

#### Steps in Problem Solving

- Goal Formulation: based on current situation and the agent's performance measure is the first step in Problem Solving
- Problem Formulation-Deciding what actions and states to consider given a goal.
- Search for different possible actions and choose the best.
- A search algorithm takes problem as input and returns a solution in the form of an action sequence.

# Steps in Problem Solving

 The action recommended by the solution can be carried out, known as Execution.

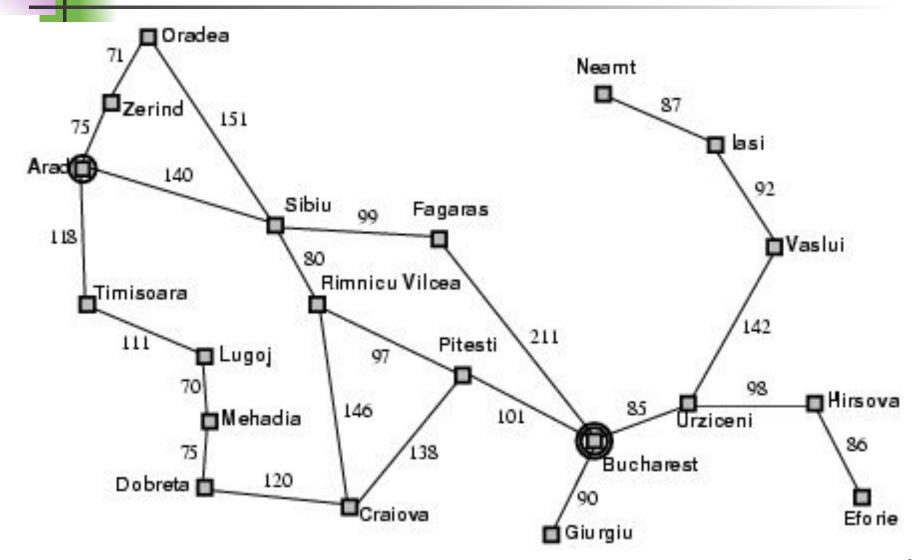
## Problem-solving agents

```
function SIMPLE-PROBLEM-SOLVING-AGENT (percept) returns an action
static: seq, an action sequence, initially empty
         state, some description of the current world state
         goal, a goal, initially null
         problem, a problem formulation
state \leftarrow \text{Update-State}(state, percept)
if seq is empty then do
     goal \leftarrow FORMULATE-GOAL(state)
     problem \leftarrow Formulate-Problem(state, goal)
     seq \leftarrow Search(problem)
action \leftarrow First(seq)
seq \leftarrow Rest(seq)
return action
```

# Example: Romania

- On holiday in Romania; currently in Arad.
- Flight leaves tomorrow from Bucharest
- Formulate goal:
  - be in Bucharest
- Formulate problem:
  - states: various cities
  - actions: drive between cities
- Find solution:
  - sequence of cities, e.g., Arad, Sibiu, Fagaras, Bucharest

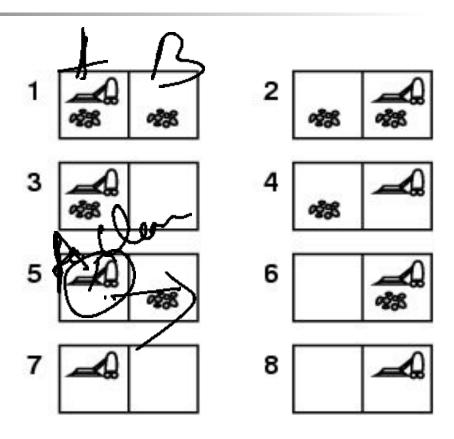
## Example: Romania



## Problem types

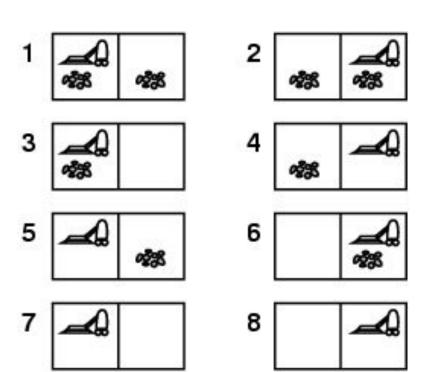
- Deterministic, fully observable □ single-state problem
  - Agent knows exactly which state it will be in; solution is a sequence
- Non-observable ☐ sensorless problem (conformant problem)
  - Agent may have no idea where it is; solution is a sequence
- Nondeterministic and/or partially observable □ contingency problem
  - percepts provide new information about current state
  - often interleave} search, execution
- Unknown state space □ exploration problem

