

# Artificial Intelligence

*For HEDSPI Project*

## Lecturer 3 - Search

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## 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

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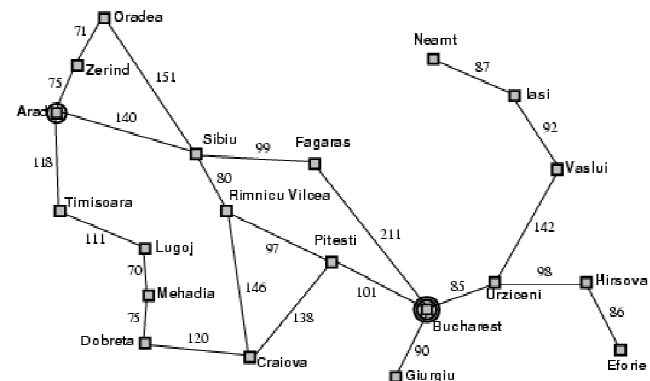
## Problem-solving agents

```
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
```

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## 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

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## 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:

$$\begin{array}{r} \text{SEND} \\ + \text{MORE} \\ \hline \text{MONEY} \end{array} \quad \begin{array}{r} \text{CROSS} \\ + \text{ROADS} \\ \hline \text{DANGER} \end{array}$$

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## 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:

$$(x, y) \rightarrow (x', y')$$

$$A \ B \quad A \ B$$

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### 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 đồ 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

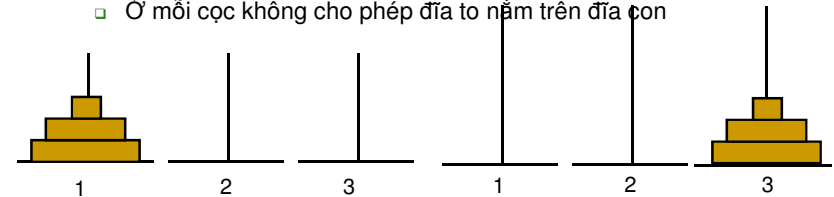
	1	2
3	4	5
6	7	8

Goal State

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### 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  $n = 3$

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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

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## 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  $\geq 0$
- A **solution** is a sequence of actions leading from the initial state to a goal state

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## Example: The 8-puzzle

7	2	4
5		6
8	3	1

Start State

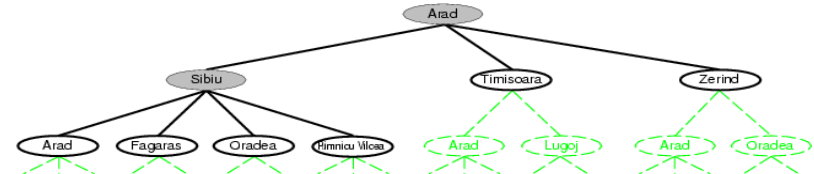
	1	2
3	4	5
6	7	8

Goal State

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

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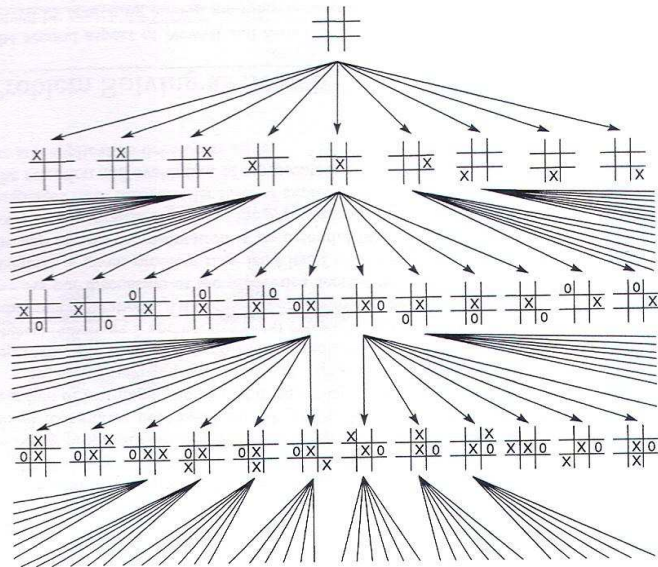
## Search tree



- 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?

