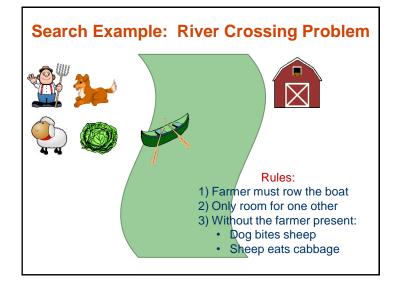
Uninformed Search

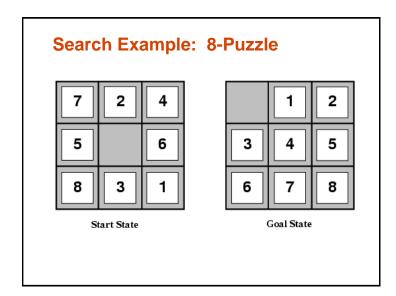
Chapter 3.1 - 3.4

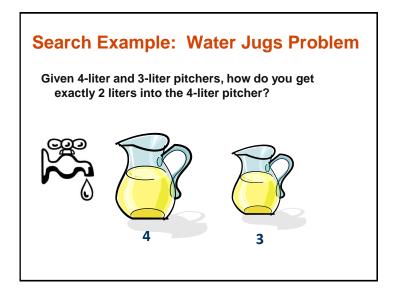
Many Al Tasks can be Formulated as Search Problems

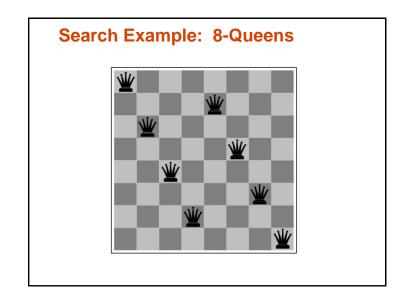
- Puzzles
- Games
- Navigation
- Assignment
- Layout
- Scheduling
- Routing

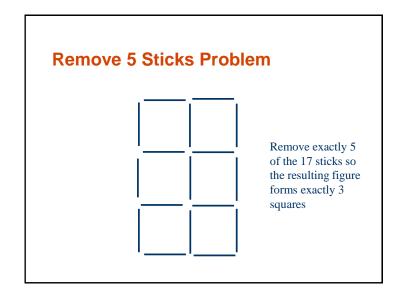












Basic Search Task Assumptions (usually, though not games)

- Fully observable
- Deterministic
- Static
- Discrete
- Single agent

How should the Environment be Represented?

- Knowledge representation problem:
 - What information from the sensors is relevant?
 - How to represent domain knowledge?
- Determining what to represent is difficult and is usually left to the system designer to specify
- Problem State = representation of all necessary information about the environment
- **State Space** (aka **Problem Space**) = all possible valid configurations of the environment

What Knowledge does the Agent Need?

- The information needs to be
 - sufficient to describe all relevant aspects for reaching the goal
 - adequate to describe the world state/situation
- Fully observable assumption, also known as the closed world assumption, means
 - All necessary information about a problem domain is accessible so that each state is a complete description of the world; there is no missing information at any point in time

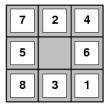
What Goal does the Agent want to Achieve?

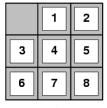
- How do you describe the goal?
 - as a task to be accomplished
 - as a state to be reached
 - as a set of properties to be satisfied
- How do you know when the goal is reached?
 - with a goal test that defines what it means to have achieved/satisfied the goal
 - or, with a set of **goal states**
- Determining the goal is usually left to the system designer or user to specify

What Actions does the Agent Need?

- Discrete and Deterministic task assumptions imply
- Given:
 - an action (aka operator or move)
 - a description of the current state of the world
- Action completely specifies:
 - if that action *can* be applied (i.e., legal)
 - what the exact state of the world will be after the action is performed in the current state (no "history" information needed to compute the successor state)

Search Example: 8-Puzzle





Start State

Goal State

- States = configurations
- Actions = up to 4 kinds of moves: up, down, left, right

Water Jugs Problem

Given 4-liter and 3-liter pitchers, how do you get exactly 2 liters into the 4-liter pitcher?







