

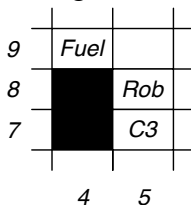
- Often we are not given an algorithm to solve a problem, but only a specification of what is a solution — we have to search for a solution.
- A typical problem is when the agent is in one state, it has a set of deterministic actions it can carry out, and wants to get to a goal state.
- Many AI problems can be abstracted into the problem of finding a path in a directed graph.
- Often there is more than one way to represent a problem as a graph.

# Directed Graphs

- A **graph** consists of a set  $N$  of **nodes** and a set  $A$  of ordered pairs of nodes, called **arcs**.
- Node  $n_2$  is a **neighbor** of  $n_1$  if there is an arc from  $n_1$  to  $n_2$ . That is, if  $\langle n_1, n_2 \rangle \in A$ .
- A **path** is a sequence of nodes  $\langle n_0, n_1, \dots, n_k \rangle$  such that  $\langle n_{i-1}, n_i \rangle \in A$ .
- The **length** of path  $\langle n_0, n_1, \dots, n_k \rangle$  is  $k$ .
- Given a set of **start nodes** and **goal nodes**, a **solution** is a path from a start node to a goal node.

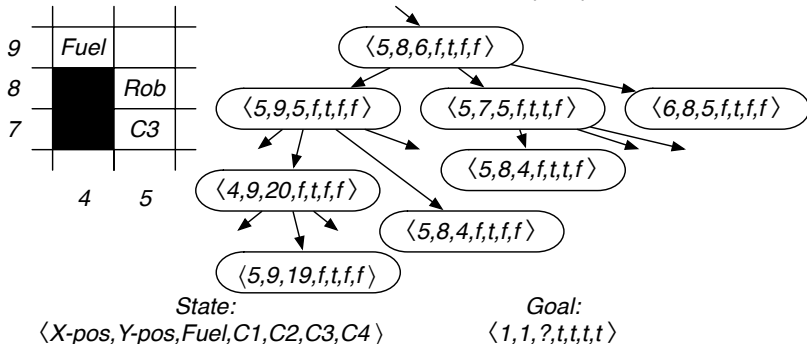
# Partial Search Space for a Video Game

Grid game: Rob needs to collect coins  $C_1$ ,  $C_2$ ,  $C_3$ ,  $C_4$ , without running out of fuel, and end up at location (1,1):



# Partial Search Space for a Video Game

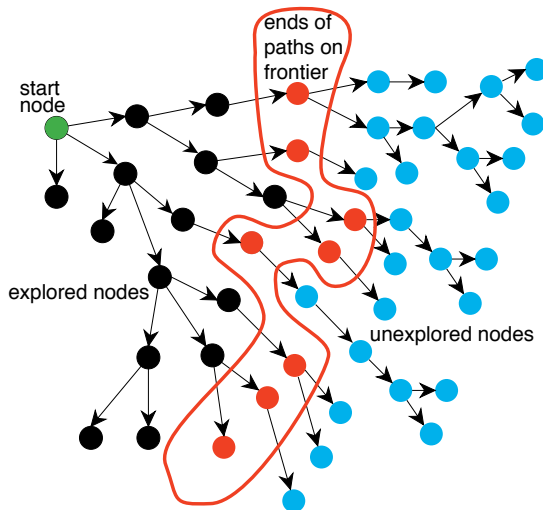
Grid game: Rob needs to collect coins  $C_1$ ,  $C_2$ ,  $C_3$ ,  $C_4$ , without running out of fuel, and end up at location (1,1):



# Graph Searching

- Generic search algorithm: given a graph, start nodes, and goal nodes, incrementally explore paths from the start nodes.
- Maintain a **frontier** of paths from the start node that have been explored.
- As search proceeds, the frontier expands into the unexplored nodes until a goal node is encountered.
- The way in which the frontier is expanded defines the **search strategy**.