Single-state, start in #5. Solution?

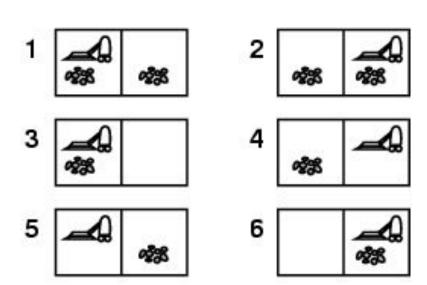


Single-state, start in #5. Solution? [Right, Suck]

Sensorless, start in {1,2,3,4,5,6,7,8} e.g., Right goes to {2,4,6,8} Solution?



Sensorless, start in {1,2,3,4,5,6,7,8} e.g., Right goes to {2,4,6,8} Solution? [Right,Suck,Left,Suck]



8

- Contingency
  - Nondeterministic: Suck may dirty a clean carpet
  - Partially observable: location, dirt at current location.
  - Percept: [L, Clean], i.e., start in #5 or #7 Solution?

Sensorless, start in {1,2,3,4,5,6,7,8} e.g., Right goes to {2,4,6,8} Solution? [Right,Suck,Left,Suck]

- 3 20
- 4
- 5
- 6

- Contingency
  - Nondeterministic: Suck may dirty a clean carpet
- 7 🕰
- 8 40
- Partially observable: location, dirt at current location.
- Percept: [L, Clean], i.e., start in #5 or #7 Solution? [Right, if dirt then Suck]



#### Assumptions for the Environment

- Observable: Knows the current state.
- Discrete: Only finitely many actions
- Known : Result of action is known.
- Deterministic : Each action has only one action.

# 1

#### Single-state problem formulation

A problem is defined by four items:

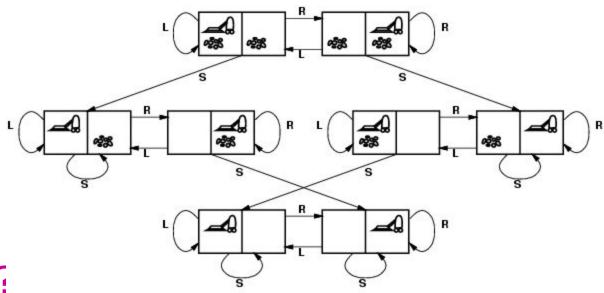
- initial state e.g., "at Arad"
- actions or successor function S(x) = set of action-state pairs
  - e.g.,  $S(Arad) = \{ \langle Arad \square Zerind, Zerind \rangle, \dots \}$
- goal test, can be
  - explicit, e.g., x = "at Bucharest"
  - implicit, e.g., Checkmate(x)
- path cost (additive)
  - e.g., sum of distances, number of actions executed, etc.
  - c(x,a,y) is the step cost, assumed to be  $\geq 0$
- A solution is a sequence of actions leading from the initial state to a goal state

# Selecting a state space

- Real world is absurdly complex
  - ☐ state space must be abstracted for problem solving
- (Abstract) state = set of real states
- (Abstract) action = complex combination of real actions
  - e.g., "Arad □ Zerind" represents a complex set of possible routes, detours, rest stops, etc.
- For guaranteed realizability, any real state "in Arad" must get to some real state "in Zerind"
- (Abstract) solution =
  - set of real paths that are solutions in the real world

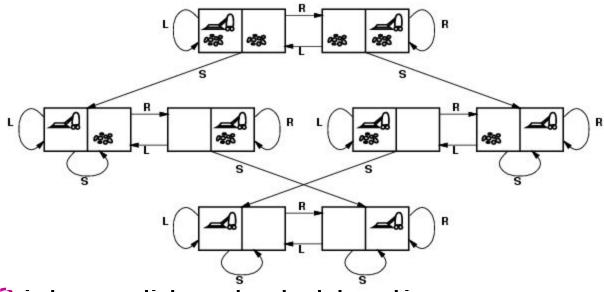


#### Vacuum world state space graph



- states:
- actions?
- goal test?
- path cost?

