

# Assignment 4

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CSCI 4100 - Machine Learning from Data

September 29, 2019

1. (50) LFD Exercise 2.4

(a)

$$X = \{1\} \times \mathbb{R}^d$$

$$X = \begin{bmatrix} x_1^T \\ x_2^T \\ \vdots \\ x_{d+1}^T \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdots & \cdots & 0 \\ 1 & 1 & \cdots & \cdots & 0 \\ \vdots & 0 & \ddots & & 0 \\ \vdots & \vdots & & \ddots & 0 \\ 1 & 0 & \cdots & 0 & 1 \end{bmatrix}, w = \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_d \end{bmatrix}, y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{d+1} \end{bmatrix} = \begin{bmatrix} \pm 1 \\ \pm 1 \\ \vdots \\ \pm 1 \end{bmatrix} = [\pm 1 \ \pm 1 \ \cdots \ \pm 1]^T$$

$X$  has all rows and columns linearly independent, giving it full rank and thus, is an invertible and nonsingular matrix.

For any  $y$ , finding a  $w$  such that  $\text{sign}(Xw) = y$ , then the perceptron can shatter  $(d+1)$  points. This is because  $Xw = y$  and  $X$  is invertible, meaning  $w = X^{-1}y$ .

(b)

$$X = [x_1, x_2, \dots, x_d]$$

So, for there to be any new vector, it must be a linear combination of the  $d+1$  vector. That means any  $d+2$  vectors that have length  $d+1$  (more points than dimensions) must be linearly dependent, or

$$x_j = \sum_{i \neq j} a_i x_i$$

Where not all of the  $a_i$ 's are equal to 0.

Now, we create a dichotomy that the perceptron learning algorithm cannot shatter

$$y_i = \begin{cases} \text{sign}(a_i), & i \neq j \\ -1, & i = j \end{cases}$$

Using the PLA, we have

$$x_j = \sum_{i \neq j} a_i x_i \rightarrow w^T x_i = \sum_{i \neq j} a_i w^T x_i$$

We know that  $a_i x_i = w^T x_i$ , so if  $y_i = \text{sign}(w^T x_i)$  and  $\text{sign}(a_i)$ , then  $a_i w^T x_i > 0$ .

Thus,

$$w^T x_i = \sum_{i \neq j} a_i w^T x_i > 0$$

Must be true since each of our classifications will be  $> 0$ . This means having a correct classification of  $y_i$  is impossible because  $y_i = \text{sign}(w^T x_i) = +1$

Thus, the PLA cannot implement this dichotomy.

$\therefore \nexists \{d+2 \text{ points}\}$  that can be shattered by the PLA

## 2. (300) LFD Problem 2.3

(a) Positive or negative ray:

As previously computed in exercise 2.1, we know that the growth function for positive rays is equal to  $N + 1$ .

For positive rays, we have  $m_H(N) = N + 1$

For negative rays, we have  $m_H(N) = N - 1$

The sum of these two rays gives us

$$m_H(N) = (N + 1) + (N - 1) = 2N.$$

The largest value of  $N$  we can use where  $m_H(N) = 2^N$  is 2.

$$m_H(N) = 2^N$$

$$m_H(2) = 2(2) = 4 \text{ \& } 2^2 = 4, \rightarrow 4 = 2^2$$

$$m_H(3) = 2(3) = 6 \text{ \& } 2^3 = 8, \rightarrow 6 < 2^3$$

$$\therefore d_{vc} = 2$$

(b) Positive or negative interval:

As previously computed in exercise 2.1, we know that the growth function for positive intervals is equal to  $\binom{N+1}{2} + 1 = \frac{1}{2} N^2 + \frac{1}{2} N + 1$ .

For positive intervals, we have  $m_H(N) = \frac{1}{2} N^2 + \frac{1}{2} N + 1$

For negative intervals, we have  $\binom{N-1}{2}$  The +1 is included in the positive intervals

-We get  $N-2$  new intervals giving us  $N+1-2 = N-1$

The sum of these two intervals give us

$$m_H(N) = \binom{N+1}{2} + \binom{N-1}{2} = N^2 - N + 2$$

The largest value of  $N$  we can use where  $m_H(N) = 2^N$  is 3

$$m_H(N) = 2^N$$

$$m_H(2) = (2)^2 - 2 + 2 = 4 \text{ \& } 2^2 = 4, \rightarrow 4 = 2^2$$

$$m_H(3) = (3)^2 - 3 + 2 = 8 \text{ \& } 2^3 = 8, \rightarrow 8 = 2^3$$

$$m_H(4) = (4)^2 - 4 + 2 = 10 \text{ \& } 2^4 = 16, \rightarrow 10 < 2^4$$

$$\therefore d_{vc} = 3$$

(c) Two concentric spheres in  $\mathbb{R}^d$ :

This problem can be viewed as solving the problem of positive intervals. This can be done by mapping a point in  $\mathbb{R}^d$  into a point  $y \in \mathbb{R}$  giving us

$$\text{Map: } \{x_1, \dots, x_d\} \rightarrow y = \sqrt{x_1^2 + \dots + x_d^2} \text{ in } 0 \leq y \leq \infty$$

