# Machine Learning, 2024 Spring Assignment 5

## **Notice**

Plagiarizer will get 0 points.

LATEX is highly recommended. Otherwise you should write as legibly as possible.

## Problem 1

(15 points) Which of the following are possible growth functions  $m_{\mathcal{H}}(N)$  for some hypothesis set:

$$1 + N; 1 + N + \tfrac{N(N-1)}{2}; 2^N; 2^{\lfloor \sqrt{N} \rfloor}; 2^{\lfloor N/2 \rfloor}; 1 + N + \tfrac{N(N-1)(N-2)}{6}.$$

#### Solution:

It is known that  $m_{\mathcal{H}}(N)$  is either equal to  $2^N$  or has a polynomial upper bound. Therefore, all the functions except  $2^{\lfloor \sqrt{N} \rfloor}, 2^{\lfloor N/2 \rfloor}$  are possible growth functions  $m_{\mathcal{H}}(N)$ .

# Problem 2

(15 points)For an  $\mathcal{H}$  with  $d_{vc}=10$ , what sample size do you need (as prescribed by the generalization bound) to have a 95% confidence that your generalization error is at most 0.05?

#### Solution:

From 
$$E_{\mathrm{out}}\left(g\right) \leq E_{\mathrm{in}}\left(g\right) + \sqrt{\frac{8}{N}\ln\left(\frac{4\left((2N)^{d_{\mathrm{vc}}}+1\right)}{\delta}\right)}$$
, we can get that

$$\sqrt{\frac{8}{N}\ln\left(\frac{4\left(\left(2N\right)^{d_{\text{vc}}}+1\right)}{\delta}\right)} \le 0.05$$

By solving the inequality, we can get the smallest N=452957.

### Problem 3

(15 points)Let  $\mathcal{H} = \{h_1, h_2, \dots, h_M\}$  with some finite M. Prove that  $d_{vc}(\mathcal{H}) \leq \log_2 M$ .

#### Solution:

Since there are M hypotheses, a total of M scenarios can be distinguished.  $d_{\rm vc}(\mathcal{H})$  means that for  $n=d_{\rm vc}(\mathcal{H})$  sets of data, these M hypotheses can distinguish all  $2^{d_{\rm vc}(\mathcal{H})}$  cases, and at most M cases in total, thus

$$2^{d_{\text{vc}}(\mathcal{H})} \le M$$
$$d_{\text{vc}}(\mathcal{H}) \le \log_2 M$$

#### Problem 4

(15 points)Let  $\mathcal{H}_1, \mathcal{H}_2, \dots, \mathcal{H}_K$  be K hypothesis sets with finite VC dimension  $d_{vc}$ . Let  $\mathcal{H} = \mathcal{H}_1 \cup \mathcal{H}_2 \cup \dots \cup \mathcal{H}_K$  be the union of these models. Show that  $d_{vc}(\mathcal{H}) < K(d_{vc}+1)$ .

#### Solution:

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First, we prove a conclusion

$$d_{\text{vc}}\left(\bigcup_{k=1}^{K} \mathcal{H}_{k}\right) \leq K - 1 + \sum_{k=1}^{K} d_{\text{vc}}\left(\mathcal{H}_{k}\right)$$

We prove this conclusion by mathematical induction.

Proof: when K = 2, we can get that

$$d_{\text{vc}}\left(\bigcup_{k=1}^{2} \mathcal{H}_{k}\right) \leq 1 + \sum_{k=1}^{2} d_{\text{vc}}\left(\mathcal{H}_{k}\right)$$

Suppose that

$$d_{\text{vc}}\left(\bigcup_{k=1}^{2} \mathcal{H}_{k}\right) \ge 2 + \sum_{k=1}^{2} d_{\text{vc}}\left(\mathcal{H}_{k}\right)$$

then

$$m_{\mathcal{H}_1 \bigcup \mathcal{H}_2} (d_1 + d_2 + 2) \ge 2^{d_1 + d_2 + 2}$$

