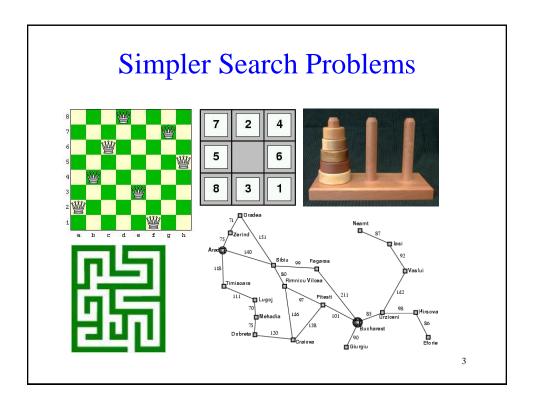
# CS 331: Artificial Intelligence Uninformed Search

1

# Real World Search Problems No Observed Hill Rd North Abary County Etaks Revisitory Opinger Opin



# Assumptions About Our Environment

- Static
- Observable
- Discrete
- Deterministic
- Single-agent

#### Search Problem Formulation

A search problem has 5 components:

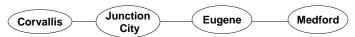
- 1. A finite set of states S
- 2. A non-empty set of initial states  $I \subseteq S$
- 3. A non-empty set of goal states  $G \subseteq S$
- 4. A successor function succ(s) which takes a state s as input and returns as output the set of states you can reach from state s in one step.
- 5. A cost function *cost(s,s')* which returns the non-negative one-step cost of travelling from state *s* to *s'*. The cost function is only defined if *s'* is a successor state of *s*.

5

#### Example: Oregon Portland S = {Coos Bay, Newport, Corvallis, Junction City, **McMinnville** Eugene, Medford, Albany, (Salem) Lebanon, Salem, **Initial State** Portland, McMinnville} Lebanon Albany I = {Corvallis} Newport G={Medford} Junction Coos Succ(Corvallis)={Albany, Eugene Newport, McMinnville, Junction City) Cost(s,s') = 1 for all transitions **Goal State** 6

#### Results of a Search Problem

• Solution
Path from initial state to goal state



- Solution quality
  Path cost (3 in this case)
- Optimal solution
   Lowest path cost among all solutions (In this case, we found the optimal solution)

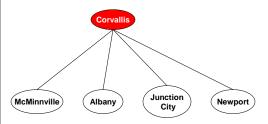
7

#### Search Tree



Start with Initial State

#### Search Tree

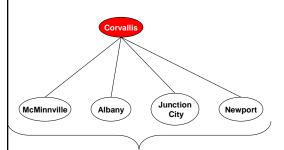


Is initial state the goal?

- Yes, return solution
- No, apply Successor() function

9

#### Search Tree



These nodes have not been expanded yet. Call them the fringe. We'll put them in a queue.

Apply Successor() function

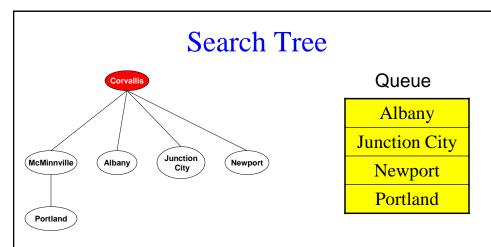
Queue

McMinnville

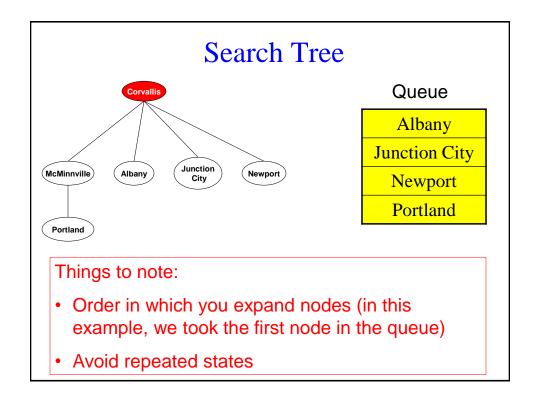
Albany

**Junction City** 

Newport



Now remove a node from the queue. If it's a goal state, return the solution. Otherwise, call Successor() on it, and put the results in the queue. Repeat.



