

## Algorithmic Game Theory

Winter Term 2020/21

### Exercise Set 4

**Exercise 1:** (4+4 Points)

Consider the following regret-minimization-algorithm.

GREEDY

- Set  $p_1^1 = 1$  and  $p_j^1 = 0$  for all  $j \neq 1$ .

- In each round  $t = 1, \dots, T$ :

Let  $L_{min}^t = \min_{i \in N} L_i^t$  and  $S^t = \{i \in N \mid L_i^t = L_{min}^t\}$ .

Set  $p_i^{t+1} = 1$  for  $i = \min S^t$  and  $p_j^{t+1} = 0$  otherwise.

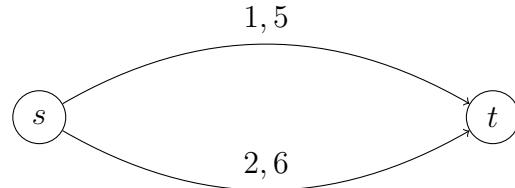
- Show that the costs of GREEDY are at most  $N \cdot L_{min}^T + (N - 1)$ .
- State a scheme for an example such that the stated upper bound of (a) is tight for an infinite number of values  $T$ .

**Exercise 2:** (1+3+2 Points)

Referring to the price of anarchy from Lecture 8 we can introduce a more optimistic point of view called the *price of stability*. For an equilibrium concept  $\mathbf{Eq}$ , it is defined as

$$PoS_{\mathbf{Eq}} = \frac{\min_{p \in \mathbf{Eq}} SC(p)}{\min_{s \in S} SC(s)} .$$

Consider the following symmetric network congestion game with two players:



- What is the Price of Anarchy and the Price of Stability of pure Nash equilibria?

- What is the Price of Anarchy and the Price of Stability of mixed Nash equilibria?

**Hint:** First of all, determine all mixed Nash equilibria. You might start with a sentence like “Let  $\sigma$  be a mixed Nash equilibrium with  $\sigma_1 = (\lambda_1, 1 - \lambda_1)$ ,  $\sigma_2 = (\lambda_2, 1 - \lambda_2)$ ” and subsequently derive properties of  $\lambda_1$  and  $\lambda_2$ .

- What is the best upper bound for the Price of Anarchy that can be shown via smoothness?

**Exercise 3:**

(4 Points)

Consider a  $(\lambda, \mu)$ -smooth game with  $N$  players and let  $s^{(1)}, \dots, s^{(T)}$  be a sequence of states such that the external regret of every player is at most  $R^{(T)}$ . Moreover, let  $s^*$  denote a state that minimizes the social cost. We want to upper bound the average social cost of the sequence of states. To this end, show the following bound

$$\frac{1}{T} \sum_{t=1}^T SC(s^{(t)}) \leq \frac{N \cdot R^{(T)}}{(1 - \mu)T} + \frac{\lambda}{1 - \mu} SC(s^*).$$