## Algorithmic game theory – Tutorial 5\*

November 28, 2022

## 1 Regret minimization

There are N available actions  $X = \{1, \dots, N\}$  and at each time step t the online algorithm A selects a probability distribution  $p^t = (p_1^t, \dots, p_N^t)$  over X. After the distribution  $p^t$  is chosen at time step t, the adversary chooses a loss vector  $\ell^t = (\ell_1^t, \dots, \ell_N^t) \in [-1, 1]^N$ , where the number  $\ell_i^t$  is the loss of action i in time t. The algorithm A then experiences loss  $\ell_A^t = \sum_{i=1}^N p_i^t \ell_i^t$ . After T steps, the loss of action i is  $L_i^T = \sum_{t=1}^T \ell_i^t$  and the loss of A is  $L_A^T = \sum_{t=1}^T \ell_A^t$ . The external regret of A is  $R_A^T = \max_{i \in X} \{L_A^T - L_i^T\}$ .

**Exercise 1.** Let A be an algorithm with parameter  $\eta \in (0, 1/2]$  and with external regret at most  $\alpha/\eta + \beta\eta T$  for some constants  $\alpha, \beta$  (that may depend on the number N of actions). We showed that choosing  $\eta = \sqrt{\alpha/(T\beta)}$  minimizes the bound. Modify this algorithm so that we obtain an external regret bound that is at most O(1)-times larger that the original bound for any T. In particular, you cannot run A with a parameter  $\eta$  that depends on T.

Hint: Partition the set  $\{1, ..., T\}$  into suitable intervals  $I_m$  for m = 0, 1, 2, ... and run A with a suitable parameter  $\eta_m$  in every step from  $I_m$ .

Given the sequence  $(p^t)_{t=1}^T$  of the probability distributions used by A and a modification rule F, we define a modified sequence  $(f^t)_{t=1}^T = (F^t(p^t))_{t=1}^T$ , where  $f^t = (f_t^t, \dots, f_N^t)$  and  $f_i^t = \sum_{j: F^t(j)=i} p_j^t$ . The loss of the modified sequence is  $L_{A,F}^T = \sum_{t=1}^T \sum_{i=1}^N f_i^t \ell_i^t$ . Given a sequence  $\ell^t$  of loss vectors, the regret of A with respect to F is  $R_{A,\mathcal{F}}^T = \max_{F \in \mathcal{F}} \left\{ L_A^T - L_{A,F}^T \right\}$ . The external regret of A is then  $R_{A,\mathcal{F}^{ex}}^T$  for  $\mathcal{F}^{ex} = \{F_i : i \in X\}$  of N modification rules  $F_i = (F_i^t)_{t=1}^T$ , where each  $F_i^t$  always outputs action i. The internal regret of A is  $R_{A,\mathcal{F}^{in}}^T$  for the set  $\mathcal{F}^{in} = \{F_{i,j} : (i,j) \in X \times X, i \neq j\}$  of N(N-1) modification rules  $F_{i,j} = (F_{i,j}^t)_{t=1}^T$ , where, for every time step t,  $F_{i,j}^t(i) = j$  and  $F_{i,j}^t(i') = i'$  for each  $i' \neq i$ . The swap regret of A is  $R_{A,\mathcal{F}^{sw}}^T$  for the set  $\mathcal{F}^{sw}$  of all modification rules  $F: X \to X$ .

**Exercise 2.** Show that the swap regret is at most N times larger than the internal regret.

**Exercise 3.** Show an example with N=3 where the external regret is zero and the swap regret goes to infinity with T.

Clarification: you need to choose only a sequence of actions  $a^1, \ldots, a^T$ ,  $a^i \in X = \{1, 2, 3\}$ , and a loss sequence  $\ell_a^1, \ldots, \ell_a^T$  for every  $a \in X$ .

For a normal-form game G=(P,A,C) of n players, a probability distribution p(a) on A is a correlated equilibrium in G if  $\sum_{a_{-i}\in A_{-i}}C_i(a_i;a_{-i})p(a_i;a_{-i})\leq \sum_{a_{-i}\in A_{-i}}C_i(a_i';a_{-i})p(a_i;a_{-i})$  for every player  $i\in P$  and all  $a_i,a_i'\in A_i$ . A probability distribution p(a) on A is a coarse correlated equilibrium in G if  $\sum_{a\in A}C_i(a)p(a)\leq \sum_{a\in A}C_i(a_i';a_{-i})p(a)$  for every player  $i\in P$  and every  $a_i'\in A_i$ .

Exercise 4. Show formally that every correlated equilibrium is a coarse correlated equilibrium.

<sup>\*</sup>Information about the course can be found at http://kam.mff.cuni.cz/~balko/