However,

$$m_{\mathcal{H}_1 \bigcup \mathcal{H}_2} (d_1 + d_2 + 2) \le m_{\mathcal{H}_1} (d_1 + d_2 + 2) + m_{\mathcal{H}_2} (d_1 + d_2 + 2)$$

$$\le \sum_{i=0}^{d_1} {d_1 + d_2 + 2 \choose i} + \sum_{i=0}^{d_2} {d_1 + d_2 + 2 \choose i}$$

$$= 2^{d_1 + d_2 + 2} - {d_1 + d_2 + 2 \choose d_1 + 1} < 2^{d_1 + d_2 + 2}$$

Therefore,  $d_{\text{vc}}\left(\bigcup_{k=1}^{2}\mathcal{H}_{k}\right)\leq1+\sum_{k=1}^{2}d_{\text{vc}}\left(\mathcal{H}_{k}\right)$ . Suppose that when K=n, the conclusion is valid, then for K=n+1

$$d_{\text{vc}}\left(\bigcup_{k=1}^{n+1} \mathcal{H}_{k}\right) = d_{\text{vc}}\left(\left(\bigcup_{k=1}^{n} \mathcal{H}_{k}\right) \bigcup \mathcal{H}_{n+1}\right)$$

$$\leq 1 + d_{\text{vc}}\left(\bigcup_{k=1}^{n} \mathcal{H}_{k}\right) + d_{\text{vc}}\left(\mathcal{H}_{n+1}\right)$$

$$\leq 1 + n - 1 + \sum_{k=1}^{n} d_{\text{vc}}\left(\mathcal{H}_{k}\right) + d_{\text{vc}}\left(\mathcal{H}_{n+1}\right)$$

$$= n + \sum_{k=1}^{n+1} d_{\text{vc}}\left(\mathcal{H}_{k}\right)$$

Therefore, for K = n + 1, the conclusion is still valid.

As 
$$\mathcal{H} = \mathcal{H}_1 \cup \mathcal{H}_2 \cup \cdots \cup \mathcal{H}_K$$
,  $d_{vc}(\mathcal{H}_k) = d_{vc} \,\forall \, k = 1, 2, ..., K$ , we can get that 
$$d_{vc}(\mathcal{H}) < K - 1 + K d_{vc} < K \, (d_{vc} + 1)$$

### Problem 5

(40 points)In this part, you need to complete some mathematical proofs about VC dimension. Suppose the hypothesis set

$$\mathcal{H} = \{ f(x, \alpha) = \operatorname{sign}(\sin(\alpha x)) \mid, \alpha \in \mathbb{R} \}$$

where x and f are feature and label, respectively.

• Show that  $\mathcal{H}$  cannot shatter the points  $x_1 = 1, x_2 = 2, x_3 = 3, x_4 = 4$ .

(Key: Mathematically, you need to show that there exists  $y_1, y_2, y_3, y_4$ , for any  $\alpha \in \mathbb{R}$ ,  $f(x_i) \neq y_i, i = 1, 2, 3, 4$ , for example +1, +1, -1, +1)

• Show that the VC dimension of  $\mathcal{H}$  is  $\infty$ . (Note the difference between it and the first question)

(Key: Mathematically, you have to prove that for any label sets  $y_1, \dots, y_m, m \in \mathbb{N}$ , there exists  $\alpha \in \mathbb{R}$  and  $x_i, i = 1, 2, \dots, m$  such that  $f(x; \alpha)$  can generate this set of labels. Consider the points  $x_i = 10^{-i} \dots$ )