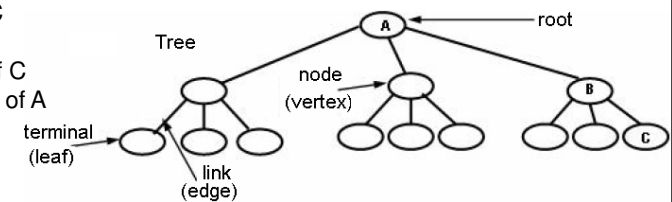
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## Search space of the game Tic-Tac-Toe

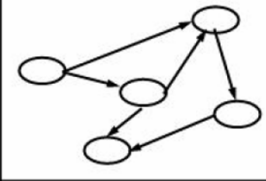


## Tree and graph

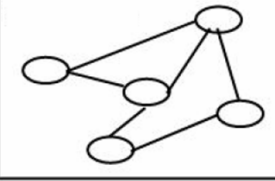
B is parent of C  
C is child of B  
A is ancestor of C  
C is descendant of A



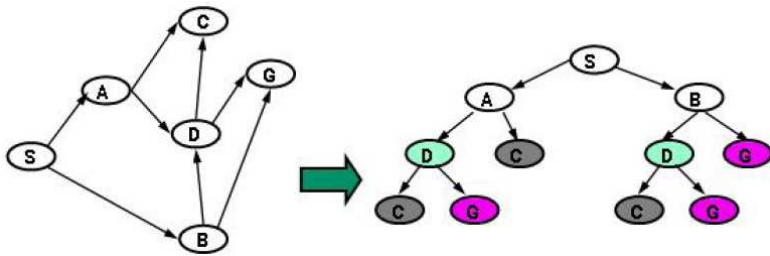
Directed graph  
(one-way street)



Undirected graph  
(two-way streets)



## Convert from search graph to search tree

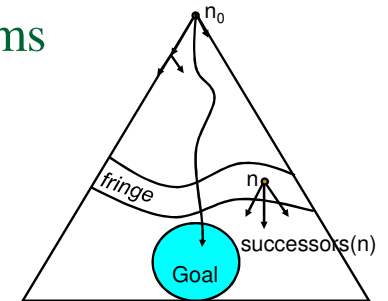


- We can turn graph search problems into tree search problems by:
  - replacing undirected links by 2 directed links
  - avoiding loops in path (or keeping track of visited nodes globally)

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## Tree search algorithms

- Basic idea:
  - offline, 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
    
```

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## 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 ← INSERT-ALL( EXPAND( node, problem ), fringe )

```

```

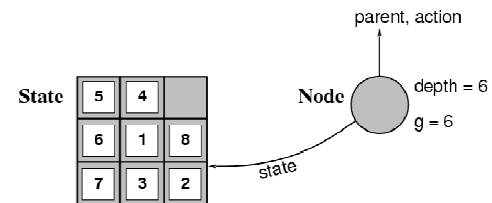
function EXPAND( node, problem ) returns a set of nodes
  successors ← the empty set
  for each action, result in SUCCESSOR-FN[problem]( STATE[node] ) do
    s ← a new NODE
    PARENT-NODE[s] ← node; ACTION[s] ← action; STATE[s] ← result
    PATH-COST[s] ← PATH-COST[node] + STEP-COST( node, action, s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

```

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## Implementation: states vs. nodes

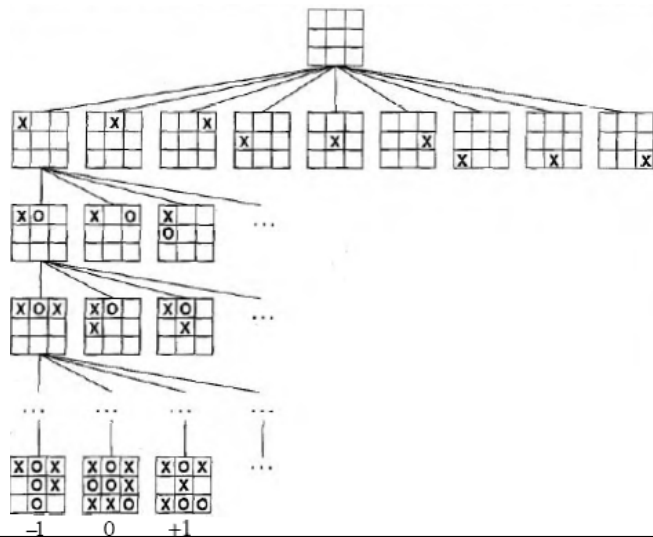
- A **state** is a (representation of) a physical configuration
- A **node** is a data structure constituting part of a search tree includes **state**, **parent node**, **action**, **path cost**  $g(x)$ , **depth**



