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In [1]: import numpy as np
        import random
In [3]: # Maze environment
        # 0 = free space, -1 = obstacle, 1 = goal
        maze = np.array([
            [0, 0, 0, 0],
            [0, -1, 0, 0],
            [0, 0, 0, -1],
            [0, 0, 0, 1]
        1)
        # Display the maze
        print("Maze layout (0=free, -1=obstacle, 1=goal):")
        print(maze)
       Maze layout (0=free, -1=obstacle, 1=goal):
       [[ 0 0 0 0]
        [ 0 -1 0 0]
        [000-1]
        [0001]]
In [5]: # Q-learning parameters
        alpha = 0.7  # Learning rate
gamma = 0.9  # Discount factor
        epsilon = 0.3  # Exploration rate
        episodes = 500  # Training episodes
        # Possible actions
        actions = ['up', 'down', 'left', 'right']
        # Initialize Q-table
        q_table = np.zeros((4, 4, len(actions)))
In [7]: def is_valid(state):
            x, y = state
            return 0 <= x < 4 and 0 <= y < 4 and maze[x, y] != -1
        def get_next_state(state, action):
            x, y = state
            if action == 'up': x -= 1
            if action == 'down': x += 1
            if action == 'left': y -= 1
            if action == 'right': y += 1
            if not is_valid((x, y)):
                return state, -5 # invalid move penalty
            if maze[x, y] == 1:
                return (x, y), 10 # goal reward
            return (x, y), -1 # normal step cost
In [9]: for episode in range(episodes):
            state = (0, 0)
            done = False
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while not done:
                 # ε-greedy action selection
                  if random.uniform(0, 1) < epsilon:</pre>
                      action = random.choice(actions)
                  else:
                     action = actions[np.argmax(q table[state[0], state[1]])]
                  next_state, reward = get_next_state(state, action)
                  # Q-value update
                  old_value = q_table[state[0], state[1], actions.index(action)]
                  next_max = np.max(q_table[next_state[0], next_state[1]])
                  new_value = old_value + alpha * (reward + gamma * next_max - old_value)
                  q table[state[0], state[1], actions.index(action)] = new value
                  state = next_state
                  # Stop if goal is reached
                 if maze[state[0], state[1]] == 1:
                     done = True
         print(" Training completed!")
        Training completed!
In [11]: state = (0, 0)
         path = [state]
         while maze[state[0], state[1]] != 1:
             action = actions[np.argmax(q_table[state[0], state[1]])]
             state, _ = get_next_state(state, action)
             if state in path:
                  break # prevent infinite loop
             path.append(state)
         print("\nLearned path to goal:")
         print(path)
        Learned path to goal:
        [(0, 0), (1, 0), (2, 0), (3, 0), (3, 1), (3, 2), (3, 3)]
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