## cs104-denis-practical-8-q-learning

## October 3, 2024

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[4]: import numpy as np
     import random
     # Environment: Simple Grid
     class GridEnvironment:
         def __init__(self, grid_size=(4, 4), start=(0, 0), goal=(3, 3)):
             self.grid_size = grid_size
             self.start = start
             self.goal = goal
             self.state = start
             self.done = False
         def reset(self):
             self.state = self.start
             self.done = False
             return self.state
         def step(self, action):
             if self.done:
                 raise Exception("Episode is done. Please reset the environment.")
             next_state = list(self.state)
             if action == 0: # Up
                 next_state[0] = max(0, self.state[0] - 1)
             elif action == 1: # Down
                 next_state[0] = min(self.grid_size[0] - 1, self.state[0] + 1)
             elif action == 2: # Left
                 next_state[1] = max(0, self.state[1] - 1)
             elif action == 3: # Right
                 next_state[1] = min(self.grid_size[1] - 1, self.state[1] + 1)
             self.state = tuple(next_state)
             if self.state == self.goal:
                 self.done = True
                 reward = 1 # Reward for reaching the goal
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[5]: # Q-Learning Algorithm
     class QLearningAgent:
         def __init__(self, actions, learning_rate=0.1, discount_factor=0.9,_

exploration_rate=1.0, exploration_decay=0.99, min_exploration_rate=0.01):
             self.q_table = {}
             self.actions = actions
             self.learning_rate = learning_rate
             self.discount_factor = discount_factor
             self.exploration_rate = exploration_rate
             self.exploration_decay = exploration_decay
             self.min_exploration_rate = min_exploration_rate
         def get_q_value(self, state, action):
             return self.q_table.get((state, action), 0.0)
         def update_q_value(self, state, action, reward, next_state):
             best_next_q = max(self.get_q value(next_state, a) for a in range(self.
      →actions))
             current_q = self.get_q_value(state, action)
             new_q = current_q + self.learning_rate * (reward + self.discount_factor⊔
      →* best_next_q - current_q)
             self.q_table[(state, action)] = new_q
         def choose_action(self, state):
             if random.uniform(0, 1) < self.exploration_rate:</pre>
                 return random.choice(range(self.actions)) # Explore
             else:
                 return np.argmax([self.get_q_value(state, a) for a in range(self.
      ⇔actions)]) # Exploit
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def decay_exploration(self):
    self.exploration_rate = max(self.min_exploration_rate, self.
exploration_rate * self.exploration_decay)
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[6]: # Main loop to run Q-learning
def train_agent(episodes=1000):
    env = GridEnvironment()
    agent = QLearningAgent(actions=4)

for episode in range(episodes):
    state = env.reset()

while True:
    action = agent.choose_action(state)
    next_state, reward, done = env.step(action)
    agent.update_q_value(state, action, reward, next_state)
    state = next_state

    if done:
        break

    agent.decay_exploration()

return agent
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Total Reward: 0.5