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

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## Implementation: states vs. nodes



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## 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  $\infty$ )

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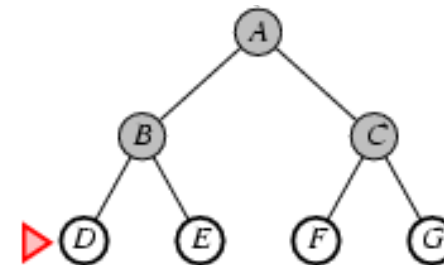
## Uninformed search strategies

- **Uninformed** search strategies use only the information available in the problem definition
  - Breadth-first search
    - Expand shallowest unexpanded node
    - *fringe* = queue (FIFO)
- Diagram illustrating Breadth-First Search (BFS) using a queue:
- ```
graph LR
    in[in] --> n[n]
    n --> fringe[fringe]
    fringe --> out[out]
    fringe --> successors[successors(n)]
    successors --> in
```
- Depth-first search
    - Expand deepest unexpanded node
    - *fringe* = stack (LIFO)
- Diagram illustrating Depth-First Search (DFS) using a stack:
- ```
graph LR
    in[in] --> n[n]
    n --> fringe[fringe]
    fringe --> out[out]
    fringe --> successors[successors(n)]
    successors --> in
```
- Depth-limited search: depth-first search with depth limit
  - Iterative deepening search

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## Breadth-first search

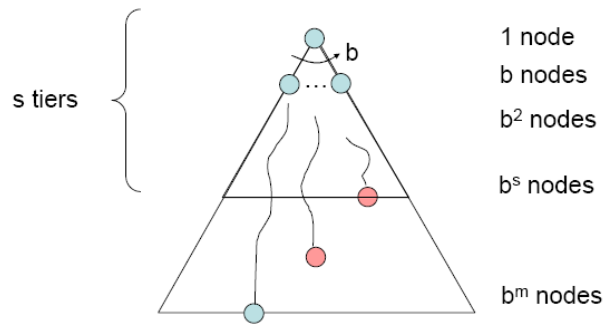
- Expand shallowest unexpanded node



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## Breadth-first search (con't)

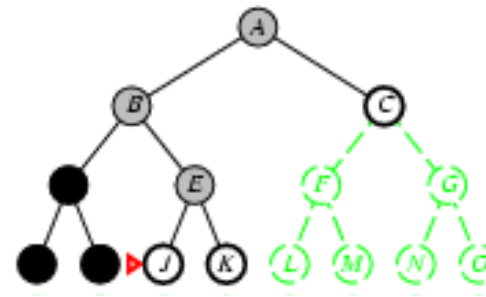
- Complete? Yes (if  $b$  is finite)
- Time?  $1+b+b^2+b^3+\dots+b^d + b(b^d-1) = O(b^{d+1})$
- Space?  $O(b^{d+1})$  (keeps every node in memory)
- Optimal? Yes (if cost = 1 per step)



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## Depth-first search

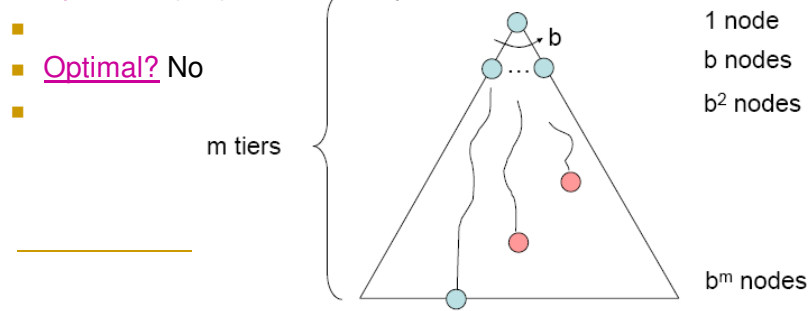
- Expand deepest unexpanded node



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## Depth-first search (con't)

- **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  $m$  is much larger than  $d$ 
  - but if solutions are dense, may be much faster than breadth-first
- **Space?**  $O(bm)$ , i.e., linear space!



