COS30019-Introduction to Artificial Intelligence

Assignment 1 – Tree Based Search Report

HAI NAM NGO

STUDENT ID: 103488515

Page Count: 14 Pages (Excluding Cover pages, TOC, References)



Table of Contents

INSTURCTIONS	2
INTRODUCTION	ZETHMS 3 arch (BFS) 3 ch (DFS) 3 dearch (DLS- custom search 1) 3 st Search (GBFS) 4 aS) 5 ing A Star Search (IDA- custom search 2) 5 tween algorithms 5 aN 6 MISSING 13 15 15
SEARCH ALGORITHMS	3
Breadth First Search (BFS)	3
Depth First Search (DFS)	3
Depth Limited Search (DLS- custom search 1)	3
Greedy Best First Search (GBFS)	4
A Star Search (AS)	5
Iterative Deepening A Star Search (IDA- custom search 2)	5
Comparison between algorithms	5
IMPLEMENTATION	6
TESTING	8
FEATURES/ BUGS/ MISSING	13
RESEARCH	15
CONCLUSION	15
ACKNOWLEDGEMENTS/ RESOURCES	15
REFERENCES (APA7)	16

INSTURCTIONS

Step 1: Make sure that the laptop has downloaded python and pygame.

To download pygame, this command line should be used:

PS <u>C:\Users\ngoha</u>> pip install pygame

Step 2: Open the folder that has my "search.py" source code and the test case (txt file) that will be used (They need to be in the same folder) with the command format: "cd [the path file]".

PS C:\Users\ngoha> cd 'C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1'

Step 3: Run the command line with the following format:

python search.py [filename.txt] [the shorten name of the search algorithm]

C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1> python search.py .\RobotNav-test.txt BFS

There are 6 options for the search algorithm available in the program:

- 1. BFS Breath First Search
- 2. DFS Depth First Search
- 3. DLS- Depth Limited Search
- 4. AS- A* Search
- 5. GBFS- Greedy Best First Search
- 6. IDA- Iterative Deepening A* Search

Step 4: If you want to try other algorithms with the same test cases, you can use the GUI provided in the program (details will be explained later in the report).

Step 5: If you want to test other test cases, please quit the program, and use the command line to open the new test case, instructions are given in Step 2 and Step 3.

INTRODUCTION

Robotics aims to enable robots to find the most effective routes from beginning to end while avoiding obstacles to improve autonomous navigation across a variety of environments. This is also known as the description for **The Robot Navigation Problem** and is essential for applications in automated vehicles and robotics.

In general, an agent (robot in this case) must take through an environment to get from a starting point (start node) to a destination (goal node). Usually, this environment is shown as a graph, with nodes representing potential robot locations and edges representing viable routes between these locations. The key objectives of solving this problem are to guarantee that the robot avoids any obstacles that are represented in the graph as impassable nodes or edges, and that the path is optimal or nearly optimal in terms of cost or distance.

Basic Graph and Tree Concepts

Graph: A graph is made up of nodes and the lines that join node pairs. Nodes in robot navigation refer to locations, and lines indicate paths between these locations.

Tree: This is a unique kind of graph. There are no cycles in trees, they are made up of a root node and child nodes, each of which may have additional child nodes.

This program will be more like a graph search, and the report will analyze the problem based on the collected data from the actual implementation. Other terminology will be explained when it is used in the report.

SEARCH ALGORITHMS

Breadth First Search (BFS)

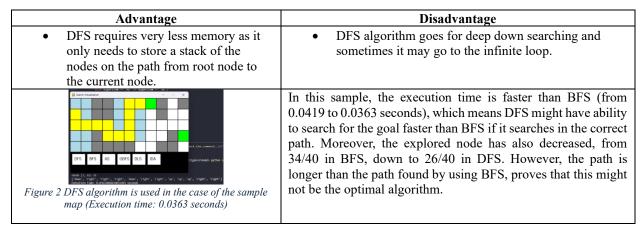
According to the research (*Uninformed Search Algorithms - Javatpoint*, n.d.), An algorithm for graph traversal called Breadth First Search (BFS) investigates every node in a graph at the depth it is currently at before going on to the next depth level. It visits each of its neighbors after beginning at a given node and progressing to the next level of neighbors. BFS is frequently utilized in graph-based pathfinding, connected components, and shortest path algorithms.

Advantage	Disadvantage
 BFS will offer a solution if one exists. If there are multiple solutions to a given problem, BFS will select the simplest solution with the fewest number of steps. 	 It requires a large amount of memory because each level of the tree must be saved in memory before proceeding to the next level. BFS requires a significant amount of time if the solution is located far from the initial state.
Figure 1 BFS algorithm is used in the case of the sample map (Execution time: 0.0419 seconds)	For example, in this sample map (provided by Assignment 1), the BFS algorithm has successfully found the shortest path to the nearest goal state. Optimal solution, however, we can see that it needs to explore nearly every node in the map (34/40), to find the path successfully. This proves that BFS needs lots of data related to explored node in, which is not as fast as other methods.

