Uninformed Search

CMPUT 366: Intelligent Systems

P&M §3.5

Logistics

- TA office hours begin next week
 - See eClass page for times and meeting links
- Assignment #1 released next week
- Python tutorials next week during TA office hours
 - Further details forthcoming

Recap: Graph Search

- - them all!
- function
- search graph with costs

• Many AI tasks can be represented as **search problems**

• A single generic graph search algorithm can then solve

• A search problem consists of states, actions, start states, a successor function, a goal function, optionally a cost

Solution quality can be represented by labelling arcs of the

Recap: Generic Graph Search Algorithm

Input: a graph; a set of start nodes; a goal function

- frontier := $\{ \langle s \rangle | s \text{ is a start node} \}$ while *frontier* is not empty: **select** a path *<n*₁*, n*₂*, ..., n_k>* from *frontier* **remove** <*n*₁, *n*₂, ..., *n_k*> from *frontier* if $goal(n_k)$: **return** <*n*₁, *n*₂, ..., *n*_k> for each neighbour *n* of n_k : **add** <*n*₁, *n*₂, ..., *n_k*, *n*> to frontier end while



https://artint.info/2e/html/ArtInt2e.Ch3.S4.html

(i.e., **expand** node n_k)

Which value is selected from the frontier defines the search strategy

Lecture Outline

- Logistics & Recap 1.
- 2. Properties of Algorithms and Search Graphs
- 3. Depth First and Breadth First Search
- Iterative Deepening Search 4.

Algorithm Properties

What properties of algorithms do we want to analyze?

- A search algorithm is complete if it is guaranteed to find a solution within a finite amount of time whenever a solution exists.
- The time complexity of a search algorithm is a measure of how much time the algorithm will take to run, in the worst case.
 - In this section we measure by total number of paths added to the frontier.
- The space complexity of a search algorithm is a measure of how much **space** the algorithm will use, in the **worst case**.
 - We measure by maximum number of paths in the frontier at one time.

Search Graph Properties

- Forward branch factor: Maximum number of neighbours Notation: *b*
- Maximum path length. (Could be infinite!) Notation: *m*
- Presence of **cycles**
- Length of the **shortest** path to a **goal** node

What properties of the search graph do algorithmic properties depend on?

Depth First Search

Input: a graph; a set of start nodes; a goal function

frontier := $\{\langle s \rangle \mid s \text{ is a start node}\}$ while *frontier* is not empty: select the newest path $\langle n_0, \ldots, n_k \rangle$ from frontier **remove** $\langle n_0, \ldots, n_k \rangle$ from *frontier* if $goal(n_k)$: return $\langle n_0, \ldots, n_k \rangle$ for each neighbour n of n_k : add $\langle n_0, \ldots, n_k, n \rangle$ to *frontier* end while

Question:

What data structure for the frontier implements this search strategy?



Depth First Search

Depth-first search always removes one of the **longest** paths from the frontier.

Example: Frontier: $[p_1, p_2, p_3, p_4]$ $successors(p_1) = \{n_1, n_2, n_3\}$

What happens?

- 1. Remove p_1 ; test p_1 for goal
- 2. Add { $\langle p_1, n_1 \rangle$, $\langle p1, n2 \rangle$, $\langle p1, n3 \rangle$ } to **front** of frontier
- 3. New frontier: $[\langle p_1, n_1 \rangle, \langle p_1, n_2 \rangle]$

Question: When is $\langle p_1, n_3 \rangle$ selected?



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$$\rangle, \langle p1, n3 \rangle, p_2, p_3, p_4]$$

4. p_2 is selected only after all paths starting with p_1 have been explored

Depth First Search Analysis

For a search graph with maximum branch factor b and maximum path length m...

- 1. What is the worst-case time complexity?
 - [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]
- 2. When is depth-first search complete?
- 3. What is the worst-case **space complexity**?



