

## Algorithms and Uncertainty

Summer Term 2021

Exercise Set 10

### Exercise 1:

(3 Points)

Consider  $f : \mathbb{R}^2 \rightarrow \mathbb{R} : \mathbf{x} \mapsto \mathbf{x}^T \mathbf{x}$ . Show that  $f$  is convex.

### Exercise 2:

(3 Points)

Let  $\mathcal{X}$  be a convex set. Prove the following statement: If a function  $f : \mathcal{X} \rightarrow \mathbb{R}$  is convex, then any local minimum of  $f$  in  $\mathcal{X}$  is also a global minimum.

### Exercise 3:

(5 Points)

Show that Follow-the-Regularized-Leader with Entropical regularization in the experts setting is equivalent to the Multiplicative Weights algorithm.

*Hint:* It can be helpful to use a Lagrange multiplier, which works in this special case as follows: For  $\mathbf{x}$  to be a local optimum of  $F$  subject to  $\sum_{i=1}^d x_i = 1$ , it is necessary that there exists a  $\lambda \in \mathbb{R}$  such that  $\frac{\partial F}{\partial x_i}(\mathbf{x}) - \lambda = 0$  for all  $i$ .