

Machine Learning (911.236)

Exercise sheet C

Exercise 1.

5 P.

In the NFL lecture, we enumerated all $k = (2m)^m$ possible training sets of size m (out of the set C of size $2m$), i.e., S_1, \dots, S_k . If labeled by f_i , we added the superscript i , e.g., S_1^i, \dots, S_k^i . Later in the proof of the NFL theorem, we wrote

$$\mathbb{E}_{S \sim \mathcal{D}_i^m} [L_{\mathcal{D}_i}(A(S))] = \frac{1}{k} \sum_{j=1}^k L_{\mathcal{D}_i}(A(S_j^i))$$

for the expected generalization error wrt. \mathcal{D}_i . As an argument for why we can actually do this, we remarked that “...all sets are equally likely” (meaning the sets S_1^i, \dots, S_k^i are equally likely to be drawn). *Where do we actually see this? Provide a (formal) argument, remembering how we constructed the \mathcal{D}_i ’s.*

Exercise 2.

5 P.

In the NFL theorem from the lecture, we assumed $m < |X|/2$. If one would now assume $|X| \geq km$ with integer $k \geq 2$ and let $m < |X|/k$, we would get

$$\mathbb{E}_{S \sim \mathcal{D}^m} [L_{\mathcal{D}}(A(S))] \geq \frac{1}{2} - \frac{1}{2k},$$

i.e., the lower bound of $1/4$ is replaced by $1/2 - 1/2k$.

In the NFL proof, which part has to be modified (and how) to show this? What happens as k gets large?