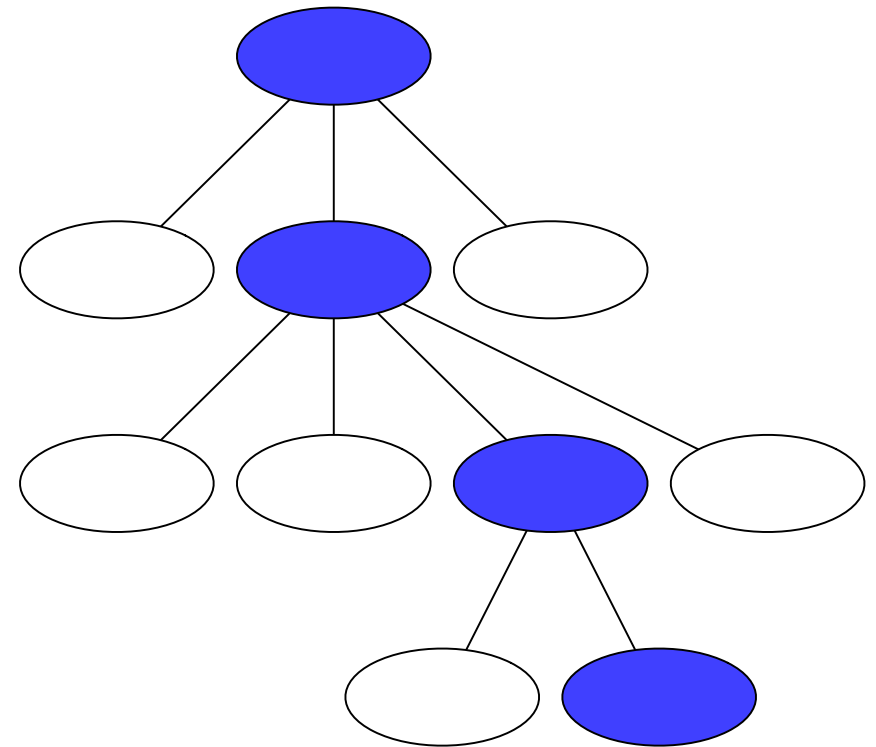


# Depth First Search

- Expand one node at the deepest level reached so far
- Implementation:
  - Implement the frontier as a stack, i.e. insert newly generated states at the front of the open list (frontier)
  - Can be implemented by recursive function calls, where call stack maintains open list
- In depth-first search, like breadth-first, the order in which the paths are expanded does not depend on the goal.