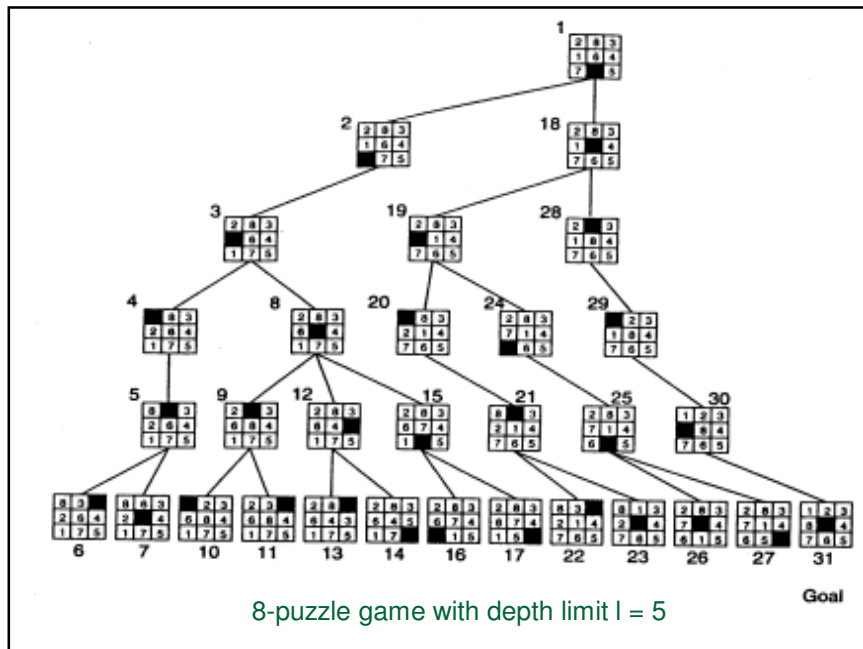
## Depth-limited search

- 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

```

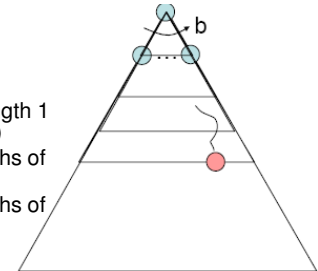
function DEPTH-LIMITED-SEARCH( problem, limit ) returns soln/fail/cutoff
    RECURSIVE-DLS(MAKE-NODE(INITIAL-STATE[problem]), problem, limit)

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
    
```



## Iterative deepening search

- Problem with depth-limited search: if the shallowest 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 or less. (DFS gives up on any path of length 2)
  2. If "1" failed, do a DFS which only searches paths of length 2 or less.
  3. If "2" failed, do a DFS which only searches paths of length 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  $\infty$  **do**

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

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

## Iterative deepening search (con't)

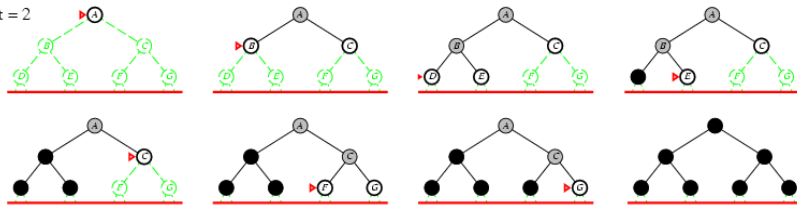
Limit = 0



Limit = 1



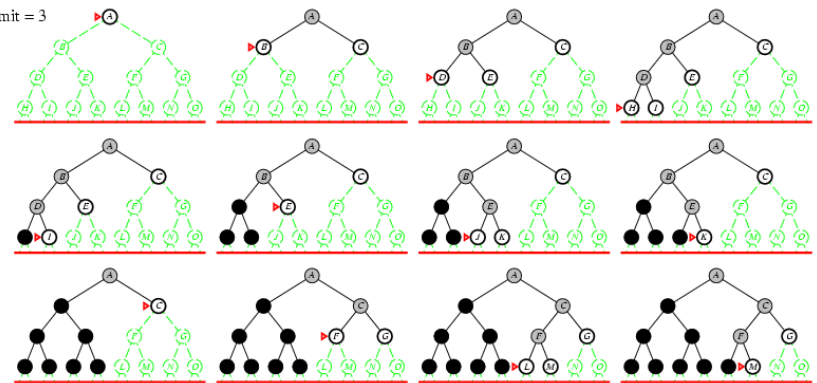
Limit = 2



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## Iterative deepening search (con't)

Limit = 3



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

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## 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

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## 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