# Problem Solving by Graph Searching



# Graph Search Algorithm

**Input:** a graph,  
a set of start nodes,  
Boolean procedure  $goal(n)$  that tests if  $n$  is a goal node.  
 $frontier := \{\langle s \rangle : s \text{ is a start node}\};$   
**while**  $frontier$  is not empty:  
    **select and remove** path  $\langle n_0, \dots, n_k \rangle$  from  $frontier$ ;  
    **if**  $goal(n_k)$   
        **return**  $\langle n_0, \dots, n_k \rangle$ ;  
    **for every** neighbor  $n$  of  $n_k$   
        **add**  $\langle n_0, \dots, n_k, n \rangle$  to  $frontier$ ;  
**end while**

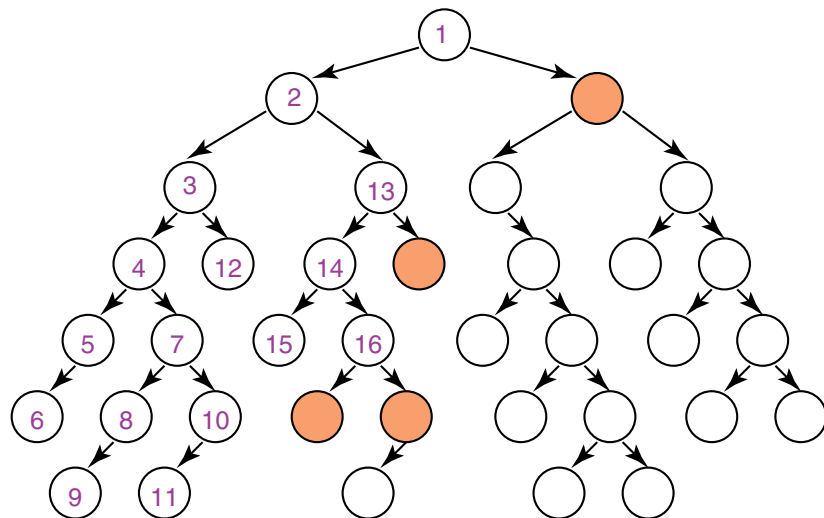
- Which value is selected from the frontier at each stage defines the search strategy.
- The neighbors define the graph.
- *goal* defines what is a solution.
- If more than one answer is required, the search can continue from the return.



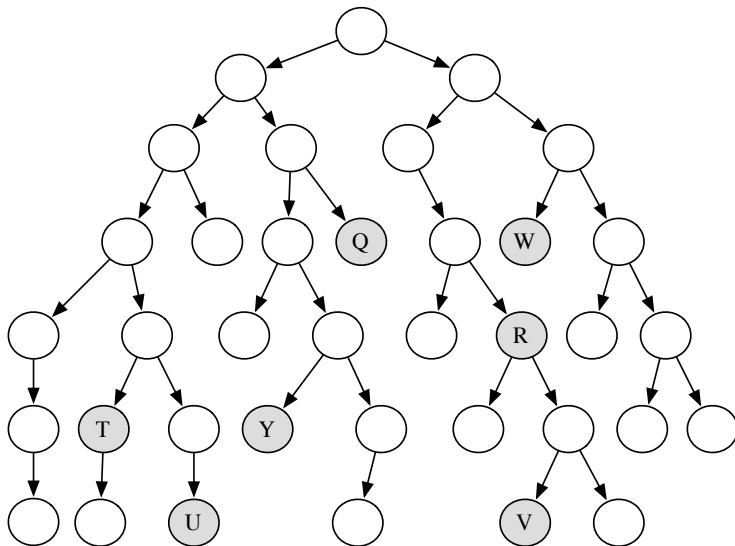
# Depth-first Search

- **Depth-first search** treats the frontier as a stack
- It always selects one of the last elements added to the frontier.
- If the list of paths on the frontier is  $[p_1, p_2, \dots]$ 
  - ▶  $p_1$  is selected. Paths that extend  $p_1$  are added to the front of the stack (in front of  $p_2$ ).
  - ▶  $p_2$  is only selected when all paths from  $p_1$  have been explored.

# Illustrative Graph — Depth-first Search



Which shaded goal will a depth-first search find first?



