
Online and approximation algorithms

Due June 18, 2014 before class!

Exercise 1 (EXPO - 10 points)

Recall the algorithm *EXPO* from the lecture. It can be modified to work even without knowledge about ϕ . Let $\mu = \{q(i)\}_{i=0}^{\infty}$ be a probability distribution over the natural numbers (i.e. i is chosen with probability $q(i)$).

EXPO $_{\mu}$ chooses the reservation price $p_1 2^i$ with probability $q(i)$ for $i = 0, 1, \dots$ where p_1 is the first price that is revealed.

Prove that *EXPO* $_{\mu}$ is $\frac{2}{q(\lfloor \log \phi \rfloor)}$ -competitive against an oblivious adversary, where ϕ is the posteriori global fluctuation ratio.

Exercise 2 (EXPO II - 10 points)

Extend algorithm *EXPO* and its analysis to the case in which ϕ is not a power of 2.

Exercise 3 (k-Server - 10 points)

Recall the k-server problem from the lecture. Algorithm *Greedy* always finds the nearest server to each request and moves it to the request location.

- (a) Show that *Greedy* is not competitive against an oblivious adversary.
- (b) Show that metrical task systems generalize the k-server problem in finite spaces.

Exercise 4 (Lower envelope - 10 points)

Recall the Lower Envelope Algorithm (LEA) from the lecture. In the lecture we showed, that the algorithm is $3 - 2\sqrt{2}$ competitive for general state systems.

We consider the special case where the costs for powering down are additive, i.e. for $i < j < k$, powering down from state s_i to s_j and from s_j to s_k is equally expensive as powering down from s_i to s_k .

Prove that in this setting LEA is 2-competitive.

Hint: Fold the cost for powering down into the cost for powering up. This yields a system where powering down is free and only transitions from low-power states to higher-power states create costs.

Exercise 5 (Energy efficiency - 10 points)

Recall the Energy-efficiency problem from the lecture. In the lecture we showed, that given S, R and D , an online algorithm with the competitive ratio $c^* + \epsilon$ can be constructed in exponential time, where c^* is the optimal competitive ratio possible for this system.

The proof in the lecture introduced the notion of eagerness, now we add the notion of earliness. We define $E(s, p)$ as the earliest transition time at which any online algorithm can transition to state s while still being p -eager (e.g. $E(s_0, p) = 0$). A transition to state s that happens at time $E(s, p)$ is called p -early. A p -early strategy consists exclusively of p -early transitions.

Use dynamic programming to define a decision procedure *EXISTS* that takes S, R, D as well as a constant p and determines whether a p -competitive strategy for this system exists or not. Show that your procedure runs in polynomial time.

Hint: Without proof use the following lemma: If there is a p -competitive strategy A , then there exists a p -eager and p -early p -competitive strategy.