## REINFORCEMENT LEARNING

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[1]: import math
     import random
     class UCBGame:
         def __init__(self, num_arms):
             Initialize the UCBGame instance.
             :param num_arms: Number of arms in the multi-armed bandit.
             self.num_arms = num_arms
             self.counts = [0] * num_arms # Number of times each arm has been played
             self.values = [0.0] * num_arms # Average reward for each arm
             self.total_plays = 0
         def select_arm(self):
             Selects the arm to play based on the Upper Confidence Bound (UCB)_{\sqcup}
      \hookrightarrow algorithm.
             if self.total_plays < self.num_arms:</pre>
                  # Play each arm at least once
                  return self.total_plays
             # Calculate UCB values for all arms
             ucb values = [
                  self.values[i] + math.sqrt((2 * math.log(self.total_plays)) / self.
      ⇔counts[i])
                  for i in range(self.num_arms)
             # Return the arm with the maximum UCB value
             return ucb_values.index(max(ucb_values))
         def update(self, chosen_arm, reward):
             Updates the counts and values for the chosen arm based on the received
      \hookrightarrow reward.
             :param chosen_arm: The arm that was chosen.
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:param reward: The reward received from the chosen arm.
       self.counts[chosen_arm] += 1
       self.total_plays += 1
       n = self.counts[chosen_arm]
       value = self.values[chosen_arm]
       # Update the value using incremental formula
       self.values[chosen_arm] = ((n - 1) / n) * value + (1 / n) * reward
def simulate_game(num_arms, num_rounds, reward_probabilities):
   Simulates a game where a player uses the UCB algorithm to maximize rewards.
    :param num arms: Number of actions (arms) available to the player.
    :param num_rounds: Total number of rounds to play.
    :param reward probabilities: List of probabilities for each arm to give a_{\sqcup}
    ⇔values for each arm.
   game = UCBGame(num_arms)
   total_rewards = 0
   for _ in range(num_rounds):
       chosen_arm = game.select_arm()
       # Simulate reward: 1 with given probability, 0 otherwise
       reward = 1 if random.random() < reward_probabilities[chosen_arm] else 0</pre>
       game.update(chosen_arm, reward)
       total_rewards += reward
   return total_rewards, game.counts, game.values
if __name__ == "__main__":
   # Define game parameters
   num_arms = 5
   num_rounds = 1000
   reward_probabilities = [0.1, 0.2, 0.3, 0.5, 0.8] # True probabilities of
 \rightarrowrewards
    # Run simulation
   total_rewards, counts, values = simulate_game(num_arms, num_rounds,_u
 →reward probabilities)
   # Output results
   print(f"Total rewards: {total_rewards}")
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print(f"Number of times each arm was played: {counts}")
print(f"Estimated values for each arm: {values}")
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Total rewards: 703

Number of times each arm was played: [19, 27, 55, 76, 823]

Estimated values for each arm: [0.05263157894736842, 0.18518518518518517, 0.3999999999999, 0.47368421052631576, 0.7764277035236941]

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[2]: import math
     import random
     class UCBOptimizer:
         def __init__(self, num_modes):
             Initializes the UCBOptimizer.
              :param num modes: Number of possible settings (modes) for the devices.
             self.num modes = num modes # Number of possible settings for devices
             self.counts = [0] * num_modes # Number of times each mode has been_
      \hookrightarrow selected
             self.values = [0.0] * num_modes # Average efficiency for each mode
             self.total selections = 0 # Total number of selections made
         def select_mode(self):
             Selects the mode to use based on the Upper Confidence Bound (UCB)_{\sqcup}
      \hookrightarrow algorithm.
             if self.total_selections < self.num_modes:</pre>
                  # Ensure each mode is selected at least once
                  return self.total_selections
              # Calculate UCB values for each mode
             ucb_values = [
                  self.values[i] + math.sqrt((2 * math.log(self.total_selections)) /__
      ⇒self.counts[i])
                  for i in range(self.num_modes)
             1
             # Return the mode with the highest UCB value
             return ucb_values.index(max(ucb_values))
         def update(self, chosen_mode, efficiency):
             Updates the counts and values for the chosen mode based on the observed,
      \hookrightarrow efficiency.
              :param chosen_mode: The mode that was chosen.
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:param efficiency: The observed efficiency value.
        self.counts[chosen_mode] += 1
        self.total_selections += 1
        n = self.counts[chosen_mode]
        value = self.values[chosen_mode]
        # Update the average efficiency using the incremental formula
        self.values[chosen_mode] = ((n - 1) / n) * value + (1 / n) * efficiency
def simulate_smart_home(num_modes, num_rounds, efficiency_factors):
    Simulates an IoT smart home system optimizing energy usage across multiple \Box
    :param num_modes: Number of settings (modes) available for devices.
    :param num_rounds: Total number of iterations for optimization.
    :param efficiency_factors: List of real efficiency probabilities for each ⊔
 ⇔mode.
    :return: Total efficiency score and mode selection statistics.
    optimizer = UCBOptimizer(num_modes)
    total_efficiency = 0
    for _ in range(num_rounds):
        chosen_mode = optimizer.select_mode()
        # Simulate efficiency: value based on real efficiency factor plusu
 ⇔random noise
        efficiency = efficiency factors[chosen_mode] + random.uniform(-0.05, 0.
 ⇔05)
        efficiency = max(0, min(1, efficiency)) # Clamp efficiency between O_{\square}
 \hookrightarrow and 1
        optimizer.update(chosen_mode, efficiency)
        total_efficiency += efficiency
    return total_efficiency, optimizer.counts, optimizer.values
if __name__ == "__main__":
    # Define simulation parameters
    num_modes = 4  # Example: Low, Medium, High, Auto settings for devices
    num_rounds = 500  # Number of rounds to simulate
    efficiency_factors = [0.7, 0.8, 0.9, 0.85] # Base efficiency for each mode
    # Run the simulation
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total_efficiency, counts, values = simulate_smart_home(num_modes,_u_num_rounds, efficiency_factors)

