# Title: Implementation of A\* Search and Best First Search Algorithm and Differentiation from BFS and DFS

# 1. Objectives

- The objectives of this lab are:
- To understand and implement A Search\* and Best First Search (GBFS) algorithms.
- To analyze and compare these algorithms with Breadth-First Search (BFS) and Depth-First Search (DFS).
- To evaluate their efficiency in terms of time complexity, optimality, and space utilization.

#### 2. Introduction

Search algorithms are fundamental in Artificial Intelligence for pathfinding and decision-making.

- A Search\*: Uses both cost (g(n)) and heuristic (h(n)) to find the optimal path efficiently.
- \*Best First Search (BFS)\*\*: A greedy algorithm that selects nodes based on the heuristic function alone, often leading to faster but non-optimal solutions.
- Breadth-First Search (BFS): Explores all paths level by level, guaranteeing the shortest path in unweighted graphs.
- Depth-First Search (DFS): Explores deeply before backtracking, which is memory-efficient but may not find the best solution.

## **Key Differences:**

- $A^*$  balances cost (g(n)) and heuristic (h(n)) to guarantee the shortest path.
- Best First Search uses only h(n), making it faster but sometimes misleading.
- BFS is complete and optimal for unweighted graphs but requires more memory.
- DFS is useful for deep searches but not optimal.

## 3. Lab Work Implementation

#### A Search Algorithm\*

#### Logic:

- Uses a priority queue (heapq) to always expand the most promising node based on f(n) = g(n) + h(n).
- Ensures the shortest path by balancing cost and heuristic values.

#### Python Code for A Search\*

```
import heapq
def a_star(graph, start, goal, heuristic):
  priority_queue = [(0, start, [start])]
  g_cost = \{start: 0\}
  while priority_queue:
    cost, node, path = heapq.heappop(priority_queue)
    if node == goal:
       return path
    for neighbor, weight in graph.get(node, { }).items():
       new_cost = g_cost[node] + weight
       if neighbor not in g_cost or new_cost < g_cost[neighbor]:
         g_cost[neighbor] = new_cost
         f_cost = new_cost + heuristic[neighbor]
         heapq.heappush(priority_queue, (f_cost, neighbor, path + [neighbor]))
  return None
```

#### **Best First Search Algorithm**

# Logic:

- Uses priority queue but selects nodes only based on h(n).
- Faster than A\* but does not guarantee the shortest path.

# **Python Code for Best First Search**

```
from heapq import heappop, heappush

def best_first_search(graph, start, goal, heuristic):
    priority_queue = [(heuristic[start], start, [start])]
    visited = set()

while priority_queue:
    _, node, path = heappop(priority_queue)
    if node == goal:
        return path

if node not in visited:
    visited.add(node)
    for neighbor in graph.get(node, {}):
        heappush(priority_queue, (heuristic[neighbor], neighbor, path + [neighbor]))

return None
```

# **Breadth-First Search Algorithm**

# Logic:

- Uses a queue (FIFO) to explore level by level.
- Guarantees shortest path in unweighted graphs.

# **Python Code for BFS**

```
from collections import deque
def bfs(graph, start, goal):
  queue = deque([(start, [start])])
  visited = set()
  while queue:
     node, path = queue.popleft()
    if node == goal:
       return path
     if node not in visited:
       visited.add(node)
       for neighbor in graph.get(node, {}):
          queue.append((neighbor, path + [neighbor]))
  return None
```

# **Depth-First Search Algorithm**

## Logic:

- Uses a **stack** (**LIFO**) to explore deeply before backtracking.
- Memory-efficient but **not guaranteed to find the best solution**.

# **Python Code for DFS**

```
def dfs(graph, start, goal):
    stack = [(start, [start])]
    visited = set()

while stack:
    node, path = stack.pop()
    if node == goal:
        return path

if node not in visited:
    visited.add(node)
    for neighbor in graph.get(node, { }):
        stack.append((neighbor, path + [neighbor]))

return None
```

## 4. Result Analysis and Comparison

## **Comparison of Algorithms**

Algorithm	<b>Uses Heuristic?</b>	<b>Guarantees Shortest Path?</b>	Space Complexity	Completeness
A Search*	✓ Yes	✓ Yes	High	✓ Yes
Best First Search	✓ Yes	× No	Medium	× No
BFS	× No	✓ Yes	High	✓ Yes
DFS	× No	× No	Low	× No

#### **Observations:**

- 1. A Search\* finds the shortest path most efficiently but requires more memory.
- 2. Best First Search is fast but can get trapped in local minima.
- 3. BFS guarantees the shortest path in unweighted graphs but is slower than A\*.
- 4. DFS is better for deep searches but is not guaranteed to find the optimal path.

#### **5. Discussion and Conclusion**

- A\* Search is the best choice when both efficiency and accuracy are required, making it widely used in robotics, navigation, and AI decision-making.
- Best First Search is suitable for quick searches but does not always find the optimal solution.
- BFS is preferred when shortest paths are needed in unweighted graphs.
- DFS is useful for deep exploration but may not be the best for pathfinding.

Thus, A Search remains the most reliable algorithm for optimal pathfinding in AI applications.

#### 6. References

- Stuart Russell, Peter Norvig "Artificial Intelligence: A Modern Approach"
- Wikipedia: A\* Search Algorithm
- GeeksforGeeks: Best First Search Algorithm

[NOTE: Full codes with visualization are available at github.com/ar-sayeem/AI-Lab