State: (x, y) for # liters in 4-liter and 3-liter pitchers, respectively Actions: empty, fill, pour water between pitchers

Initial state: (0, 0) Goal state: (2, *)

Actions / Successor Functions

1. $(x, y / x < 4) \rightarrow (4, y)$ Fill 4

2. $(x, y / y < 3) \rightarrow (x, 3)$ Fill 3

3. $(x, y / x > 0) \rightarrow (0, y)$ Empty 4

4. $(x, y / y > 0) \rightarrow (x, 0)$ Empty 3

 $5. (x, y / x + y \ge 4 \text{ and } y > 0) \longrightarrow (4, y - (4 - x))$

Pour from 3 to 4 until 4 is full

6. $(x, y / x + y \ge 3 \text{ and } x > 0) \longrightarrow (x - (3 - y), 3)$

Pour from 4 to 3 until 3 is full

7. $(x, y | x+y \le 4 \text{ and } y > 0) \longrightarrow (x+y, 0)$

Pour all water from 3 to 4

Formalizing Search in a State Space

- A state space is a graph: (V, E)
 - V is a set of nodes (vertices)
 - E is a set of arcs (edges)
 each arc is directed from one node to another node
- Each node is a data structure that contains:
 - a **state** description
 - other information such as:
 - link to parent node
 - name of action that generated this node (from its parent)
 - other bookkeeping data

Formalizing Search in a State Space

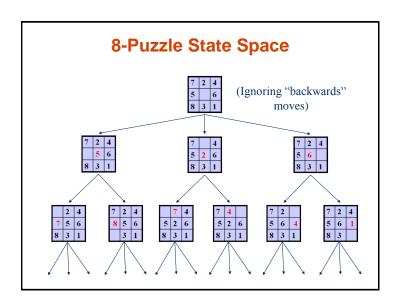
- Each arc corresponds to one of the finite number of actions:
 - when the action is applied to the state associated with the arc's source node
 - then the resulting state is the state associated with the arc's destination node
- Each arc has a fixed, positive cost:
 - corresponds to the cost of the action

Formalizing Search in a State Space

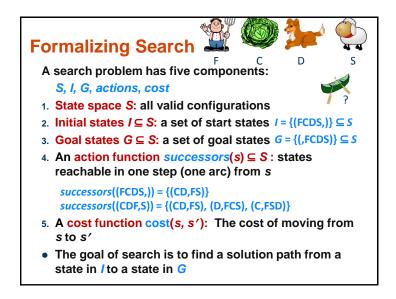
- Each node has a finite set of successor nodes:
 - corresponds to all of the legal actions that can be applied at the source node's state
- Expanding a node means:
 - generate all of the successor nodes
 - add them and their associated arcs to the statespace search tree

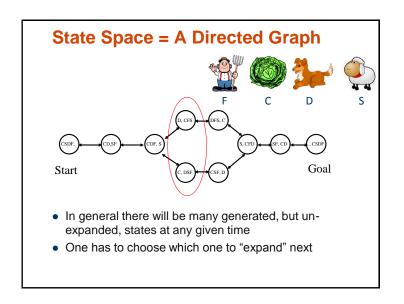
Formalizing Search in a State Space

- One or more nodes are designated as start nodes
- A goal test is applied to a node's state to determine if it is a goal node
- A solution is a sequence of actions associated with a path in the state space from a start to a goal node:
 - just the goal state (e.g., cryptarithmetic)
 - a path from start to goal state (e.g., 8-puzzle)
- The cost of a solution is the sum of the arc costs on the solution path



Sizes of State Spaces Problem Nodes **Brute-Force Search Time** (10 million nodes/second) 39 Tic-Tac-Toe 10⁵ .01 seconds 8 Puzzle • 23 Rubik's Cube 10⁶ .2 seconds 15 Puzzle 10¹³ 6 davs 3³ Rubik's Cube 10¹⁹ 68,000 years 10²⁵ 12 billion years 24 Puzzle Checkers 10⁴⁰ 10¹²⁰ Chess

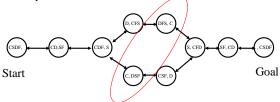




Different Search Strategies

- The generated, but not yet expanded, states define the Frontier (aka Open or Fringe) list
- The essential difference is, which one to expand first?