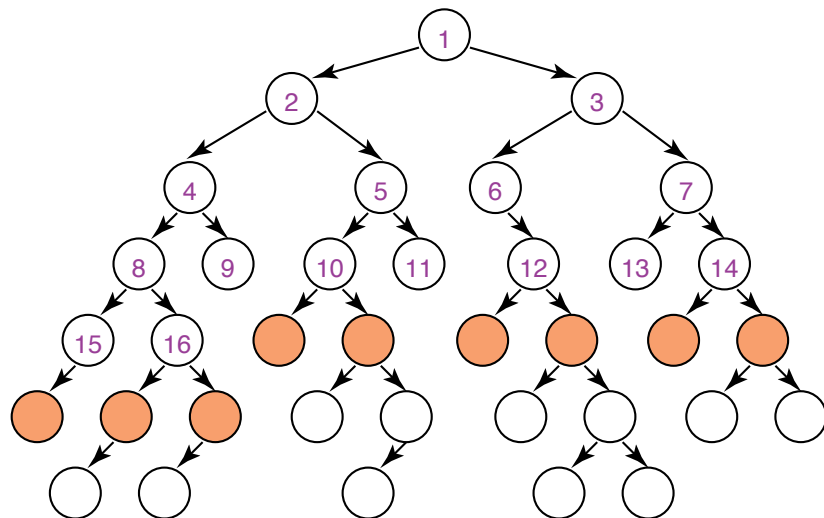
# Complexity of Depth-first Search

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

# Breadth-first Search

- Breadth-first search treats the frontier as a queue.
- It always selects one of the earliest elements added to the frontier.
- If the list of paths on the frontier is  $[p_1, p_2, \dots, p_r]$ :
  - ▶  $p_1$  is selected. Its neighbors are added to the end of the queue, after  $p_r$ .
  - ▶  $p_2$  is selected next.

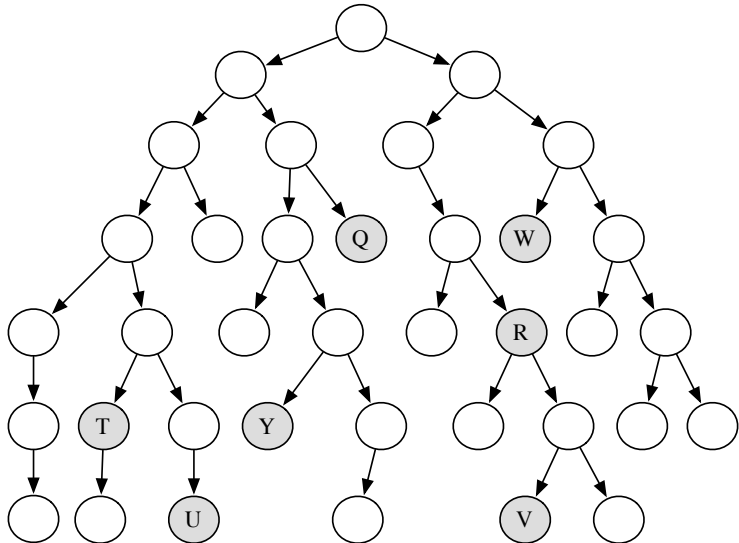
# Illustrative Graph — Breadth-first Search



# Complexity of Breadth-first Search

- Does breadth-first search guarantee to find the path with fewest arcs?
- 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?

Which shaded goal will a breadth-first search find first?





# Lowest-cost-first Search

- Sometimes there are **costs** associated with arcs. The cost of a path is the sum of the costs of its arcs.

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

An **optimal solution** is one with minimum cost.

- At each stage, lowest-cost-first search selects a path on the frontier with lowest cost.
- The frontier is a priority queue ordered by path cost.
- It finds a least-cost path to a goal node.
- When arc costs are equal  $\Rightarrow$  breadth-first search.

# Summary of Search Strategies

Strategy	Frontier Selection	Complete	Halts	Space
Depth-first	Last node added			
Breadth-first	First node added			
Lowest-cost-first	Minimal $cost(p)$			

**Complete** — if there a path to a goal, it can find one, even on infinite graphs.

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

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

# 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** — if there a path to a goal, it can find one, even on infinite graphs.

