

## Algorithms and Uncertainty

Winter Term 2023/24

### Exercise Set 10

*If you want to hand in your solutions for this problem set, please send them via email to [anna.heuser@uni-bonn.de](mailto:anna.heuser@uni-bonn.de) by Monday evening – make sure to send a pdf-file which contains your name and your email address. Of course, submitting solutions in groups is also possible.*

*If you would like to present one of the solutions in class, please also send an email to [anna.heuser@uni-bonn.de](mailto:anna.heuser@uni-bonn.de) containing the **task** which you would like to present and in **which of the tutorials** you would like to do so. Deadline for the email is Monday, 10:00 pm. Please note that the tasks will be allocated via a first-come-first-served procedure, so sending this email earlier than Monday evening is highly recommended.*

**Exercise 1:** (4 Points)

Consider the following explore-exploit algorithm. In the first  $\frac{T}{2}$  steps (so  $k = \frac{T}{2n}$ ), we explore. Afterwards, we exploit the most promising arm. Use the approach from Lecture 17 to derive an upper-bound for the expected regret of this algorithm.

**Exercise 2:** (4 Points)

Consider the modified update rule for Multiplicative Weights that sets  $w_i^{(t+1)} = w_i^{(t)} \cdot (1 - \ell_i^{(t)} \eta)$ . Show that Theorem 18.3 still holds.

**Exercise 3:** (4 Points)

We consider a generalization of the algorithm *Weighted Majority* for classifiers with  $k$  different classes. (The case covered in the lecture, binary classification, is  $k = 2$ .) In each step, the algorithm chooses a class, which is recommended by the largest number of classifiers (so the class has a plurality).

Show that this algorithm makes at most  $(2 + 2\eta) \min m_i + 2 \ln n / \eta$  errors, where  $m_i$  is the number of errors of classifier  $i$ .