

Puzzle-8: Graph Search, IDS, and Uniform Cost Backtracking

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A. Objective

The objective is to solve the 8-puzzle problem using a graph search agent. The goal is to implement:

- Graph search agent assuming uniform cost per move.
- Iterative Deepening Search (IDS).
- Backtracking to produce the solution path.
- Generation of solvable Puzzle-8 instances at a specific depth d .
- Measure memory and execution time for each instance.

B. Problem Statement

The 8-puzzle consists of 8 numbered tiles and a blank space arranged in a 3×3 grid. The goal is to move tiles until they reach the goal state:

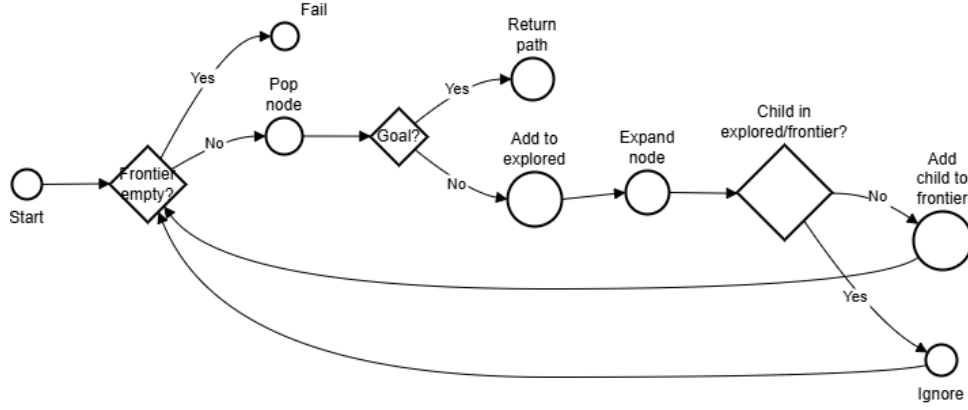
Goal state: $[1, 2, 3, 4, 5, 6, 7, 8, 0]$

where 0 represents the blank. Each move has equal cost. We are to implement graph search agents, generate instances at given depths, and record memory/time performance.

C. Methodology

- Each state is a list of 9 integers representing tiles row-wise.
- Successor states are generated by moving the blank in four directions.
- BFS implements uniform-cost search; IDS is used for memory-efficient deep search.
- Backtracking reconstructs the path from goal to start.
- Puzzle generation at depth d ensures solvable instances.

D. Flowchart



E. Algorithms

Algorithm 1 Graph Search Agent (Uniform-Cost / BFS)

```

1: Function: graph_search(initial_state, goal_state)
2: Initialize queue with initial_state and visited set
3: Initialize parent mapping: came_from[start] = None
4: while queue is not empty do
5:   Pop node from queue
6:   if node == goal_state then
7:     Backtrack using came_from to get path
8:     return path
9:   end if
10:  for each successor of node do
11:    if successor not in visited and not in queue then
12:      Add successor to queue
13:      came_from[successor] = node
14:    end if
15:  end for
16: end while
17: return failure

```

F. Results

G. Conclusion

The graph search agent successfully solves Puzzle-8 instances using uniform-cost search (BFS). IDS reduces memory requirements while guaranteeing completeness. Backtracking correctly reconstructs the solution path, and puzzle generation at depth d produces solvable instances. Memory, time, and moves vary with depth, demonstrating algorithm efficiency.

Algorithm 2 Iterative Deepening Search (IDS)

```
1: Function: IDS(start, goal)
2: Initialize depth limit  $d = 0$ 
3: repeat
4:   Perform Depth-Limited Search with limit  $d$ 
5:   if goal found then
6:     return path
7:   end if
8:    $d \leftarrow d + 1$ 
9: until goal found
10: return failure
```

Algorithm 3 Backtracking Path Retrieval

```
1: Function: backtrack(came_from, goal)
2: Initialize empty path list
3: current = goal
4: while current  $\neq$  None do
5:   Append current to path
6:   current = parent from came_from
7: end while
8: return reversed path
```

Algorithm 4 Puzzle Generation at Depth d

```
1: Function: generate_puzzle_at_depth( $d$ )
2: state = GOAL_STATE
3: for  $i = 1$  to  $d$  do
4:   Generate successors of state
5:   Randomly select one successor as new state
6: end for
7: return state
```

Depth	Memory (MB)	Time (s)	Moves
6	0.0	0.0001	0
8	0.01	0.0002	2
10	0.0	0.0	0
13	0.24	0.0065	9
15	0.03	0.0015	7
17	0.03	0.0015	7
20	0.01	0.0004	4
25	0.08	0.0043	9
40	0.01	0.0004	4

Table 1: Memory and time requirements for solving Puzzle-8 instances at depth d .