• Deep or shallow?



Formalizing Search in a State Space

- This algorithm does NOT detect goal when node is generated
- This algorithm does NOT detect loops in state space
- Each node implicitly represents
 - a partial solution path from the start node to the given node
 - cost of the partial solution path
- From this node there may be
 - many possible paths that have this partial path as a prefix
 - many possible solutions

Formalizing Search in a State Space

State-space search is the process of searching through a state space for a solution by making explicit a sufficient portion of an implicit state-space graph to include a goal node: TREE SEARCH Alg.

Frontier = $\{S\}$, where S is the start node **Loop do**

if Frontier is empty then return failure pick a node, n, from Frontier
if n is a goal node then return solution
Generate all n's successor nodes and add them all to Frontier
Remove n from Frontier

Uninformed Search on Trees

- Uninformed means we only know:
 - The goal test
 - The successors() function
- But *not* which non-goal states are better
- For now, also assume state space graph is a tree
 - That is, we won't encounter (or at least worry about) repeated states
 - We will relax this later
- Search strategies differ by what un-expanded node is expanded next

Uninformed Search Strategies

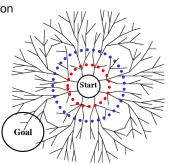
Uninformed Search: strategies that order nodes *without* using any domain specific information, i.e., doesn't use any information stored in a state

- BFS: breadth-first search
 - Queue (FIFO) used for the Frontier list
 - remove from front, add to back
- DFS: depth-first search
 - Stack (LIFO) used for the Frontier list
 - remove from front, add to front

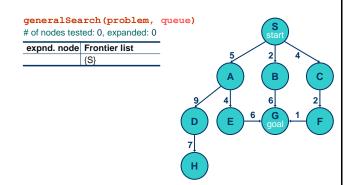
Breadth-First Search (BFS)

Expand the shallowest node first:

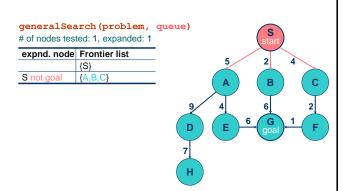
- 1. Examine states one step away from the initial states
- 2. Examine states two steps away from the initial states
- 3. and so on

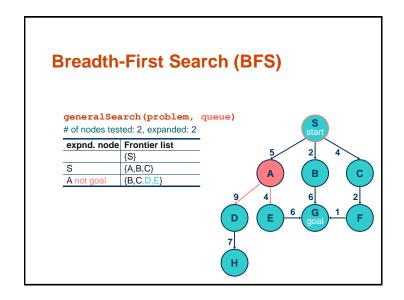


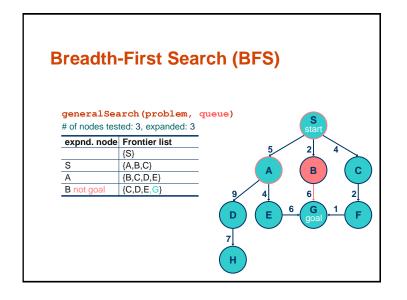
Breadth-First Search (BFS)

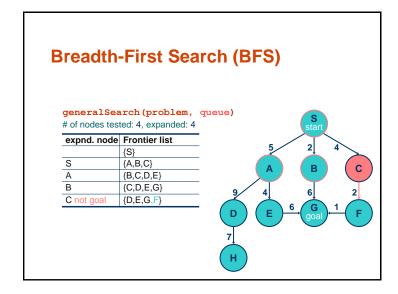


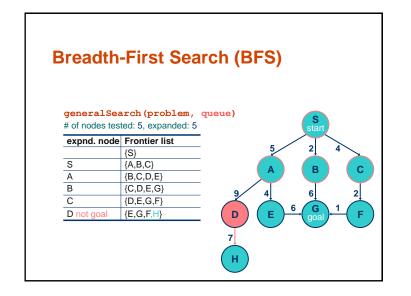
Breadth-First Search (BFS)