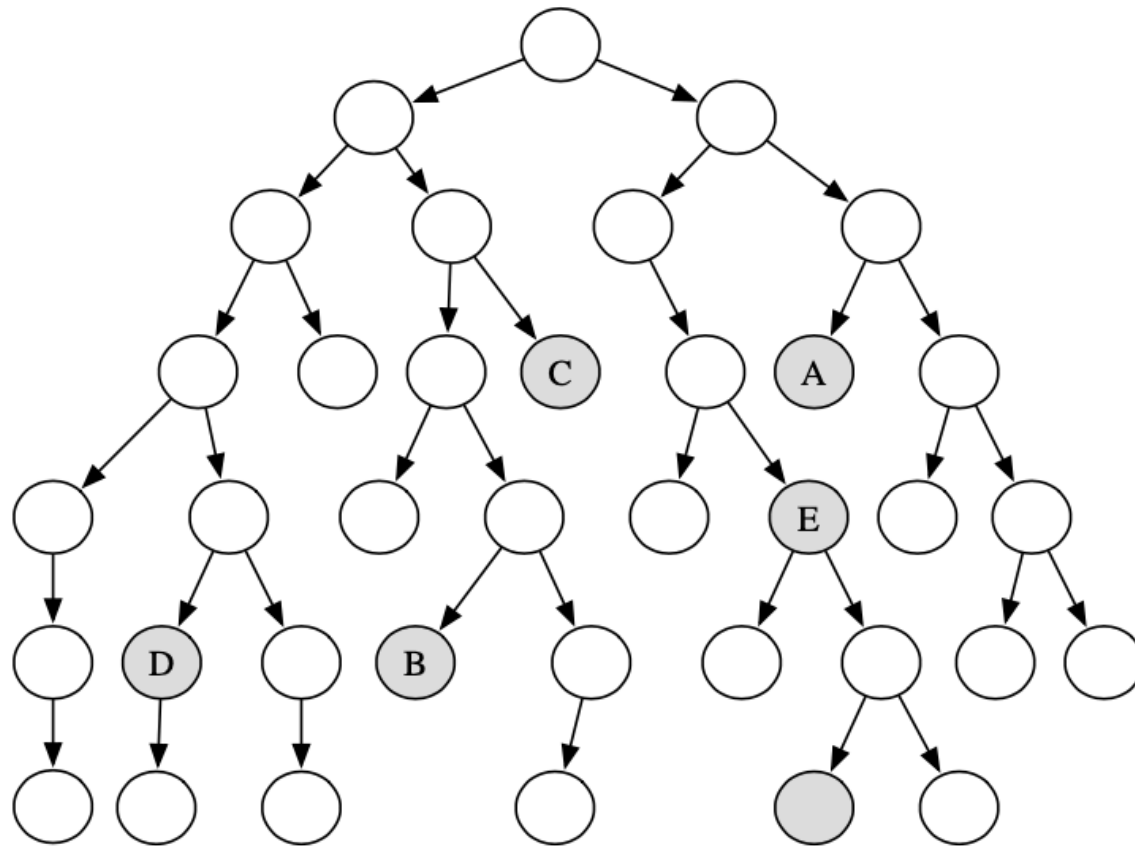
# Depth First Search

- At any point depth-first search stores single path from root to leaf, together with any remaining unexpanded siblings of nodes along path
- Stop when node with goal state is expanded
- Include check that state has not already been explored along a path – cycle checking





**Which goal (shaded) will depth-first search find first?**





# Depth-First Search Analysis

- In cases where problem has many solutions, depth-first search may outperform breadth-first search because there is a good chance it will find a solution after exploring only a small part of the space
- However, depth-first search may get stuck following a deep or infinite path even when a solution exists at a relatively shallow level
- Therefore, depth-first search is not complete and not optimal
  - Avoid depth-first search for problems with deep or infinite path

# Lowest-cost-first Search

## Uniform-Cost Search

- Sometimes transitions have a cost
- Cost of a path is the sum of the costs of its arcs:

$$\text{cost}(\langle n_0, \dots, n_k \rangle) = \sum_{i=1}^k \text{cost}(\langle n_{i-1}, n_i \rangle)$$

- An optimal solution has minimum cost
- **Delivery robot example:**
  - cost of arc may be resources (e.g., time, energy) required to execute action represented by the arc
  - aim is to reach goal using least resources



# Lowest-cost-first Search

## Uniform-Cost Search

- The simplest search method that is guaranteed to find a minimum cost path is **lowest-cost-first** search or **uniform-cost search**
- similar to breadth-first search, but instead of expanding path with least number of arcs, select path with lowest cost
- implemented by treating the frontier as a priority queue ordered by the cost function

$$cost(\langle n_0, \dots, n_k \rangle) = \sum_{i=1}^k cost(\langle n_{i-1}, n_i \rangle)$$

# Lowest-Cost Search for Delivery Robot

- Edges are labelled with cost
- e.g. distance to travel
- Sort queue by increasing cost of path to the node

