



Solving problems by searching

Chapter 3



Outline

- Problem-solving agents
- Problem types
- Problem formulation
- Example problems
- Basic search algorithms



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

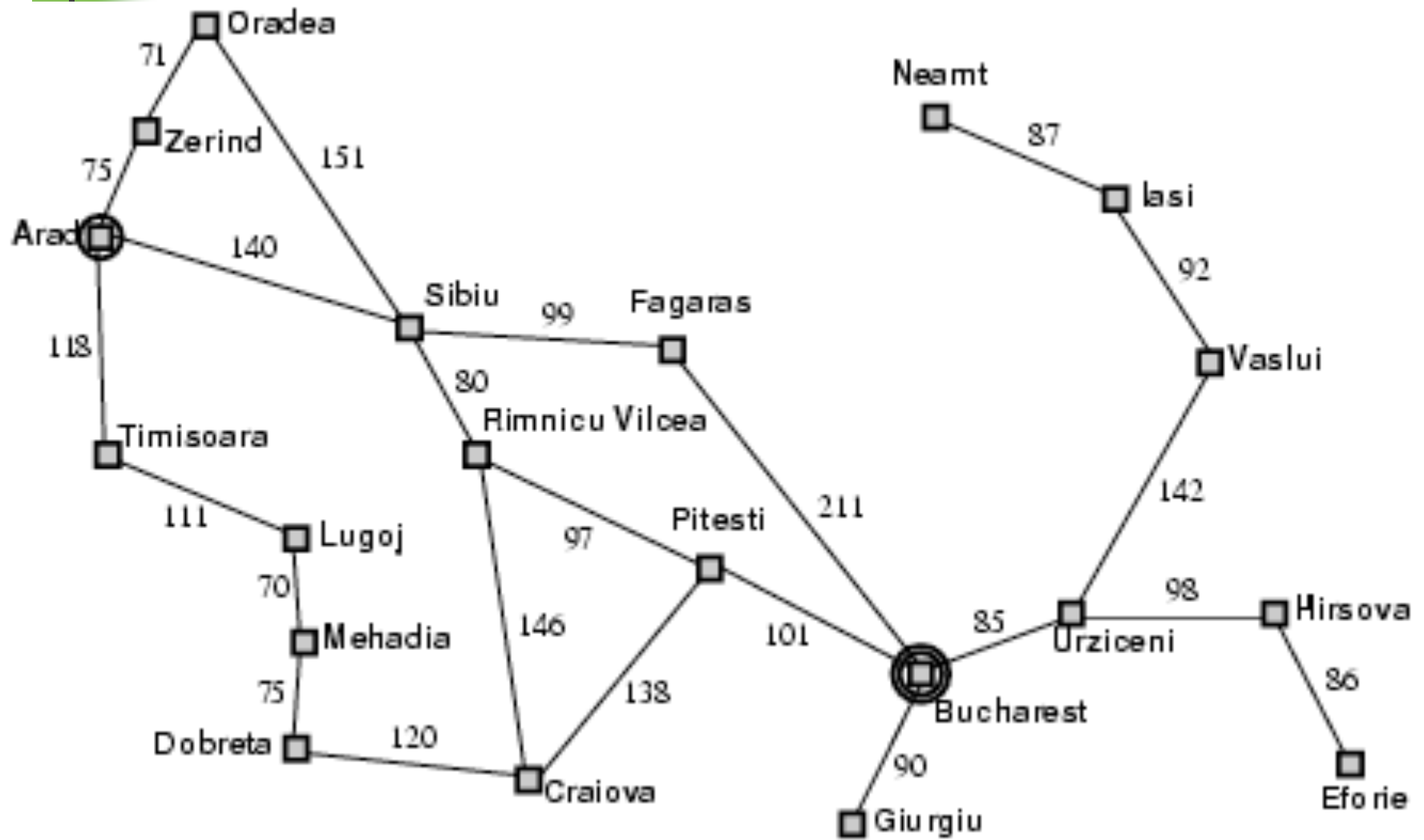
  state ← UPDATE-STATE(state, percept)
  if seq is empty then do
    goal ← FORMULATE-GOAL(state)
    problem ← FORMULATE-PROBLEM(state, goal)
    seq ← SEARCH(problem)
  action ← FIRST(seq)
  seq ← REST(seq)
  return action
```



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





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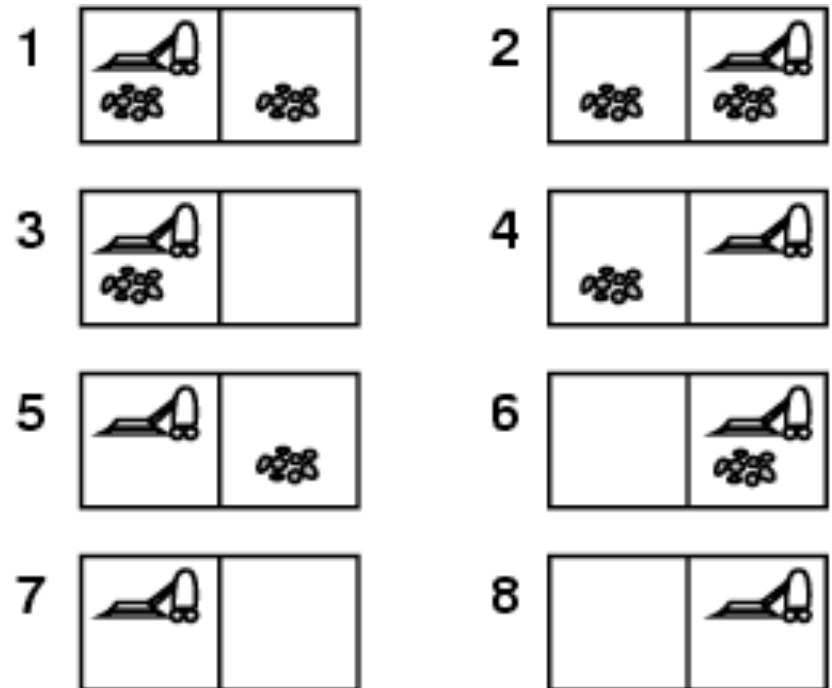
Problem types

- 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

Example: vacuum world

- Single-state, start in #5.

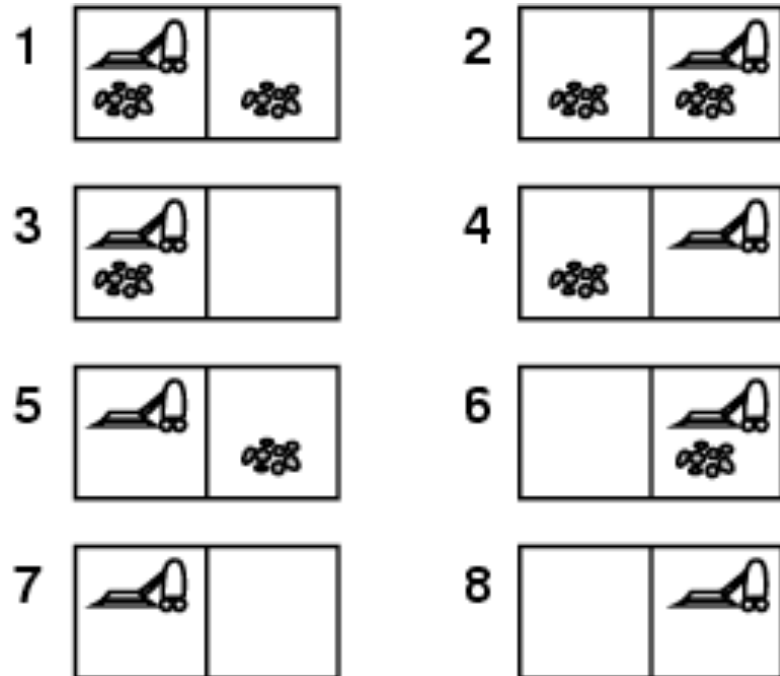
Solution?



Example: vacuum world

- 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?



Example: vacuum world

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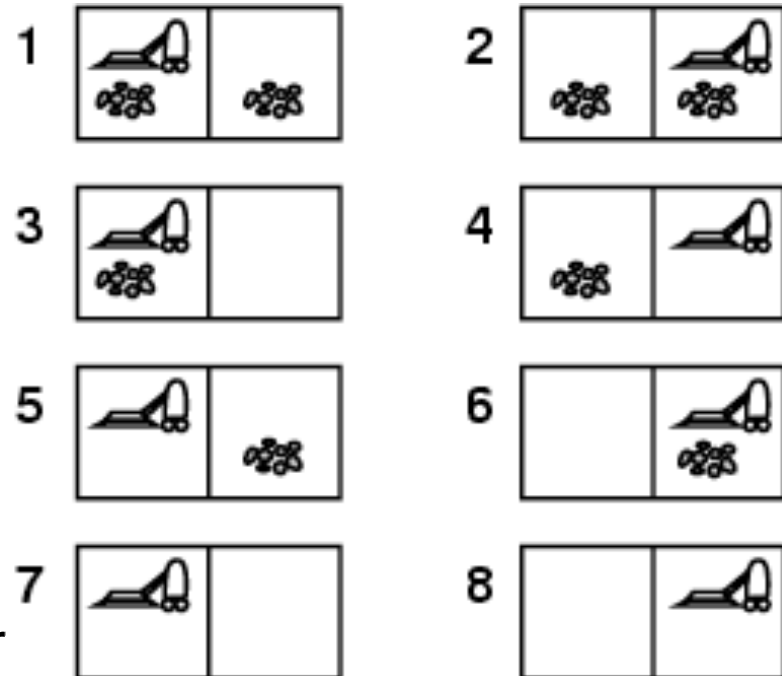
Solution?

[Right,Suck,Left,Suck]