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

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# Heuristic Search

- **Idea:** don't ignore the goal when selecting paths.
- Often there is extra knowledge that can be used to guide the search: **heuristics**.
- **$h(n)$**  is an estimate of the cost of the shortest path from node  $n$  to a goal node.
- $h(n)$  needs to be efficient to compute.
- $h$  can be extended to paths:  $h(\langle n_0, \dots, n_k \rangle) = h(n_k)$ .
- $h(n)$  is an **underestimate** if there is no path from  $n$  to a goal with cost less than  $h(n)$ .
- An **admissible heuristic** is a nonnegative heuristic function that is an underestimate of the actual cost of a path to a goal.

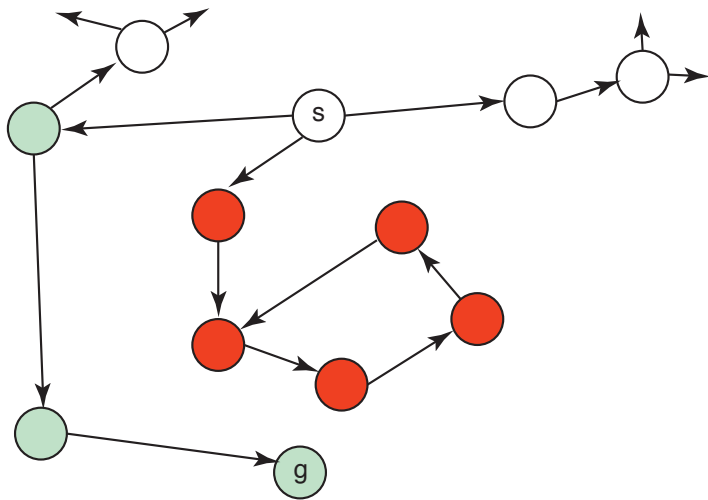
# Example Heuristic Functions

- If the nodes are points on a Euclidean plane and the cost is the distance,  $h(n)$  can be the straight-line distance from  $n$  to the closest goal.
- If the nodes are locations and cost is time, we can use the distance to a goal divided by the maximum speed.
- If the goal is to collect all of the coins and not run out of fuel, the cost is an estimate of how many steps it will take to collect the rest of the coins, refuel when necessary, and return to goal position.
- A heuristic function can be found by solving a simpler (less constrained) version of the problem.

# Best-first Search

- **Idea:** select the path whose end is closest to a goal according to the heuristic function.
- Best-first search selects a path on the frontier with minimal  $h$ -value.
- It treats the frontier as a priority queue ordered by  $h$ .

# Illustrative Graph — Best-first Search

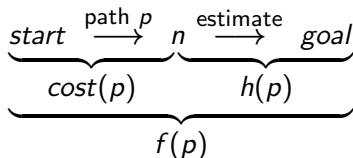


# Complexity of Best-first Search

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



- A\* search uses both path cost and heuristic values
- $cost(p)$  is the cost of path  $p$ .
- $h(p)$  estimates the cost from the end of  $p$  to a goal.
- Let  $f(p) = cost(p) + h(p)$ .  
 $f(p)$  estimates the total path cost of going from a start node to a goal via  $p$ .



# A\* Search Algorithm

- A\* is a mix of lowest-cost-first and best-first search.
- It treats the frontier as a priority queue ordered by  $f(p)$ .
- It always selects the node on the frontier with the lowest estimated distance from the start to a goal node constrained to go via that node.

# Complexity of $A^*$ Search

- Does  $A^*$  search guarantee to find the shortest path or the path with fewest arcs?
- What happens on infinite graphs or on graphs with cycles if there is a solution?
- What is the time complexity as a function of length of the path selected?
- What is the space complexity as a function of length of the path selected?
- How does the goal affect the search?

If there is a solution,  $A^*$  always finds an optimal solution —the first path to a goal selected— if

- the branching factor is finite
- arc costs are bounded above zero (there is some  $\epsilon > 0$  such that all of the arc costs are greater than  $\epsilon$ ), and
- $h(n)$  is nonnegative and an underestimate of the cost of the shortest path from  $n$  to a goal node.