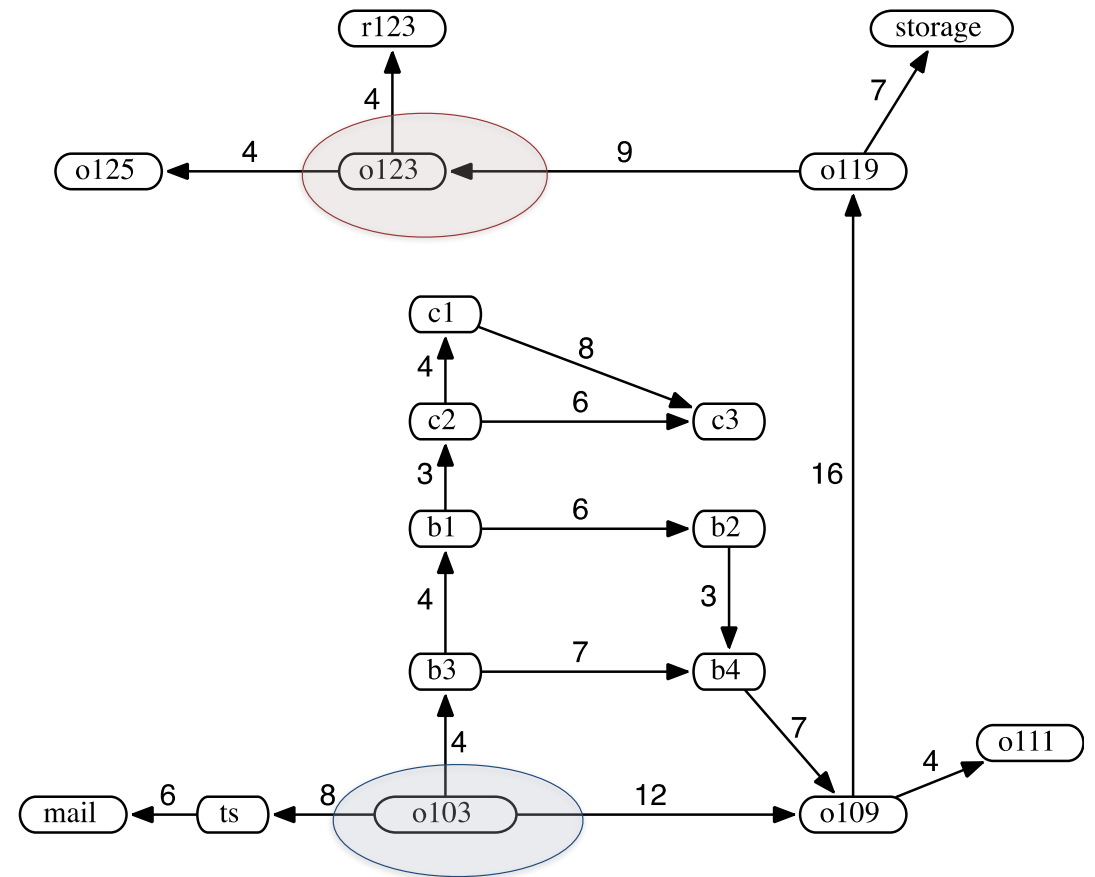
$[o103_0]$

$[b3_4, ts_8, o109_{12}]$

$[b1_8, ts_8, b4_{11}, o109_{12}]$

$[ts_8, c2_{11}, b4_{11}, o109_{12}, b2_{14}]$

$[c2_{11}, b4_{11}, o109_{12}, mail_{14}, b2_{14}]$



# Uniform-Cost Search

- Expand root first, then expand least-cost unexpanded node
- Implementation with priority queue
  - insert nodes in order of increasing path cost - lowest path cost is  $g(n)$ .
- Reduces to breadth-first search when all actions have same cost
- Finds the cheapest goal provided path cost is monotonically increasing along each path (i.e. no negative-cost steps)

