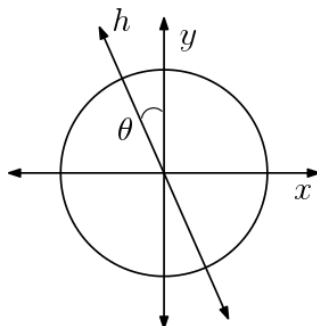


# Homework 1

Christopher Mertin  
CS6966: Theory of Machine Learning

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1. Consider the problem of classifying points in the two-dimensional plane, *i.e.*,  $\chi = \mathbb{R}^2$ . Suppose that the (unknown) true label of a point  $(x, y)$  is given by  $\text{sign}(x)$  (we define  $\text{sign}(0) = \pm 1$  for convenience). Suppose the input distribution  $\mathcal{D}$  is the uniform distribution over the unit circle centered at the origin.
  - (a) Consider the hypothesis  $h$  as shown in the figure below ( $h$  classifies all the points on the right of the line as  $+1$  and all the points to the left as  $-1$ ). Compute the risk  $L_{\mathcal{D}}(h)$ , as a function of  $\theta$  (which is, as is standard, given in radians).



**Solution:**

$$Area_{circle} = \pi r^2$$

$$Area_{slice} = \pi r^2 \frac{\theta}{2\pi} = \frac{\theta}{2} r^2$$

For a unit circle, the failure percentage for something *completely* random on the unit circle ( $r = 1$ ) would be

$$\mathcal{L}_{\mathcal{D}}(h) = \frac{\theta r^2}{\pi r^2} = \frac{\theta}{\pi}$$

- (b) Suppose we obtain  $1/\theta$  (which is given to be an integer  $\geq 2$ ) training samples (*i.e.*, samples from  $\mathcal{D}$ , along with their true labels). What is the probability that we

find a point whose label is “inconsistent” with  $h$ ? Can you bound this probability by a constant independent of  $\theta$ ?

**Solution:**

- (c) Give an example of a distribution  $\mathcal{D}$  under which  $h$  has risk zero.

**Solution:**

$$\text{Distribution} = \begin{cases} y = 1 & x \geq 0 \\ y = -1 & x < 0 \end{cases}$$

2. Suppose  $A_1, A_2, \dots, A_n$  are events in a probability space.

- (a) Suppose  $\Pr[A_i] = \frac{1}{2n}$  for all  $i$ . Then, show that the probability that none of the  $A_i$ 's occur is at least  $1/2$ .

**Solution:**

$$P(A_i) = \frac{1}{2n}$$

If independent, the probability of choosing all of them is

$$P(A_{all}) = \left(\frac{1}{2n}\right) \left(\frac{1}{2n}\right) \cdots \left(\frac{1}{2n}\right)$$

$$P(A_{all}) = \prod_{i=1}^n \left(\frac{1}{2n}\right) = \left(\frac{1}{2n}\right)^n$$

Therefore, the probability of not choosing any

$$P(A_{none}) = 1 - \left(\frac{1}{2n}\right)^n$$

We can place a lower bound for  $n = 1$

$$P(A_1) = 1 - \frac{1}{2} = \frac{1}{2}$$

Therefore, the probability of not being chosen for a given  $n$

$$P(A_i) = 1 - \left(\frac{1}{2n}\right)^n \geq \frac{1}{2}$$

- (b) Give a concrete example of events  $A_i$  for which  $\Pr[A_i] = \frac{1}{n-1}$  for all  $i$ , and the probability that none of them occur is zero.

**Solution:**

Probability of not choosing an integer at random on the infinite domain.

- (c) Suppose  $n \geq 3$ , and  $\Pr[A_i] = \frac{1}{n-1}$ , but the events are all *independent*. Show that the probability that none of them occur is  $\geq 1/8$ .

**Solution:**

Probability of not choosing one

$$P(A_i) = \left(1 - \frac{1}{n-1}\right)$$

probability of not choosing  $n$  independent

$$P(A_1, A_2, \dots, A_n) = \left(1 - \frac{1}{n-1}\right)^n = \left(\frac{2-n}{n-1}\right)^n$$

We can bound it by using  $n = 3$ , which gives

$$P(A_1, A_2, A_3) = \frac{1}{8}$$

3. In our proof of the no-free lunch theorem, we assumed the algorithm  $A$  to be deterministic. Let us now see how to allow randomized algorithms. Let  $A$  be a randomized map from set  $X$  to set  $Y$ . Formally, this means that for every  $x \in X$ ,  $A(x)$  is a random variable, that takes values in  $Y$ . Suppose  $|X| < c|Y|$ , for some constant  $c < 1$ .

- (a) Show that there exists  $y \in Y$  such that  $\max_{x \in X} \Pr[A(x) = y] \leq c$ .

**Solution:**

- (b) Show that this implies that for any distribution  $\mathcal{D}$  over  $X$ ,  $\Pr_{x \sim \mathcal{D}}(A(x) = y) \leq c$ .

**Solution:**

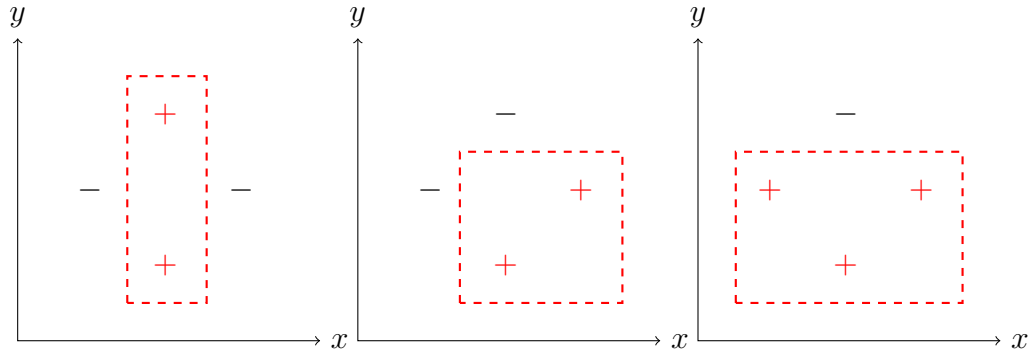
4. Recall that the VC dimension of a hypothesis class  $\mathcal{H}$  is the size of the largest set that it can “shatter.”

