

Neural Networks and Learning Theory

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Exercise Sheet 5

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Exercise 1.

Let $X := \mathbb{R}$ and the concept space be the set of all positive “half-lines”, i.e. $C := \{[a, \infty) | a \in \mathbb{R}\}$. Consider the algorithm that first, after seeing a training set S which contains m labeled examples of the form $(x_i, c(x_i))$ where $x_i \in \mathbb{R}$, define $(x_r := \max_{i, c(x_i)=0} x_i)$ and $(x_l := \max_{i, c(x_i)=1} x_i)$. The algorithm outputs an hypothesis $h = [a, \infty)$, where a is a real number arbitrarily selected from the open interval (x_r, x_l) . Prove that this algorithm satisfies the requirements of PAC learning and therefore proves that C is learnable by $H := C$.

Exercise 2.

PAC learning of hyper-rectangles: An axis-parallel hyper-rectangle in \mathbb{R}^n is a set of the form $[a_1, b_1] \times \dots \times [a_n, b_n]$. Show that axis-parallel hyper-rectangles are PAC-learnable by extending the proof given in the lecture for the case $n = 2$.

Exercise 3.

Concentric circles: Let $X := \mathbb{R}^2$ and consider the set of concepts of the form $c = \{(x, y) | x^2 + y^2 \leq r^2\}$ for some real number r . Show that this class can be PAC-learned from training data of size $m \geq (1/\epsilon) \log(1/\delta)$.

Exercise 4.

Given any set T of $n + 2$ points in \mathbb{R}^n . Show that there is a partition T_1, T_2 of T such that $\text{convhull}(T_1) \cap \text{convhull}(T_2) \neq \emptyset$.

Exercise 5.

Consider the sample space $X := \mathbb{R}$. What is the VC-dimension of

- (a) $H_1 := \{(a, \infty) | a \in \mathbb{R}\}$ and
- (b) $H_2 := \{(-\infty, a) \cup (a, \infty) | a \in \mathbb{R}\}$.