# Properties of Uniform-Cost Search

**Complete?** Yes, if  $b$  is finite and if transition  $cost \geq \epsilon$  with  $\epsilon > 0$

**Time?** Worst case,  $O(b^{\lceil C^*/\epsilon \rceil})$  where  $C^*$  = cost of the optimal solution  
every transition costs at least  $\epsilon$   
 $\therefore$  cost per step is  $\frac{C^*}{\epsilon}$

**Space?**  $O(b^{\lceil C^*/\epsilon \rceil})$ ,  $b^{\lceil C^*/\epsilon \rceil} = b^d$  if all step costs are equal

**Optimal?** Yes – nodes expanded in increasing order of  $g(n)$

# Summary of Search Strategies

Strategy	Frontier Selection	Complete	Halts	Space
Depth-first	Last node added	No	No	Linear
Breadth-first	First node added	Yes	No	Exp
Lowest-cost-first	Minimal $cost(p)$	Yes	No	Exp

**Complete:** guaranteed to find a solution if there is one (for graphs with finite number of neighbours, even on infinite graphs)

**Halts:** on finite graph (perhaps with cycles).

**Space:** as a function of the length of current path

# Depth Bounded Search

Expands nodes like Depth First Search but imposes a cutoff on the maximum depth of path.

- Complete?** Yes (no infinite loops anymore)
- Time?**  $O(b^k)$  where  $k$  is the depth limit
- Space?**  $O(bk)$ , i.e., linear space similar to DFS
- Optimal?** No, can find suboptimal solutions first.

**Problem: How to pick a good limit ?**

# Iterative Deepening Search

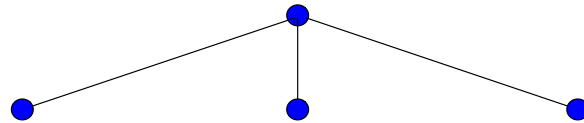
- Depth-bounded search: hard to decide on a depth bound
- Iterative deepening: Try all possible depth bounds in turn
- Combines benefits of depth-first and breadth-first search

# Iterative Deepening Search

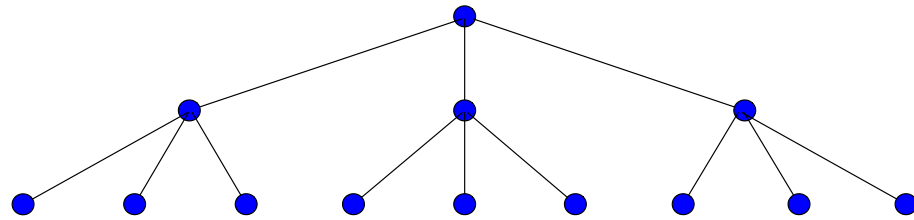
- Tries to combine the benefits of depth-first (low memory) and breadth-first (optimal and complete)
- Does a series of depth-limited depth-first searches to depth 1, 2, 3, etc.
- Early states will be expanded multiple times, but that might not matter too much because most of the nodes are near the leaves.



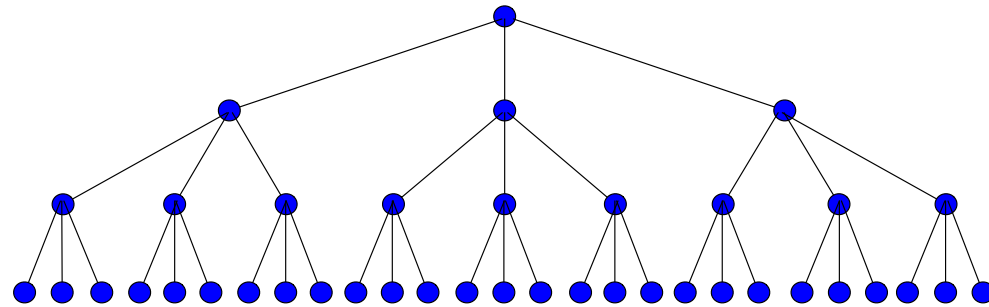
# Iterative Deepening Search



# Iterative Deepening Search



# Iterative Deepening Search



# Properties of Iterative Deepening Search

- **Complete?** Yes.
- **Time:** nodes at the bottom level are expanded once, nodes at the next level up twice, and so on:

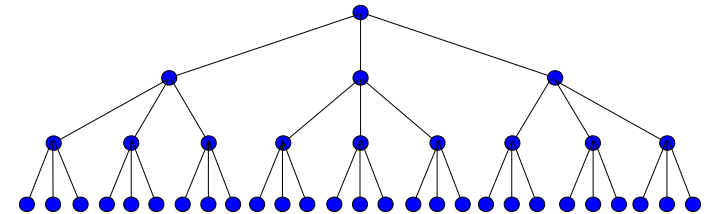
- depth-bounded:  $1 + b^2 + b^3 + \dots + b^d = \frac{b^{d+1} - 1}{b - 1} = O(b^d)$

- Iterative deepening:

$$(d + 1)b^0 + db^1 + (d - 1)b^2 + \dots + 2 \cdot b^{d-1} + 1 \cdot b^d = O(b^d)$$

- Example  $b=10, d=5$ :

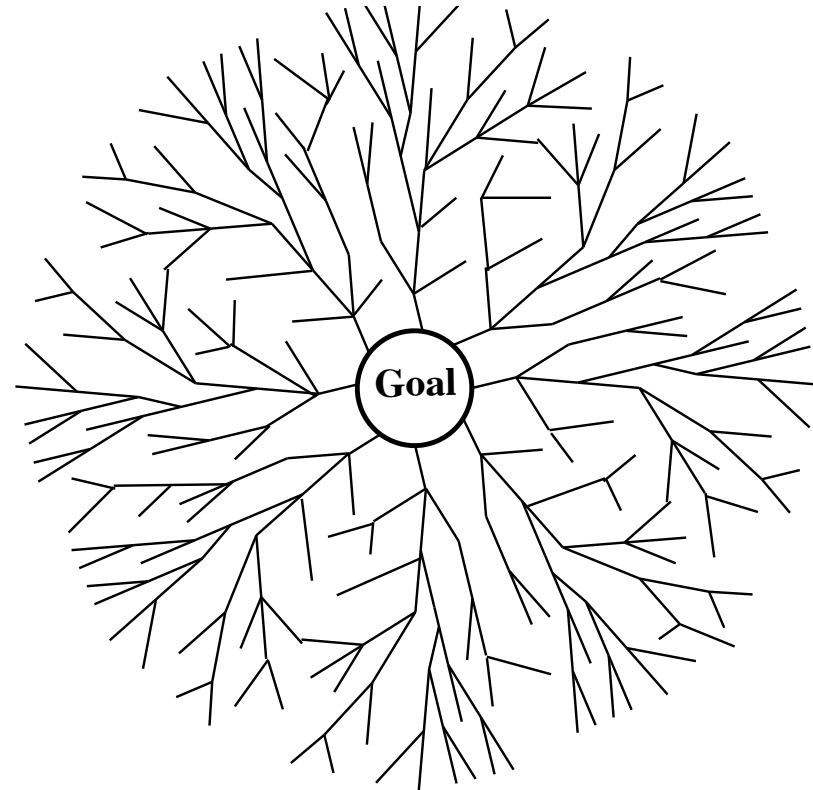
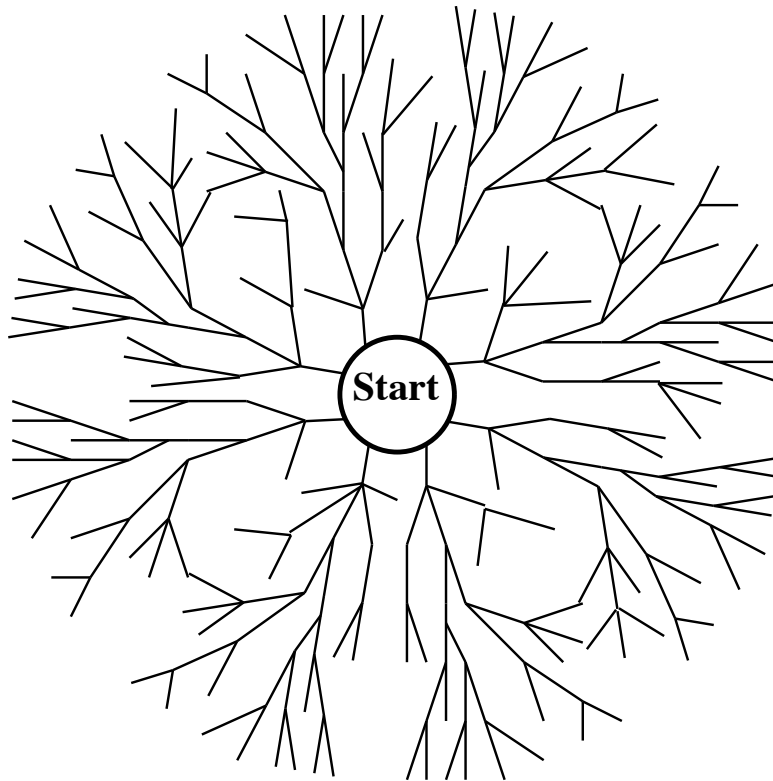
- depth-bounded:  $1 + 10 + 100 + 1,000 + 10,000 + 100,000 = 111,111$
- iterative-deepening:  $6 + 50 + 400 + 3,000 + 20,000 + 100,000 = 123,456$
- only about 11% more nodes (for  $b = 10$ ).