# Why is $A^*$ admissible?

- If a path  $p$  to a goal is selected from a frontier, can there be a shorter path to a goal?
- Suppose path  $p'$  is on the frontier. Because  $p$  was chosen before  $p'$ , and  $h(p) = 0$ :

$$\text{cost}(p) \leq \text{cost}(p') + h(p').$$

- Because  $h$  is an underestimate:

$$\text{cost}(p') + h(p') \leq \text{cost}(p'')$$

for any path  $p''$  to a goal that extends  $p'$ .

- So  $\text{cost}(p) \leq \text{cost}(p'')$  for any other path  $p''$  to a goal.

# Why is $A^*$ admissible?

$A^*$  can always find a solution if there is one:

- The frontier always contains the initial part of a path to a goal, before that goal is selected.
- $A^*$  halts, as the costs of the paths on the frontier keeps increasing, and will eventually exceed any finite number.

# How do good heuristics help?

Suppose  $c$  is the cost of an optimal solution. What happens to a path  $p$  where

- $cost(p) + h(p) < c$
- $cost(p) + h(p) = c$
- $cost(p) + h(p) > c$

How can a better heuristic function help?

# Summary of Search Strategies

Strategy	Frontier Selection	Complete	Halts	Space
Depth-first	Last node added			
Breadth-first	First node added			
Heuristic depth-first	Local min $h(p)$			
Best-first	Global min $h(p)$			
Lowest-cost-first	Minimal $cost(p)$			
$A^*$	Minimal $f(p)$			

**Complete** — if there a path to a goal, it can find one, even on infinite graphs.

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# Summary of Search Strategies

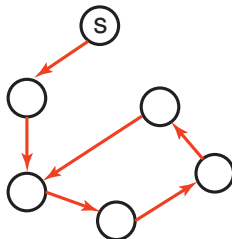
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Lowest-cost-first	Minimal $cost(p)$	Yes	No	Exp
$A^*$	Minimal $f(p)$	Yes	No	Exp

**Complete** — if there a path to a goal, it can find one, even on infinite graphs.

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

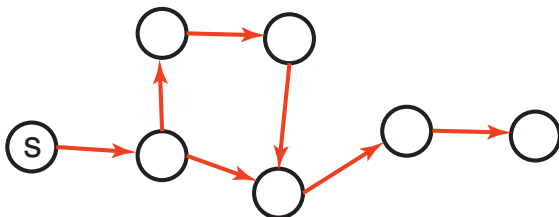
**Space** — as a function of the length of current path

# Cycle Checking



- A searcher can prune a path that ends in a node already on the path, without removing an optimal solution.
- In depth-first methods, checking for cycles can be done in \_\_\_\_\_ time in path length.
- For other methods, checking for cycles can be done in \_\_\_\_\_ time in path length.
- Does cycle checking mean the algorithms halt on finite graphs?

# Multiple-Path Pruning



- Multiple path pruning: prune a path to node  $n$  that the searcher has already found a path to.
- What needs to be stored?
- How does multiple-path pruning compare to cycle checking?
- Do search algorithms with multiple-path pruning always halt on finite graphs?
- What is the space & time overhead of multiple-path pruning?
- Can multiple-path pruning prevent an optimal solution being found?

# Multiple-Path Pruning & Optimal Solutions

**Problem:** what if a subsequent path to  $n$  is shorter than the first path to  $n$ ?

- remove all paths from the frontier that use the longer path.
- change the initial segment of the paths on the frontier to use the shorter path.
- ensure this doesn't happen. Make sure that the shortest path to a node is found first.