- Contingency

- Nondeterministic: *Suck* may dirty a clean carpet
- Partially observable: location, dir
- Percept: *[L, Clean]*, i.e., start in #5 or #7

Solution?



Example: vacuum world

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

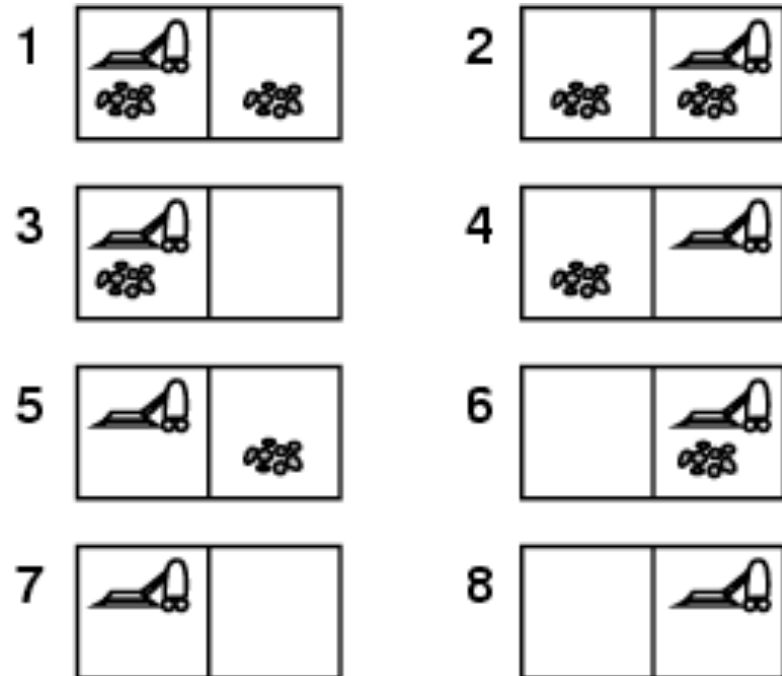
Solution?

[Right,Suck,Left,Suck]

- 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? *[Right, **if dirt then Suck]***





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Single-state problem formulation

A **problem** is defined by four items:

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



Selecting a state space

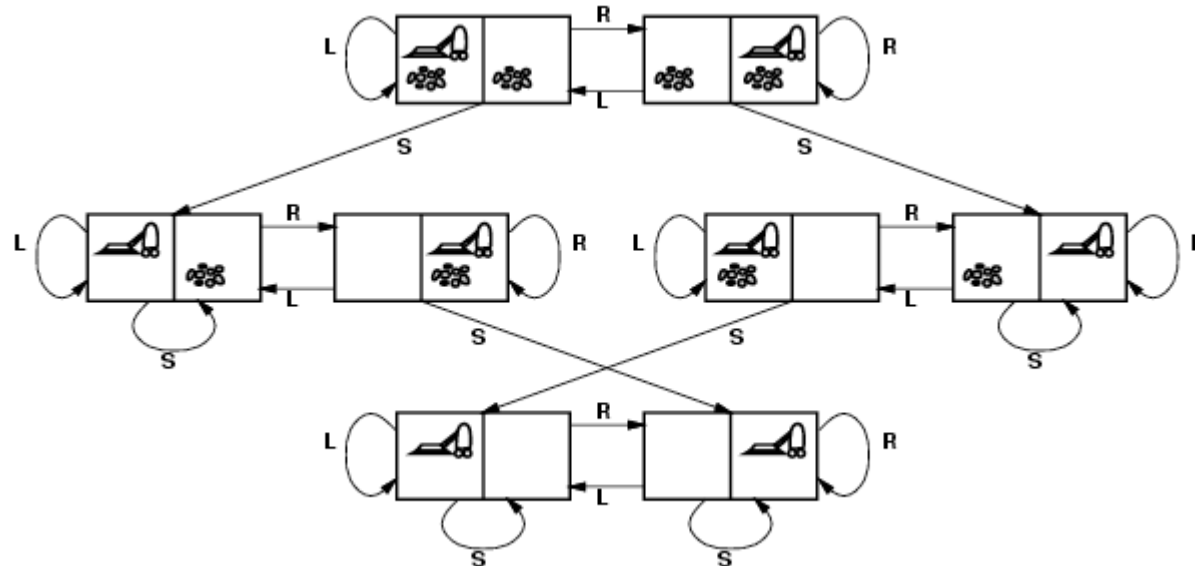
- Real world is absurdly complex
 - state space must be **abstracted** for problem solving
- (Abstract) state = set of real states
- (Abstract) action = complex combination of real actions
 - e.g., "Arad → Zerind" represents a complex set of possible routes, detours, rest stops, etc.
- For guaranteed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