#### Tree-Search Pseudocode

function TREE-SEARCH(problem, fringe) returns a solution, or failure

```
fringe \leftarrow \text{Insert}(\text{Make-Node}(\text{Initial-State}[problem]), fringe) \\ \textbf{loop do} \\ \textbf{if } fringe \textbf{is empty then return failure} \\ node \leftarrow \text{Remove-Front}(fringe) \\ \textbf{if } \text{Goal-Test}[problem](\text{State}[node]) \textbf{ then return } \text{Solution}(node) \\ fringe \leftarrow \text{InsertAll}(\text{Expand}(node, problem), fringe) \\ \hline \\ \textbf{function } \text{Expand}(node, problem) \textbf{ returns a set of nodes} \\ successors \leftarrow \textbf{ the empty set} \\ \textbf{for each } action, result \textbf{ in } \text{Successor-Fn}[problem](\text{State}[node]) \textbf{ do} \\ s \leftarrow \textbf{a new } \text{Node} \\ \text{Parent-Node}[s] \leftarrow node; \text{ Action}[s] \leftarrow action; \text{ State}[s] \leftarrow result \\ \text{Path-Cost}[s] \leftarrow \text{Path-Cost}[node] + \text{Step-Cost}(node, action, s) \\ \text{Depth}[s] \leftarrow \text{Depth}[node] + 1 \\ \text{add } s \textbf{ to } successors \\ \textbf{return } successors \\ \\ \textbf{return } successors \\ \hline \end{aligned}
```

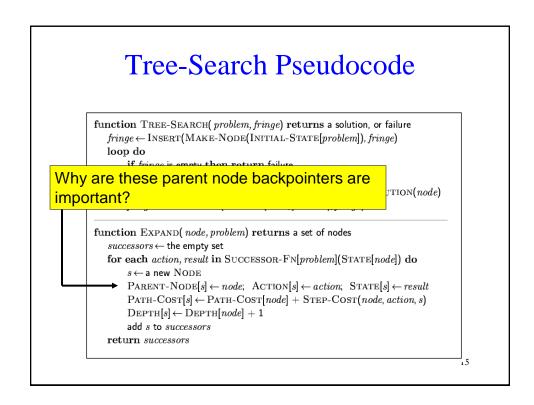
#### Tree-Search Pseudocode

```
function Tree-Search( problem, fringe) returns a solution, or failure fringe \leftarrow Insert(Make-Node(Initial-State[problem]), fringe) loop do if fringe is empty then return failure node \leftarrow Remove-Front(fringe) if Goal-Test[problem](State[node]) then return Solution(node) fringe \leftarrow InsertAll(Expand(node, problem), fringe)

function Expand( node, problem) returns a set of nodes successors \leftarrow the empty set

Note: Goal test happens after we grab a node off the queue.

Path-Cost[s] \leftarrow Path-Cost[node] + Step-Cost(node, action, s)
Depth[s] \leftarrow Depth[node] + 1
add s to successors
return successors
```



#### **Uninformed Search**

- No info about states other than generating successors and recognizing goal states
- Later on we'll talk about informed search –
  can tell if a non-goal state is more
  promising than another

## **Evaluating Uninformed Search**

- Completeness
  Is the algorithm guaranteed to find a solution when there is one?
- Optimality Does it find the optimal solution?
- Time complexity
  How long does it take to find a solution?
- Space complexity
   How much memory is needed to perform the search

17

### Complexity

- 1. Branching factor (b) maximum number of successors of any node
- 2. Depth (d) of the shallowest goal node
- 3. Maximum length (m) of any path in the search space