Depth First Search (DFS)

Depth-first search begins at the root node (initial state) and travels as far as possible along each branch before backtracking. The process of the DFS is nearly the same when comparing to BFS.

The difference between DFS and BFS: DFS uses LIFO queue (stack- Last in First Out) for its implementation, while BFS uses FIFO queue (First in First Out).



Depth Limited Search (DLS- custom search 1)

DLS algorithm is like DFS but has a specified limit. DLS can overcome the disadvantage of the infinite path in DFS. In this algorithm, the node at the depth limit will be treated as if it has no more child nodes and continue to search for the node in the next branch.

Advantage	Disadvantage
Memory efficient	 Have possibility of unable to find a solution if the goal node is deeper than the depth limit. Might not be optimal if there is more than one solution path.



Figure 3 DLS with depth limit 1000

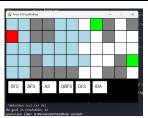


Figure 4 DLS with depth limit 7

In this sample, a solution path is found, and it is the same solution as the DFS above. However, Figure 3 is the case when the depth limit is set to 1000, if it is set to a low number like in the Figure 4, for example, the depth limit is 7, the algorithm cannot find the path, even though there is a possible solution available, because the goal state is deeper than the depth limit.

Greedy Best First Search (GBFS)

According to (*Informed Search Algorithms in AI - Javatpoint*, 2011), GBFS always chooses the most efficient node only at that time, and it will connect the efficient node list together to make up a path. It is the combination between DFS and BFS. It makes use of heuristics and searches. GBFS allows us to benefit from both algorithms. At each step, it can select the most promising node, and expands the node that is closest to the goal node, and the closest cost is estimated using a heuristic function and the cost to act.

f(n)=g(n), where g(n) is the cost to travel from the start node to the current node.

Advantage	Disadvantage
This algorithm is more efficient than	Might cause infinite loop like DFS
BFS and DFS.	Might not optimal

Since the sample map is easy to solve it is difficult to see the difference between BFS, DFS and GBFS, so a more complicated map (map3.txt in the test cases) has been created to analyze the GBFS algorithm.

Figure 5 BFS with the map3.txt	For this map, BFS needs to explore 59 nodes, to come up with the most optimal path to the goal (1,9), which is closer comparing to the other goal.
Figure 6 DFS with the map3.txt	DFS needs to explore 61 nodes, and it shows the path to the further goal, not even search for the goal that is closer to the initial state.
	GBFS might come up with the best option with the least nodes that need to explore, only 57 nodes, even though it is not that huge different from the length of explored list of the BFS and DFS, it still proves that GBFS is more efficient than both DFS and BFS.
Figure 7 GBFS with the map3.txt	Sometimes, GBFS might not be able to show the most optimal path, it is not optimal because of the natural of GBFS itself, this case is just because that path is the only option to reach the goal.

A Star Search (AS)

It employs the heuristic function h(n) and the cost to reach node n from the initial state g(n). It combines UCS features with greedy best-first search to efficiently solve the problem. The A^* search algorithm uses the heuristic function to find the shortest path through the search space. This search algorithm expands fewer search trees and delivers optimal results faster.

Advantage	Disadvantage
 can solve extremely complex 	Complexity issues
problems.	Have high chance (but not always) give the best path
Figure 8 AS with the sample map	Compared to other search algorithm above, AS come up with the optimal solution with the least node that need to be explored. This proves that AS is the best algorithm so far, although the execution time is quite long compared to other algorithm (0.065 seconds)

Iterative Deepening A Star Search (IDA- custom search 2)

IDA is a path-finding algorithm that combines the heuristic-driven search of AS with the iterative deepening approach of depth-first search to find the shortest path in a weighted graph. It functions similarly to AS, using a heuristic to estimate the cost to the goal, thereby determining which nodes to explore and at what depth.

Advantage	Disadvantage
use less memory than AS	As depth increases, there is a significant amount of redundancy and computational overhead because of each iteration's re-exploration.
Figure 9 IDA with the sample map	In this sample, IDA allows the agent to travel through the map, it can still reach the goal even though it is not that optimal.