# Display results

print(f"Total efficiency score: {total_efficiency:.2f}")

print(f"Number of times each mode was selected: {counts}")

print(f"Estimated efficiencies for each mode: {values}")
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Total efficiency score: 421.84 Number of times each mode was selected: [62, 101, 202, 135] Estimated efficiencies for each mode: [0.7020655744272584, 0.8017331823941076, 0.9048560301754909, 0.8485900024313898]

```
[3]: import random
     import math
     class PACChessGame:
         def __init__(self, board_size, max_depth, epsilon, delta):
             Initialize the PAC Chess game.
             :param board size: Size of the chessboard.
             :param max_depth: Number of moves ahead to evaluate.
             :param epsilon: Error tolerance for PAC decision-making.
             :param delta: Confidence threshold for PAC decision-making.
            self.board_size = board_size
            self.max_depth = max_depth
            self.epsilon = epsilon
             self.delta = delta
             self.board = self.initialize_board()
         def initialize_board(self):
             """Initializes a simplified chessboard."""
             board = [[None for _ in range(self.board_size)] for _ in range(self.
      ⇔board_size)]
             # Place pawns for both players
             for i in range(self.board_size):
                 board[1][i] = 'P1' # Player 1 pawns
                 board[-2][i] = 'P2' # Player 2 pawns
             # Place rooks
            board[0][0], board[0][-1] = 'R1', 'R1' # Player 1 rooks
             board[-1][0], board[-1][-1] = 'R2', 'R2' # Player 2 rooks
             # Place kings
            board[0][self.board_size // 2] = 'K1' # Player 1 king
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board[-1][self.board_size // 2] = 'K2' # Player 2 king
      return board
  def possible_moves(self, player):
       """Generates all possible moves for the given player."""
      moves = []
      for row in range(self.board_size):
           for col in range(self.board_size):
               if self.board[row][col] and self.board[row][col].
⇔endswith(player):
                   # Example: Simplified pawn movement
                   if row + 1 < self.board_size:</pre>
                       moves.append(((row, col), (row + 1, col))) # Move_
\hookrightarrow forward
      return moves
  def evaluate_board(self):
       """Evaluates the board state for scoring."""
       score = 0
      for row in self.board:
           for piece in row:
               if piece == 'P1':
                   score += 1
               elif piece == 'P2':
                   score -= 1
               elif piece == 'K1':
                   score += 10
               elif piece == 'K2':
                   score -= 10
       return score
  def pac_decision(self, moves):
       """Selects the best move using PAC principles."""
       best_move = None
      best_score = float('-inf')
       # Calculate the number of samples needed for PAC guarantee
       samples = int((1 / (self.epsilon ** 2)) * math.log(1 / self.delta))
       for move in moves:
           total_score = 0
           for _ in range(samples):
               total_score += self.simulate_move(move)
           avg_score = total_score / samples
           if avg_score > best_score:
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best_score = avg_score
                best_move = move
        return best_move
   def simulate_move(self, move):
        """Simulates the effect of a move and evaluates the board."""
       start, end = move
       piece = self.board[start[0]][start[1]]
        # Make the move
       self.board[start[0]][start[1]] = None
       self.board[end[0]][end[1]] = piece
        # Evaluate the board
        score = self.evaluate_board()
        # Undo the move
        self.board[end[0]][end[1]] = None
        self.board[start[0]][start[1]] = piece
       return score
   def play_game(self):
        """Plays a game between two PAC-based players."""
       current_player = '1'
       for turn in range(50): # Maximum number of turns
            moves = self.possible_moves(current_player)
            if not moves:
                print(f"Player {current_player} has no moves left. Game over.")
                break
            chosen_move = self.pac_decision(moves)
            print(f"Player {current_player} chooses move {chosen_move}")
            start, end = chosen_move
            piece = self.board[start[0]][start[1]]
            self.board[start[0]][start[1]] = None
            self.board[end[0]][end[1]] = piece
            current_player = '2' if current_player == '1' else '1'
       print("Game ended.")
if __name__ == "__main__":
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game = PACChessGame(board_size=6, max_depth=3, epsilon=0.1, delta=0.05)
         game.play_game()
    Player 1 chooses move ((0, 0), (1, 0))
    Player 2 chooses move ((4, 3), (5, 3))
    Player 1 chooses move ((0, 3), (1, 3))
    Player 2 chooses move ((4, 0), (5, 0))
    Player 1 chooses move ((0, 5), (1, 5))
    Player 2 chooses move ((4, 1), (5, 1))
    Player 1 chooses move ((1, 0), (2, 0))
    Player 2 chooses move ((4, 2), (5, 2))
    Player 1 chooses move ((1, 1), (2, 1))
    Player 2 chooses move ((4, 4), (5, 4))
    Player 1 chooses move ((1, 2), (2, 2))
    Player 2 chooses move ((4, 5), (5, 5))
    Player 1 chooses move ((1, 3), (2, 3))
    Player 2 has no moves left. Game over.
    Game ended.
[]:
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# Define game parameters