



ĐẠI HỌC BÁCH KHOA HÀ NỘI
VIỆN CÔNG NGHỆ THÔNG TIN VÀ TRUYỀN THÔNG

Artificial Intelligence

Lecture 3 - Search

School of Information and Communication
Technology - HUST

Outline

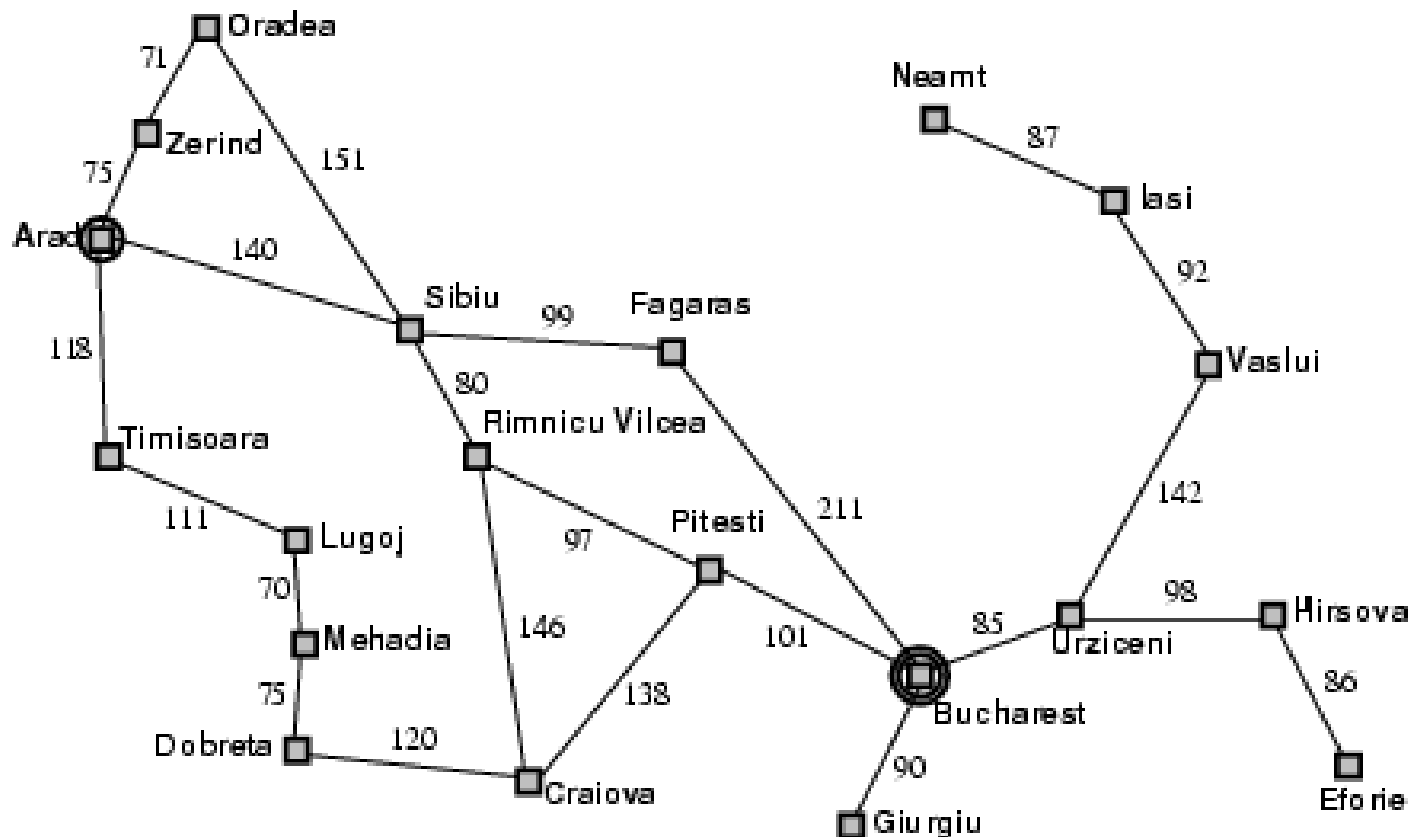
- Problem-solving agents
- Problem types
- Problem formulation
- Example problems
- Basic search algorithms
 - breadth-first search
 - depth-first search
 - depth-limited search
 - iterative deepening depth-first search

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  $\leftarrow$  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 1: Route Planning



- Performance: Get from Arad to Bucharest as quickly as possible
- Environment: The map, with cities, roads, and guaranteed travel times
- Actions: Travel a road between adjacent cities

Example 2: Finding letters

- Replace letters by numbers from 0 to 9 such as no different letter is replaced by the same number and satisfying the following constraint:

SEND	CROSS
+ MORE	+ ROADS
<hr/>	<hr/>
MONEY	DANGER

Example 3: Pouring water

- Given 2 containers A(m litres), B(n litres). Finding a method to measure k litres ($k \leq \max(m,n)$) by 2 containers A, B and a container C
- Actions (how):
 $C \rightarrow A; C \rightarrow B; A \rightarrow B; A \rightarrow C; B \rightarrow A; B \rightarrow C$
- Conditions: no overflow, pouring all water
- Eg: $m = 5, n = 6, k = 2$ (what)
- Mathematical model:
 $(x, y) \rightarrow (x', y')$
A B A B

Example 4: The 8-puzzle

- Trong bảng ô vuông n hàng, n cột, mỗi ô chứa 1 số nằm trong phạm vi từ $1 \rightarrow n^2 - 1$ sao cho không có 2 ô có cùng giá trị. Còn đúng 1 ô bị trống. Xuất phát từ 1 cách sắp xếp nào đó của các số trong bảng, hãy dịch chuyển các ô trống sang phải, sang trái, lên trên, xuống dưới để đưa về bảng:

