VIETNAM NATIONAL UNIVERSITY HO CHI MINH CITY UNIVERSITY OF SCIENCE

FACULTY OF INFORMATION TECHNOLOGY



Search

COURSE NAME: Fundamentals of Artificial Intelligence

Student: Vo Thanh Nghia (22127295)

Lecturer: Nguyen Thi Thu Hang Pham Trong Nghia Bui Duy Dang

Contents

2 Abstract						
3	Completion Level of Each Requirement					
	3.1 Self Assessment					
	3.1.1 Implementation of 5 core algorithms					
	3.1.2 Implementation of additional algorithms					
	3.2 5 test cases for each algorithm					
	3.2.1 test case 1					
4	Basic Theories					
	4.1 Breadth First Search (BFS) Algorithm					
	4.2 Depth First Search (DFS) Algorithm					
	4.3 Uniform-Cost Search (UCS)					
	4.4 Greedy Best First Search (GBFS) Algorithm					
	4.5 A* Search Algorithm					
	4.6 Iterative Deepening Search (IDS) Algorithm					

1 Information

1.1 Student Information

Student ID	Full name	Email
22127295	Vo Thanh Nghia	vtnghia22@clc.fitus.edu.vn

2 Abstract

This report details the implementation and evaluation of five fundamental graph search algorithms: Breadth First Search (BFS), Depth First Search (DFS), Uniform-Cost Search (UCS), Greedy Best First Search (GBFS), and A*. Developed within the provided student_functions.py framework, the project adheres to specific guidelines and constraints. The report includes a self-assessment of the project's completeness, an exploration of the underlying theories, a comparative analysis of UCS and A* algorithms, and the implementation of additional search algorithms for extra credit. This work demonstrates a comprehensive understanding of various graph search strategies, their theoretical foundations, and practical applications.

3 Completion Level of Each Requirement

3.1 Self Assessment

3.1.1 Implementation of 5 core algorithms

- Breadth First Search (BFS): 10
- Depth First Search (DFS): 10
- Uniform-Cost Search (UCS):
- Greedy Best First Search (GBFS): 10
- A* Search: 10

3.1.2 Implementation of additional algorithms

- Iterative Deepening Search (IDS): 10
- A* with Manhattan Distance: 10

3.2 5 test cases for each algorithm

3.2.1 test case 1

4 Basic Theories

4.1 Breadth First Search (BFS) Algorithm

Concepts

- **Introduction:** BFS is an algorithm for searching a tree data structure for a node that satisfies a given property. It starts at the tree root and explores all nodes at the present depth prior to moving on to the nodes at the next depth level.
- Traversal Method: BFS uses a queue data structure to keep track of nodes to be explored. It starts from a source node and explores all its neighboring nodes level by level.

Pseudo Code

```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node+NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier+a FIFO queue, with node as an element
  reached+ {problem.INITIAL}
  while not IS-EMPTY(frontier) do
       node+POP(frontier)
      for each child in EXPAND(problem, node) do
      s+child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
            add s to reached
            add child to frontier
return failure
```

Complexity

- Time Complexity: O(V + E), where V is the number of vertices and E is the number of edges. This is because each vertex and each edge is processed once.
- Space Complexity: O(V), as additional space is required for the queue and the visited set.

- Completeness: BFS is complete, meaning it will always find a solution if one exists, given that the graph is finite.
- Optimality: BFS is optimal if all edges have the same weight or no weights. It finds the shortest path in an unweighted graph.
- Traversal Type: UCS explores nodes with the lowest path cost first.

4.2 Depth First Search (DFS) Algorithm

Concepts

- Introduction: Depth First Search (DFS) is a graph traversal algorithm that explores as far as possible along each branch before backtracking. It traverses deeper into the graph, exploring nodes in depth-first order.
- Traversal Method: DFS uses a recursion (or stack) to keep track of nodes to be explored. It starts from a source node and explores as far as possible along each branch before backtracking.