# Multiple-Path Pruning & $A^*$

- Suppose path  $p$  to  $n$  was selected, but there is a shorter path to  $n$ . Suppose this shorter path is via path  $p'$  on the frontier.
- Suppose path  $p'$  ends at node  $n'$ .
- $p$  was selected before  $p'$ , so:  
$$\text{cost}(p) + h(n) \leq \text{cost}(p') + h(n').$$
- Suppose  $\text{cost}(n', n)$  is the actual cost of a path from  $n'$  to  $n$ . The path to  $n$  via  $p'$  is shorter than  $p$  so:  
$$\text{cost}(p') + \text{cost}(n', n) < \text{cost}(p).$$

$$\text{cost}(n', n) < \text{cost}(p) - \text{cost}(p') \leq h(n') - h(n).$$

We can ensure this doesn't occur if

$$|h(n') - h(n)| \leq \text{cost}(n', n).$$

# Monotone Restriction

- Heuristic function  $h$  satisfies the **monotone restriction** if  $|h(m) - h(n)| \leq \text{cost}(m, n)$  for every arc  $\langle m, n \rangle$ .
- If  $h$  satisfies the monotone restriction,  $A^*$  with multiple path pruning always finds the shortest path to a goal.
- This is a strengthening of the admissibility criterion.

# Direction of Search

- The definition of searching is symmetric: find path from start nodes to goal node or from goal node to start nodes.
- **Forward branching factor:** number of arcs out of a node.
- **Backward branching factor:** number of arcs into a node.
- Search complexity is  $b^n$ . Should use forward search if forward branching factor is less than backward branching factor, and vice versa.
- Note: when graph is dynamically constructed, the backwards graph may not be available.

# Bidirectional Search

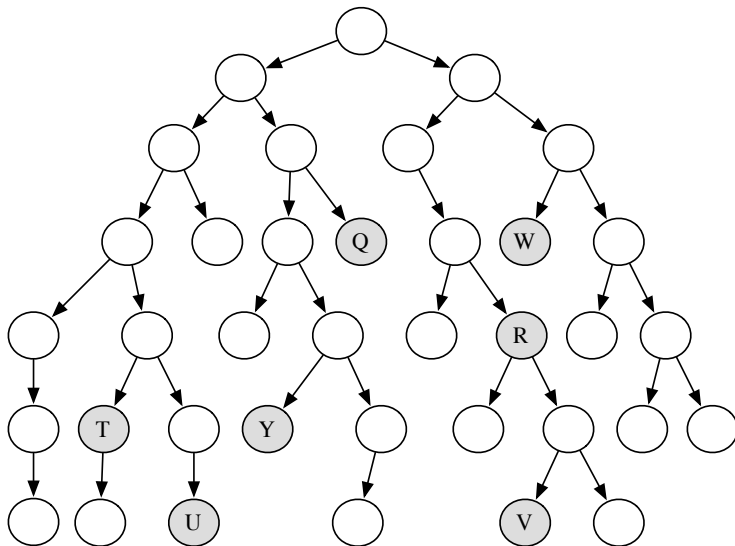
- Idea: search backward from the goal and forward from the start simultaneously.
- This wins as  $2b^{k/2} \ll b^k$ . This can result in an exponential saving in time and space.
- The main problem is making sure the frontiers meet.
- This is often used with one breadth-first method that builds a set of locations that can lead to the goal. In the other direction another method can be used to find a path to these interesting locations.



# Bounded Depth-first search

- A bounded depth-first search takes a bound (cost or depth) and does not expand paths that exceed the bound.
  - ▶ explores part of the search graph
  - ▶ uses space linear in the depth of the search.
- How does this relate to other searches?
- How can this be extended to be complete?

Which shaded goal will a depth-bounded search find first?



# Iterative-deepening search

- Iterative-deepening search:
  - ▶ Start with a bound  $b = 0$ .
  - ▶ Do a bounded depth-first search with bound  $b$
  - ▶ If a solution is found return that solution
  - ▶ Otherwise increment  $b$  and repeat.
- This will find the same first solution as what other method?
- How much space is used?
- What happens if there is no path to a goal?
- Surely recomputing paths is wasteful!!!

# Iterative Deepening Complexity

Complexity with solution at depth  $k$  & branching factor  $b$ :

level	breadth-first	iterative deepening	# nodes
1	1	$k$	$b$
2	1	$k - 1$	$b^2$
...	...	...	...
$k - 1$	1	2	$b^{k-1}$
$k$	1	1	$b^k$
total	$\geq b^k$	$\leq b^k \left( \frac{b}{b-1} \right)^2$	