7	2	4
5		6
8	3	1

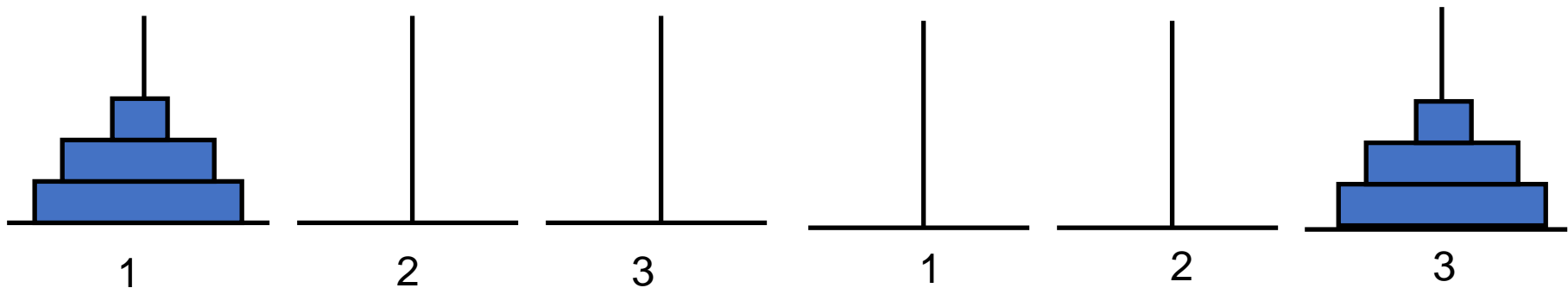
Start State

	1	2
3	4	5
6	7	8

Goal State

Example 5: Hà Nội tower

- Cho 3 cọc 1,2,3. Ở cọc 1 ban đầu có n đĩa, sắp theo thứ tự to dần từ trên xuống dưới. Hãy tìm cách chuyển n đĩa đó sang cọc 3 sao cho:
 - Mỗi lần chỉ chuyển 1 đĩa
 - Ở mỗi cọc không cho phép đĩa to nằm trên đĩa con



Bài toán tháp Hà Nội với $n = 3$

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

Search Problem Definition

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

1. **initial state**: e.g., 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., $x = \text{Bucharest}$
 - **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

Example: The 8-puzzle

7	2	4
5		6
8	3	1

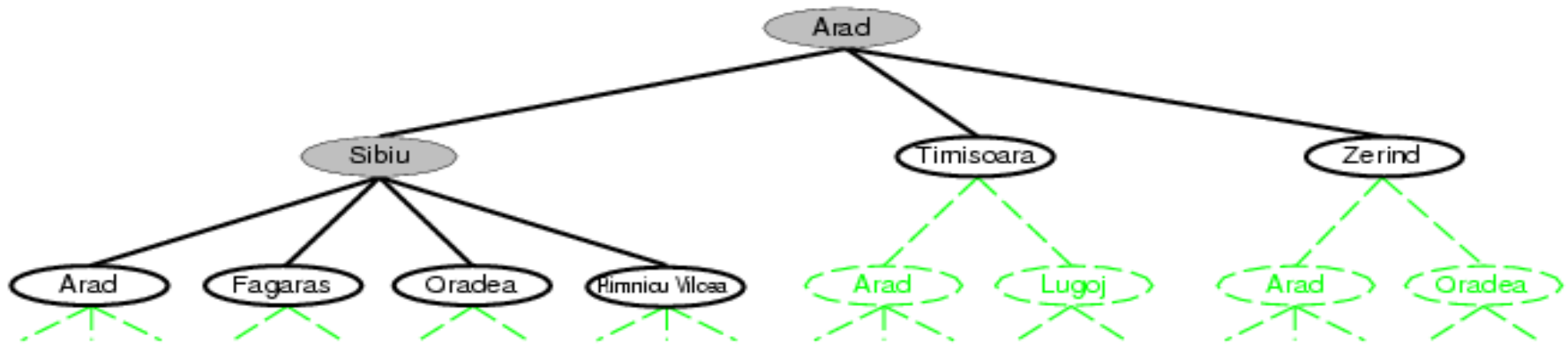
Start State

	1	2
3	4	5
6	7	8

Goal State

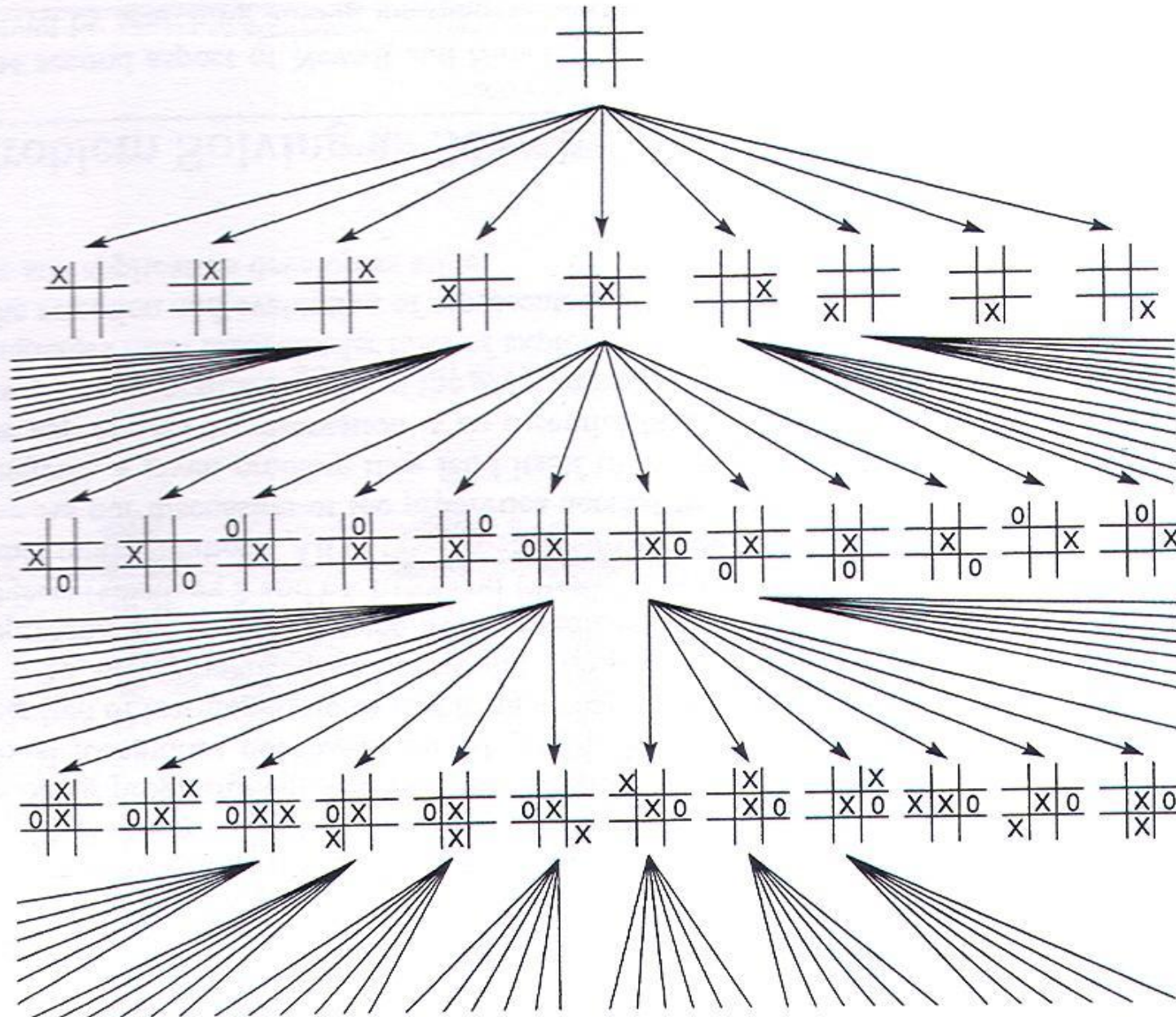
- states? locations of tiles
- actions? move blank left, right, up, down
- goal test? = goal state (given)
- path cost? 1 per move

