

## Algorithms and Uncertainty

Summer Term 2020

### Exercise Set 4

*Due to the public holiday (Christi Himmelfahrt) on Thursday, May 21, there will be no tutorial sessions next week. As a consequence, this sheet is only due in two weeks, but also covers content from the lecture on Monday, May 18 (exercises 4 + 5). We will discuss this sheet in the tutorials on May 28.*

**Exercise 1:** (4 Points)

Let us consider a generalization of the version of Markov decision processes covered in the lecture. For every state  $s \in \mathcal{S}$ , only a subset of the actions  $\mathcal{A}_s \subseteq \mathcal{A}$ ,  $\mathcal{A}_s \neq \emptyset$ , is available. Devise an algorithm that computes an optimal policy for a finite time horizon  $T$ , show its correctness, and give a bound on its running time.

**Exercise 2:** (1+5 Points)

We consider a stochastic decision problem similar to the one with the envelopes we solved in class. There are  $n$  boxes; box  $i$  contains a prize of 1 Euro with probability  $q_i$  and is empty otherwise. The game ends when we have found a non-empty box. That is, the final prize is either 0 Euros or 1 Euro. At each point in time, we can also decide to stop playing. We can open as many boxes as we like but opening box  $i$  costs  $c_i$  Euros.

- (a) Model this problem as a Markov decision process. In particular, give the state and action sets as well as transition probabilities and rewards.
- (b) Find an optimal policy.

**Hint:** It can be useful to consider the cases  $n = 1$  and  $n = 2$  first.

**Exercise 3:** (2 Points)

Consider the cost-minimization variant of the optimal stopping problem. In step  $i$ , we can stop the sequence at cost  $c_i$ . We have to stop the sequence at some point and want to minimize the cost for doing so.

Show that there is **no**  $\alpha < \infty$  such that for all distributions the optimal policy has cost at most  $\alpha \mathbf{E}[\min_i c_i]$ .

**Hint:** It suffices to consider  $n = 2$ .

**Exercise 4:** (2 Points)

Consider the following distribution for the prize of box  $i$ : the prize  $v_i$  is equal to  $w_i$  with probability  $q_i$  and is 0 else. Compute the fair cap.

**Exercise 5:**

(3 Points)

In order to generalize the Pandora's Box setup from the lecture, suppose we would like to match people  $i \in [n]$  to boxes  $j \in [m]$  (each person can take at most one prize home). We know that person  $i$ 's value  $v_{ij}$  for the prize in box  $j$  is independently drawn from a distribution  $\mathcal{D}_{ij}$ , but it costs  $c_{ij}$  to inspect the exact value of the box  $v_{ij}$ . Consider  $A_{ij}$ ,  $I_{ij}$ ,  $\sigma_{ij}$ ,  $\kappa_{ij}$  and  $b_{ij}$  to be the corresponding generalizations of the variables introduced in the lecture.

Show that for any policy  $\pi$ , the expected value is given by

$$V(\pi) = \sum_{i,j} \mathbf{E} [A_{ij}\kappa_{ij} - (I_{ij} - A_{ij})b_{ij}] .$$