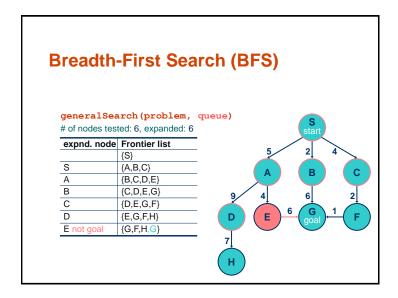


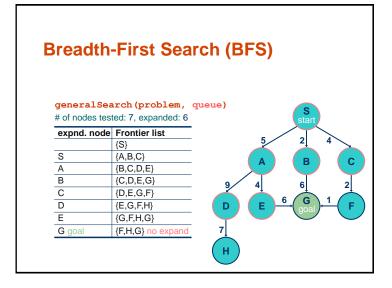


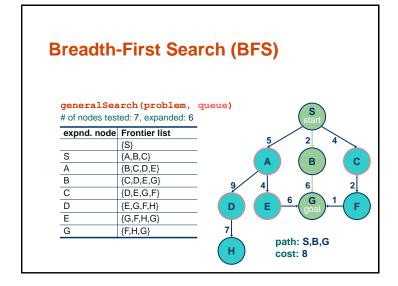












Evaluating Search Strategies

• Completeness

If a solution exists, will it be found?

- a complete algorithm will find **a** solution (not all)

• Optimality / Admissibility

If a solution is found, is it guaranteed to be optimal?

an admissible algorithm will find a solution with minimum cost

Evaluating Search Strategies

• Time Complexity

How long does it take to find a solution?

- usually measured for worst case
- measured by counting **number of nodes expanded**

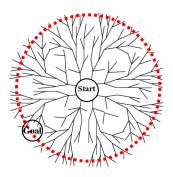
Space Complexity

How much space is used by the algorithm?

 measured in terms of the maximum size of the Frontier list during the search

What's in the Frontier (Queue) for BFS?

 If goal is at depth d, how big is the frontier (worst case)?



Breadth-First Search (BFS)

- Complete
- Optimal / Admissible
 - Yes, if all operators (i.e., arcs) have the same constant cost, or costs are positive, non-decreasing with depth
 - otherwise, not optimal but does guarantee finding solution of shortest *length* (i.e., fewest arcs)

Breadth-First Search (BFS)

- Time and space complexity: $O(b^d)$ (i.e., exponential)
 - *d* is the depth of the solution
 - -b is the branching factor at each non-leaf node
- Very slow to find solutions with a large number of steps because must look at all shorter length possibilities first

Breadth-First Search (BFS)

- A complete search tree has a total # of nodes =
 - $1 + b + b^2 + ... + b^d = (b^{(d+1)} 1) / (b-1)$
 - d: the tree's depth
 - *b*: the branching factor at each non-leaf node
- For example: d = 12, b = 10

$$1 + 10 + 100 + ... + 10^{12} = (10^{13} - 1)/9 = O(10^{12})$$

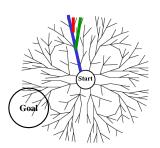
 If BFS expands 1,000 nodes/sec and each node uses 100 bytes of storage, then BFS will take 35 years to run in the worst case, and it will use 111 terabytes of memory!

Depth-First Search

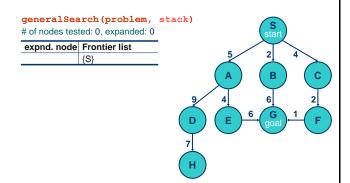
Expand the *deepest* node first

- 1. Select a direction, go deep to the end
- 2. Slightly change the end
- 3. Slightly change the end some more...