Search tree



- Search trees:
 - Represent the branching paths through a state graph.
 - Usually **much** larger than the state graph.
 - Can a finite state graph give an infinite search tree?

Search space of the game Tic-Tac-Toe



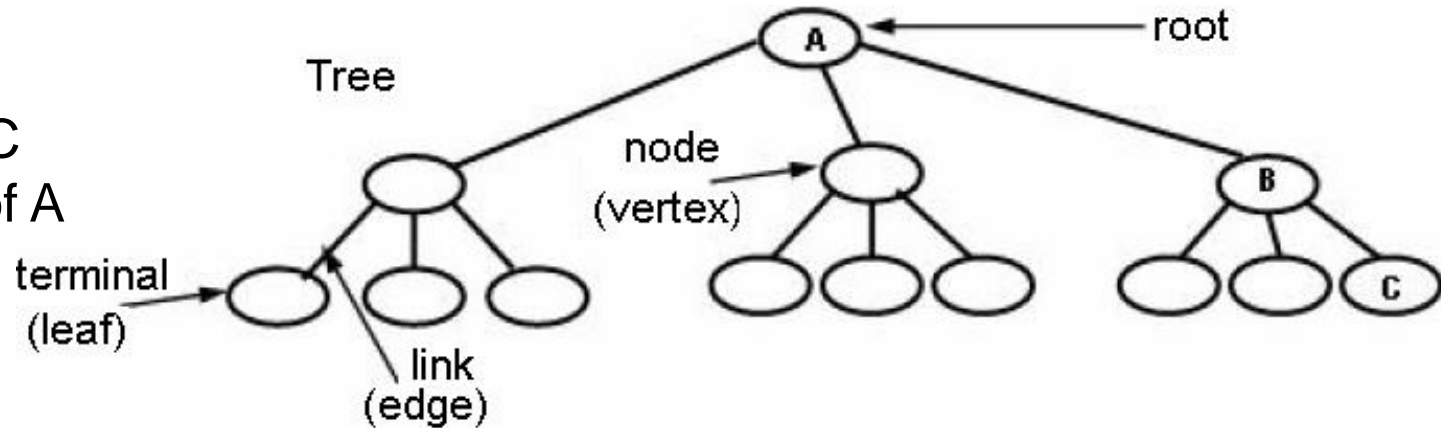
Tree and graph

B is parent of C

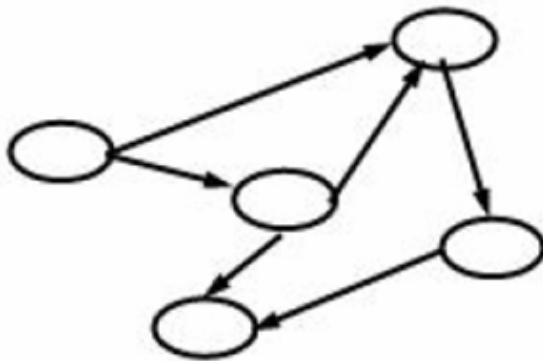
C is child of B

A is ancestor of C

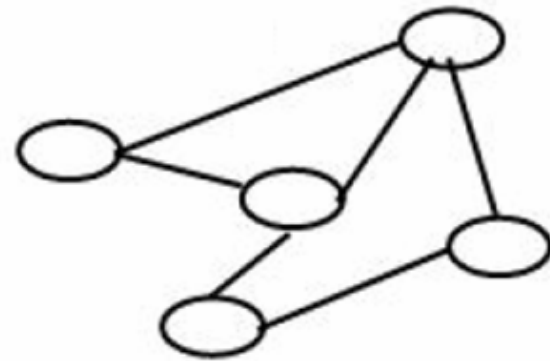
C is descendant of A



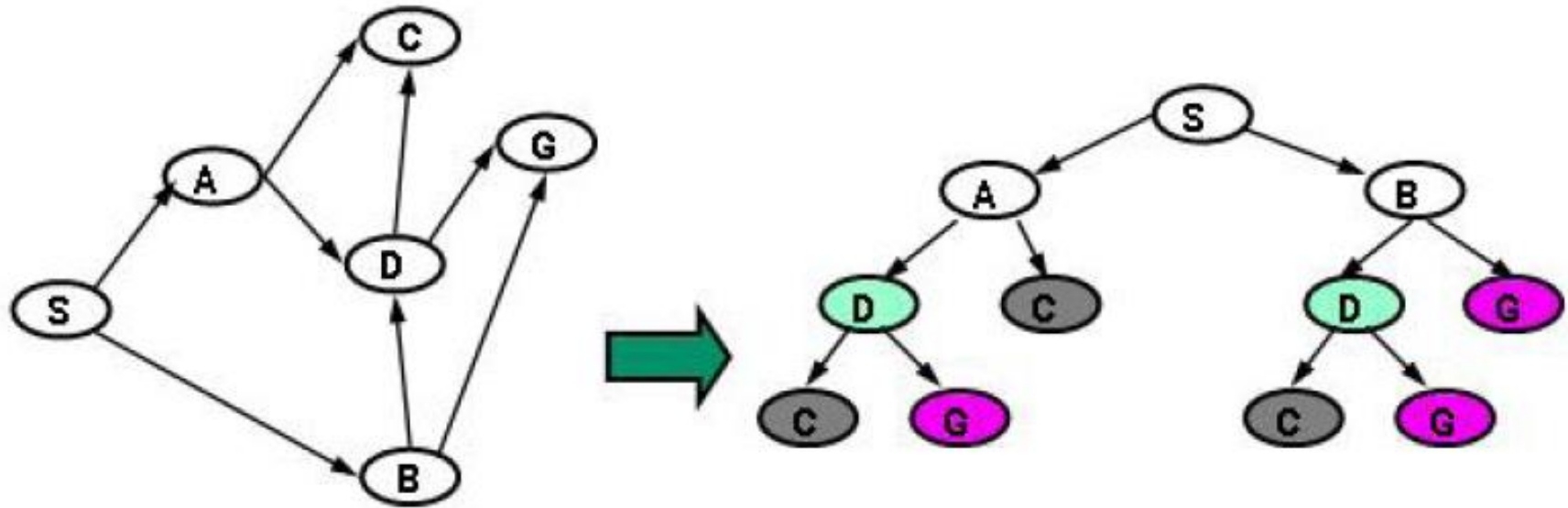
Directed graph
(one-way street)