#### Vacuum world state space graph

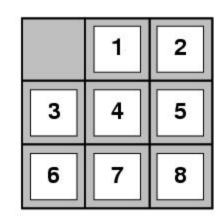


- <u>states?</u> integer dirt and robot location
- <u>actions?</u> Left, Right, Suck
- goal test? no dirt at all locations
- path cost? 1 per action



#### Example: The 8-puzzle





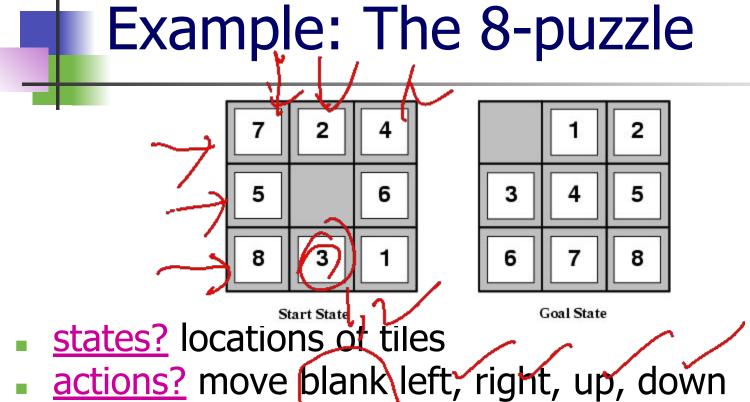
states?

actions?

goal test?

path cost?

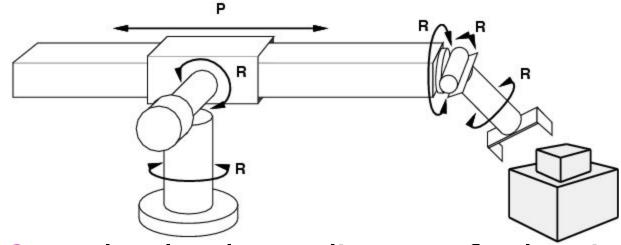
Start State Goal State



- goal test? = goal state (given)
- path cost? 1 per move

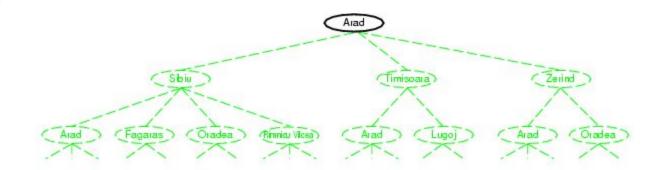
[Note: optimal solution of *n*-Puzzle family is NP-hard]

## Example: robotic assembly

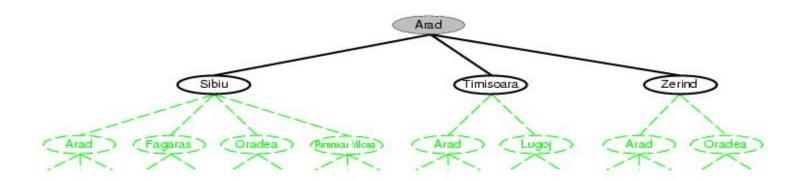


- <u>states?</u>: real-valued coordinates of robot joint angles parts of the object to be assembled
- <u>actions?</u>: continuous motions of robot joints
- goal test?: complete assembly
- path cost?: time to execute