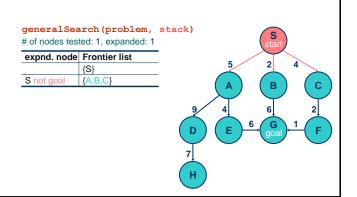
Use a Stack to order nodes on the Frontier list

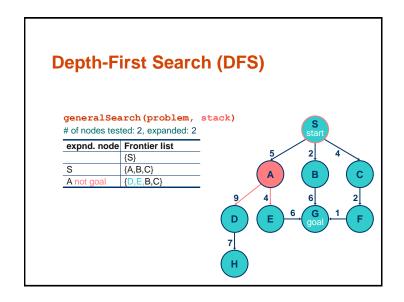


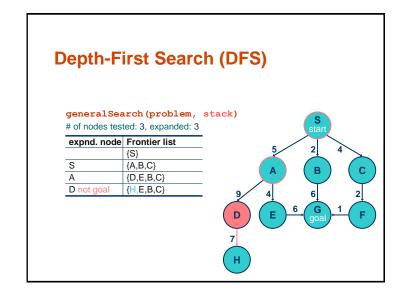
Depth-First Search (DFS)

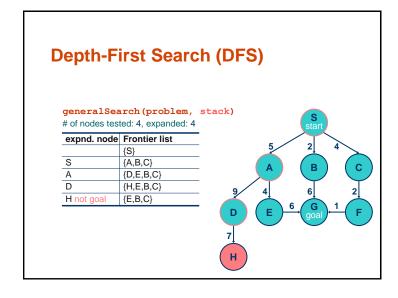


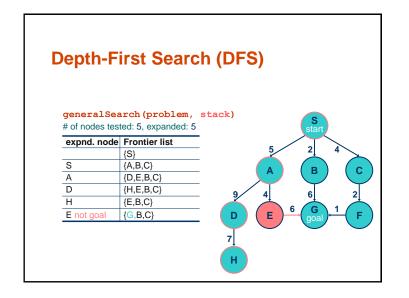
Depth-First Search (DFS)

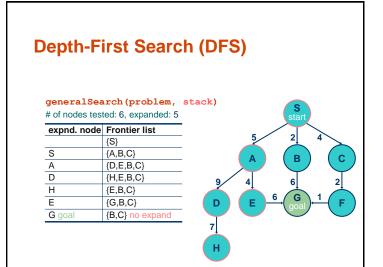


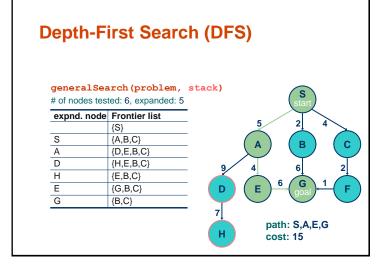












Depth-First Search (DFS)

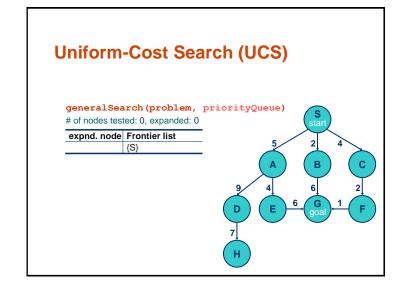
- May not terminate without a depth bound i.e., cutting off search below a fixed depth, D
- Not complete
 - with or without cycle detection
 - and, with or without a depth cutoff
- Not optimal / admissible
- Can find long solutions quickly if lucky

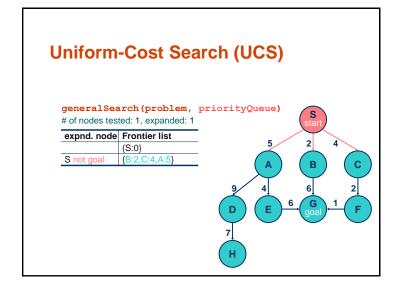
Depth-First Search (DFS)

