Search Strategies

Informatics 2D: Reasoning and Agents Lecture 3

Adapted from slides provided by Dr Petros Papapanagiotou



Search strategies

A search strategy is defined by picking the order of node expansion.
Nodes are taken from the *frontier*.

Evaluating search strategies



completeness: does it always find a solution if one exists?



time complexity: number of nodes generated / expanded



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 ∞)

Recall: Tree Search

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





Repeated states

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

Graph search

Augment TREE-SEARCH with a new data-structure:

- the **explored set** (closed list), which remembers every expanded node
- newly expanded nodes already in explored set are discarded



Expand **shallowest** unexpanded node

Implementation:





Expand **shallowest** unexpanded node

Implementation:





Expand **shallowest** unexpanded node

Implementation:





Expand **shallowest** unexpanded node

Implementation:



function BREADTH-FIRST-SEARCH(*problem*) returns a solution, or failure $node \leftarrow a node with STATE = problem.INITIAL-STATE, PATH-COST = 0$ if *problem*.GOAL-TEST(*node*.STATE) then return SOLUTION(*node*) frontier \leftarrow a FIFO queue with *node* as the only element explored \leftarrow an empty set loop do if EMPTY?(*frontier*) then return failure $node \leftarrow POP(frontier)$ /* chooses the shallowest node in frontier */ add node.STATE to explored for each action in problem.ACTIONS(*node*.STATE) do $child \leftarrow CHILD-NODE(problem, node, action)$ if child.STATE is not in explored or frontier then if problem.GOAL-TEST(child.STATE) then return SOLUTION(child) frontier \leftarrow INSERT(child, frontier) Breadthfirst search algorithm















Space is the bigger problem (more than time)

Depth	Nodes		Time	Ν	Aemory
2	110	.11	milliseconds	107	kilobytes
4	11,110	11	milliseconds	10.6	megabytes
6	10^{6}	1.1	seconds	1	gigabyte
8	10^{8}	2	minutes	103	gigabytes
10	10^{10}	3	hours	10	terabytes
12	10^{12}	13	days	1	petabyte
14	10^{14}	3.5	years	99	petabytes
16	10^{16}	350	years	10	exabytes

assume branching factor b = 10; 1 million nodes/second; 1000 bytes/node.



Expand **deepest** unexpanded node

Implementation:





Expand **deepest** unexpanded node

Implementation:





Expand **deepest** unexpanded node

Implementation:





Expand **deepest** unexpanded node

Implementation:





Expand **deepest** unexpanded node

Implementation:



Expand **deepest** unexpanded node

Implementation:



In Out J K E C

Expand **deepest** unexpanded node

Implementation:



In

Out

J

K

Expand **deepest** unexpanded node

Implementation:



In Out K C

Expand **deepest** unexpanded node

Implementation:





Expand **deepest** unexpanded node

Implementation:

















Mid-Lecture Exercise



Mid-Lecture Exercise

BREADTH-FIRST

DEPTH-FIRST

- When completeness is important.
- When optimal solutions are important.

• When solutions are dense and low-cost is important, especially space costs.

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

 $cutoff_occurred? \leftarrow false$ **for each** action **in** problem.ACTIONS(node.STATE) **do** $child \leftarrow CHILD-NODE(problem, node, action)$ $result \leftarrow RECURSIVE-DLS(child, problem, limit - 1)$ **if** result = cutoff **then** $cutoff_occurred? \leftarrow$ true **else if** $result \neq failure$ **then return** result**if** $cutoff_occurred?$ **then return** cutoff **else return** failure Depth-limited search

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

Properties of depth-limited tree search



Iterative deepening search

... or how to improve depth-first search

Iterative deepening search

 $\begin{array}{l} \textbf{function ITERATIVE-DEEPENING-SEARCH}(\textit{problem}) \textbf{ returns} a \ solution, \ or \ failure \\ \textbf{for } \textit{depth} = 0 \ \textbf{to} \ \infty \ \textbf{do} \\ \textit{result} \leftarrow \text{DEPTH-LIMITED-SEARCH}(\textit{problem}, \textit{depth}) \\ \textbf{if } \textit{result} \neq \text{cutoff } \textbf{then } \textbf{return } \textit{result} \end{array}$

Iterative deepening search I = 0



Iterative deepening search / =1



Iterative deepening search I = 2



Iterative deepening search I = 3



Iterative deepening search

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

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

Some cost associated with generating upper levels multiple times

Example: For b = 10, d = 5, $\circ N_{BFS} = 10 + 100 + 1,000 + 10,000 + 100,000 = 111,110$ $\circ N_{IDS} = 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450$

Overhead = (123, 450 - 111, 110)/111, 110 = 11%

	Complete?
Č	Time complexity?
	Space complexity?
	Optimal?









Criterion	Breadth- First	Uniform- Cost	Depth- First	Depth- Limited	Iterative Deepening	Bidirectional (if applicable)
Complete? Time Space Optimal?	$\begin{array}{c} \operatorname{Yes}^a \\ O(b^d) \\ O(b^d) \\ \operatorname{Yes}^c \end{array}$	$egin{array}{l} \operatorname{Yes}^{a,b} & \ O(b^{1+\lfloor C^*/\epsilon floor}) & \ O(b^{1+\lfloor C^*/\epsilon floor}) & \ O(b^{1+\lfloor C^*/\epsilon floor}) & \ \mathrm{Yes} & \end{array}$	$egin{array}{c} { m No} \ O(b^m) \ O(bm) \ { m No} \ { m No} \end{array}$	$egin{array}{c} { m No} \ O(b^\ell) \ O(b\ell) \ { m No} \end{array}$	$\begin{array}{c} \operatorname{Yes}^a \\ O(b^d) \\ O(bd) \\ \operatorname{Yes}^c \end{array}$	$egin{array}{l} \operatorname{Yes}^{a,d} & \ O(b^{d/2}) & \ O(b^{d/2}) & \ \operatorname{Yes}^{c,d} & \end{array}$

Summary of algorithms

Summary

Variety of uninformed search strategies:
breadth-first, depth-first, depth-limited, iterative deepening

Iterative deepening search uses only linear space and not much more time than other uninformed algorithms

Why?

- Very common algorithms.
- Used whenever we are looking for a path in a tree or graph.
 - Anywhere from games to programming languages.
- Properties matter!
 - time and/or space complexity.
- Understanding which algorithm to use in what circumstances.