Time Complexity: number of nodes generated during search

Space Complexity: maximum number of nodes stored in memory

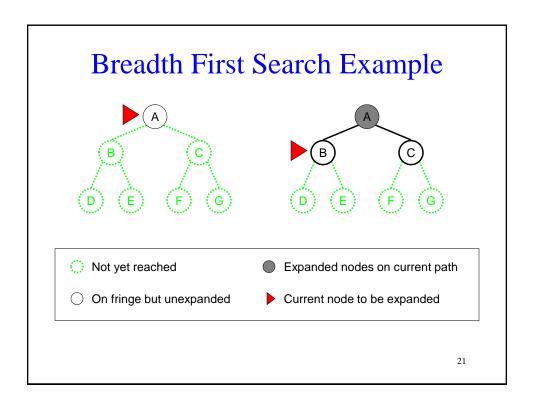
# **Uninformed Search Algorithms**

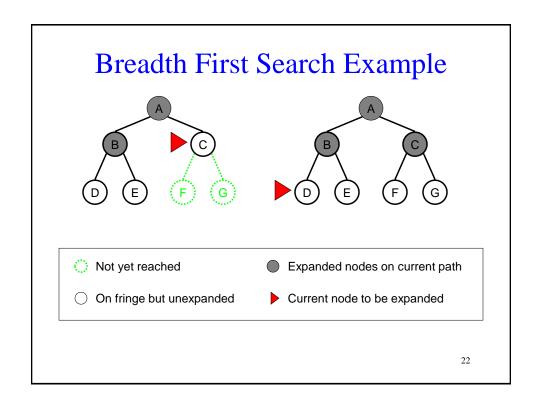
- Breadth-first search
- Uniform-cost search
- Depth-first search
- Depth-limited search
- Iterative Deepening Depth-first Search
- Bidirectional search

19

#### **Breadth-First Search**

- Expand all nodes at a given depth before any nodes at the next level are expanded
- Implement with a FIFO queue





# **Evaluating BFS**

Complete?	
Optimal?	
Time Complexity	
Space Complexity	

23

# **Evaluating BFS**

Complete?	Yes provided branching factor is
	finite
Optimal?	Yes if step costs are identical
Time Complexity	$b+b^2+b^3++b^d+(b^{d+1}-b)=$
	$O(b^{d+1})$
Space Complexity	$O(b^{d+1})$

Exponential time and space complexity make BFS impractical for all but the smallest problems

#### **Uniform-cost Search**

- What if step costs are not equal?
- Recall that BFS expands the shallowest node
- Now we expand the node with the lowest path cost
- Uses priority queues

Note: Gets stuck if there is a zero-cost action leading back to the same state.

For completeness and optimality, we require the cost of every step to be  $\geq \epsilon$ 

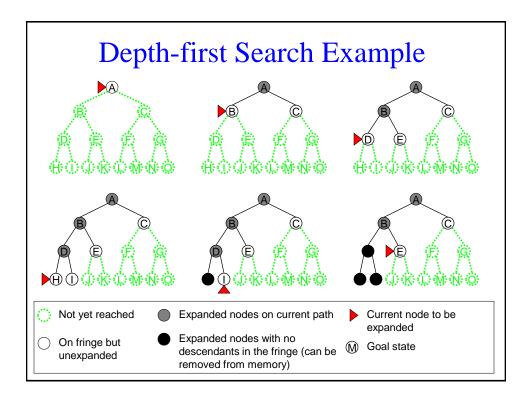
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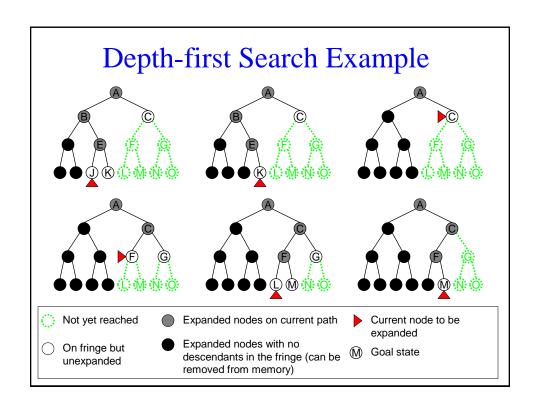
# **Evaluating Uniform-cost Search**