- (a) Consider the task of classifying points on a 2D plane, and let  $\mathcal{H}$  be the class of axis parallel rectangles (points inside the rectangle are “+” and the points outside are “−”). Prove that the VC dimension of  $\mathcal{H}$  is 4.

In order to prove VC dimension, we need to prove two things

- i. There exists ( $d = 4$ ) points which can be shattered

**Solution:**

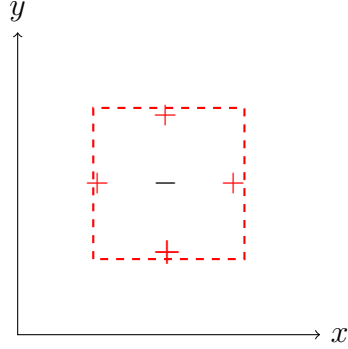


With the last case of all 4 being positive being trivial. These are also trivially applicable when flipping between  $+$ 's and  $-$ 's.

- ii. No set of 5 points can be shattered

**Solution:**

The minimum enclosing rectangle is defined where 1 point is each edge



Therefore, the 5<sup>th</sup> point must lie on the edge or inside of the rectangle.

- (b) This time, let  $\chi = \mathbb{R}^d \setminus \{0\}$  (origin excluded), and let  $\mathcal{H}$  be the set of all hyperplanes through the origin (points on one side are “ $+$ ” and the other side are “ $-$ ”). Prove that the VC dimension of  $\mathcal{H}$  is  $\leq d$ .

*Hint:* Consider *any* set of  $d+1$  points. They need to be linearly dependent. Now, could it happen that  $u, v$  are “ $+$ ”, but  $\alpha u + \beta v$  is “ $-$ ” for  $\alpha, \beta \geq 0$ ? Can you generalize this?

**Solution:**

Consider  $d$  unit base vectors in  $\mathbb{R}^d$

$$(1, 0, \dots, 0), (0, 1, 0, \dots, 0), \dots, (0, \dots, 0, 1)$$

It is trivially seen that this set can be defined by  $d$  hyper planes through the origin. We can now show that there are no  $d+1$  vectors in  $\mathbb{R}^d$  that can be shattered by hyperplanes through the origin. We can do this by *proof by contradiction*.

Suppose that  $u_1, \dots, u_{d+1}$  can be shattered. This implies that there exists  $2^{d+1}$  vectors  $a_i \in \mathbb{R}^d, i = \{1, \dots, 2^{d+1}\}$  such that the matrix of inner products, denoted by  $z_{i,j} = u_i^T a_j$  has columns with all possible combination of signs. Therefore, we have the matrix inner products of

$$A = \begin{pmatrix} z_{1,1} & \cdots & z_{1,2^{d+1}} \\ \vdots & \ddots & \vdots \\ z_{(d+1),1} & \cdots & z_{(d+1),2^{d+1}} \end{pmatrix}$$

which has all  $2^{d+1}$  possible combinations of signs.

$$\text{sign}(A) = \begin{pmatrix} - & - & \cdots & - & + \\ - & \cdot & \cdots & \cdot & + \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ - & + & \cdots & \cdot & + \end{pmatrix}$$

Then, the rows of  $A$  are linearly independent as there are no constants  $c$  such that  $\sum_{i=1}^{d+1} c_i z_{i,v} = 0$  as for any value of  $c_i$  there is a column with the same sign, which makes it always non-zero. This implies that  $d+1$  vectors in  $\mathbb{R}^d$  are linearly independent but it is a false statement. This contradiction proves there are no  $d+1$  vectors in  $\mathbb{R}^d$  that can be shattered by hyperplanes through the origin. Thus, the VC dimension is  $d$

- (c) **(BONUS)** Let  $\chi$  be the points on the real line, and let  $\mathcal{H}$  be the class of hypotheses of the form  $\text{sign}(p(x))$ , where  $p(x)$  is a polynomial of degree at most  $d$  (for convenience, define  $\text{sign}(0) = +1$ ). Prove that the VC dimension of this class is  $d+1$ .

*Hint:* The tricky part is the upper bound. Here, suppose  $d = 2$ , and suppose we consider any four points  $x_1 < x_2 < x_3 < x_4$ . Can the sign pattern  $+, -, +, -$  arise from a degree 2 polynomial?

5. In the examples above (and in general), a good rule of thumb for VC dimension of a function class is the *number of parameters* involved in defining a function in that class. However, this is not universally true, as illustrated in this problem: Let  $\chi$  be the points on the real line, and define  $\mathcal{H}$  to be the class of functions of the form  $h_\theta := \text{sign}(\sin(\theta x))$ , for  $\theta \in \mathbb{R}$ . Note that each hypothesis is defined by the single parameter  $\theta$ .

Prove that the VC dimension of  $\mathcal{H}$  is infinity.

**Solution:**

So where does the “complexity” of the function class come from? **(BONUS)** Prove that if we restrict  $\theta$  to be a rational number whose numerator and denominator have at most  $n$  bits, then the VC dimension is  $\mathcal{O}(n)$ .