Q_Learning_From_Scratch

March 15, 2024

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1 1.) Define Enviornment

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[1]: import numpy as np
     import matplotlib.pyplot as plt
     from IPython.display import clear_output
     import time
     grid_size = 5
     actions = ['up', 'down', 'left', 'right']
     num_actions = len(actions)
     agent_position = [0, 0]
     goal_position = [4, 4]
     # Rewards
     rewards = {'goal': 1, 'other': -0.01} # Minor negative reward to encourage_
      \hookrightarrow exploration
     # Initialize Q-table
     Q_table = np.zeros((grid_size, grid_size, num_actions))
     # Learning parameters
     learning_rate = 0.1
     discount_factor = 0.95
     episodes = 1000
     epsilon = 0.1 # Exploration rate
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2 2.) Define Action Rewards

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[2]: # Visualization setup

def plot_episode(steps, episode):
    clear_output(wait=True)
    plt.figure(figsize=(5, 5))
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plt.title(f"Episode: {episode}, Steps: {steps}")
    plt.xlim(-0.5, grid_size-0.5)
    plt.ylim(-0.5, grid_size-0.5)
    plt.grid()
    for i in range(grid_size):
        for j in range(grid_size):
            if [i, j] == agent_position:
                plt.text(j, grid_size-1-i, 'A', ha='center', va='center')
            elif [i, j] == goal_position:
                plt.text(j, grid_size-1-i, 'G', ha='center', va='center')
                plt.text(j, grid_size-1-i, '.', ha='center', va='center')
    plt.show()
def move_agent(agent_position, action_index):
    if actions[action_index] == 'up' and agent_position[0] > 0:
        agent_position[0] -= 1
    elif actions[action_index] == 'down' and agent_position[0] < grid_size - 1:</pre>
        agent_position[0] += 1
    elif actions[action_index] == 'left' and agent_position[1] > 0:
        agent_position[1] -= 1
    elif actions[action_index] == 'right' and agent_position[1] < grid_size - 1:</pre>
        agent_position[1] += 1
    return agent_position
def get_reward(agent_position):
    if agent_position == goal_position:
        return rewards['goal']
    else:
        return rewards['other']
```

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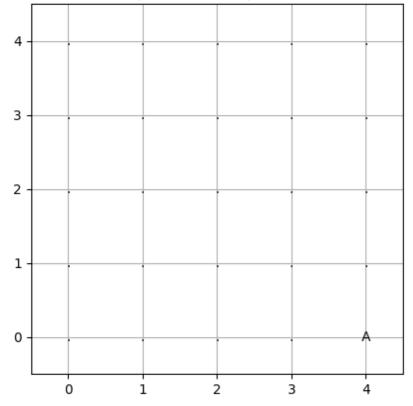
3 3.) Implement Basic Q learning

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[3]: for episode in range(episodes):
    agent_position = [0, 0] # Reset position at start of each episode
    steps = 0

while agent_position != goal_position:
    steps += 1
    if np.random.rand() < epsilon: # Explore
        action = np.random.randint(num_actions)
    else: # Exploit</pre>
```

```
action = np.argmax(Q_table[agent_position[0], agent_position[1], :])
      old_position = list(agent_position)
     new_position = move_agent(list(agent_position), action)
     reward = get_reward(new_position)
      # Update Q-table
     old_q_value = Q_table[old_position[0], old_position[1], action]
     future_q_value = np.max(Q_table[new_position[0], new_position[1], :])
     →learning_rate * (reward + discount_factor * future_q_value - old_q_value)
     agent_position = new_position
      # Visualization every 100 episodes
      if episode % 100 == 0:
         plot_episode(steps, episode)
         time.sleep(0.1) # Slow down the visualization
  if steps <= grid_size * 2: # Early stop if it finds a reasonably good path
     break
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