#### Tree search example

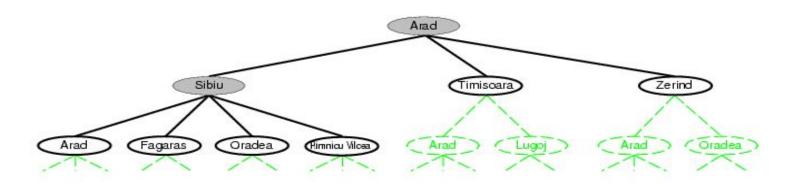


# Tree search example





#### Tree search example





- function TREE-SEARCH(problem) returns a solution, or failure
- initialize the frontier using the initial state of problem
- loop do
- if the frontier is empty then return failure
- choose a leaf node and remove it from the frontier
- if the node contains a goal state then return the corresponding solution
- expand the chosen node, adding the resulting nodes to the frontier

 \* frontier –The set of all leaf nodes available for expansion at any given point

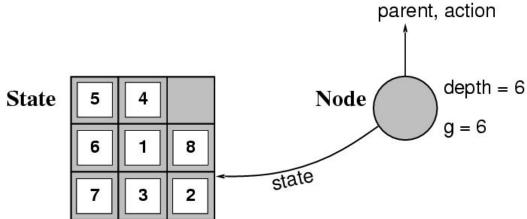


- function GRAPH-SEARCH(problem) returns a solution, or failure
- initialize the frontier using the initial state of problem
- initialize the explored set to be empty
- loop do
- if the frontier is empty then return failure
- choose a leaf node and remove it from the frontier
- if the node contains a goal state then return the corresponding solution
- add the node to the explored set
- expand the chosen node, adding the resulting nodes to the frontier
- only if not in the frontier or explored set



#### Implementation: states vs. nodes

- A state is a (representation of) a physical configuration
- A node is a data structure constituting part of a search tree includes state, parent node, action, path cost g(x), depth



The Expand function creates new nodes, filling in the various fields and using the SuccessorFn of the problem to create the corresponding states.



- function CHILD-NODE(problem, parent, action) returns a node
- return a node with
- STATE = problem.RESULT(parent.STATE, action),
- PARENT = parent, ACTION = action,
- PATH-COST = parent.PATH-COST + problem.STEP-COST(parent.STATE, action)

# Search strategies

- A search strategy is defined by picking the order of node expansion
- Strategies are evaluated along the following dimensions:
  - completeness: does it always find a solution if one exists?
  - time complexity: number of nodes generated
  - space complexity: maximum number of nodes in memory
  - optimality: does it always find a least-cost solution?
- Time and space complexity are measured in terms of
  - b: maximum branching factor of the search tree// no.of children at each node
  - d: depth of the least-cost solution
  - m: maximum depth of the state space (may be  $\infty$ )

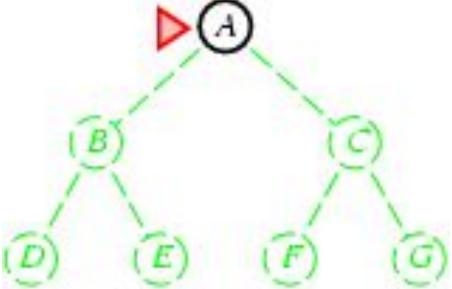
# Uninformed search strategies

- Uninformed search strategies use only the information available in the problem definition
- Breadth-first search
- Uniform-cost search
- Depth-first search
- Depth-limited search
- Iterative deepening search

Expand shallowest unexpanded node

Implementation:

 fringe is a FIFO queue, i.e., new successors go at end

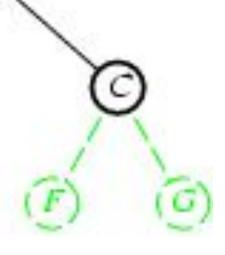


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at end



Expand shallowest unexpanded node

Implementation:

• fringe is a FIFO queue, i.e., new successors go







function BREADTH-FIRST-SEARCH(problem) returns a solution, or failure

- •node ←a node with STATE = problem.INITIAL-STATE, PATH-COST = 0
- if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
- frontier ← a FIFO queue with node as the only element
- explored ← an empty set
- loop do if EMPTY?(frontier) then return failure
- •node ← POP(frontier ) /\* chooses the shallowest node in frontier \*/