Comparison between algorithms

Ranking (Personal Idea)	Algorithm	Time Complexity	Space Complexity	Complete	Optimal
Top2	BFS	$O(b^d)$	$O(b^d)$	Complete (finite nodes)	Optimal
Тор3	DFS	$O(b^m)$	$O(b^m)$	Complete (finite nodes)	No (might require more step to reach the goal)
Top5	DLS	$O(b^l)$	$O(b^l)$	Complete (if the goal state is above the depth limit)	Not optimal even when l>d.
Top4	GBFS	$O(b^m)$	$O(b^m)$	Incomplete	Not optimal
Top1	AS	$O(b^d)$	$O(b^d)$	Complete (finite nodes)	Optimal (If the heuristic function is valid)
Тор6	IDA	$O(b^d)$	<i>O</i> (<i>b</i>)	Incomplete (if heuristic function is not consistent)	Not optimal (if heuristic function is not consistent)

b: maximum child node can have; d: the depth; m: maximum depth; l: depth limit

IMPLEMENTATION

```
Breadth First Search (BFS)
                                                                                                                            Method __init__(my_map, initial_state, goal_states)

Call parent class constructor with my_map, initial_state, goal_states
                                                                                                                                   Initialize explored with initial_state
Initialize frontier with initial_state
Initialize bfsPath as an empty dictionary
                                                                                                                                   While frontier is not empty

current_state = Remove the oldest element from frontier (FIFO)
                                                                                                                                          If current_state is a goal state
Set self.explored to explored
Return generate_output(explored, bfsPath)
                                                                                                                                          neighbors = find_neighbor(current_state, directions)
For each neighbor in neighbors

If neighbor is in my_map.path_list and not in explored
                                                                                                                                                         Add neighbor to explored
Add neighbor to the end of frontier (FIFO)
Set bfsPath[neighbor] to current_state # Map the neighbor back to current state
                                                                                                                                   Set self.explored to explored
Return the number of explored nodes
                                                                                                                                                            Figure 10 Pseudocode for BFS implementation
                                                                                                                           ss DFS inherits Search
Method __init__(my_map, initial_state, goal_states)
Call parent class constructor with my_map, initial_state, goal_states
  Depth First Search (DFS)
                                                                                                                           Method search()
                                                                                                                                  Initialize explored with initial_state
Initialize frontier with initial_state
Initialize dfsPath as an empty dictionary
                                                                                                                                  While frontier is not empty
current_state = Remove the last element from frontier (LIFO)
If current_state is a goal state
Set self.explored to explored
Return generate_output(explored, dfsPath)
                                                                                                                                        Define directions as list of tuples [(1, 0), (0, 1), (-1, 0), (0, -1)] # RIGHT, DOWN, LEFT, UP neighbors = find_neighbor(current_state, directions)

for each neighbor is neighbors

If neighbor is in my_map.path_list and not in explored

Add neighbor to explored

Add neighbor to frontier (LIFO)

Set dfsPath[neighbor] to current_state # Map the neighbor back to current state
                                                                                                                                   Set self.explored to explored
Return the number of explored node
                                                                                                                                                            Figure 11 Pseudocode for DFS implementation
                                                                                                                         lass DLS inherits Search
       Depth Limited Search
                                                                                                                              Method __init__(my_map, initial_state, goal_states)
Call parent class constructor with my_map, initial_state, goal_states
    (DLS- custom search 1)
                                                                                                                                    nou search()
Initialize explored as a list with initial_state
Initialize frontier as a stack with (initial_state, 0) # Node with its depth
Initialize disPath as an empty dictionary
Set depth_limit to 7 # Define a maximum depth limit
                                                                                                                                     While frontier is not empty

current_state, depth = Remove the last element from frontier (LIFO)
                                                                                                                                            If depth is greater than depth_limit
Continue to the next iteration of the loop
                                                                                                                                            If check_goal_state(current_state) is true
Set self.explored to explored
Return generate_output(explored, dlsPath) # Generate and return output if goal is found
                                                                                                                                            directions = [(1, \theta), (\theta, 1), (-1, \theta), (\theta, -1)] # RIGHT, DOWN, LEFT, UP neighbors = find_neighbor(current_state, directions) # Get valid neighbors
                                                                                                                                             For each neighbor in neighbors
                                                                                                                                                    each neighbor in neighbors
If neighbor is in my_map.path_list and not in explored
Add neighbor to explored
Add (neighbor to explored
Add (neighbor, depth + 1) to frontier # Push neighbor with incremented depth
Set dlsPath[neighbor] to current_state # Map neighbor back to current state
                                                                                                                                      Set self.explored to explored
Return the number of explored nodes # Return count if no goal is found
                                                                                                                                                             Figure 12 Pseudocode for DLS implementation
```

```
Greedy Best First Search (GBFS)
                                                                                                                                                                                                                                                                                                                                                                                  Method __init__(my_map, initial_state, goal_states)
Call parent class constructor with my_map, initial_state, goal_states
                                                                                                                                                                                                                                                                                                                                                                               Method search()
Initialize unvisited with all cells in my_map.path_list set to infinity
set unvisited[initial_state] to 0
Initialize explored with initial_state
Initialize gbfsPath as an empty dictionary
                                                                                                                                                                                                                                                                                                                                                                                             While unvisited is not empty

current_state = Get the state with the minimum value in unvisited
                                                                                                                                                                                                                                                                                                                                                                                                          directions = [(0, -1), (-1, 0), (0, 1), (1, 0)] \neq UP, LEFT, DOWN, RIGHT neighbors = find_neighbor(current_state, directions)
                                                                                                                                                                                                                                                                                                                                                                                                                    each neighbor in neighbors
If neighbor is in my_map.path.list and net in explored
templist = unwisited[current_state] + 1
If templist = unwisited[neighbor]
Add neighbor to explored
Set unwisited[neighbor] to templist
Set gb/Santh[neighbor] to current_state
                                                                                                                                                                                                                                                                                                                                                                         Figure 13 Pseudocode for GBFS implementation
                                                                                                                                                                                                                                                                                                                                                         ss AS inherits Search
Method <u>init (my m</u>ap, initial_state, goal_states)
Call parent class constructor with my_map, initial_state, goal_states
                                                                     A Star Search (AS)
                                                                                                                                                                                                                                                                                                                                                                   thod search()
Initialize explored with initial_state
Initialize explored with initial_state
Initialize explored with infinite values
Initialize priority queue 'open' and dictionaries g_score and f_score with infinite values
                                                                                                                                                                                                                                                                                                                                                                     For each goal in goal states

Set g_score[start] to 0

Set f_score[start] to manhattan distance from start to goal

Add start to "open" with (f_score[start], manhattan distance, start)
                                                                                                                                                                                                                                                                                                                                                                                  While open is not empty \label{eq:current_state} \mbox{current\_state} = \mbox{get the node from open with the lowest } \mbox{f\_score}
                                                                                                                                                                                                                                                                                                                                                                                             If current_state is a goal state
Set self.explored to explored
Return generate_output(explored, aPath)
                                                                                                                                                                                                                                                                                                                                                                                             \label{eq:directions} \begin{split} &\text{directions} = [\,(\theta,\,-1),\,(-1,\,\theta),\,(\theta,\,1),\,(1,\,\theta)\,] &\text{ $\#$ UP, LEFT, DOWN, RIGHT neighbors} &- &\text{find_neighbor(current_state, directions)} \end{split}
                                                                                                                                                                                                                                                                                                                                                                                             For each neighbor in neighbors
If neighbor is valid and not in explored
Add neighbor to explored
                                                                                                                                                                                                                                                                                                                                                                                                                   Calculate temp_g_score and temp_f_score
If temp_f_score < f_score[neighbor]
Update g_score[neighbor] and f_score[neighbor]
Add neighbor to open with its new f_score
Update aPath mapping neighbor to current_state
                                                                                                                                                                                                                                                                                                                                                                      If no valid neighbor found and open is empty
Break|
Set self-explored to explored
Return the length of explored nodes
                                                                                                                                                                                                                                                                                                                                                                    Figure 14 Pseudocode for AS implementation
Iterative Deepening A Star Search (IDA-
                                                                       custom search 2)
                                                                                                                                                                                                                                                                                                                                                                                                             e path_cost as 1
# explored with initial_state
# idaPath as an empty dictionary
# start as initial_state
                                                                                                                                                                                                                                                                                                                                                                                                   Indian allowages as follows an engine of a single state of the same party law of the sam
                                                                                                                                                                                                                                                                                                                                                                                Figure 15 Pseudocode for IDA implementation
```