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Vacuum world state space graph



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

Example: The 8-puzzle

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

Goal State

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

Example: The 8-puzzle

7	2	4
5		6
8	3	1

Start State

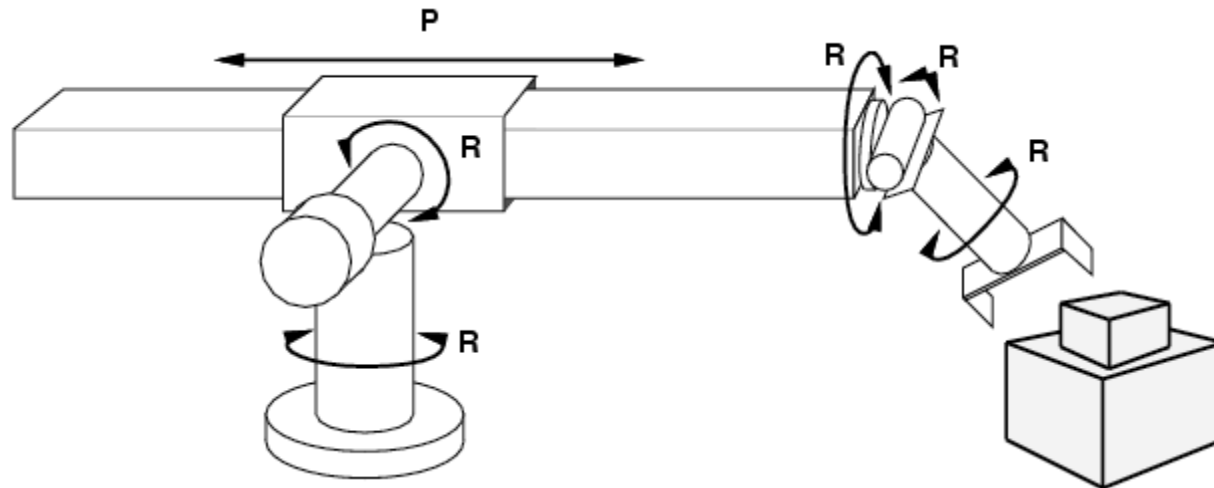
	1	2
3	4	5
6	7	8

Goal State

- states? locations of tiles
- actions? move blank left, right, up, down
- goal test? = goal state (given)
- path cost? 1 per move

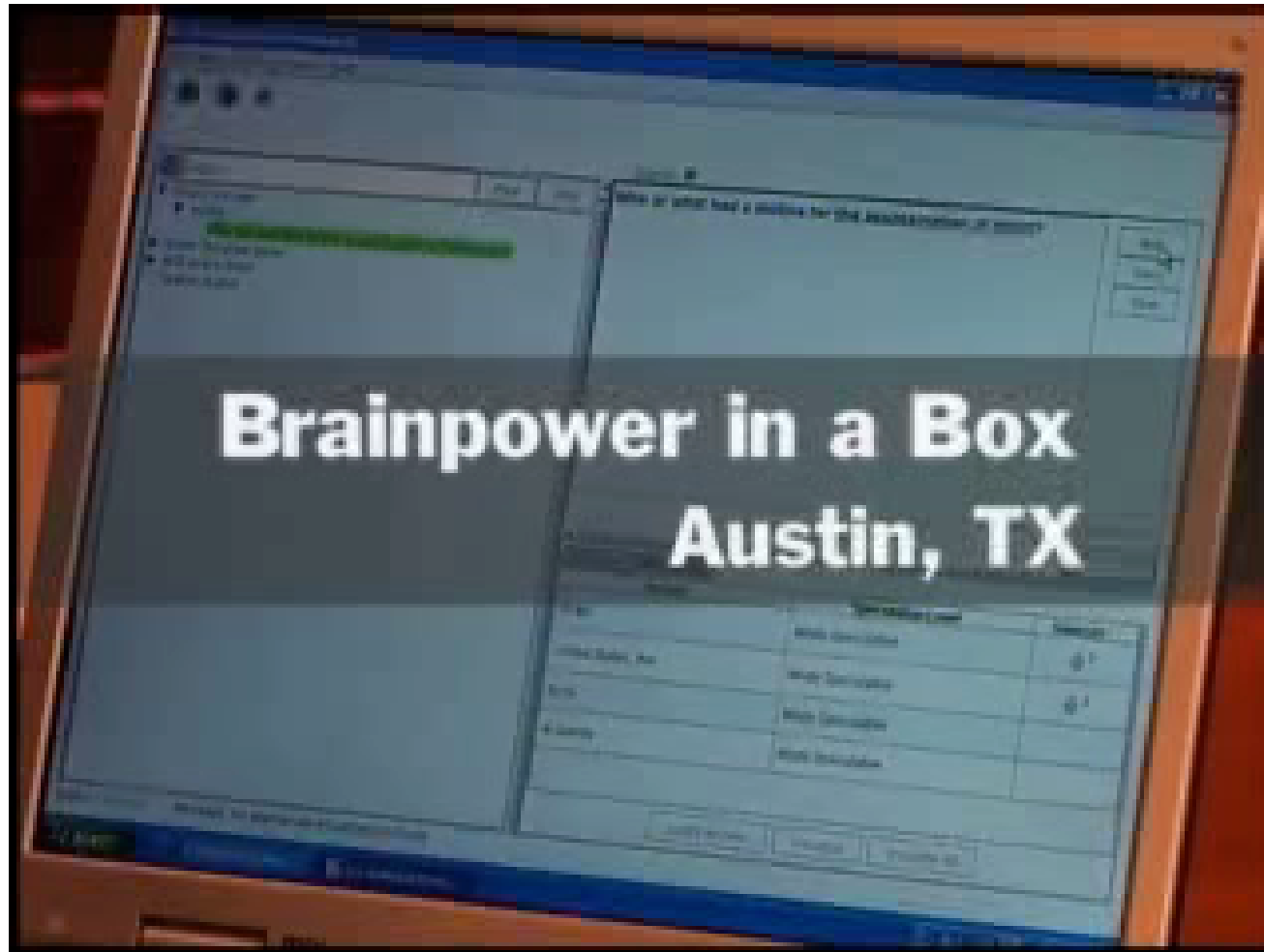
[Note: optimal solution of n -Puzzle family is NP-hard]

Example: robotic assembly



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

Computers & Common Sense: Cyc





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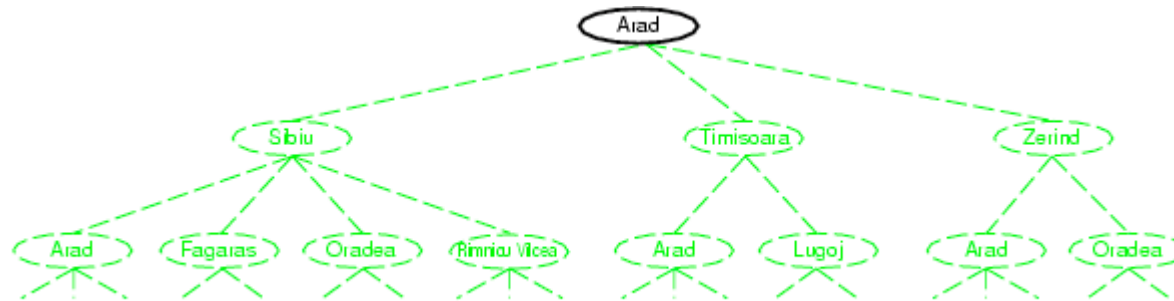


Tree search algorithms

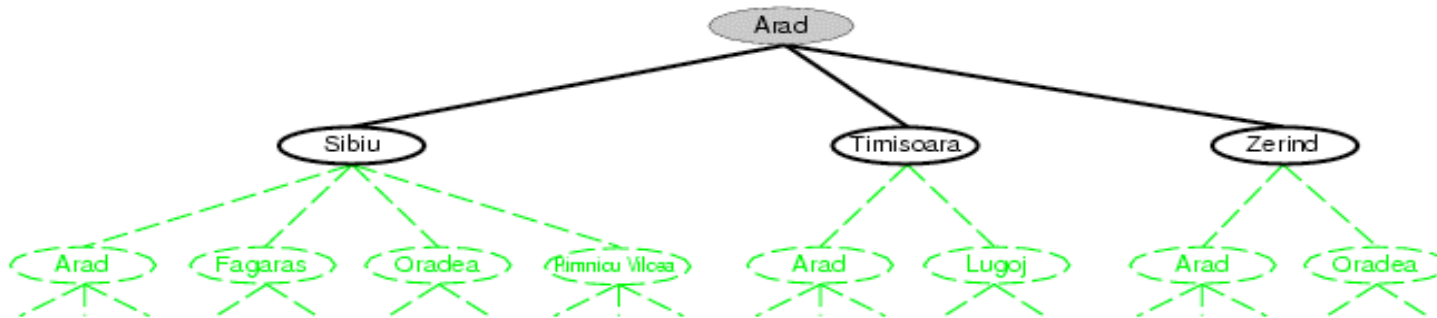
- Basic idea:
 - offline, simulated exploration of state space by generating successors of already-explored states (a.k.a. ~**expanding** states)