- add node.STATE to explored
- for each action in problem.ACTIONS(node.STATE) do
- child ← CHILD-NODE(problem, node, action)
- if child.STATE is not in explored or frontier then
- if problem.GOAL-TEST(child.STATE) then return SOLUTION(child)
- frontier ← INSERT(child,frontier )

#### Properties of breadth-first search

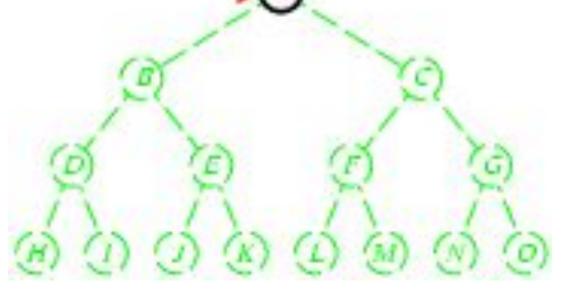
- Complete? Yes (if b is finite)
- Time?  $1+b+b^2+b^3+...+b^d+b(b^d-1)=O(b^{d+1})$
- Space?  $O(b^{d+1})$  (keeps every node in memory)
- Optimal? Yes (if cost = 1 per step)

Space is the bigger problem (more than time)

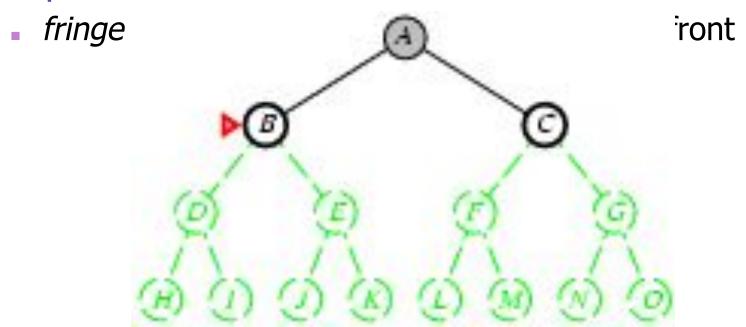
## Depth-first search

- Expand deepest unexpanded node
- Implementation:

fringe = LIFO queue, i.e., put successors at front

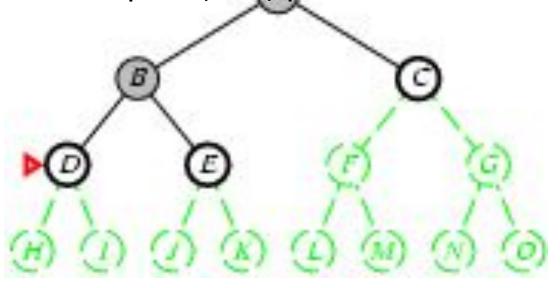


- Expand deepest unexpanded node
- Implementation:



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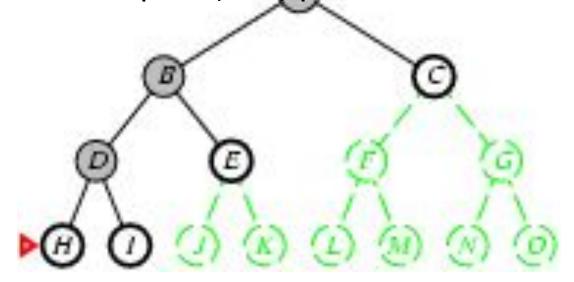
fringe = LIFO queue, i.e., put successors at front



