

CSE623: Machine Learning Theory and Practice

Report-7

Group 1

| Name | Enr No. |
|----------------|-----------|
| Dhaivat Patel | AU2240022 |
| Dhyey Patel | AU2240054 |
| Tirthraj Raval | AU2240079 |
| Anusha Jain | AU2240092 |
| Sloka Thakkar | AU2240103 |

Work Done

Firstly, we defined a **custom environment class** LineupEnv to simulate lineup building:

- Maintains a list of available players and current lineup.
- Uses opposition lineup as fixed input.
- Calculates team and opponent stats dynamically as players are added or removed.
- Allows actions to add or remove a player.
- Limits the lineup to 5 players maximum.
- Computes a **reward only when the lineup is complete**, based on win probability predicted by a logistic regression model.

Developed the **Q-Learning algorithm**:

- Initializes a Q-table to store values for each (state, action) pair.
- Repeats simulation over **1000 episodes**.
- Uses ε -greedy policy for action selection:
 - With probability ε , selects a random action (exploration).
 - Otherwise, selects the best-known action (exploitation).
 - Updates Q-values using the Bellman equation:

$$Q(s,a) = (1-lpha) \cdot Q(s,a) + lpha \cdot (r + \gamma \cdot \max_a Q(s',a))$$

where:

- \bullet α = learning rate
- r = reward
- \bullet s, s' = current and next state

- \blacksquare a = action
- After training:
 - Starts from an empty lineup.
 - Chooses the best action at each step based on learned Q-values.
 - Builds the **optimal lineup**.
- Finally, outputs:
 - The best 5-player lineup.
 - The **predicted win probability** for that lineup.

```
opp_lineup = ['Alyssa Geary', 'A\']ah Davis', 'A\'Riana Gray', 'Player4', 'Player5'] # Example opposition lineup
```

Logistic Regression Model Accuracy: 0.9242

Best Lineup: ['Aaliyah Gayles', 'Aaliyah Frazier', "A'Riana Gray", "A'Jah Davis", "A'Moni Waiters"]

Predicted Win Probability: 0.5376