```
function TREE-SEARCH(problem, strategy) returns a solution, or failure
  initialize the search tree using the initial state of problem
  loop do
    if there are no candidates for expansion then return failure
    choose a leaf node for expansion according to strategy
    if the node contains a goal state then return the corresponding solution
    else expand the node and add the resulting nodes to the search tree
```

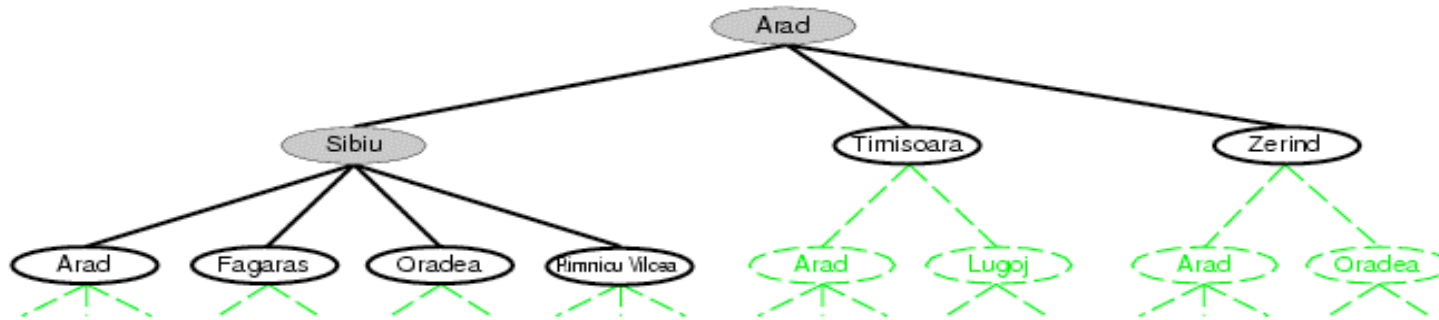
Tree search example



Tree search example



Tree search example





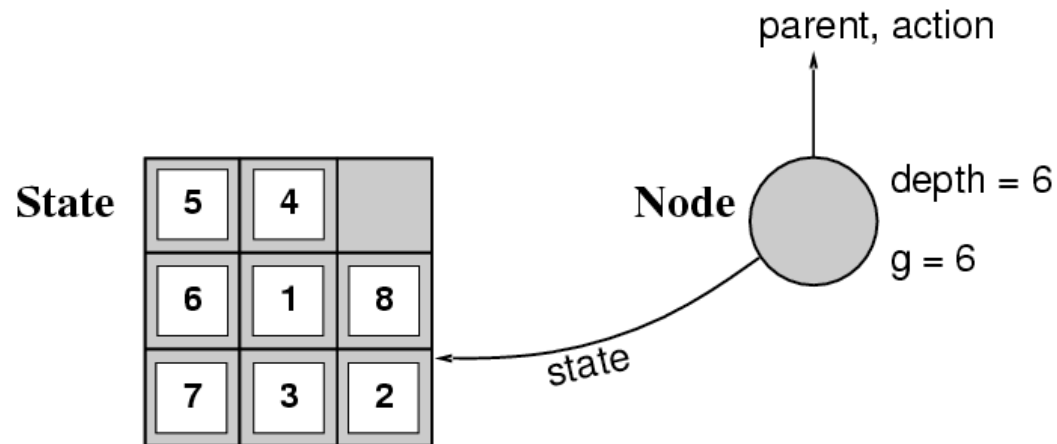
Implementation: general tree search