Expand deepest unexpanded node

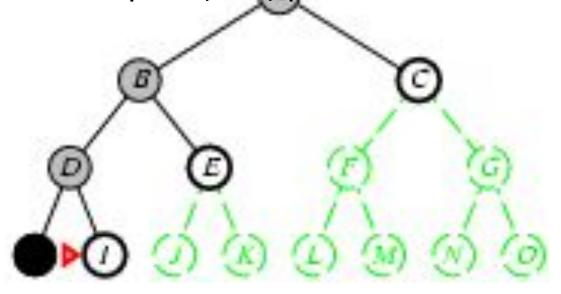
#### Implementation:

fringe = LIFO queue, i.e., put successors at front

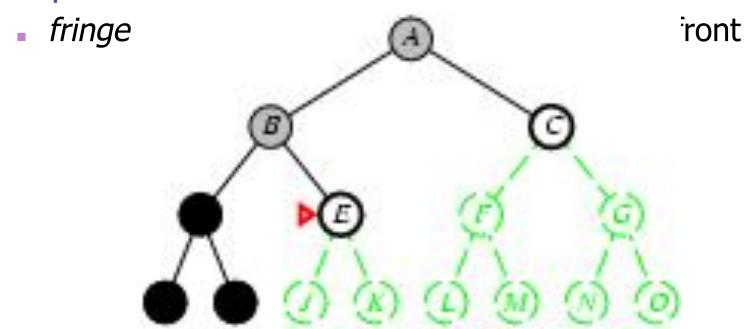


- Expand deepest unexpanded node
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- Expand deepest unexpanded node
- Implementation:

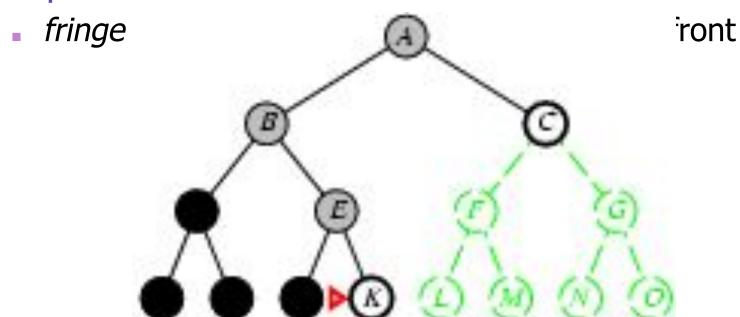


Expand deepest unexpanded node

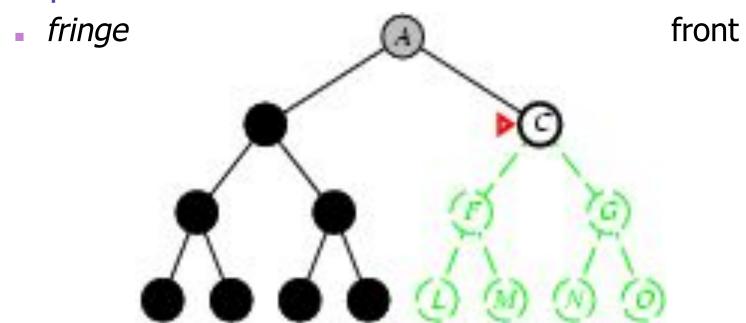
#### Implementation:

Fringe

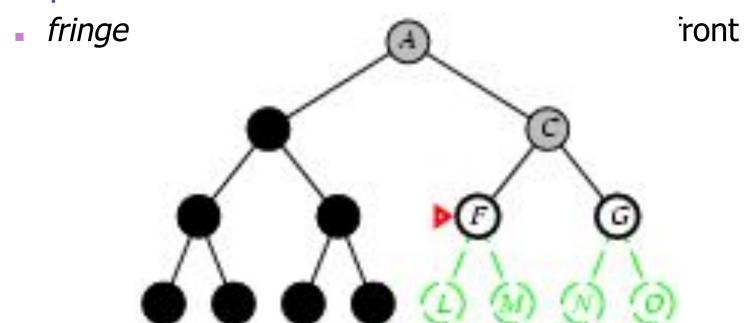
- Expand deepest unexpanded node
- Implementation:



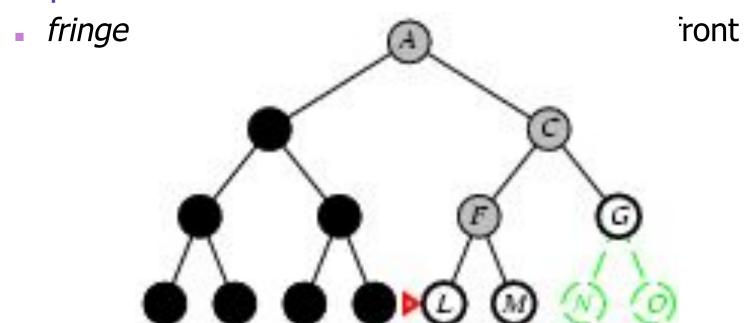
- Expand deepest unexpanded node
- Implementation:



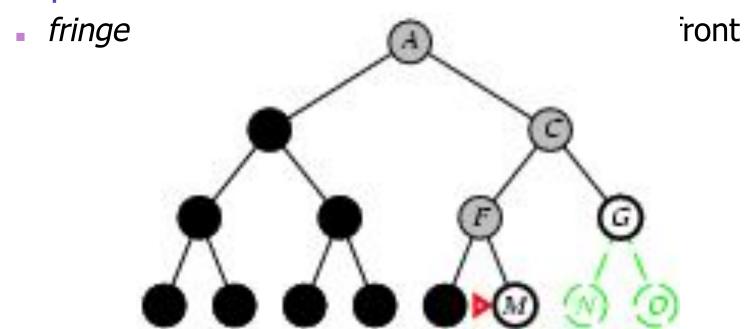
- Expand deepest unexpanded node
- Implementation:



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- Implementation:



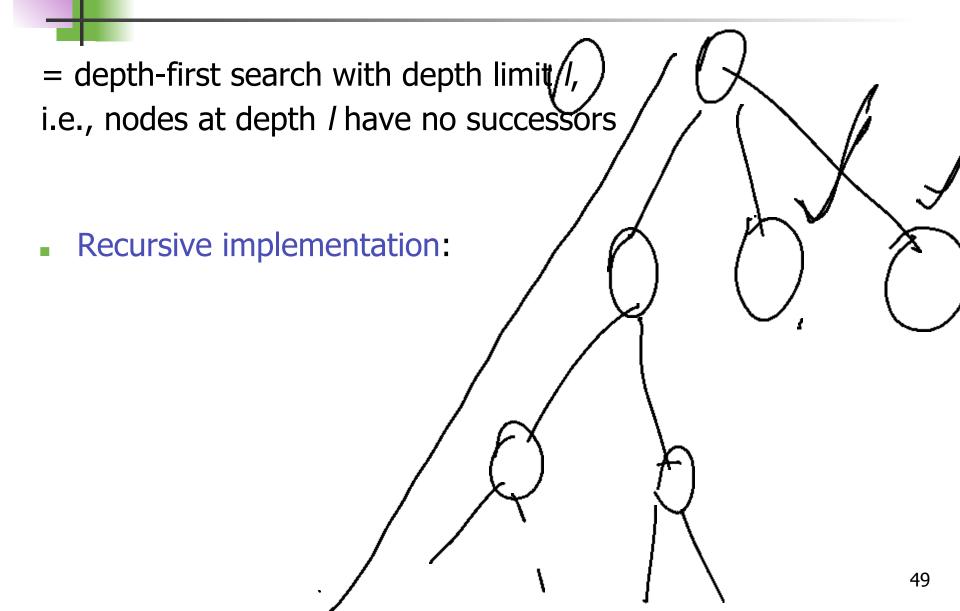
- Expand deepest unexpanded node
- Implementation:



## Properties of depth-first search

- <u>Complete?</u> No: fails in infinite-depth spaces, spaces with loops
  - Modify to avoid repeated states along path
    - ☐ complete in finite spaces
- Time?  $O(b^m)$ : terrible if m is much larger than d
  - but if solutions are dense, may be much faster than breadth-first
- Space? O(bm), i.e., linear space!