Undirected graph
(two-way streets)



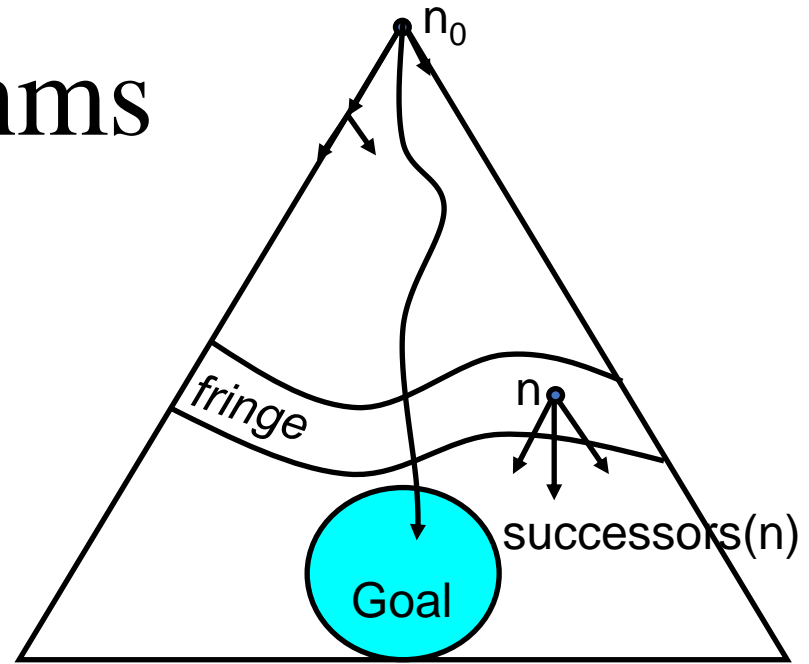
Convert from search graph to search tree



- We can turn graph search problems into tree search problems by:
 - replacing undirected links by 2 directed links
 - avoiding loops in path (or keeping track of visited nodes globally)

Tree search algorithms

- Basic idea:
 - simulated exploration of state space by generating successors of already-explored 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
```

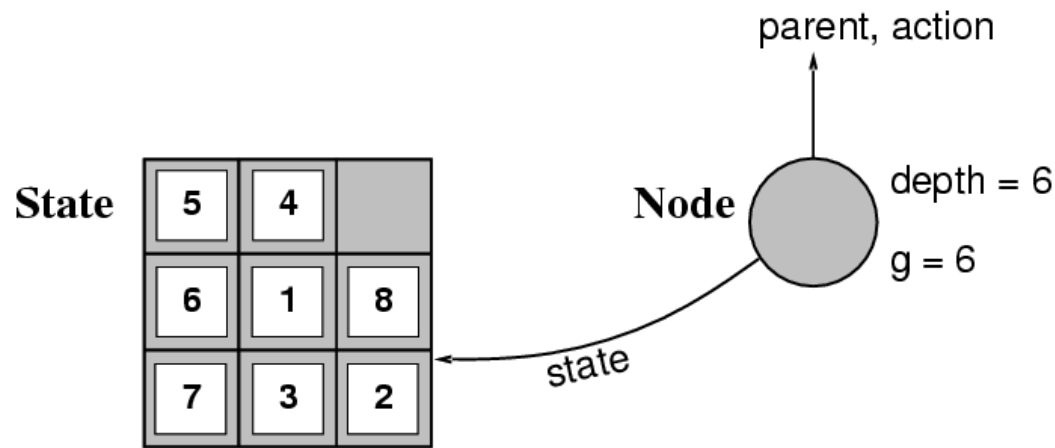

Implementation: general tree search