```
function TREE-SEARCH(problem, fringe) returns a solution, or failure  
fringe ← INSERT(MAKE-NODE(INITIAL-STATE[problem]), fringe)  
loop do  
  if fringe is empty then return failure  
  node ← REMOVE-FRONT(fringe)  
  if GOAL-TEST[problem](STATE[node]) then return SOLUTION(node)  
  fringe ← INSERTALL(EXPAND(node, problem), fringe)
```

```
function EXPAND(node, problem) returns a set of nodes  
successors ← the empty set  
for each action, result in SUCCESSOR-FN[problem](STATE[node]) do  
  s ← a new NODE  
  PARENT-NODE[s] ← node; ACTION[s] ← action; STATE[s] ← result  
  PATH-COST[s] ← PATH-COST[node] + STEP-COST(node, action, s)  
  DEPTH[s] ← DEPTH[node] + 1  
  add s to successors  
return successors
```

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.



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
 - *d*: depth of the least-cost solution
 - *m*: maximum depth of the state space (may be ∞)

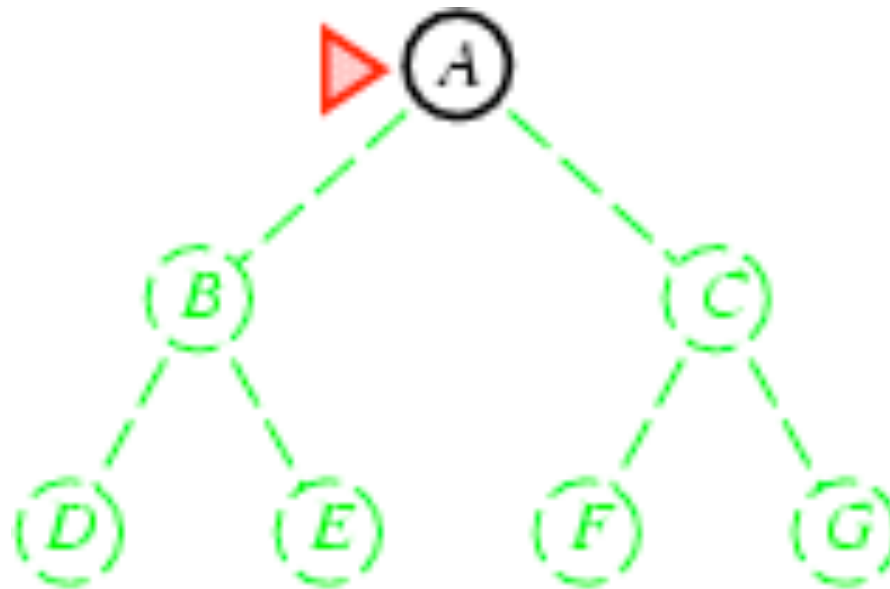


Uninformed search strategies

- **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

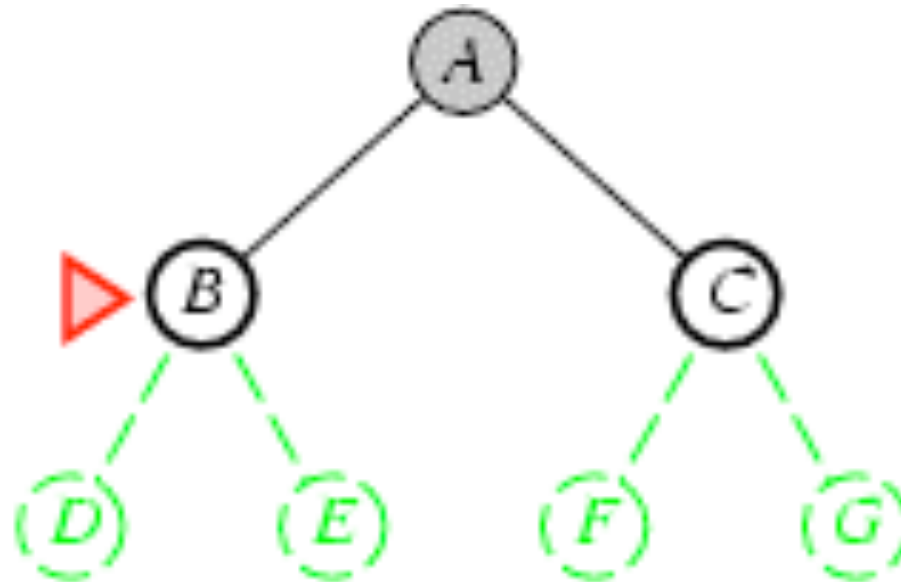
Breadth-first search

- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a FIFO queue, i.e., new successors go at end