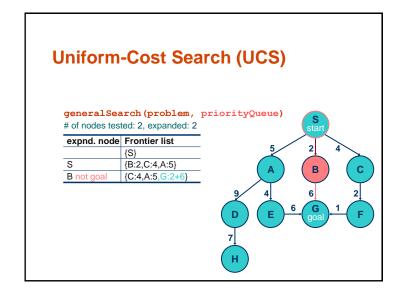
- Time complexity: $O(b^d)$ exponential Space complexity: O(bd) linear
 - *d* is the depth of the solution
 - -b is the branching factor at each non-leaf node
- Performs "chronological backtracking"
 - i.e., when search hits a dead end, backs up one level at a time
 - problematic if the mistake occurs because of a bad action choice near the top of search tree

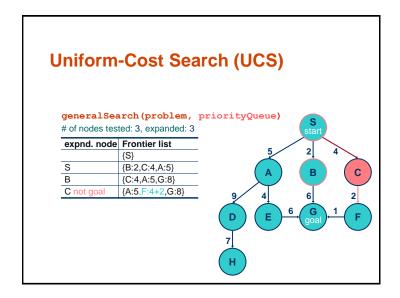
Uniform-Cost Search (UCS)

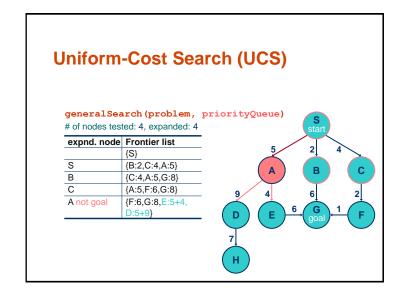
- Use a "Priority Queue" to order nodes on the Frontier list, sorted by path cost
- Let g(n) = cost of path from start node s to current node n
- Sort nodes by increasing value of g

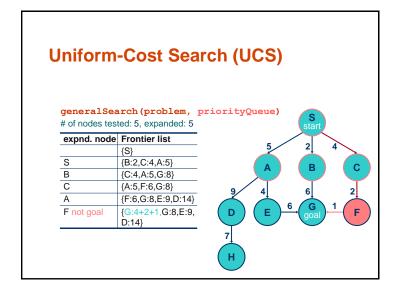


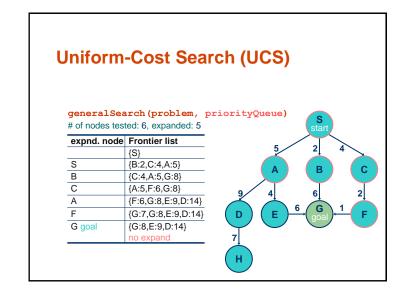




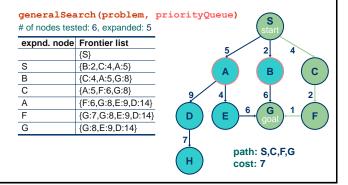








Uniform-Cost Search (UCS)



Uniform-Cost Search (UCS)

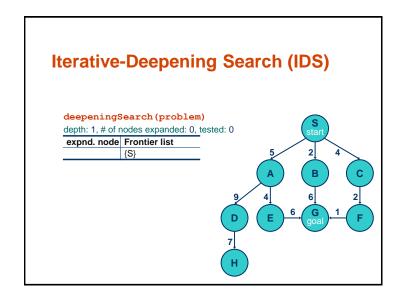
- Called *Dijkstra's Algorithm* in the algorithms literature
- Similar to *Branch and Bound Algorithm* in Operations Research literature
- Complete
- Optimal / Admissible
 - requires that the goal test is done when a node is removed from the Frontier list rather than when the node is generated by its parent node

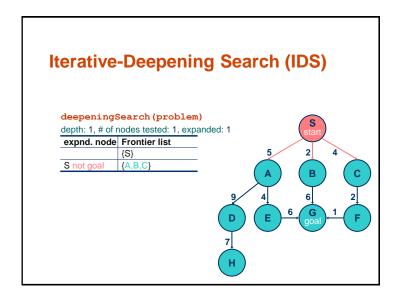
Uniform-Cost Search (UCS)

