## Problem Set 1

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## 1 Definition of game environments

**Problem 1.** (15 pt) Given a two-player normal-form game  $(2, A_1 \times A_2, r_{\text{joint}})$  with player 2's policy  $\pi \in \Delta(A_2)$  fixed. Construct a single-agent environment from the player 1's point of view.

Hint: The model of the single agent environment is also called multi-armed bandit defined as (2, A, r), where the reduced action space A and the corresponding reward r should be defined by you.

**Problem 2.** (15 pt) Reformulate the team Markov game  $(N, H, S, s_0, A_{\text{joint}}, \mathbb{P}, r, \gamma)$ , where  $|S| < \infty$ ,  $|A_{\text{joint}}| < \infty$  into a indentical-interest normal-form game  $(N, A'_{\text{joint}}, r')$ . Show that the size of the joint action space  $|A'_{\text{joint}}|$  grows exponentially with respect to |S|.

**Problem 3.** (30 pt) Extend the definition of stochastic game into partially observable setting (usually called partially observable stochastic game), based on the definition you just created

- 1. Define policies and cumulative reward.
- 2. Define regrets for player 1 given other player's joint policies.
- 3. Reformulate the *partially observable stochastic game* into a single agent environment in player 1's perspective given other player's joint policies. The single agent environment is called partially observable Markov decision process.

## 2 Regret analysis

**Problem 4.** (15 pt) Construct an example using normal-form game that shows the regret minimization of player 1 returns a better strategy than the cumulative reward maximization.

Hint: Construct a two-player identical interest game. Carefully design the reward so the reward varies dramatically with respect to player 2's action.

## 3 Markov decision process

**Problem 5.** (25 pt) Prove that the optimal policy of an infinite  $(H = \infty)$  Markov decision process is stationary, i.e.,  $\mu^*: S \to \Delta(A)$  instead of  $\mu^*: S \times \mathbb{N}_+ \to \Delta(A)$ .