Assignment 1

Deadline: Friday, 22^{nd} of April 2022

Upload your solutions at: https://tinyurl.com/AML-2022-ASSIGNMENT1

- 1. (0.5 points) Give an example of a finite hypothesis class \mathcal{H} with $VCdim(\mathcal{H}) = 2022$. Justify your choice.
- 2. (0.5 points) What is the maximum value of the natural even number n, n = 2m, such that there exists a hypothesis class \mathcal{H} with n elements that shatters a set C of $m = \frac{n}{2}$ points? Give an example of such an \mathcal{H} and C. Justify your answer.
- 3. (0.75 points) Let $\mathcal{X} = \mathbb{R}^2$ and consider \mathcal{H} the set of axis aligned rectangles with the center in origin O(0, 0). Compute the $VCdim(\mathcal{H})$.
- 4. (1 point) Let $\mathcal{X} = \mathbb{R}^2$ and consider \mathcal{H}_{α} the set of concepts defined by the area inside a right triangle ABC with two catheti AB and AC parallel to the axes (Ox and Oy), and with the ratio AB/AC = α (fixed constant > 0). Consider the realizability assumption. Show that the class \mathcal{H}_{α} is (ϵ, δ) -PAC learnable by giving an algorithm A and determining an upper bound on the sample complexity $m_H(\epsilon, \delta)$ such that the definition of PAC-learnability is satisfied.
 - 5. (1.25 points) Consider $\mathcal{H} = \mathcal{H}_1 \cup \mathcal{H}_2 \cup \mathcal{H}_3$, where:

$$\mathcal{H}_{1} = \{h_{\theta_{1}} : \mathbb{R} \to \{0,1\} \mid h_{\theta_{1}}(x) = \mathbf{1}_{[x \geq \theta_{1}]}(x) = \mathbf{1}_{[\theta_{1},+\infty)}(x), \theta_{1} \in \mathbb{R} \},$$

$$\mathcal{H}_{2} = \{h_{\theta_{2}} : \mathbb{R} \to \{0,1\} \mid h_{\theta_{2}}(x) = \mathbf{1}_{[x < \theta_{2}]}(x) = \mathbf{1}_{(-\infty,\theta_{2})}(x), \theta_{2} \in \mathbb{R} \},$$

$$\mathcal{H}_{3} = \{h_{\theta_{1},\theta_{2}} : \mathbb{R} \to \{0,1\} \mid h_{\theta_{1},\theta_{2}}(x) = \mathbf{1}_{[\theta_{1} \leq x \leq \theta_{2}]}(x) = \mathbf{1}_{[\theta_{1},\theta_{2}]}(x), \theta_{1}, \theta_{2} \in \mathbb{R} \}.$$

Consider the realizability assumption.

- a) Compute $VCdim(\mathcal{H})$.
- b) Show that \mathcal{H} is PAC-learnable.
- c) Give an algorithm A and determine an upper bound on the sample complexity $m_{\mathcal{H}}(\epsilon, \delta)$ such that the definition of PAC-learnability is satisfied.

6. (1 point) A decision list may be thought of as an ordered sequence of if-then-else statements. The sequence of conditions in the decision list is tested in order, and the answer associated with the first satisfied condition is output.

More formally, a k-decision list over the boolean variables x_1, x_2, \ldots, x_n is an ordered sequence $L = \{(c_1, b_1), (c_2, b_2), \ldots, (c_l, b_l)\}$ and a bit b, in which each c_i is a conjunction of at most k literals over x_1, x_2, \ldots, x_n and each $b_i \in \{0, 1\}$. For any input $a \in \{0, 1\}^n$, the value L(a) is defined to be b_j where j is the smallest index satisfying $c_j(a) = 1$; if no such index exists, then L(a) = b. Thus, b is the "default" value in case a falls off the end of the list. We call b_i the bit associated with the condition c_i .

The next figure shows an example of a 2-decision list along with its evaluation on a particular input.

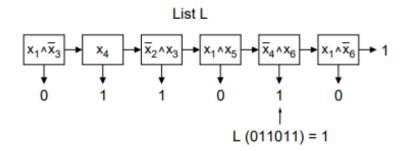


Figure 1: A 2-decision list and the path followed by an input. Evaluation starts at the leftmost item and continues to the right until the first condition is satisfied, at which point the binary value below becomes the final result of the evaluation.

Show that the VC dimension of 1-decision lists over $\{0,1\}^n$ is lower and upper bounded by linear functions, by showing that there exists $\alpha, \beta, \gamma, \delta \in \mathbb{R}$ such that:

$$\alpha \cdot n + \beta \leq VCdim(\mathcal{H}_{1-decision\ list}) \leq \gamma \cdot n + \delta$$

Hint: Show that 1-decision lists over $\{0,1\}^n$ compute linearly separable functions (halfspaces).

Ex-officio: 0.5 points