# Properties of Iterative Deepening Search

- Complete? Yes.
- Time:  $O(b^d)$
- Space?  $O(bd)$
- Optimal? Yes, if step costs are identical.
- In general, iterative deepening is the preferred search strategy for a large search space where depth of solution is not known

# Bidirectional Search



# Bidirectional Search

- Search both forward from the initial state and backward from the goal
  - stop when the two searches meet in the middle.
- Need efficient way to check if a new node appears in the other half of the search.
  - Complexity analysis assumes this can be done in constant time, using a hash table.
- Assume branching factor =  $b$  in both directions and that there is a solution at depth =  $d$ :
  - Then bidirectional search finds a solution in  $O(2b^{d/2}) = O(b^{d/2})$  time steps.

# Bidirectional Search Analysis

- If solution exists at depth  $d$  then bidirectional search requires time

$$O(2b^{\frac{d}{2}}) = O(b^{\frac{d}{2}})$$

- (assuming constant time checking of intersection)
- To check for intersection must have all states from one of the searches in memory, therefore space complexity is  $O(b^{\frac{d}{2}})$



# Summary

- Problem formulation usually requires abstracting away real-world details to define a state space that can feasibly be explored.
- Variety of Uninformed search strategies
- Iterative Deepening Search uses only linear space and not much more time than other Uninformed algorithms.

# Complexity Results for Uninformed Search

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening
Time	$O(b^d)$	$O(b^{\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^k)$	$O(b^d)$
Space	$O(b^d)$	$O(b^{\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(bk)$	$O(bd)$
Complete?	Yes <sup>1</sup>	Yes <sup>2</sup>	No	No	Yes <sup>1</sup>
Optimal ?	Yes <sup>3</sup>	Yes	No	No	Yes <sup>3</sup>

$b$  = branching factor,  $d$  = depth of the shallowest solution,  
 $m$  = maximum depth of the search tree,  $k$  = depth limit.

1 = complete if  $b$  is finite.

2 = complete if  $b$  is finite and step costs  $\geq \epsilon$  with  $\epsilon > 0$ .

3 = optimal if actions all have the same cost.