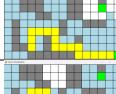
Algorithm	Approach	Implementation
BFS	Explores nodes in layers, ensuring that all nodes	Uses a queue to manage the frontier, ensuring nodes are
	at the current depth are explored before moving	explored in the order they were discovered.
	to the next depth.	
DFS	Explores as deep as possible along each branch	Uses a stack to manage the frontier, allowing the
	before backtracking.	algorithm to dive deep into the graph by processing the
		most recently discovered node first.
DLS	Like DFS but with a predetermined limit on the	Uses a stack with an additional depth parameter for each
	depth of the search, preventing it from going	node to control how deep the search can go.
	infinitely deep.	
GBFS	Uses a heuristic to prioritize nodes that are	Like AS, it uses a priority queue but only considers the
	estimated to be closest to the goal, ignoring the	heuristic value (h(n)), making it less optimal but often
	cost from the start node.	faster.
AS	Uses a heuristic to estimate the cost from the	Employs a priority queue based on the f-score of nodes,
	current node to the goal, combined with the cost	ensuring that the path with the lowest estimated total cost
	from the start node to the current node	is explored first.
IDA	Combines the depth-first exploration of DLS	Repeatedly performs depth-first searches, increasing the
	with the heuristic evaluation of AS.	depth threshold based on the f-score until the goal is
		found.