Complete?	Yes provided branching factor is
	finite and step costs $\geq \varepsilon$ for small
	positive ε
Optimal?	Yes
Time Complexity	$O(b^{1+floor(C^*/\epsilon)})$ where $C^*$ is the
	cost of the optimal solution
Space Complexity	$O(b^{1+floor(C^*/\epsilon)})$ where C* is the
	cost of the optimal solution

# Depth-first Search

- Expands the deepest node in the current fringe of the search tree
- Implemented with a LIFO queue





# **Evaluating Depth-first Search**

Complete?	
Optimal?	
Time Complexity	
Space Complexity	

# **Evaluating Depth-first Search**

Complete?	No (Might not terminate if it goes down an infinite path with no solutions)
Optimal?	No (Could expand a much longer path than the optimal one first)
Time Complexity	O(b <sup>m</sup> )
Space Complexity	O(bm)

31

## Depth-limited Search

- Solves infinite path problem by using predetermined depth limit *l*
- Nodes at depth l are treated as if they have no successors
- Can use knowledge of the problem to determine *l* (but in general you don't know this in advance)

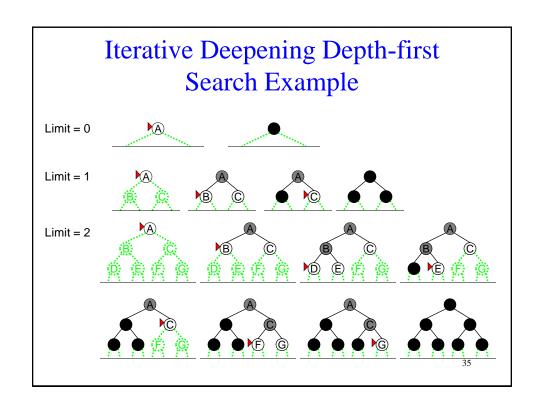
# **Evaluating Depth-limited Search**

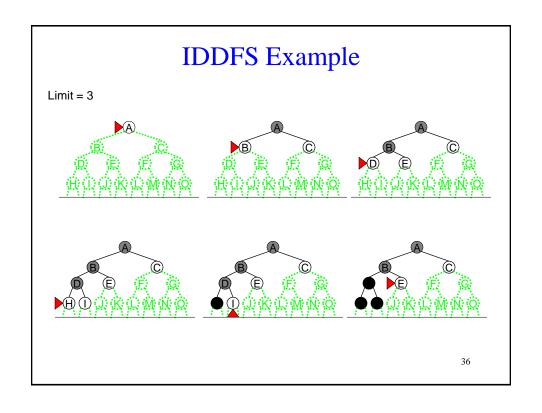
Complete?	No (If shallowest goal node beyond depth limit)
Optimal?	No (If depth limit > depth of shallowest goal node and we expand a much longer path than the optimal one first)
Time Complexity	$O(b^l)$
Space Complexity	O(bl)

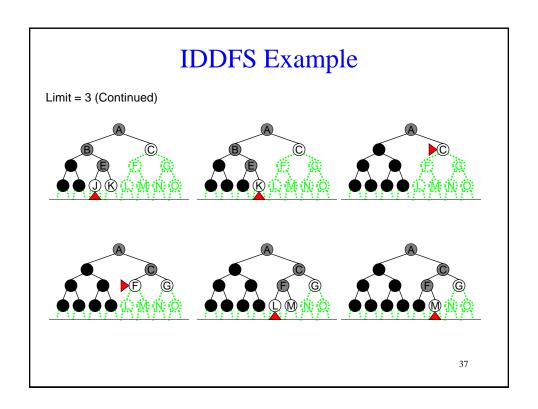
33

# Iterative Deepening Depth-first Search

- Do DFS with depth limit 0, 1, 2, ... until a goal is found
- Combines benefits of both DFS and BFS