```
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  
  for each action, result in SUCCESSOR-FN[problem](STATE[node]) do  
    s  $\leftarrow$  a new NODE  
    PARENT-NODE[s]  $\leftarrow$  node; ACTION[s]  $\leftarrow$  action; STATE[s]  $\leftarrow$  result  
    PATH-COST[s]  $\leftarrow$  PATH-COST[node] + STEP-COST(node, action, s)  
    DEPTH[s]  $\leftarrow$  DEPTH[node] + 1  
    add s to successors  
  return successors
```

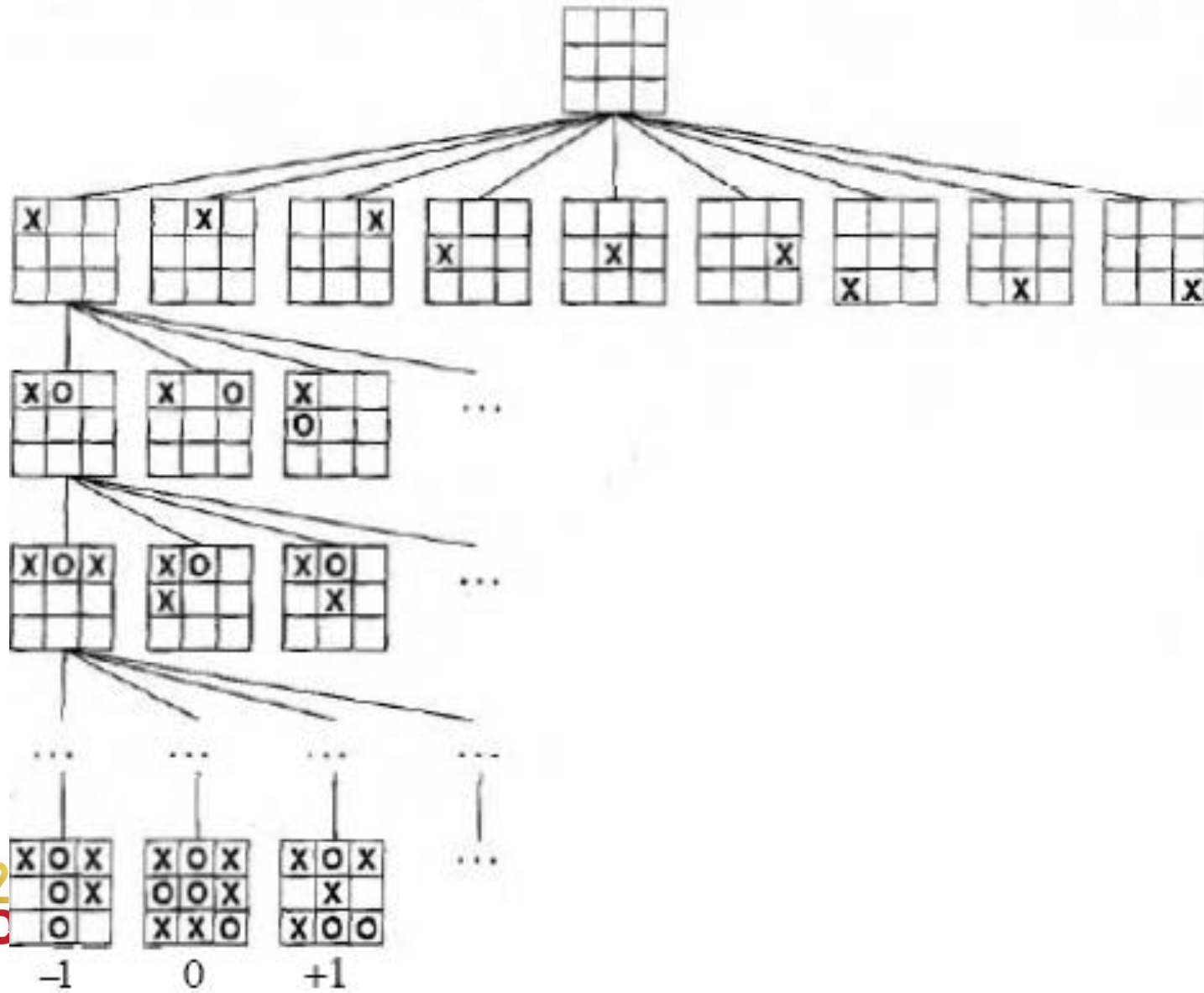
Implementation: states vs. nodes

- 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.
-

Implementation: states vs. nodes



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 ∞)