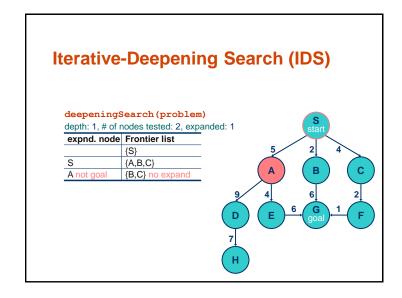
- ullet Time and space complexity: $O(b^d)$ (i.e., exponential)
 - *d* is the depth of the solution
 - -b is the branching factor at each non-leaf node
- More precisely, time and space complexity is $O(b^{C^*/\epsilon})$ where all edge costs $\geq \epsilon > 0$, and C^* is the best goal path cost

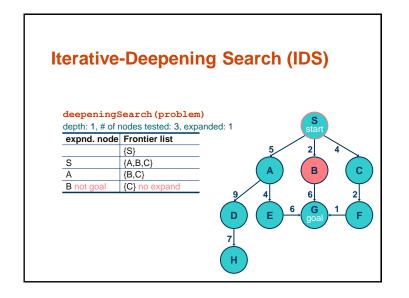
Iterative-Deepening Search (IDS)

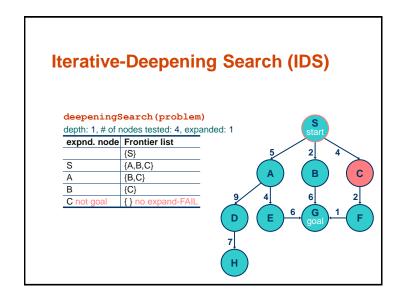
- requires modification to DFS search algorithm:
 - do DFS to depth 1 and treat all children of the start node as leaves
 - $\,$ $\,$ if no solution found, do DFS to depth 2 $\,$
 - repeat by increasing "depth bound" until a solution found
- Start node is at depth 0

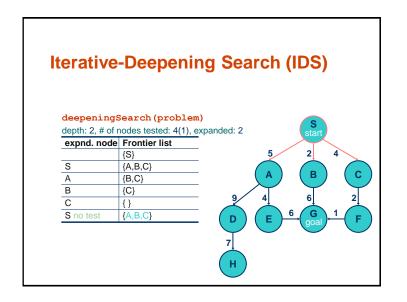


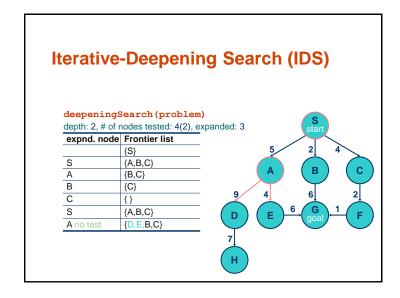


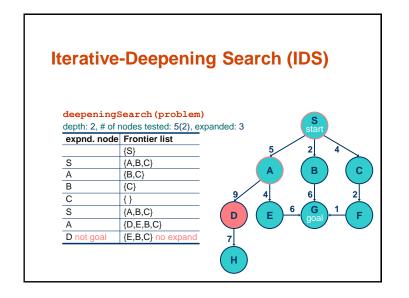


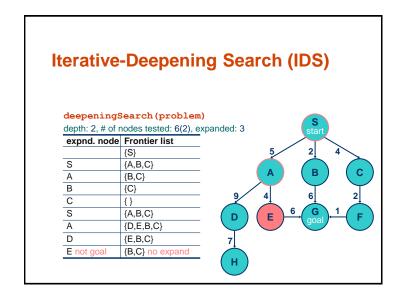


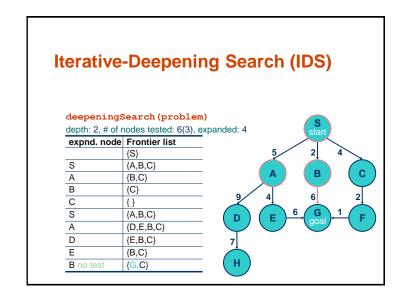


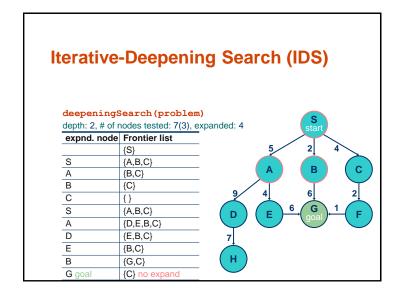


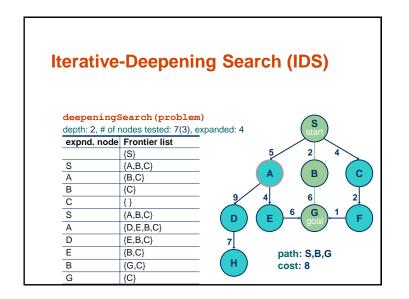












Iterative-Deepening Search (IDS)

