```
import numpy as np
num rows = 4
num\_cols = 4
num_states = num_rows * num_cols # 4x4 grid
num_actions = 4 # Up, Down, Left, Right
num_episodes = 500
alpha = 0.6 # Learning rate
gamma = 0.9 # Discount factor
0 = np.zeros((num states, num actions), dtvpe=float)
rewards = np.random.randint(1, 10, num_states)
# Define the goal state and obstacles
goal_state = num_states - 1
obstacles = [5, 7, 10]
# Update rewards for goal state and obstacles
rewards[goal_state] = 10
for obstacle in obstacles:
   rewards[obstacle] = -10
# List to store the path taken during each episode
episode_paths = []
\label{lem:def-qlearning} \mbox{\tt def q_learning(num\_states, num\_actions, num\_episodes, alpha, gamma):}
    for episode in range(num_episodes):
        state = np.random.randint(0, num_states)
        episode_path = [state] # Initialize path for this episode
        while True:
            action = np.argmax(Q[state, :]) \ if \ np.random.rand() < 0.9 \ else \ np.random.randint(0, \ num\_actions)
            next state = (state + action) % num states
            reward = rewards[next state]
            Q[state, action] = (1 - alpha) * Q[state, action] + alpha * (reward + gamma * np.max(Q[next_state, :]))
            state = next_state
            episode_path.append(state) # Add the next state to the episode path
            if state == num_states - 1:
                episode_paths.append(episode_path) # Store the episode path
                break
    return 0
learned_Q_values = q_learning(num_states, num_actions, num_episodes, alpha, gamma)
print("Learned Q-values:")
print(learned_Q_values)

    → Learned Q-values:

     [[78.88996516 80.78980777 76.8897647 82.1
      [80.78998346 76.88971639 82.1
                                           83.1
      [46.13399819 82.1
                              77.7816
                                           62.53471715]
      [82.09995945 83.09994519 70.09872244 89.
      [82.42707128 70.09264948 89.
                                           67,399147861
                              56.61585753 51.59793908]
                  89.
      Γ-6.
                  67.39999998 86. 90.
      [89.
      [50.6130068 86. 54.
                                           0.
      [70.93463195 90.
                              42.6
                                           76.8951732 ]
      [90.
                  71.
                               85.
                                           85.
                          84.99942957 90.
      [70.86278347 85.
      [84.99999946 84.99989418 90.
                                           76.70226534]
      [84.99999866 90. 76.6599301 79.17371568]
                  76.711
                               79.20099524 78.89
      [90.
      [76.66936199 79.13216411 78.87556821 80.79
      [16.9416
                              47.934
                                          76.88999471]]
                   0.
```

```
# Extract the shortest path from the recorded episode paths
if episode_paths:
    shortest_path = min(episode_paths, key=len)
    print("Shortest path to goal state while learning:")
    print(shortest_path)

# Display the states visited along the shortest path
    print("States visited along the shortest path:")
    for state in shortest_path:
        row = state // num_cols
        col = state % num_cols
        print(f"State: ({row}, {col})")
else:
    print("No paths recorded during learning.")
```

Shortest path to goal state while learning: [15, 15]
States visited along the shortest path:
State: (3, 3)
State: (3, 3)