

**Machine Learning (911.236)**

## Exercise sheet C

**Exercise 1.**

5 P.

In the proof of the No-Free-Lunch (NFL) theorem, we started with a subset  $C \subset \mathcal{X}$  of size  $2m$  and studied what happens if our learning algorithm only sees training samples of size  $m$ . Now, assume that  $C$  is of size  $km$  with  $k \geq 2$  and we only observe  $|C|/k$  samples. Proof that the original lower bound of  $1/4$  changes to  $1/2 - 1/2k$ . This means that there exists a distribution  $\mathcal{D}$  over  $\mathcal{X} \times \{0, 1\}$  such that (1) there exists a function  $f : \mathcal{X} \rightarrow \{0, 1\}$  with  $L_{\mathcal{D}}(f) = 0$  and (2)  $\mathbb{E}_{S \sim \mathcal{D}^m} [L_{\mathcal{D}}(A(S))] \geq 1/2 - 1/2k$ . Interpret this result.

*Hint:* You need to go through the proof of the NFL theorem and find the appropriate place where to inject these changes.

**Exercise 2.**

4 P.

Assume that someone gives you a coin and he/she tells you that the coin is *biased*. How many flips of the coin do you need to decide the direction of the bias? Consider the outcome of the coin flip as a sequence of independent Bernoulli random variables  $X_1, \dots, X_n$  with success probability  $p$  (say success means “heads” equals 1) and you estimate  $p$  with the empirical average  $\hat{p}_n = 1/n \sum_i X_i$ . If  $\hat{p}_n \geq 1/2$  you decide a bias towards “heads”, and “tails” otherwise. Show that if

$$n > \frac{1}{2\epsilon^2} \log \left( \frac{1}{\delta} \right), \quad (1)$$

then you decide correctly with probability  $1 - \delta$  for  $\delta \in (0, 1)$  fixed. Remember that if  $p = 1/2 - \epsilon$ ,  $\epsilon > 0$ , you make an error if  $\hat{p}_n \geq 1/2$  (and vice versa).