## CS8803-O03 Reinforcement learning Project 3 report

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## I. INTRODUCTION

Multi-agent Q-learning is a nontrivial departure from regular (deterministic action, single Q function) Q learning in two ways: (1) state values are functions of multiple Q's (2) optimal policies are allowed to be stochastic. The Greenwald paper [1] demonstrates the inadequacy of regular Q learning and convergence of multi-agent Q learning learning. Project 3 aims to deepen our understanding of multi-agent (adversarial or cooperative) Q learning through reproduction of Greenwald's results for a  $2 \times 4$  grid game called soccer.

## II. QUICK THEORY TOUR

The paper presents four multi-agent Q learning options:

1) **Regular:** Ignore the opponent's Q function and actions. Agents are only coupled through rewards. The policy is deterministic and Q function at each state for each agent is simply a vector of n values (n=5 in our case). The state value is the max for  $Q_i$ :

$$V_i = \max_{a \in A_i} Q_i(s, a) \tag{1}$$

2) **Friend Q:** Ignore the opponent's Q function, but consider its actions. Optimal policy is deterministic, and Q function at each stage is a  $n \times n$  matrix. The value for each state is the max across this matrix:

$$V_i = \max_{\vec{a} \in A_1 \times A_2} Q_i(s, \vec{a}) \tag{2}$$

3) Foe Q: Ignore opponent Q function, consider its actions, but calculate the value function using **maximin** instead of simple max. This requires linear programming (LP). Maximin also allows **stochastic** optimal policies represented as a probability distribution over n action values. The Q is an  $n \times n$  matrix and the state value is:

$$V_i = \max_{\pi \in PD(A)} \min_{o \in O} \sum_{a \in A} \pi_a Q_i(s, a, o)$$
 (3)

4) **CE Q:** Consider joint actions and compute state value as a function of agents' Q values

$$V_i = f_i(Q_1, Q_2) \tag{4}$$

where the functions  $f_i$  are linear combinations of  $Q_i$  and determined through linear programming.

## III. IMPLEMENTATION METHODOLOGY

- A. Overview
- B. Software dependencies

IV. EXPERIMENTS
SUMMARY
REFERENCES

[1] A. Greenwald and K. Hall, "Correlated Q learning," ICML, 2001.