TESTING

Test case 1: map1.txt

1631 6436	i: map i.txt
	.\map1.txt BFS <node (8,="" 8)=""> 58 ['down', 'left', 'down', 'down', 'down', 'right', 'right', 'right', 'right', 'down', 'down', 'down'] Execution time: 0.04405331611633301 seconds</node>
	.\map1.txt DFS (DLS also have the same output) <node (8,="" 8)=""> 51 ['down', 'left', 'down', 'down', 'down', 'left', 'down', 'left', 'down', 'down', 'down', 'right', 'right', 'right', 'right', 'right', 'right', 'right', 'right', 'down', 'down', 'down', 'down'] Execution time: 0.003448486328125 seconds</node>
	.\map1.txt AS <node (8,="" 8)=""> 36 ['down', 'down', 'left', 'down', 'down', 'right', 'right', 'right', 'right', 'down', 'down', 'down'] Execution time: 0.0025932788848876953 seconds</node>
	.\map1.txt GBFS <node (8,="" 8)=""> 58 ['down', 'left', 'down', 'down', 'down', 'right', 'right', 'right', 'right', 'down', 'down', 'down'] Execution time: 0.0022513866424560547 seconds</node>
	.\map1.txt IDA <node (8,="" 8)=""> 44 ['down', 'left', 'down', 'left', 'down', 'left', 'down', 'left', 'down', 'down', 'down', 'right', 'up', 'right', 'up', 'right', 'up', 'right', 'up', 'right', 'down', 'right', 'down', 'right', 'down', 'dow</node>

Test case 2: map2.txt



.\map2.txt BFS

<Node (14, 7)> 73

['up', 'up', 'right', 'right', 'right', 'down', 'down', 'right', 'right', 'right', 'right', 'right', 'right', 'right', 'right', 'right'] Execution time: 0.0010006427764892578 seconds

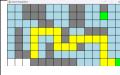


.\map2.txt DFS (DLS also have the same output)

<Node (14, 7)> 77

['up', 'up', 'right', 'right', 'down', 'right', 'down', 'down', 'right', 'right', 'up', 'up', 'right', 'right', 'down', 'down', 'down', 'right', 'r 'down', 'right', 'right', 'up', 'up', 'up', 'right', 'right', 'down', 'down', 'down']

Execution time: 0.0010089874267578125 seconds

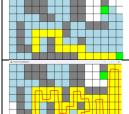


.\map2.txt AS

<Node (14, 7)> 82

['up', 'up', 'up', 'right', 'right', 'right', 'down', 'down', 'right', 'right', 'right', 'right', 'right', 'right', 'down', 'down', 'down', 'right', 'right']

Execution time: 0.003039121627807617 seconds



.\map2.txt GBFS

<Node (14, 7)> 76

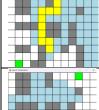
['up', 'up', 'right', 'right', 'right', 'down', 'down', 'right', ' Execution time: 0.0023317337036132812 seconds

.\map2.txt IDA

<Node (14, 7)> 73

['up', 'up', 'up', 'right', 'down', 'right', 'up', 'right', 'down', 'down', 'right', 'up', 'right', 'up', 'up', 'right', 'up', 'u 'down', 'right', 'down', 'down', 'left', 'down', 'right', 'right', 'up', 'up', 'up', 'right', 'down', 'down', 'down', 'right', 'up', 'up', 'up', 'up', 'up', 'up', 'up', 'right', 'down', 'down', 'down', 'down', 'down', 'down', 'down'] Execution time: 0.002196788787841797 seconds

Test case 3: map3.txt

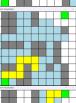


.\map3.txt DFS (DLS also have the same output)

<Node (9, 0)>61

['up', 'left', 'up', 'up', 'up', 'right', 'right', 'up', 'up', 'right', 'right', 'right']

Execution time: 0.0010082721710205078 seconds



.\map3.txt BFS

<Node (1, 9)> 59

['up', 'left', 'left', 'down', 'down', 'left', 'left', 'down']

Execution time: 0.00269317626953125 seconds

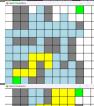


.\map3.txt AS

<Node (9, 0)> 22

['up', 'right', 'right', 'up', 'up', 'up', 'up', 'up', 'right', 'right']

Execution time: 0.001009225845336914 seconds



.\map2.txt GBFS

<Node (14, 7)> 76

['up', 'up', 'right', 'right', 'right', 'down', 'down', 'right', ' Execution time: 0.0023317337036132812 seconds



.\map3.txt IDA

<Node (9, 0)> 44

['up', 'left', 'up', 'up', 'up', 'right', 'down', 'right', 'up', 'up', 'up', 'right', 'down', 'right', 'up', 'up', 'up', 'right', 'up', 'u 'right']

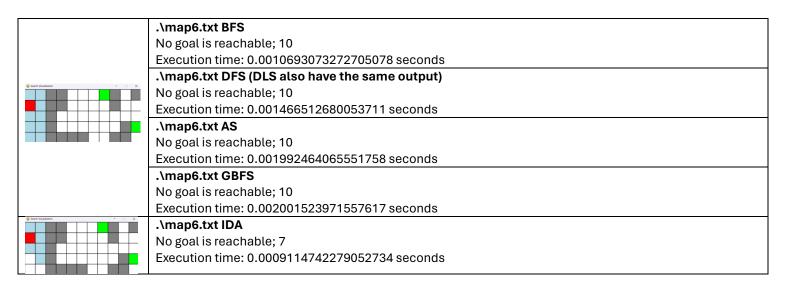
Execution time: 0.0010058879852294922 seconds