### Depth-limited search



- function DEPTH-LIMITED-SEARCH(problem, limit) returns a solution, or
- failure/cutoff
- return RECURSIVE-DLS(MAKE-NODE(problem.INITIAL-STATE), problem, limit
- function RECURSIVE-DLS(node,problem,limit) returns a solution, or
- failure/cutoff
- if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
- else if limit = 0 then return cutoff
- else
- cutoff occurred? ← false
- for each action in problem.ACTIONS(node.STATE) do
- child ← CHILD-NODE(problem,node,action)
- result ← RECURSIVE-DLS(child,problem,limit 1)
- if result = cutoff then
- cutoffoccurred? ← true
- **else if** result ≠failure **then** return result
- if cutoff occurred? then return cutoff else return failure

### Iterative deepening search

```
function Iterative-Deepening-Search (problem) returns a solution, or failure  \begin{array}{c} \text{inputs: } problem, \text{ a problem} \\ \text{for } depth \leftarrow \text{ 0 to } \infty \text{ do} \\ result \leftarrow \text{Depth-Limited-Search} (problem, depth) \\ \text{if } result \neq \text{cutoff then return } result \end{array}
```

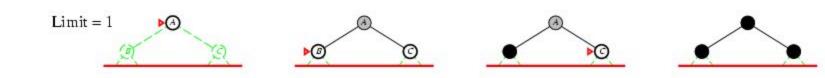
# 4

### Iterative deepening search I = 0

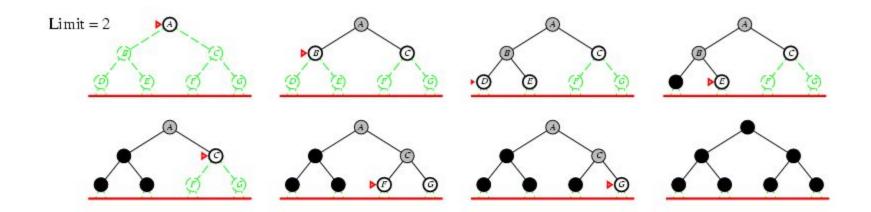
Limit = 0



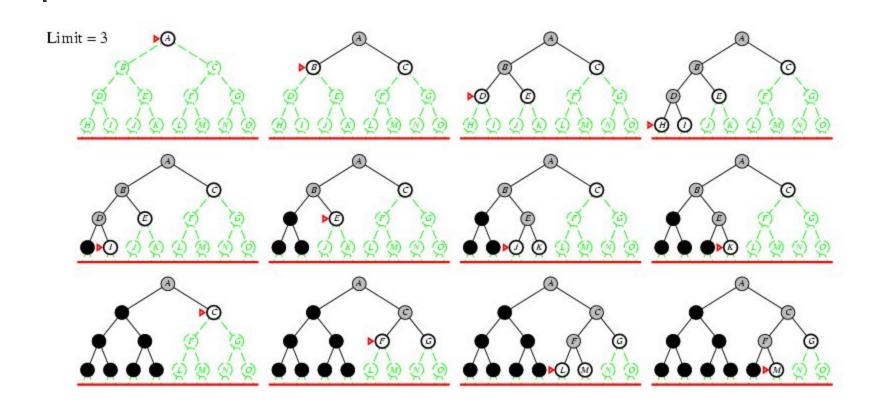
### Iterative deepening search l=1



### Iterative deepening search l = 2



### Iterative deepening search I = 3



### Iterative deepening search

 Number of nodes generated in a depth-limited search to depth d with branching factor b:

$$N_{D/S} = b^0 + b^1 + b^2 + ... + b^{d-2} + b^{d-1} + b^d$$

 Number of nodes generated in an iterative deepening search to depth d with branching factor b:

$$N_{IDS} = (d+1)b^{0} + db^{-1} + (d-1)b^{-2} + ... + 3b^{d-2} + 2b^{d-1} + 1b^{d}$$
  
//last term is bottom level

- For b = 10, d = 5,
  - $N_{DIS} = 1 + 10 + 100 + 1,000 + 10,000 + 100,000 = 111,111$
  - $N_{IDS} = 6 + 50 + 400 + 3,000 + 20,000 + 100,000 = 123,456$

# deepening search

Complete? Yes

• Time? 
$$(d+1)b^0 + db^1 + (d-1)b^2 + ... + b^d = O(b^d)$$

- Space? O(bd)
- Optimal? Yes, if step cost = 1