## Solution:

- 1) Shattering means that for any possible labeling of the points, there exists a parameter  $\alpha$  such that  $f(x,\alpha)=\mathrm{sign}(\sin(\alpha x))$  can produce those labels. Consider the labeling  $y_1=+1,y_2=+1,y_3=-1,y_4=+1$  for  $x_1=1,x_2=2,x_3=3,x_4=4$ . The sine function,  $\sin(\alpha x)$  is is periodic with a period of  $2\pi$ . This means that it repeats its values every  $2\pi$  units. For the chosen labeling, the sign of  $\sin(\alpha x)$  must change between  $x_2$  and  $x_3$ , and then again between  $x_3$  and  $x_4$ . Given the periodic nature of the sine function and the distances between the points, its impossible to choose an  $\alpha$  that will result in the sine function having the required sign changes between these specific points. The sine function would need to complete more than half a period between  $x_2$  and  $x_3$  and less than half a period between  $x_3$  and  $x_4$ , which is contradictory given the uniform spacing of the points. Therefore, it is impossible to find an  $\alpha$  that allows  $\mathcal H$  to shatter points  $x_1=1,x_2=2,x_3=3,x_4=4$  with the labeling  $x_1=1,x_2=1$ .
- 2) Consider the labeled data set  $(2\pi 10^{-i}, y_i)_{i=1}^n$  and choose, for any such data set, the parameter  $\alpha = \frac{1}{2} \left(1 + \sum_{i=1}^n \frac{1-y_i}{2} 10^i\right)$ . We observe that, for any point  $x_j = 2\pi 10^{-j}$  in the considered data set such that  $y_j = -1$ , the term  $10^j$  appears in the sum. This leads to

$$\alpha x_j = \pi 10^{-j} \left( 1 + \sum_{i:y_i = -1} 10^i \right)$$

$$= \pi \left( 10^{-j} + 1 + \sum_{i:y_i = -1, i > j} 10^{i-j} + \sum_{i:y_i = -1, i < j} 10^{i-j} \right)$$

For all i > j, the terms  $10^{i-j}$  are positive powers of 10 and thus are even numbers that can be written as  $2k_i$  for some  $k_i \in \mathbb{N}$ . Therefore, we have

$$\sum_{i:y_i=-1,i>j} 10^{i-j} = \sum_{i:y_i=-1,i>j} 2k_i = 2k$$

for some  $k \in \mathbb{N}$ , which gives

$$\alpha x_j = \pi \left( 10^{-j} + 1 + \sum_{i:y_i = -1, i < j} 10^{i-j} \right) + 2k\pi$$

Regarding the remaining sum, we have

$$\sum_{i:y_i=-1,i< j} 10^{i-j} < \sum_{i=1}^{+\infty} 10^{-i} = \sum_{i=0}^{+\infty} 10^{-i} - 1 = \frac{1}{9}$$

Let define  $\epsilon=10^{-j}+\sum_{i:y_i=-1,i< j}10^{i-j}$  and rewrite  $\alpha x_j$  as  $\alpha x_j=\pi(1+\epsilon)+2k\pi$ . Since  $0<\epsilon<1$ , thus  $\pi<\pi(1+\epsilon)<2\pi$  and  $\sin(\alpha x_j)<0$ . Hence, the classifier correctly predicts all negative labels  $y_j=-1=\mathrm{sign}\left(\sin\left(\alpha x_j\right)\right)=f(x_j)$ .

The same steps can be reproduce with positive labels  $y_j = +1$  with the difference that the term  $10^j$  does not appear in the sum defining  $\alpha$ . This leads to

$$\alpha x_j = \pi 10^{-j} \left( 1 + \sum_{i:y_i = -1, i \neq j} 10^i \right)$$

$$= \pi \left( 10^{-j} + \sum_{i:y_i = -1, i > j} 10^{i-j} + \sum_{i:y_i = -1, i < j} 10^{i-j} \right)$$

$$= \pi \epsilon + 2k\pi$$

with  $0 < \pi \epsilon < \pi$  and  $\sin(\alpha x_i) > 0$ .

Thus, all positively labeled points are also correctly classified by f using the particular choice of  $\alpha$ . Since the steps above are valid for any labeling of the points, we proved that  $\mathcal H$  shatters the set of points. In addition, the proof is valid for any number of points n, which shows that  $\mathcal H$  can shatter sets of points of any size and thus has infinite VC-dimension.