Test case 4: map4.txt

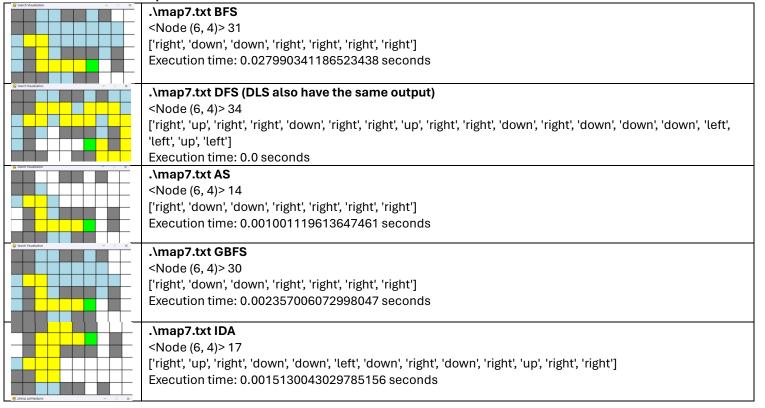
	ise +. map+.txt
Some Visualization - X	.\map4.txt BFS
	<node (7,="" 0)=""> 57</node>
	['up', 'up', 'right', 'right', 'right', 'right', 'up', 'right']
	Execution time: 0.023509502410888672 seconds
§ Searth-Vesalization - □ X	.\map4.txt DFS (DLS also have the same output)
	<node (7,="" 0)=""> 19</node>
	['up', 'up', 'right', 'right', 'right', 'right', 'up', 'right']
	Execution time: 0.0012240409851074219 seconds
Search Visualization	.\map4.txt AS
	<node (7,="" 0)=""> 19</node>
	['up', 'up', 'right', 'right', 'right', 'right', 'up', 'right']
	Execution time: 0.001680135726928711 seconds
	Excedition time: 0.001000100720320711 00001103
© Smith Venduction − X	.\map4.txt GBFS
	<node (7,="" 0)=""> 57</node>
	['up', 'up', 'right', 'right', 'right', 'right', 'up', 'right']
	Execution time: 0.0012960433959960938 seconds
© Cart Washing	
and the second	.\map4.txt IDA
	<node (7,="" 0)=""> 35</node>
	['up', 'up', 'right', 'right', 'up', 'right', 'down', 'down', 'left', 'down', 'down', 'down', 'right', 'up', 'right', 'down', 'down',
	'right', 'up', 'up', 'right', 'up', 'up', 'up', 'up', 'left']
	Execution time: 0.0022673606872558594 seconds

Test case 5: map5.txt

1000 000	se 3. map3.txt
	.\map5.txt BFS <node (4,="" 0)=""> 65 ['up', 'up', 'right', 'right', 'up', 'up', 'up', 'up'] Execution time: 0.002506732940673828 seconds</node>
	.\map5.txt DFS (DLS also have the same output) <node (4,="" 0)=""> 34 ['up', 'up', 'right', 'right', 'up', 'up', 'up', 'up'] Execution time: 0.0019555091857910156 seconds</node>
B was speaked and the speaked	.\map5.txt AS <node (4,="" 0)=""> 24 ['up', 'up', 'right', 'right', 'up', 'up', 'up', 'up'] Execution time: 0.001996755599975586 seconds</node>
	\map5.txt GBFS <node (4,="" 0)=""> 65 ['up', 'up', 'right', 'right', 'up', 'up', 'up', 'up'] Execution time: 0.004678249359130859 seconds</node>
\$ and the second	.\map5.txt IDA <node (4,="" 0)=""> 32 ['up', 'up', 'right', 'down', 'down', 'right', 'up', 'up', 'up', 'right', 'up', 'up', 'up', 'up', 'up'] Execution time: 0.00225830078125 seconds</node>