• [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]



When to Use Depth First Search

- When is depth-first search appropriate?
 - Memory is restricted
 - All solutions at same approximate depth
 - Order in which neighbours are searched can be tuned to find solution quickly
- When is depth-first search inappropriate?
 - Infinite paths exist
 - When there are likely to be shallow solutions
 - Especially if some other solutions are very deep



Breadth First Search

Input: a graph; a set of start nodes; a goal function

frontier := $\{\langle s \rangle \mid s \text{ is a start node}\}$ while *frontier* is not empty: select the oldest path $\langle n_0, \ldots, n_k \rangle$ from frontier **remove** $\langle n_0, \ldots, n_k \rangle$ from *frontier* if $goal(n_k)$: return $\langle n_0, \ldots, n_k \rangle$ for each neighbour n of n_k : add $\langle n_0, \ldots, n_k, n \rangle$ to *frontier* end while

Question:

What data structure for the frontier implements this search strategy?



Breadth First Search

Breadth-first search always removes one of the shortest paths from the frontier.

Example:

Frontier: $[p_1, p_2, p_3, p_4]$ $successors(p_1) = \{n_1, n_2, n_3\}$

What happens?

- 1. Remove p_1 ; test p_1 for goal
- 2. Add { $\langle p_1, n_1 \rangle$, $\langle p1, n2 \rangle$, $\langle p1, n3 \rangle$ } to **end** of frontier:
- 4. p_2 is selected **next**



3. New frontier: $[p_2, p_3, p_4, \langle p_1, n_1 \rangle, \langle p_1, n_2 \rangle, \langle p_1, n_3 \rangle]$

For a search graph with maximum branch factor b and maximum path length m...

- 1. What is the worst-case time complexity?
 - [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]
- 2. When is breadth-first search complete?
- 3. What is the worst-case **space complexity**?

Breadth First Search Analysis



• [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]

When to Use Breadth First Search

- When is breadth-first search appropriate?
 - When there might be infinite paths
 - When there are likely to be shallow solutions, or
 - When we want to guarantee a solution with fewest arcs
- When is breadth-first search inappropriate?
 - Large branching factor
 - All solutions located deep in the tree
 - Memory is restricted





Time complexity

- Run depth-first search to a maximum depth \bullet
 - then try again with a larger maximum •
 - until either goal found or graph completely searched

Comparing DFS vs. BFS

epth-first	Breadth-first
nly for finite	Complete

graphs	Complete
O(mb)	<i>O(b^m)</i>
<i>O(b^m)</i>	<i>O(b^m)</i>

• Can we get the space benefits of depth-first search without giving up completeness?

Iterative Deepening Search

Input: a *graph*; a set of *start nodes*; a *goal* function

for max_depth from 1 to ∞:
 Perform depth-first search to a maximum depth max_depth
end for

Iterative Deepening Search

Input: a graph; a set of start nodes; a goal function

for max_depth from 1 to ∞ : *more_nodes* := False frontier := $\{\langle s \rangle \mid s \text{ is a start node}\}$ while *frontier* is not empty: **remove** $\langle n_0, \ldots, n_k \rangle$ from *frontier* if $goal(n_k)$: return $\langle n_0, \ldots, n_k \rangle$ if k < max_depth: for each neighbour n of n_k : add $\langle n_0, \ldots, n_k, n \rangle$ to frontier else if n_k has neighbours: more_nodes := True end-while if more_nodes = False: return None

select the newest path $\langle n_0, ..., n_k \rangle$ from *frontier*

Iterative Deepening Search Analysis

For a search graph with maximum branch factor b and maximum path length m...

- What is the worst-case **time complexity**?
 - [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]
- When is iterative deepening search **complete**? 2.
- What is the worst-case **space complexity**? З.
 - [A: O(m)] [B: O(mb)] [C: $O(b^m)$] [D: it depends]

When to Use Iterative Deepening Search

- When is iterative deepening search **appropriate**?
 - Memory is limited, and
 - Both deep and shallow solutions may exist \bullet
 - or we prefer shallow ones
 - Tree may contain infinite paths \bullet

Summary

Different search strategies have different properties and behaviour

- **Depth first search** is space-efficient but not always complete or time-efficient
- **Breadth first search** is complete and always finds the shortest path to a goal, but is not space-efficient
- Iterative deepening search can provide the benefits of both, at the expense of some time-efficiency
- All three strategies must potentially expand every node, and are not guaranteed to return an optimal solution