# Evaluating Iterative Deepening Depth-first Search

Complete?	
Optimal?	
Time Complexity	
Space Complexity	

# Evaluating Iterative Deepening Depth-first Search

Complete?	Yes provided branching factor is finite
Optimal?	Yes if the path cost is a nondecreasing function of the depth of the node
Time Complexity	O(b <sup>d</sup> )
Space Complexity	O(bd)

39

## Isn't Iterative Deepening Wasteful?

- Actually, no! Most of the nodes are at the bottom level, doesn't matter that upper levels are generated multiple times.
- To see this, add up the 4th column below:

Depth	# of nodes	# of times generated	Total # of nodes generated at depth d
1	b	d	(d)b
2	$b^2$	d-1	(d-1)b <sup>2</sup>
:	:	:	:
d	b <sup>d</sup>	1	(1)b <sup>d</sup>

#### Is Iterative Deepening Wasteful?

Total # of nodes generated by iterative deepening:

Total # of nodes generated by BFS:

$$b + b^2 + ... + b^d + (b^{d+1}-b) = O(b^{d+1})$$

In general, iterative deepening is the preferred uninformed search method when there is a large search space and the depth of the solution is not known

41

#### **Bidirectional Search**

- Run one search forward from the initial state
- Run another search backward from the goal
- Stop when the two searches meet in the middle

#### **Bidirectional Search**

- Needs an efficiently computable Predecessor() function
- What if there are several goal states?
  - Create a new dummy goal state whose predecessors are the actual goal states
- Problematic if no efficient way to generate the set of all goal states and check for them in the forward search eg. "All states that lead to checkmate by move m<sub>1</sub>"

43

### **Evaluating Bidirectional Search**

Complete?	Yes provided branching factor is finite and both directions use BFS
Optimal?	Yes if the step costs are all identical and both directions use BFS
Time Complexity	$O(b^{d/2})$
Space Complexity	O(b <sup>d/2</sup> ) (At least one search tree must be kept in memory for the membership check)

#### **Avoiding Repeated States**

- Tradeoff between space and time!
- Need a closed list which stores every expanded node (memory requirements could make search infeasible)
- If the current node matches a node on the closed list, discard it (ie. discard the newly discovered path)
- We'll refer to this algorithm as GRAPH-SEARCH
- Is this optimal? Only for uniform-cost search or breadth-first search with constant step costs.

45

#### **GRAPH-SEARCH**

```
function Graph-Search(problem, fringe) returns a solution, or failure  closed \leftarrow \text{an empty set} \\ fringe \leftarrow \text{INSERT}(\text{Make-Node}(\text{Initial-State}[problem]), fringe) \\ \textbf{loop do} \\ \textbf{if } fringe \textbf{ is empty then return failure} \\ node \leftarrow \text{Remove-Front}(fringe) \\ \textbf{if } \text{Goal-Test}[problem](\text{State}[node]) \textbf{ then return } \text{Solution}(node) \\ \textbf{if } \text{State}[node] \textbf{ is not in } closed \textbf{ then} \\ \textbf{add } \text{State}[node] \textbf{ to } closed \\ fringe \leftarrow \text{INSERTALL}(\text{Expand}(node, problem), fringe) \\ \end{aligned}
```

# Things You Should Know

- How to formalize a search problem
- How BFS, UCS, DFS, DLS, IDS and Bidirectional search work
- Whether the above searches are complete and optimal plus their time and space complexity
- The pros and cons of the above searches