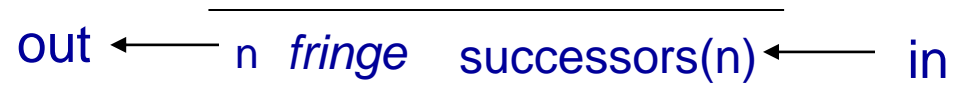
Uninformed search strategies

- **Uninformed** search strategies use only the information available in the problem definition

- Breadth-first search

- Expand shallowest unexpanded node

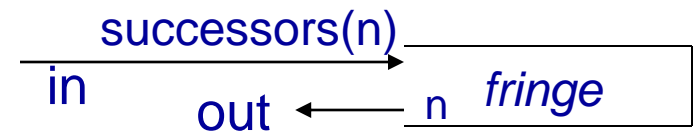
- *fringe* = queue (FIFO)



- Depth-first search

- Expand deepest unexpanded node

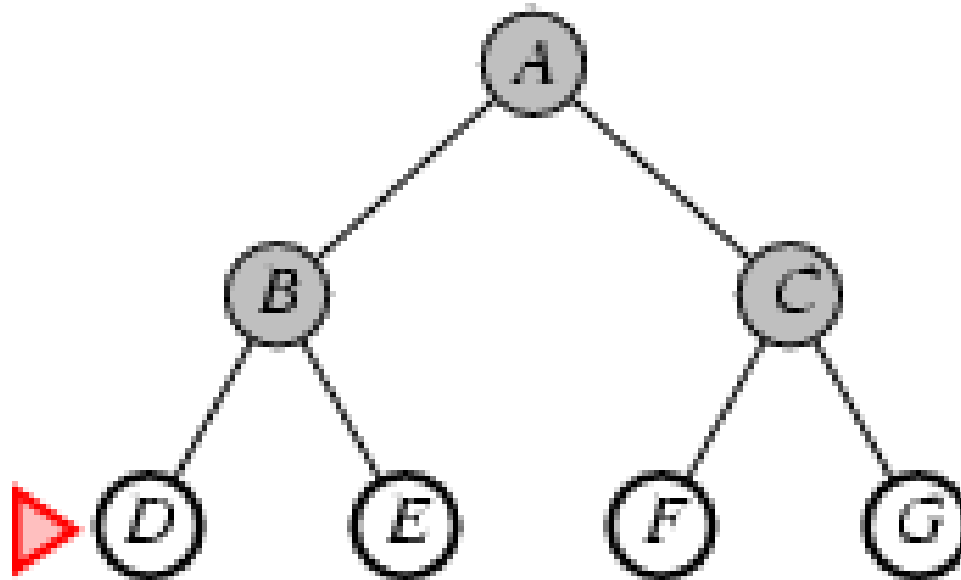
- *fringe* = stack (LIFO)



- Depth-limited search: depth-first search with depth limit
- Iterative deepening search

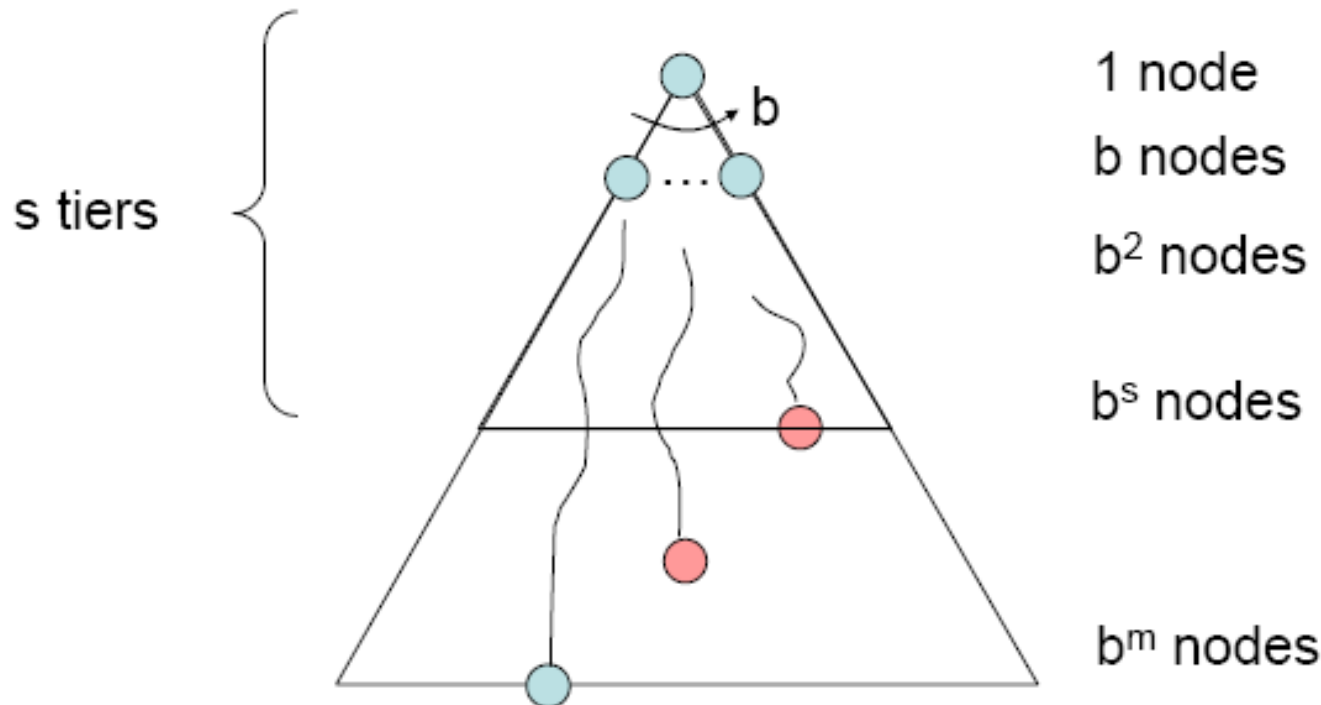
Breadth-first search

- Expand shallowest unexpanded node



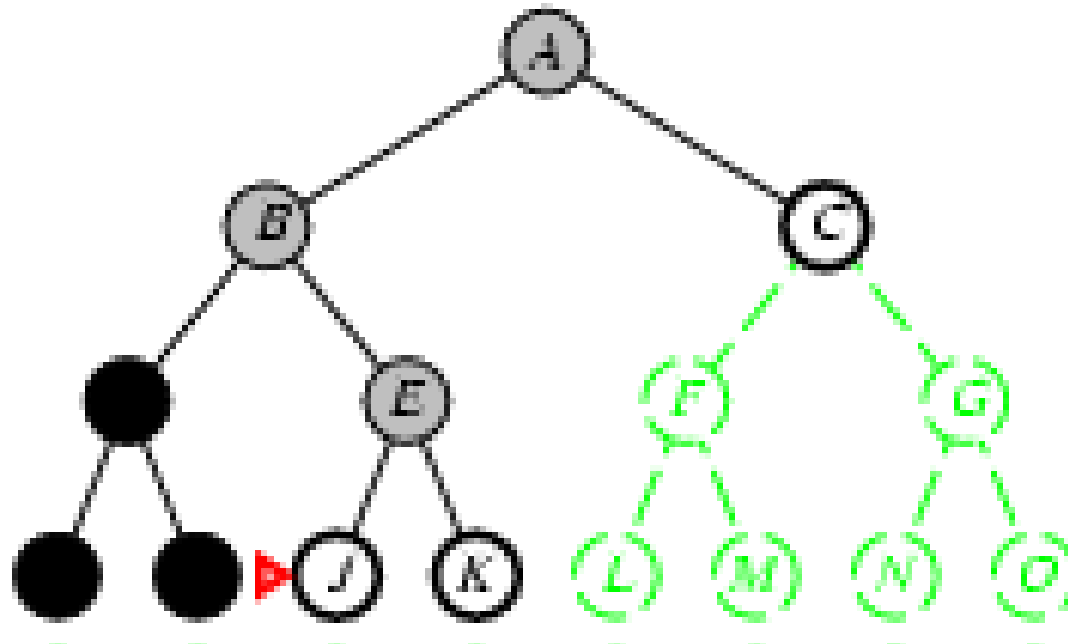
Breadth-first search (con't)