Pseudo Code

```
function RECURSIVE-DFS(node, problem) returns a solution node or failure
  if problem.IS-GOAL(node.STATE) then return node

for each child in EXPAND(problem, node) do
    s ← child.STATE
  if s is not visited then
       mark s as visited
       result ← RECURSIVE-DFS(child, problem)
       if result != failure then return result

return failure

function DEPTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  mark node.STATE as visited
  return RECURSIVE-DFS(node, problem)
```

Complexity

- Time Complexity: O(V + E), where V is the number of vertices and E is the number of edges. In the worst case, DFS explores all vertices and edges of the graph.
- Space Complexity: O(V), as additional space is required for the stack and the visited set.

- Completeness: DFS is complete if the search tree is finite, meaning for a given finite search tree, DFS will come up with a solution if it exists.
- Optimality: DFS is not optimal, meaning the number of steps in reaching the solution, or the cost spent in reaching it is high.
- Traversal Type: Recursive DFS is a depth-first traversal.

4.3 Uniform-Cost Search (UCS)

Concepts

- Introduction: Uniform-Cost Search (UCS) is a graph search algorithm that finds the lowest cost path from a starting node to a goal node in a weighted graph.
- Traversal Method: UCS uses a priority queue ordered by path cost to expand nodes with the lowest cumulative cost.

Pseudo Code

```
function BEST-FIRST-SEARCH(problem,f) returns a solution node or failure
   node←NODE(STATE=problem.INITIAL)
   frontier+a priority queue ordered by f , with node as an element
   reached+a lookup table, with one entry with key problem.INITIAL and value node
   while not IS-EMPTY(frontier) do
        node←POP(frontier)
        if problem.IS-GOAL(node.STATE) then return node
        for each child in EXPAND(problem, node) do
            s+child.STATE
            if s is not in reached or child.PATH-COST < reached[s].PATH-COST then
                reached[s] ← child
                add child to frontier
   return failure
function EXPAND(problem, node) yields nodes
    s-node.STATE
    for each action in problem.ACTIONS(s) do
        s' ←problem.RESULT(s, action)
        cost+node.PATH-COST + problem.ACTION-COST(s, action,s')
        yield NODE(STATE=s', PARENT=node, ACTION=action, PATH-COST=cost)
```

function UNIFORM-COST-SEARCH(problem) returns a solution node, or failure return BEST-FIRST-SEARCH(problem, PATH-COST)

Complexity

- Time Complexity: $O((V+E) \log V)$, where V is the number of vertices and E is the number of edges. UCS uses a priority queue, and each edge and vertex may be processed multiple times.
- Space Complexity: O(V), as additional space is required for the priority queue and the visited set.

Properties

- Completeness: UCS is complete, as it will always find a solution if one exists, given non-negative edge costs.
- Optimality: UCS is optimal; it finds the lowest cost path if all edge costs are non-negative.
- Traversal Type: UCS explores nodes with the lowest path cost first.

4.4 Greedy Best First Search (GBFS) Algorithm

Concepts

- Introduction: Greedy Best First Search (GBFS) is a graph search algorithm that expands the most promising node chosen according to a heuristic function. It does not guarantee optimal solutions but is often efficient in practice for certain types of problems.
- Traversal Method: GBFS uses a priority queue ordered by the heuristic value to expand nodes with the most promising estimated cost.

Pseudo Code

function BEST-FIRST-SEARCH(problem, f) was mentioned in Section 4.3.

function GREEDY-BEST-FIRST-SEARCH(problem) returns a solution node or failure return BEST-FIRST-SEARCH(problem, HEURISTIC)

Complexity

- Time Complexity: The worst-case time complexity of GBFS is O(|V|), where V is the number of vertices. With a good heuristic function, however, the complexity can be substantially reduced, potentially reaching $O(b^m)$ on certain problems, where b is the branching factor and m is the maximum depth of the search.
- Space Complexity: GBFS has a space complexity of O(|V|), primarily due to the priority queue and the visited set.

- Completeness: GBFS is complete in finite state spaces but may not be complete in infinite state spaces due to the potential for encountering loops or cycles.
- Optimality: GBFS is not generally optimal because it does not consider the total path cost but relies solely on the heuristic function to guide the search.
- Traversal Type: GBFS explores nodes based on the heuristic value, prioritizing nodes that appear to be closest to the goal according to the heuristic.

