7/22/24, 12:42 PM ML practical 6

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In [2]: '''
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        COURSE: AI&DS, SUB:ML(Machine Learning)
        CLASS: BE
Out[2]: '\nNAME: MANE SHIVRAJ PANDURANG\nROLL NO.37\nCOURSE: AI&DS, SUB:ML(Machine Learning)\nCLASS: BE \n'
In [3]: '''
            PRACTICAL.NO:06
            Build a Tic-Tac-Toe game using reinforcement learning in Python by using following tasks
                a. Setting up the environment
                b. Defining the Tic-Tac-Toe game
                c. Building the reinforcement learning model
                d. Training the model
                e. Testing the model
Out[3]: '\n
                                     Build a Tic-Tac-Toe game using reinforcement learning in Python by using following tasks
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                  a. Setting up the environment\n
                                                         b. Defining the Tic-Tac-Toe game\n
                                                                                                   c. Building the reinforce
         \n
         ment learning model\n
                                      d. Training the model\n
                                                                     e. Testing the model\n'
In [4]: import numpy as np
In [5]: class TicTacToeEnvironment:
          def __init__(self):
           self.state = [0] * 9
           self.is_terminal = False
          def reset(self):
           self.state = [0] * 9
           self.is_terminal = False
          def get_available_moves(self):
          return [i for i, mark in enumerate(self.state) if mark == 0]
          def make_move(self, move, player_mark):
           self.state[move] = player_mark
          def check_win(self, player_mark):
            winning_states = [
                [0, 1, 2], [3, 4, 5], [6, 7, 8], # rows
                [0, 3, 6], [1, 4, 7], [2, 5, 8], # columns
                [0, 4, 8], [2, 4, 6] # diagonals
            for state_indices in winning_states:
              if all(self.state[i] == player_mark for i in state_indices):
                self.is terminal = True
                return True
            return False
          def is_draw(self):
            return 0 not in self.state
In [6]: class QLearningAgent:
          def __init__(self, learning_rate=0.9, discount_factor=0.9, exploration_rate=0.3):
            self.learning_rate = learning_rate
            self.discount_factor = discount_factor
            self.exploration rate = exploration rate
            self.q_table = np.zeros((3**9, 9))
          def get_state_index(self, state):
            state_index = 0
            for i, mark in enumerate(state):
              state_index += (3 ** i) * (mark + 1)
              return state index
          def choose_action(self, state, available_moves):
            state_index = self.get_state_index(state)
            if np.random.random() < self.exploration_rate:</pre>
              return np.random.choice(available_moves)
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return np.argmax(self.q_table[state_index, available_moves])
          def update_q_table(self, state, action, next_state, reward):
            state index = self.get state index(state)
            next_state_index = self.get_state_index(next_state) if next_state is not None else None
            max_q_value = np.max(self.q_table[next_state_index]) if next_state is not None else 0
            self.q_table[state_index, action] = (1 - self.learning_rate) * self.q_table[state_index, action] + \
            self.learning_rate * (reward + self.discount_factor * max_q_value)
In [7]: def evaluate_agents(agent1, agent2, num_episodes=1000):
          environment = TicTacToeEnvironment()
          agent1_wins = 0
          agent2_wins = 0
          draws = 0
          for _ in range(num_episodes):
            environment.reset()
            current_agent = agent1
            while not environment.is_terminal:
               available_moves = environment.get_available_moves()
               current_state = environment.state.copy()
               action = current_agent.choose_action(current_state, available_moves)
               environment.make_move(action, 1 if current_agent == agent1 else -1)
              if environment.check_win(1 if current_agent == agent1 else -1):
                current_agent.update_q_table(current_state, action, None, 10)
                if current_agent == agent1:
                  agent1_wins += 1
                else:
                  agent2_wins += 1
                  break
               elif environment.is_draw():
                \verb|current_agent.update_q_table(current_state, action, \verb|None|, 0|)|
                draws += 1
                break
               next_state = environment.state.copy()
               reward = 0
               if environment.check_win(1 if current_agent == agent1 else -1):
                reward = -10
                current_agent.update_q_table(current_state, action, next_state, reward)
                current_agent = agent2 if current_agent == agent1 else agent1
          return agent1_wins, agent2_wins, draws
In [8]: # Create agents
        agent1 = QLearningAgent()
        agent2 = QLearningAgent()
        # Evaluate agents
        agent1_wins, agent2_wins, draws = evaluate_agents(agent1, agent2)
        # Print results
        print(f"Agent 1 wins: {agent1_wins}")
        print(f"Agent 2 wins: {agent2_wins}")
        print(f"Draws: {draws}")
        # Agent 1 wins: 458
        # # Agent 2 wins: 470
        # Draws: 72
       Agent 1 wins: 1000
       Agent 2 wins: 0
       Draws: 0
```