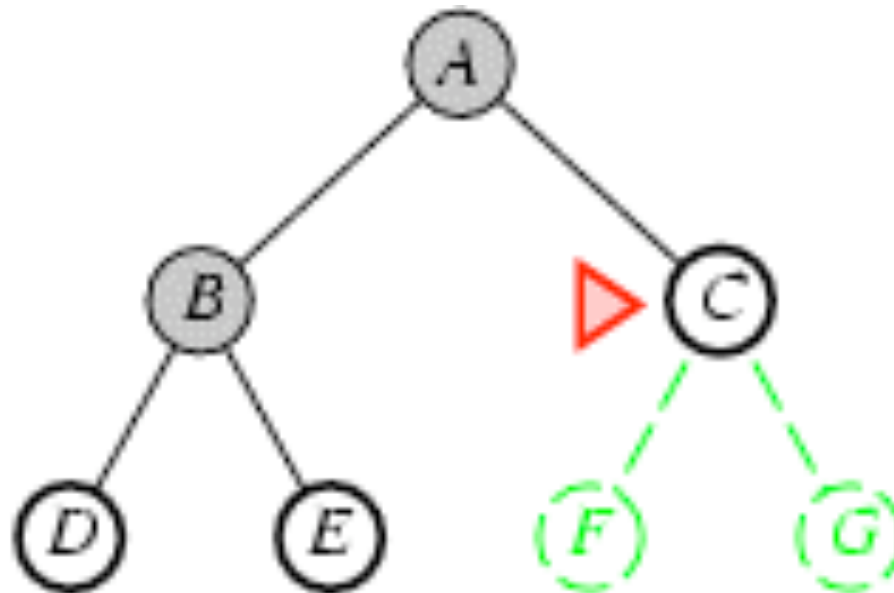
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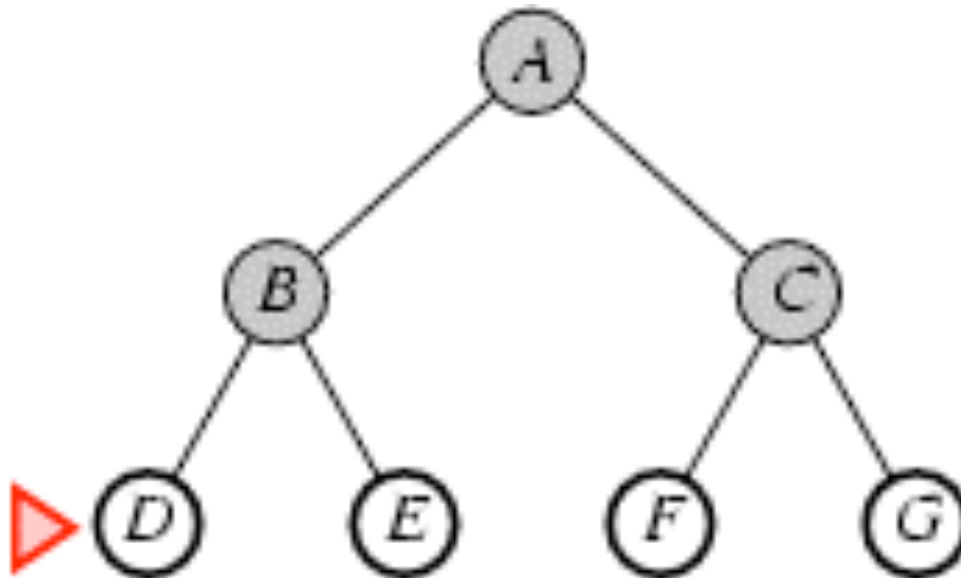
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Breadth-first search

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Properties of breadth-first search

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

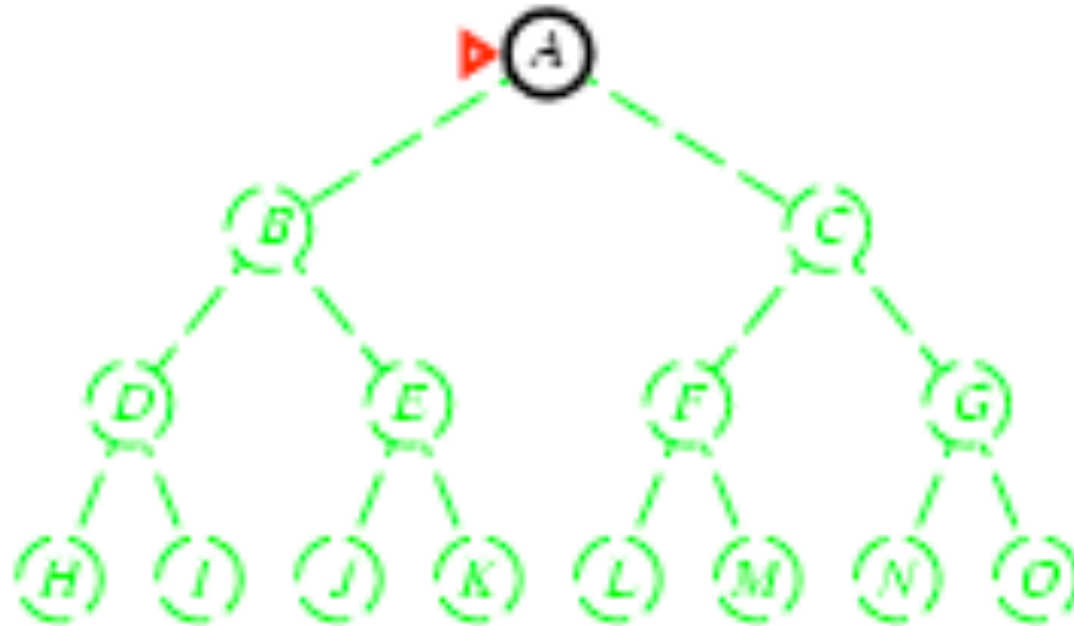
- **Space** is the bigger problem (more than time)

Uniform-cost search

- Expand least-cost unexpanded node
- **Implementation:**
 - *fringe* = queue ordered by path cost
- Equivalent to breadth-first if step costs all equal
- g is the optimum cost from init state to current state
- **Complete?** Yes, if step cost $\geq \epsilon$
- **Time?** # of nodes with $g \leq$ cost of optimal solution, $O(b^{\text{ceiling}(C^*/\epsilon)})$ where C^* is the cost of the optimal solution
- **Space?** # of nodes with $g \leq$ cost of optimal solution, $O(b^{\text{ceiling}(C^*/\epsilon)})$
- **Optimal?** Yes – nodes expanded in increasing order of $g(n)$

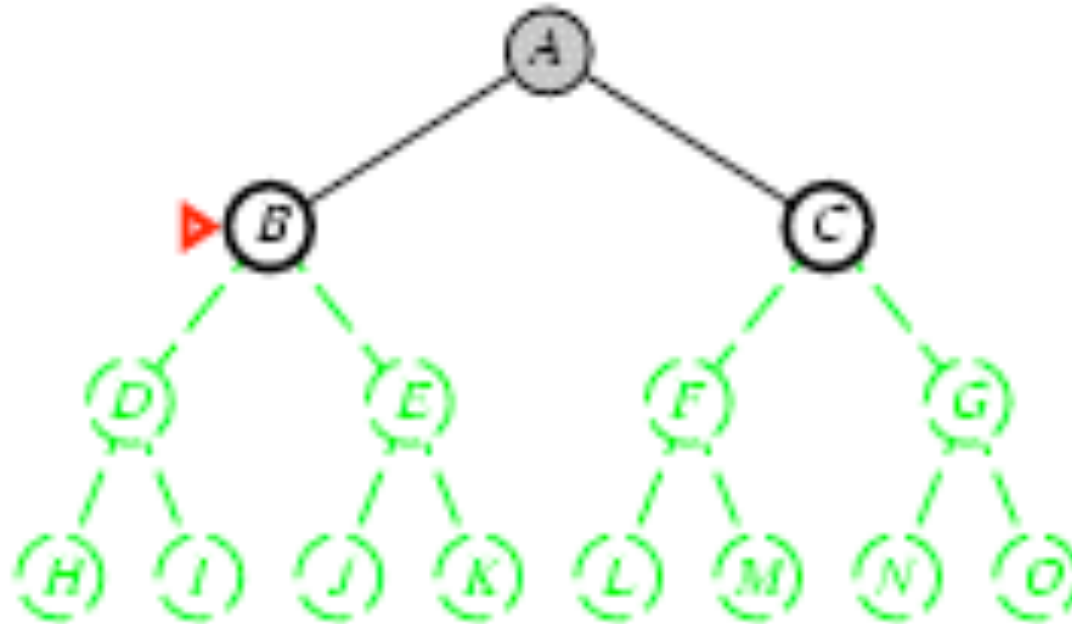
Depth-first search

- Expand deepest unexpanded node
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 - *fringe* = LIFO queue, i.e., put successors at front