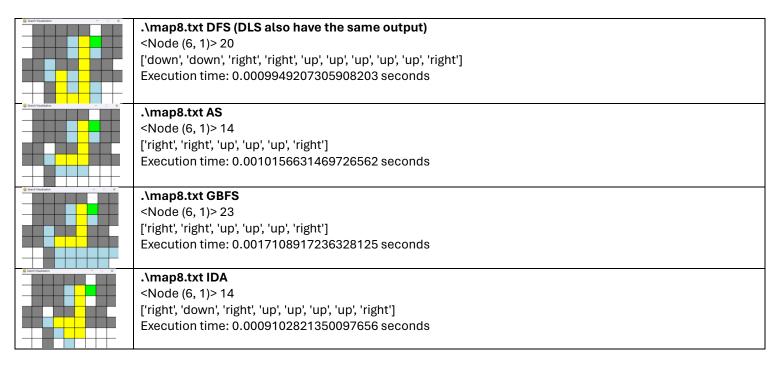


Test case 7: map7.txt



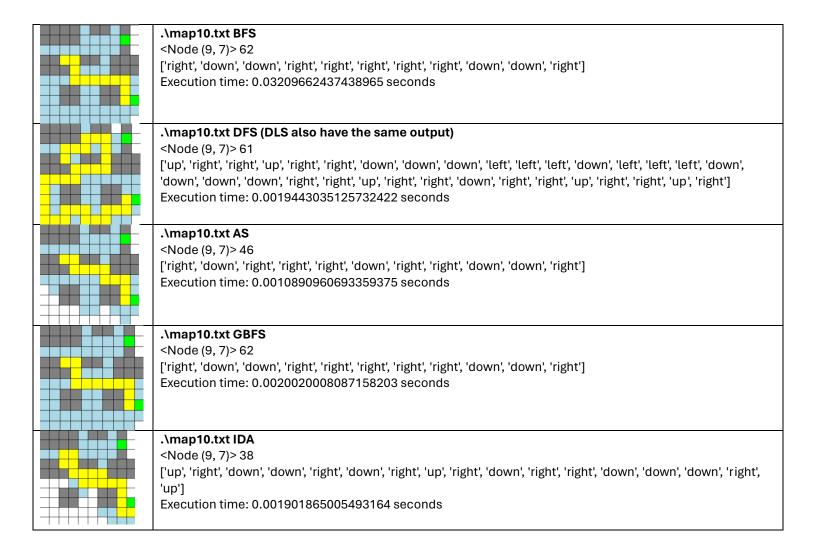
Test case 8: map8.txt

© Search Visualization — □ X	.\map8.txt BFS
	<node (6,="" 1)=""> 24</node>
	['right', 'right', 'up', 'up', 'up', 'right']
	Execution time: 0.026655912399291992 seconds



Test case 9: map9.txt

rest duse of maporixe	
	.\map9.txt BFS <node (8,="" 7)=""> 51 ['down', 'down', 'right', 'right', 'right'] Execution time: 0.030934810638427734 seconds</node>
	.\map9.txt DFS (DLS also have the same output) <node (8,="" 7)=""> 59 ['down', 'down', 'down', 'right', 'right', 'up', 'up', 'right', 'right', 'down', 'down', 'right'] Execution time: 0.0010154247283935547 seconds</node>
	.\map9.txt AS <node (8,="" 7)=""> 48 ['right', 'right', 'down', 'right', 'down', 'down', 'right'] Execution time: 0.002474546432495117 seconds</node>
	.\map9.txt GBFS <node (8,="" 7)=""> 51 ['down', 'down', 'right', 'right', 'right', 'right'] Execution time: 0.0010273456573486328 seconds</node>
	.\map9.txt IDA <node (8,="" 7)=""> 31 ['right', 'down', 'right', 'up', 'right', 'down', 'right'] Execution time: 0.0012373924255371094 seconds</node>



FEATURES/ BUGS/ MISSING

Error Handling Feature: To make sure that the program can run smoothly without errors that come from command line and text file. Some error checking has been created to check if the command line has the correct format or not, and to see if the text file is a valid one or not. If it is not satisfactory with the condition, guidance will be given for the user to check the error.

Situation 1: The user uses the wrong command format

```
PS C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1> python search.py dasd dasdas asdas pygame 2.5.2 (SDL 2.28.3, Python 3.12.0)

Hello from the pygame community. https://www.pygame.org/contribute.html

Error: You need to use the following format "python search.py [map_file.txt] [Search algorithm]". Please check the command.

PS C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1>
```

Figure 16 Situation 1 Demonstration

Situation 2: The user uses the wrong map file

```
PS C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1> python search.py dasdas BFS pygame 2.5.2 (SDL 2.28.3, Python 3.12.0)

Hello from the pygame community. https://www.pygame.org/contribute.html

Error: The map file does not exist. Please check the file path.
```

Figure 17 Situation 2 Demonstration

Situation 3: The user uses the wrong algorithm (the algorithm which is not exist in the program)

```
PS C:\Users\ngoha\OneDrive\Desktop\COS30019-Introduction to Artificial Intelligence\Asm1> python search.py map1.txt NONE pygame 2.5.2 (SDL 2.28.3, Python 3.12.0)
Hello from the pygame community. https://www.pygame.org/contribute.html
map1.txt NONE

Error: Wrong search algorithm input. Please check the command.
```

Figure 17 Situation 3 Demonstration

Situation 4: When the text file is not valid (wrong format)