- 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)



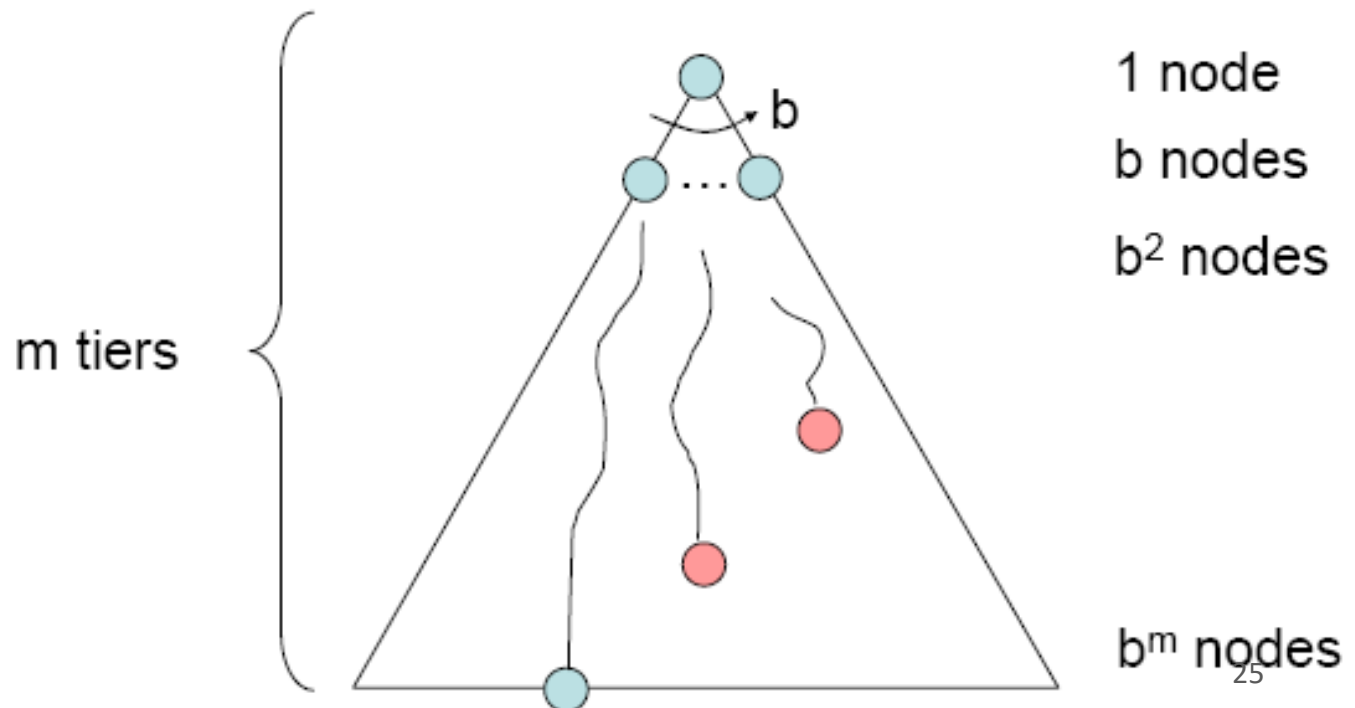
Depth-first search

- Expand deepest unexpanded node



Depth-first search (con't)

- Complete? No: fails in infinite-depth spaces, spaces with loops
 - Modify to avoid repeated states along path \rightarrow 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!
-
- Optimal? No
-

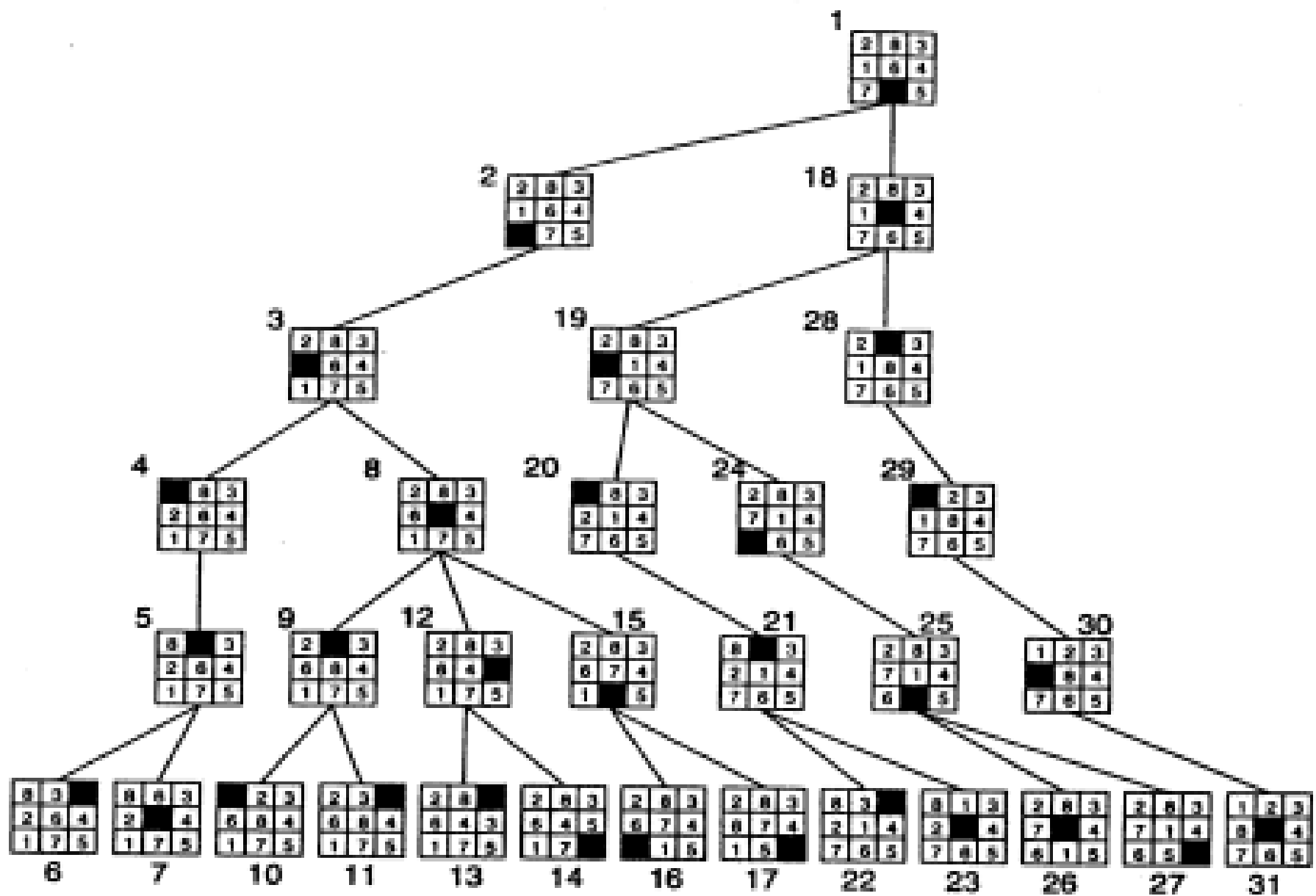


Depth-limited search

- Depth-first search can get stuck on infinite path when a different choice would lead to a solution
- ⇒ Depth-limited search = depth-first search with depth limit l , i.e., nodes at depth l have no successors