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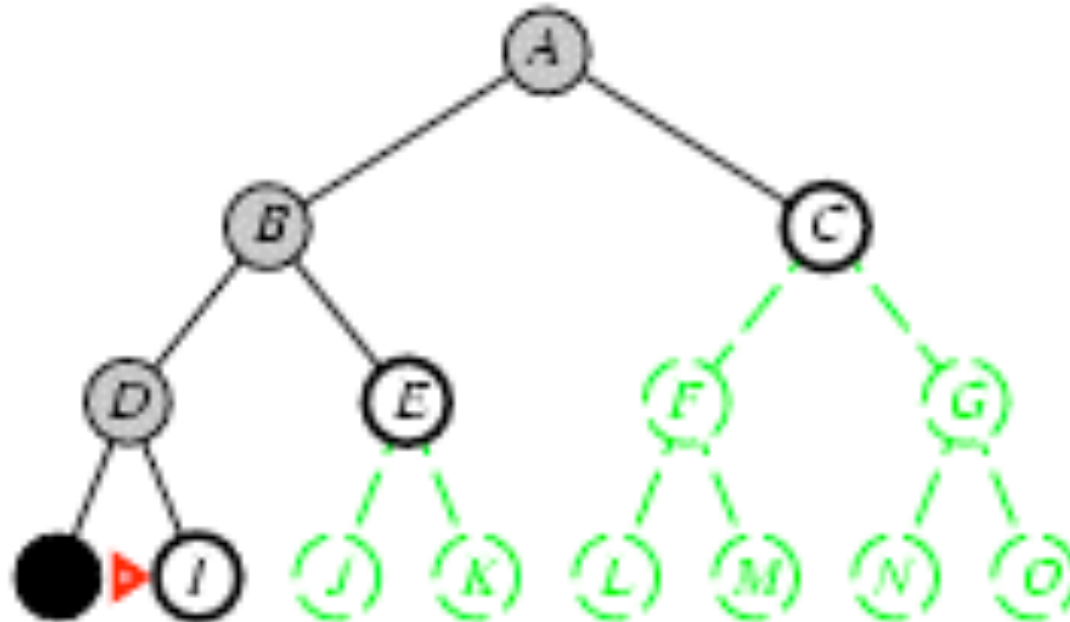
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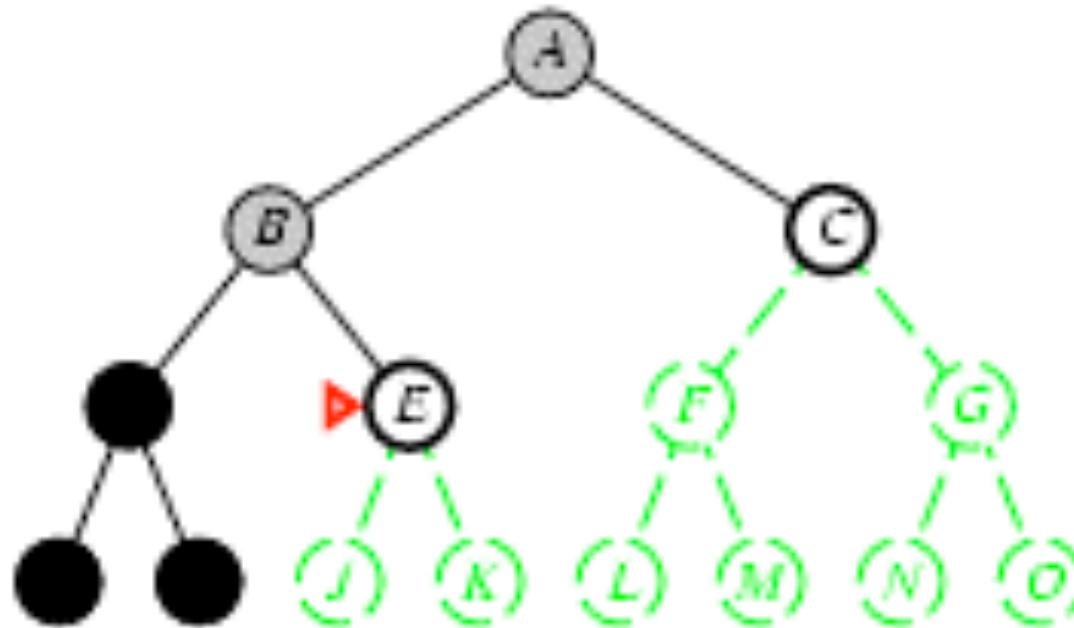
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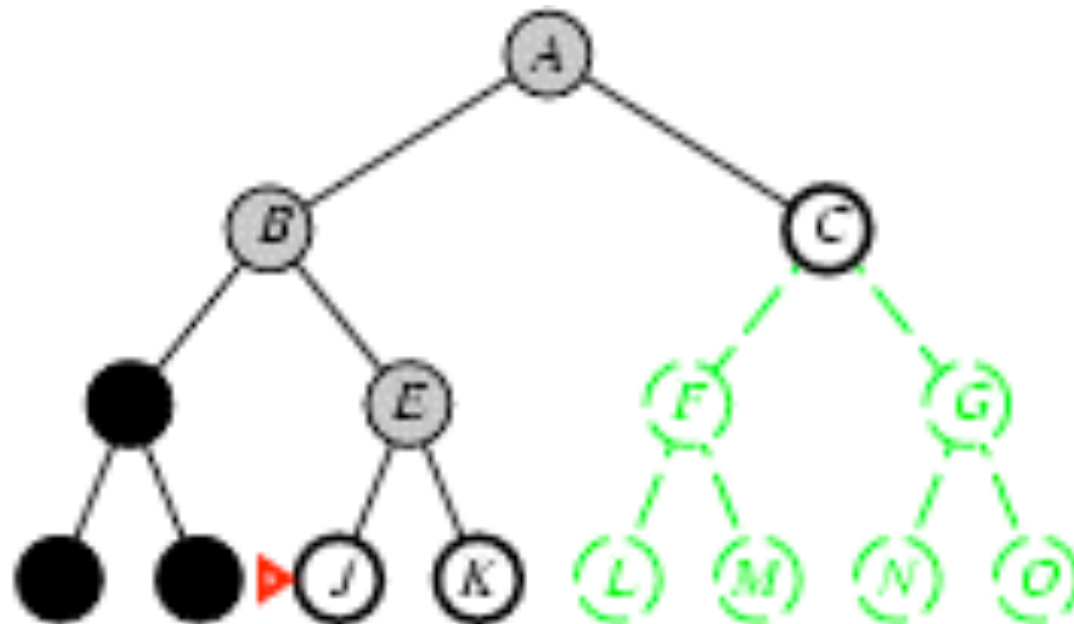
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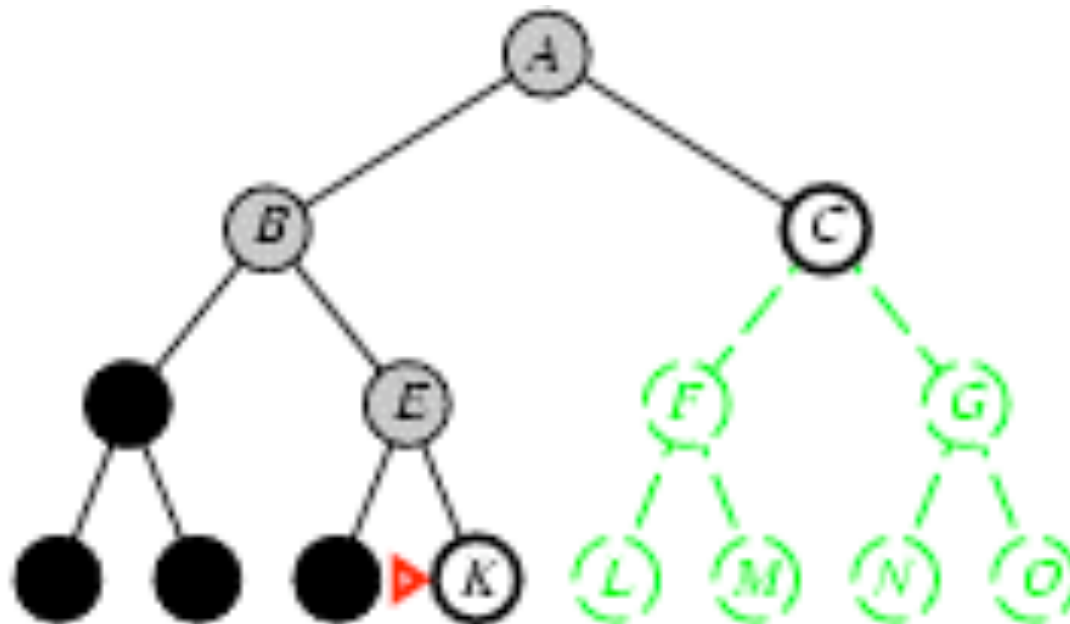
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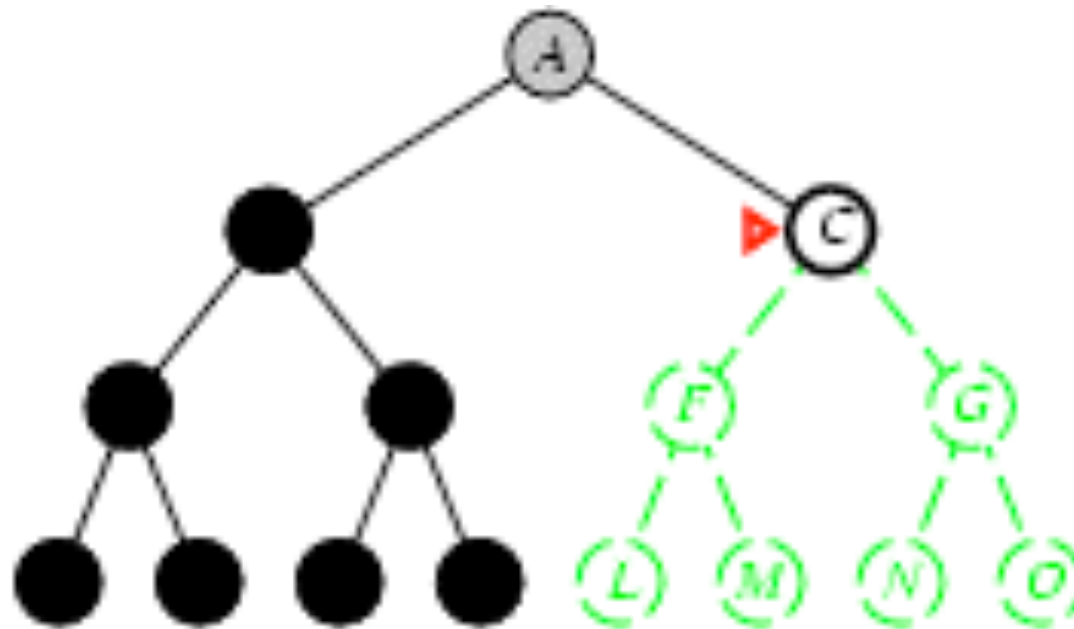
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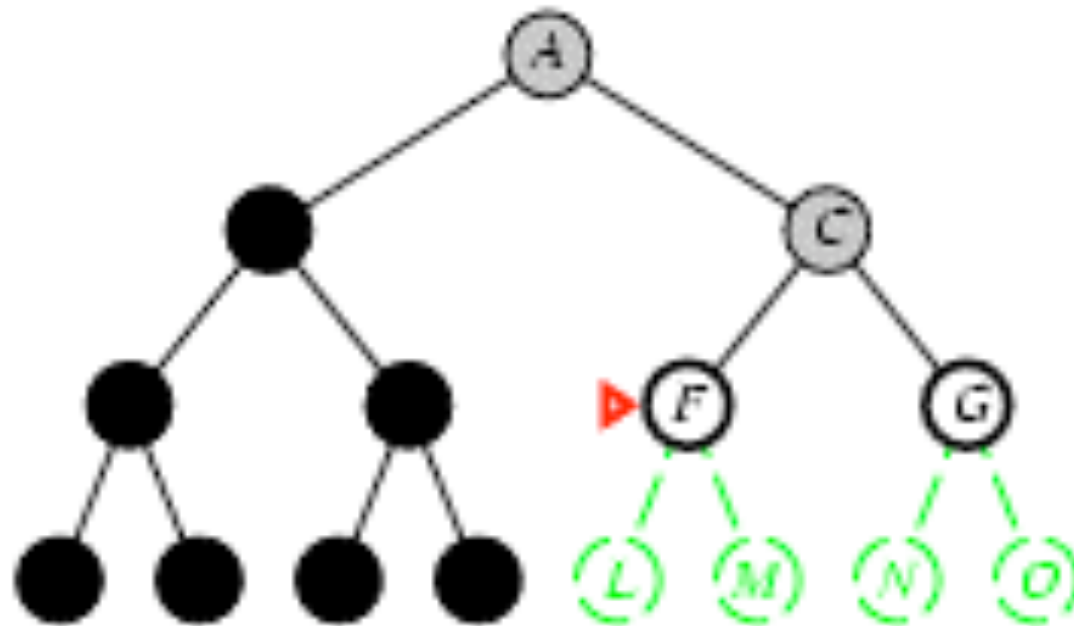
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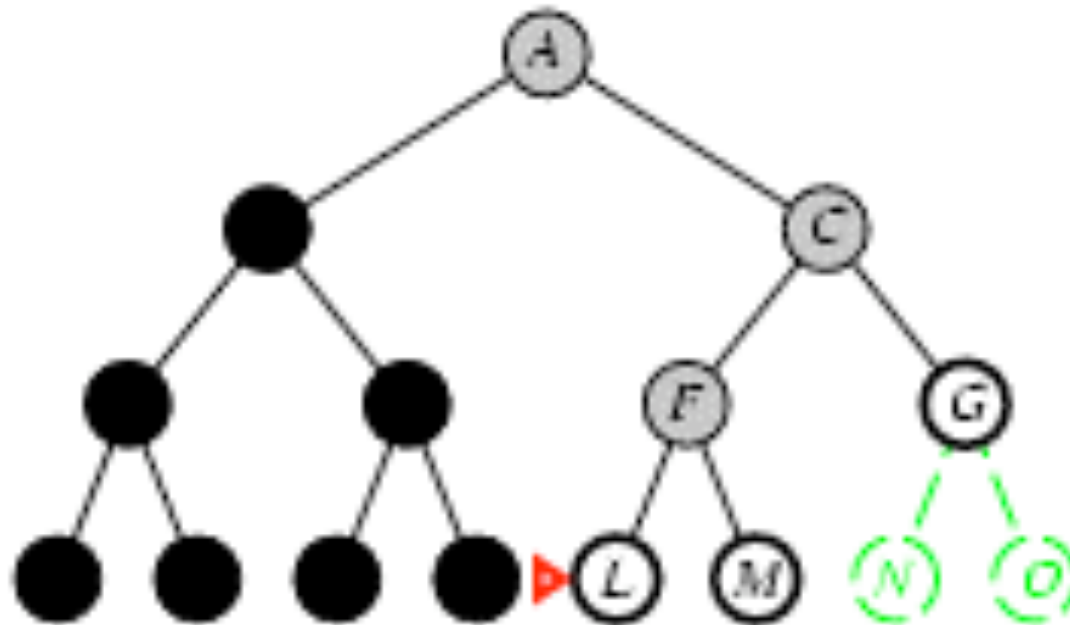
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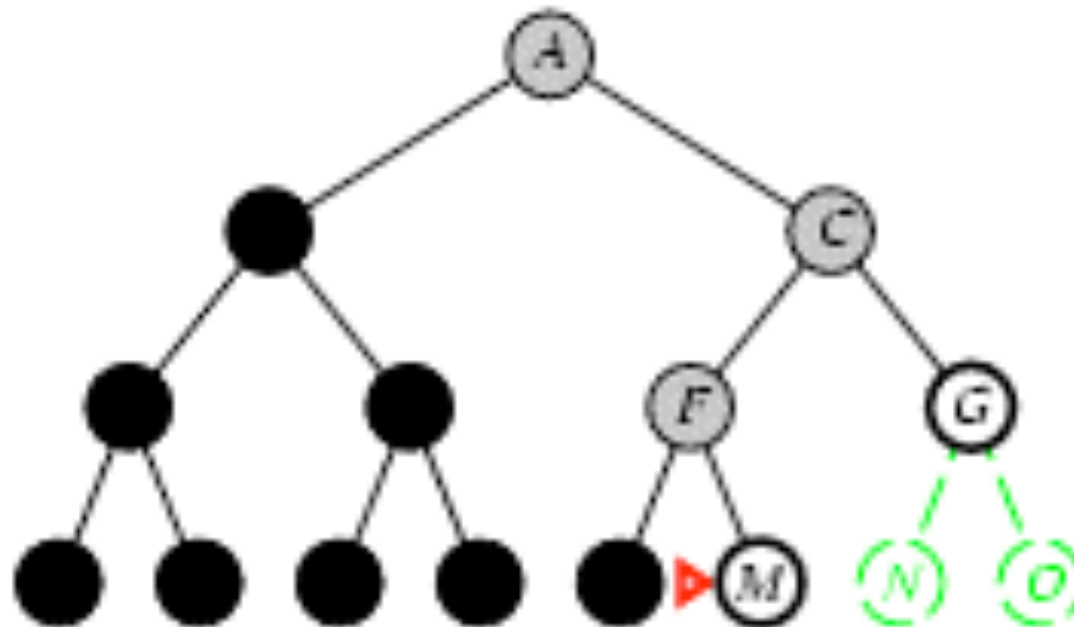
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Properties of depth-first search

- Complete? 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 (length of longest path in search space) is much larger than d
 - but if solutions are dense, may be much faster than breadth-first
- Space? $O(bm)$, i.e., linear space!
- Optimal? No



Depth-limited search

= depth-first search with depth limit l ,
i.e., nodes at depth l have no successors

- Recursive implementation:

```
function DEPTH-LIMITED-SEARCH(problem, limit) returns soln/fail/cutoff
  RECURSIVE-DLS(MAKE-NODE(INITIAL-STATE[problem]), problem, limit)

function RECURSIVE-DLS(node, problem, limit) returns soln/fail/cutoff
  cutoff-occurred? ← false
  if GOAL-TEST[problem](STATE[node]) then return SOLUTION(node)
  else if DEPTH[node] = limit then return cutoff
  else for each successor in EXPAND(node, problem) do
    result ← RECURSIVE-DLS(successor, problem, limit)
    if result = cutoff then cutoff-occurred? ← 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

inputs: *problem*, a problem

for *depth* \leftarrow 0 to ∞ **do**

result \leftarrow DEPTH-LIMITED-SEARCH(*problem*, *depth*)

if *result* \neq cutoff **then return** *result*

Iterative deepening search $l = 0$

Limit = 0



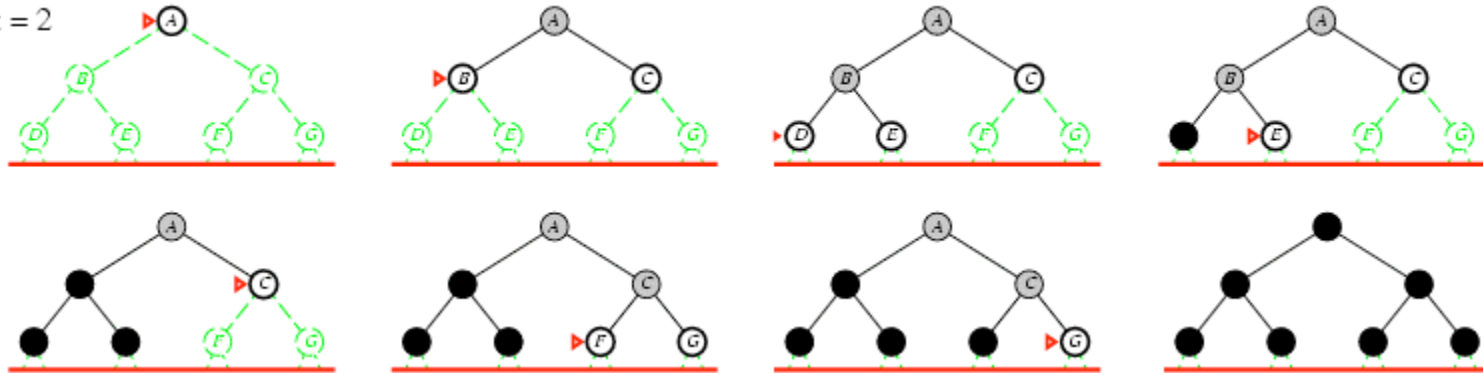
Iterative deepening search $l = 1$

Limit = 1



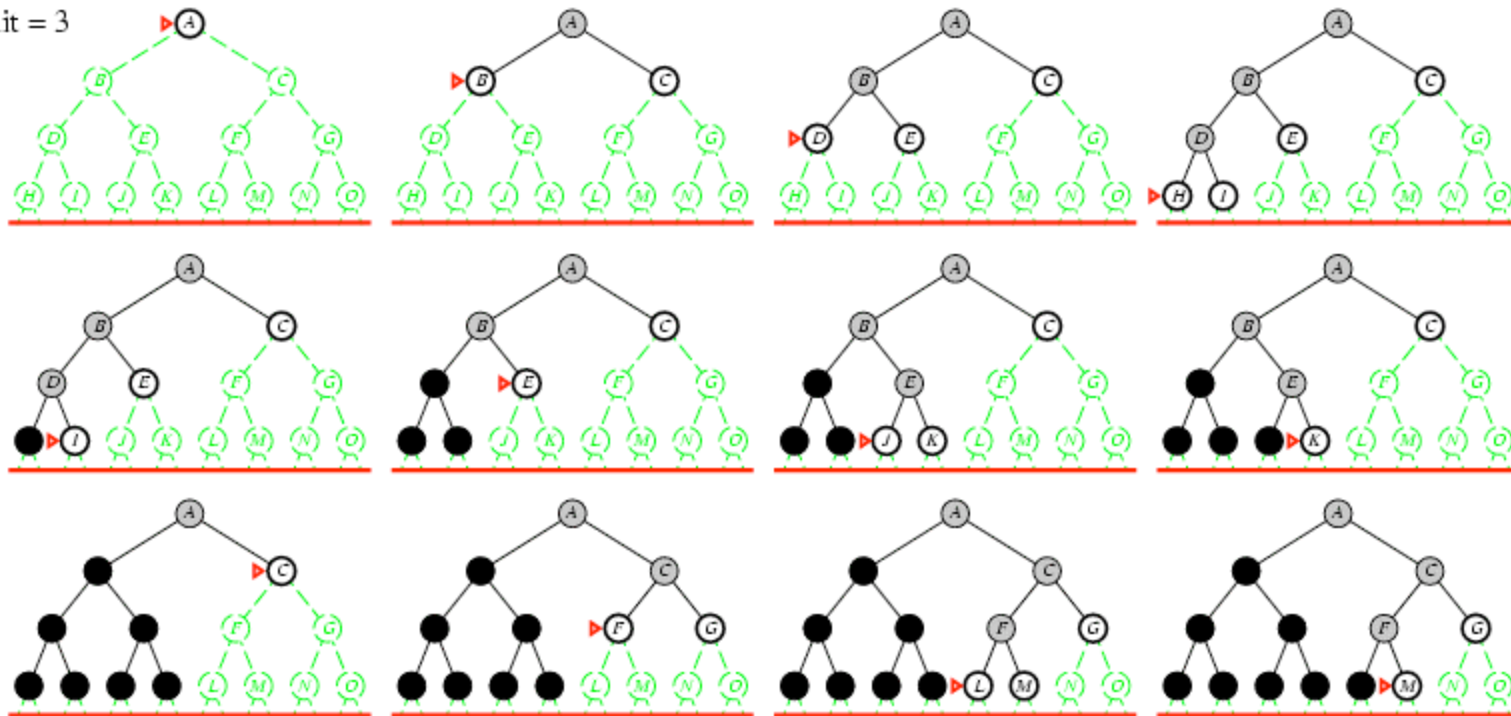
Iterative deepening search $l = 2$

Limit = 2



Iterative deepening search / =3

Limit = 3



Iterative deepening search

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

$$N_{DLS} = b^0 + b^1 + b^2 + \dots + b^{d-2} + b^{d-1} + b^d$$

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

$$N_{IDS} = (d+1)b^0 + d b^1 + (d-1)b^2 + \dots + 3b^{d-2} + 2b^{d-1} + 1b^d$$

- For $b = 10, d = 5,$

- $N_{DLS} = 1 + 10 + 100 + 1,000 + 10,000 + 100,000 = 111,111$

- $N_{IDS} = 6 + 50 + 400 + 3,000 + 20,000 + 100,000 = 123,456$

- Overhead = $(123,456 - 111,111)/111,111 = 11\%$



Properties of iterative deepening search

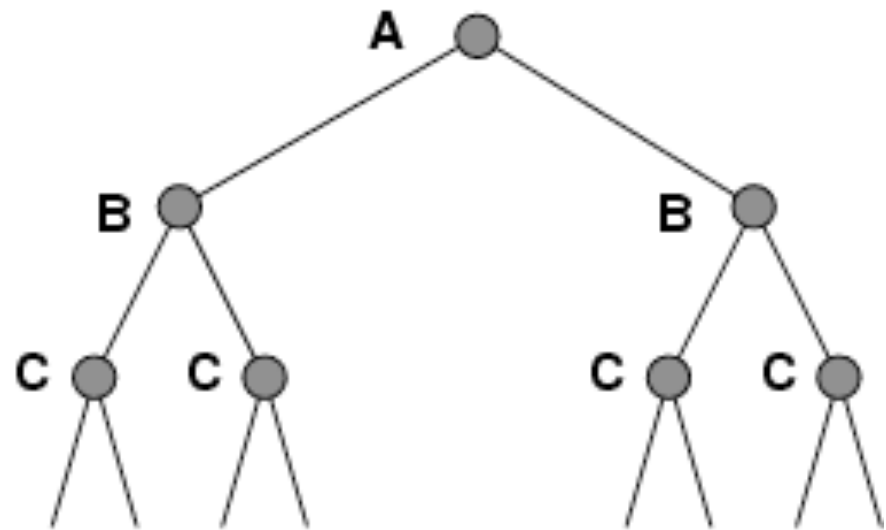
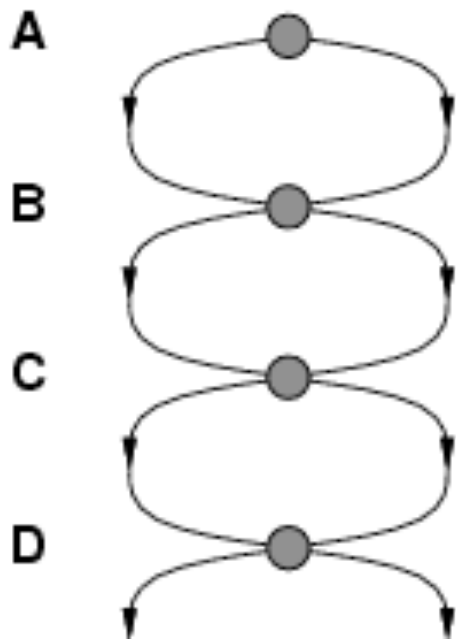
- Complete? Yes
- Time? $(d+1)b^0 + d b^1 + (d-1)b^2 + \dots + b^d = O(b^d)$
- Space? $O(bd)$
- Optimal? Yes, if step cost = 1

Summary of algorithms

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening
Complete?	Yes	Yes	No	No	Yes
Time	$O(b^{d+1})$	$O(b^{\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^l)$	$O(b^d)$
Space	$O(b^{d+1})$	$O(b^{\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(bl)$	$O(bd)$
Optimal?	Yes	Yes	No	No	Yes

Repeated states

- Failure to detect repeated states can turn a linear problem into an exponential one!





Graph search

```
function GRAPH-SEARCH(problem, fringe) returns a solution, or failure
  closed ← an empty set
  fringe ← INSERT(MAKE-NODE(INITIAL-STATE[problem]), fringe)
  loop do
    if fringe is empty then return failure
    node ← REMOVE-FRONT(fringe)
    if GOAL-TEST[problem](STATE[node]) then return SOLUTION(node)
    if STATE[node] is not in closed then
      add STATE[node] to closed
      fringe ← INSERTALL(EXPAND(node, problem), fringe)
```



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