- Has advantages of BFS
 - completeness
 - optimality as stated for BFS
- Has advantages of DFS
 - limited space
 - in practice, even with redundant effort it still finds longer paths more quickly than BFS

Iterative-Deepening Search (IDS)

- Space complexity: O(bd) (i.e., linear like DFS)
- Time complexity is a little worse than BFS or DFS
 - because nodes near the top of the search tree are generated multiple times (redundant effort)
- Worst case time complexity: $O(b^d)$ exponential
 - because most nodes are near the bottom of tree

Iterative-Deepening Search (IDS)

How much redundant effort is done?

• The number of times the nodes are generated:

$$1b^d + 2b^{(d-1)} + \dots + db \ \le \ b^d \, / \, (1 - 1/b)^2 = O(b^d)$$

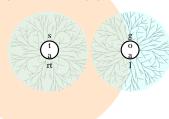
- *d*: the solution's depth
- For example: b = 4

$$4^{d} / (1 - \frac{1}{4})^{2} = 4^{d} / (.75)^{2} = 1.78 \times 4^{d}$$

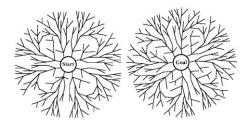
- in the worst case, 78% more nodes are searched (redundant effort) than exist at depth $\it d$
- as b increases, this % decreases

Bidirectional Search

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



Which Direction Should We Search?



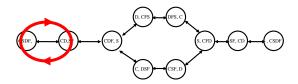
Our choices: Forward, backwards, or bidirectional

The issues: How many start and goal states are there?

Branching factors in each direction How much work is it to compare states?

If State Space is Not a Tree

• The problem: repeated states



- Ignoring repeated states: wasteful (BFS) or impossible (DFS). Why?
- How to prevent these problems?

Performance of Search Algorithms on Trees

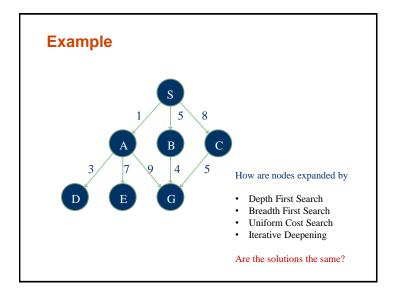
b: branching factor (assume finite) d: goal depth m: graph depth

| | Complete | optimal | time | space |
|-----------------------------------|----------|--------------------|-----------------------|-----------------------|
| Breadth-first search | Y | Y, if ¹ | O(b ^d) | O(b ^d) |
| Uniform-cost search ² | Υ | Υ | O(b ^{C*/ε}) | $O(b^{C^*/\epsilon})$ |
| Depth-first search | N | N | O(b ^m) | O(bm) |
| Iterative deepening | Y | Y, if ¹ | O(bd) | O(bd) |
| Bidirectional search ³ | Υ | Y, if ¹ | O(b ^{d/2}) | O(b ^{d/2}) |

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

If State Space is Not a Tree

- We have to remember already-expanded states (called *Explored* (aka *Closed*) list) too
- Why?
- When we pick a state from Frontier
 - Remove it from *Frontier*
 - Add it to Explored
 - Expand node, generating all successors
 - For each successor, child,
 - If child is in Explored, throw child away
 - Otherwise, check whether *child* is in *Frontier*
 - If no, add it to Frontier
 - If yes and path-cost(child) < path-cost of node already in Frontier, then replace that Frontier node with child



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
- Iterative-Deepening Search: S A B C S A D E G Solution found: S A G