```
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
```



8-puzzle game with depth limit $l = 5$

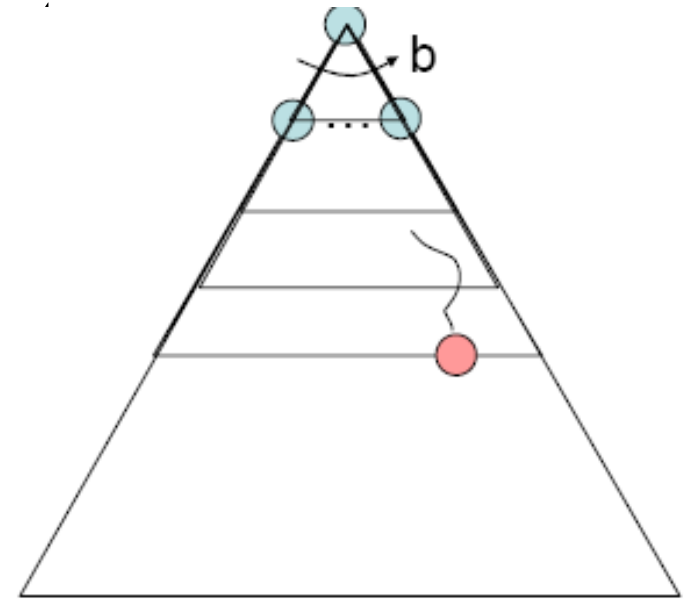
Goal

Iterative deepening search

- Problem with depth-limited search: if the shallow goal is beyond the depth limit, no solution is found

⇒ Iterative deepening search:

1. Do a DFS which only searches for paths of length 1 (DFS gives up on any path of length 2)
2. If “1” failed, do a DFS which only searches paths of 2 or less.
3. If “2” failed, do a DFS which only searches paths of 3 or less.
4.and so on.



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*)

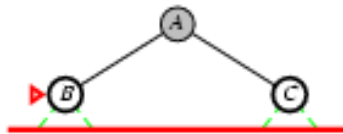
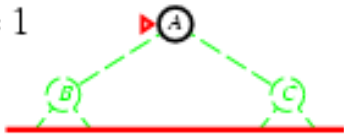
if *result* \neq cutoff **then return** *result*

Iterative deepening search (con't)

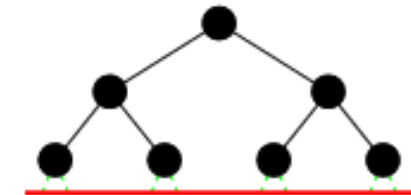
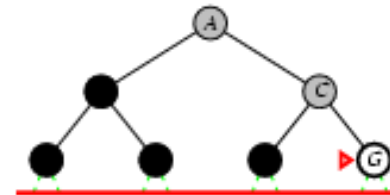
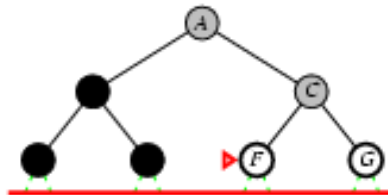
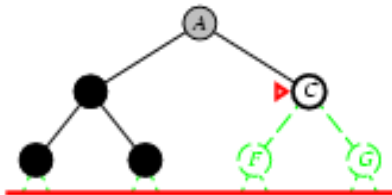
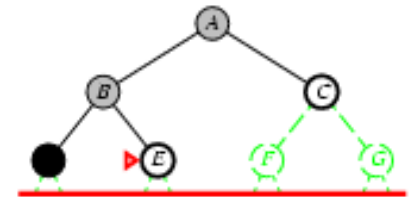
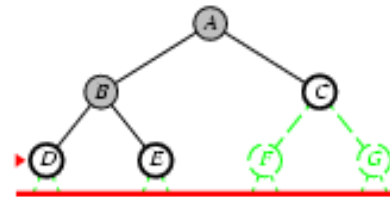
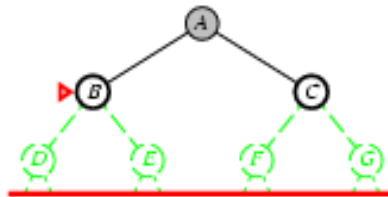
Limit = 0



Limit = 1

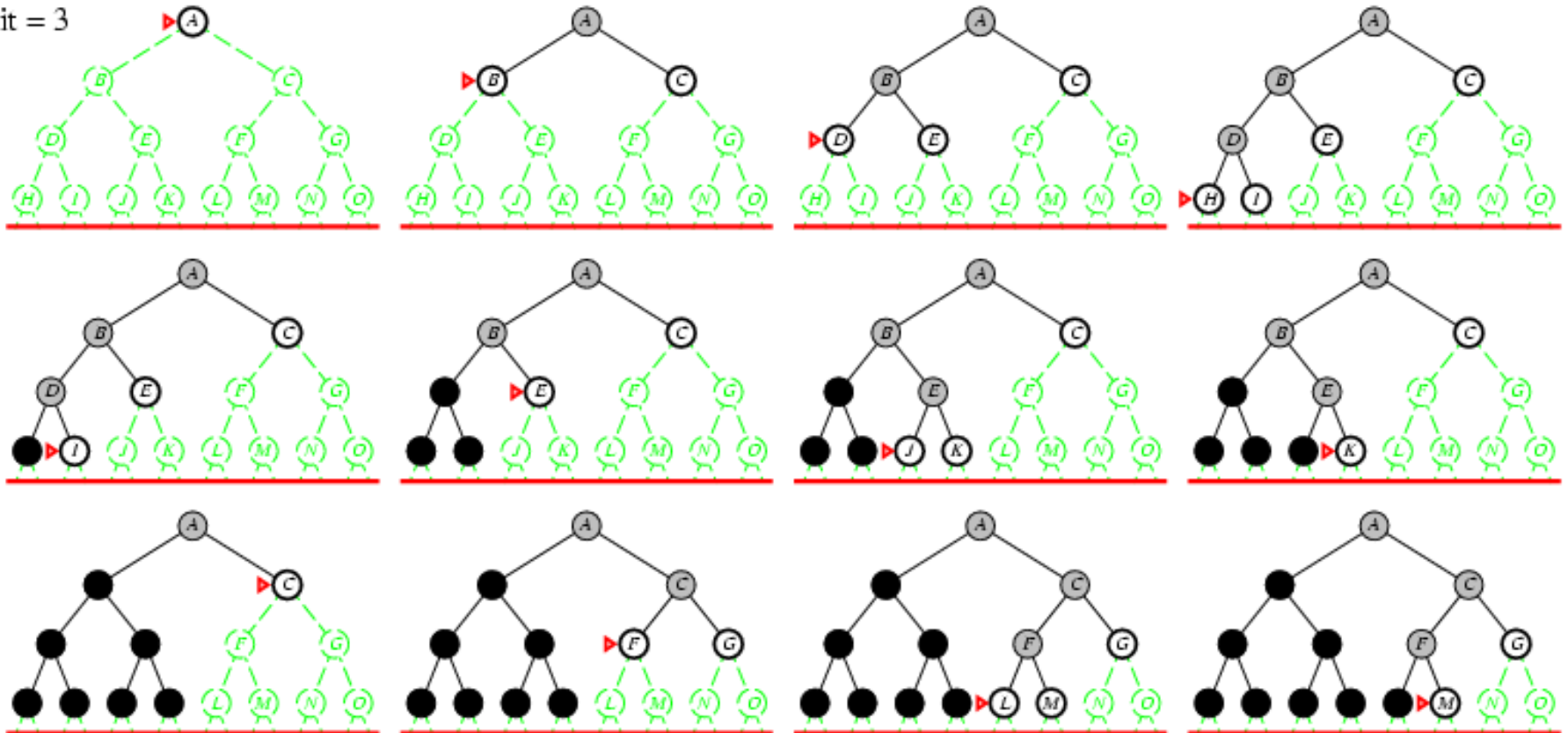


Limit = 2



Iterative deepening search (con't)

Limit = 3



Iterative deepening search (con't)

- 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