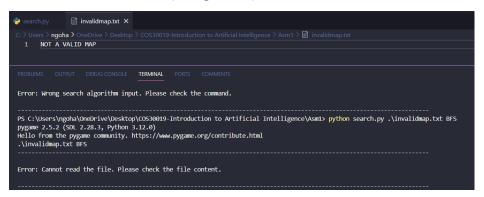


Figure 18 Situation 4 Demonstration

Execution time feature: The Python time library is used to implement this feature. An algorithm's running time from the starting state to one of the goal states can be estimated using this feature. This execution time is only an estimate and may vary each time the code runs. However, the user can still use this to compare the execution time between algorithm in the same map.

```
.\RobotNav-test.txt BFS

<Node (7, 0)> 34

['down', 'right', 'right', 'right', 'up', 'up', 'right', 'right', 'right']

Execution time: 0.045629024505615234 seconds
```

Figure 16 Execution time feature Demonstration

About custom search 2- IDA:

Even though it is successfully implemented, sometimes it also cannot find an available path even though there is one available. The reason might be that it also applies the natural of DLS, make the search is limited to specific number of nodes (that is what I think now). This issue is still in the process of fixing and analyzing, however, it does not affect the program, and is not considered as a bug.

RESEARCH

For further research, implementing a GUI for the program is the option that applied to this assignment.

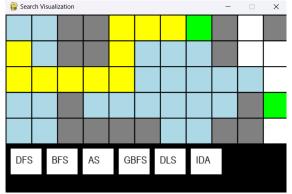


Figure 17 Appearance of the GUI

The source code for this visualization can be found in the visualization class. This function is made with the help of the pygame library. This function's logic is to print the static map first and then receive the result output from the algorithm once it has finished. The explored list (blue nodes) will then be printed. Ultimately, the completed route will be printed (with yellow nodes).

Click on any of the six buttons below the map to see a different algorithm displayed on the same map. The six buttons correspond to the six algorithms that are available in the program.

Note: The user should wait for the drawing process to finish and not press any buttons while the GUI is in this state. When a

button is clicked while the GUI is drawing, the request for that button will be visualized immediately after the current action is completed, preventing the user from seeing the results of the previous drawing process.

CONCLUSION

In conclusion, after thorough analysis and implementation of the Robot Navigation Program, it's recommended to use Breadth First Search (BFS) for uninformed search and A Star Search (AS) for informed search. BFS and AS consistently provide optimal solutions, with AS often requiring the least number of explored nodes. While the program's current performance is satisfactory, there is room for improvement in addressing issues like the reliability of Iterative Deepening A Star Search (IDA) and enhancing the graphical user interface (GUI) for better quality, for example, it can actually print out all neighbor nodes at the same time, not each of them at a time, if that is successfully implemented, it will be more efficient.

ACKNOWLEDGEMENTS/ RESOURCES

Java point website: This website has full information of both informed search and uninformed search in the program. This source helps me to understand more about the algorithm implementation, their logic and the advantages and disadvantages of each algorithm.

Learning Orbit's YouTube Channel: This is a channel that guides me through the pseudo code of BFS, DFS, AS and GBFS (in 4 different videos). And, how to implement using Python. In the tutorial, he used the library "pyamaze" which is created by himself., however, I did not use that but create everything by myself.

Algorithms & Technologies' Website: This website helps me to know how to implement IDA in Python, however my way of implementing is still different from this website's code.

ChatGPT: When my first source code isn't working well, this AI helps me identify and correct errors. It also suggests alternative approaches to the problem and evaluates the quality of my source code. I did not, however, copy and paste AI's source code; instead, I used the inspiration it provided to create a solid foundation for my own source code. To ensure that the generated map is entirely random, ChatGPT also provides most of the test cases, to make sure everything is randomly created.

The URLs will be cited in the references list below.

REFERENCES (APA7)

- 1. *Informed Search Algorithms in AI Javatpoint*. (2011). Www.javatpoint.com. https://www.javatpoint.com/ai-informed-search-algorithms
- Iterative Deepening A Star in Python. (2019, December 14). Www.algorithms-And-Technologies.com. https://www.algorithms-andtechnologies.com/iterative_deepening_a_star/python
- Learning Orbis. (2021a, September 18). Breadth First Search (BFS) in Python [Python Maze World- pyamaze]. Www.youtube.com.
 https://www.youtube.com/watch?v=D14YK-0MtcQ&t=319s
- Learning Orbis. (2021b, September 21). Depth First Search (DFS) in Python [Python Maze World- pyamaze]. Www.youtube.com.
 https://www.youtube.com/watch?v=sTRK9mQgYuc&t=0s
- 5. Learning Orbis. (2021c, October 2). *A-Star A* Search in Python [Python Maze World-pyamaze]*. Www.youtube.com. https://www.youtube.com/watch?v=W9zSr9jnoqY&t=0s
- 6. *Uninformed Search Algorithms Javatpoint*. (n.d.). Www.javatpoint.com. Retrieved April 16, 2024, from https://www.javatpoint.com/ai-uninformed-search-algorithms#:~:text=Uninformed%20search%20is%20a%20class