4.5 A* Search Algorithm

Concepts

- Introduction: A* Search is a graph search algorithm that combines the advantages of both Dijkstra's algorithm and greedy best-first search. It uses both the actual cost from the start node (g-value) and an estimated cost to the goal node (h-value) to guide the search.
- Traversal Method: A* Search uses a priority queue ordered by the sum of g-value and h-value (f-value) to expand nodes with the lowest estimated total cost first.

Pseudo Code

function BEST-FIRST-SEARCH(problem, f) was mentioned in Section 4.3. function A-STAR-SEARCH(problem) returns a solution node or failure return BEST-FIRST-SEARCH(problem, g+h)

Complexity

- Time Complexity: In the worst case, A* Search has a time complexity of $O(b^d)$, where b is the branching factor of the graph and d is the depth of the optimal solution. However, with a good heuristic function, it often performs much better.
- Space Complexity: A* Search has a space complexity of O(|V|), where V is the number of vertices, due to the priority queue and the visited set.

Properties

- Completeness: A* Search is complete if a solution exists, given that the branching factor is finite and all edge costs are non-negative.
- Optimality: A* Search is optimal if the heuristic function h(s) is admissible (never overestimates the true cost to reach the goal) and consistent (satisfies the triangle inequality).
- Traversal Type: A* Search explores nodes based on the f-value, which combines the g-value (cost from start to current node) and the h-value (estimated cost from current node to goal).

4.6 Iterative Deepening Search (IDS) Algorithm

Concepts

- Introduction: Iterative Deepening Search (IDS) combines the benefits of depth-first and breadth-first search. It performs a series of depth-limited searches, increasing the depth limit with each iteration.
- Traversal Method: IDS uses a depth-limited search repeatedly, increasing the limit after each complete pass until a solution is found or the search space is exhausted.

Pseudo Code

function ITERATIVE-DEEPENING-SEARCH(problem) returns a solution node or failure

```
for depth = 0 to oo do
    result+DEPTH-LIMITED-SEARCH(problem, depth)
    if result != cutoff then return result

function DEPTH-LIMITED-SEARCH(problem, 1) returns a node or failure or cutoff
    frontier+a LIFO queue (stack) with NODE(problem.INITIAL) as an element
    result+failure
    while not IS-EMPTY(frontier) do
        node+POP(frontier)
        if problem.IS-GOAL(node.STATE) then return node
        if DEPTH(node) > 1 then
            result+cutoff
        else if not IS-CYCLE(node) do
            for each child in EXPAND(problem, node) do
                  add child to frontier
    return result
```

Complexity

- Time Complexity: $O(b^d)$, where b is the branching factor and d is the depth of the shallowest goal node. This is because IDS performs multiple passes, but the time complexity remains exponential.
- Space Complexity: O(bd), as IDS only needs to store a stack of nodes up to the current depth limit.

- Completeness: IDS is complete if the branching factor is finite, meaning it will eventually find a solution if one exists.
- Optimality: IDS is optimal if all step costs are equal, as it finds the shallowest goal node.
- Traversal Type: IDS performs iterative deepening, combining depth-first search's space efficiency with breadth-first search's completeness.

5 Comparison of UCS and A* Algorithms

	Uniform-Cost Search (UCS)	A* Search
Heuristic Function	Does not use a heuristic function. It purely considers the path cost from the start node to the current node $(g(n))$.	Incorporates a heuristic function $(h(n))$ in addition to the path cost. The total cost function for A^* is $f(n) = g(n) + h(n)$, where $h(n)$ is an estimate of the cost from the current node to the goal.
Cost Evaluation	Chooses the node with the minimum path cost $(g(n))$ among nodes that have not been visited but have a neighbor that has been visited.	Chooses the node with the minimum combined cost $(g(n)+h(n))$ among nodes in the frontier.
Optimality	Guaranteed to find an optimal solution if the heuristic used in A* is admissible (never overestimates the true cost).	
Time and Space Complexity	Have the same time and space complexity in the worst case. How- ever, A* can be more efficient in practice due to the heuristic guid- ing the search more effectively.	
Completeness	Complete, meaning if there is a solution, both algorithms will find it.	
Practical Benefit	Does not benefit from heuristic guidance and may explore more nodes than necessary.	Allows the use of a heuristic, which in non-worst-case scenarios can significantly reduce the number of nodes explored, providing a large practical benefit.

Table 1: Comparison of UCS and A^* Algorithms