We have successfully transformed the problem from a  $\mathbb{R}^d$  to a  $\mathbb{R}$  problem where  $y \in [0, \dots, \infty)$ . This means that  $m_H(N)$  no longer depends on  $d$ .

$$\text{Thus, } m_H(N) = \frac{1}{2} N^2 + \frac{1}{2} N + 1$$

The largest value of  $N$  we can use where  $m_H(N) = 2^N$  is 2

$$m_H(N) = 2^N$$

$$\begin{aligned}
m_H(2) &= (1/2)(2)^2 + (2/2) + 1 = 4 \text{ \& } 2^2 = 4, \rightarrow 4 = 2^2 \\
m_H(3) &= (1/2)(3)^2 + (3/2) + 1 = 7 \text{ \& } 2^3 = 8, \rightarrow 7 < 2^3 \\
&\therefore d_{vc} = 2
\end{aligned}$$

### 3. (200) LFD Problem 2.8

We know since  $m_H(N) \leq 2^N$ , there can only exist two cases. The first is that if  $m_H(N) = 2^N$  for all  $N$ , then  $d_{vc}(H) = \infty$ . The second is  $d_{vc}(H)$  is a finite value, meaning that  $m_H(k) \leq 2^k$  for some  $k$ , and  $m_H(N)$  is bounded, such that

$$m_H(N) \leq N^{d_{vc}} + 1 \quad \text{Def. 2.5 \& (2.10) | pg.50 in LFD}$$

$$m_H(N) = 1 + N: \checkmark$$

$$m_H(1) = 2 \text{ \& } 2^1 = 2 \rightarrow 2 = 2^1$$

$$m_H(2) = 3 \text{ \& } 2^2 = 4 \rightarrow 3 < 2^2 [\therefore d_{vc} = 1]$$

This means that it must be bounded by  $1 + N$ ,  $\forall N$ . This is a possible growth function

$$m_H(N) = 1 + N + \frac{N(N-1)}{2}: \checkmark$$

$$m_H(1) = 2 \text{ \& } 2^1 = 2 \rightarrow 2 = 2^1$$

$$m_H(2) = 4 \text{ \& } 2^2 = 4 \rightarrow 4 = 2^2$$

$$m_H(3) = 7 \text{ \& } 2^3 = 8 \rightarrow 7 < 2^3 [\therefore d_{vc} = 2]$$

This means that it must be bounded by  $1 + N$ ,  $\forall N$ . This is a possible growth function

$$m_H(N) = 2^N: \checkmark$$

This one is trivially a possible growth function as  $2^N \leq 2^N$

$$m_H(N) = 2^{\lfloor \sqrt{N} \rfloor}: \times$$

$$m_H(1) = 2^1 \text{ \& } 2^1 = 2 \rightarrow 2 = 2^1$$

$$m_H(2) = 2^1 \text{ \& } 2^1 = 2 \rightarrow 2 = 2^1 [\therefore d_{vc} = 1]$$

$$m_H(N) \leq N^{d_{vc}} + 1$$

$$m_H(N) \leq N + 1$$

Thus, this must be bounded by  $N+1 \forall N$  but that is not true.

$$\text{Ex: When } N = 16 [m_H(16) = 2^4 = 16 > 5]$$

This is not a possible growth function.

$$m_H(N) = 2^{\lfloor \sqrt{N/2} \rfloor}: \times$$

$$m_H(1) = 2^0 \text{ \& } 2^1 = 2 \rightarrow 1 < 2^1 [\therefore d_{vc} = 0]$$

$$m_H(N) \leq N^{d_{vc}} + 1$$

$$m_H(N) \leq N^0 + 1 = 1+1 = 2$$

Thus, this must be bounded by  $1+1=2 \forall N$  but that is not true.

$$\text{Ex: When } N = 32 [m_H(32) = 2^4 = 16 > 2]$$

This is not a possible growth function.

$$m_H(N) = 1 + N + \frac{N(N-1)(N-2)}{6}: \times$$

$$\begin{aligned}
m_H(1) &= 2 \text{ \& } 2^1 = 2 \rightarrow 2 = 2^1 \\
m_H(2) &= 3 \text{ \& } 2^2 = 4 \rightarrow 3 < 2^2 [\therefore d_{vc} = 1] \\
m_H(N) &\leq N^{d_{vc}} + 1 \\
m_H(N) &\leq N^1 + 1 = N+1
\end{aligned}$$

Thus, this must be bounded by  $N+1 \forall N$  but that is not true.

Ex: When  $N = 3$  [ $m_H(3) = 5$  &  $N+1 = 4 \mid 5 > 4$ ]

This is not a possible growth function.

#### 4. (100) LFD Problem 2.10

$$\text{Show } m_H(2N) \leq m_H(N)^2$$

Suppose that there exist  $2N$  points. Partition the  $2N$  points into two sets of  $N$  points. Based on the definition of the growth function, we know that each set of  $N$  points cannot generate greater than  $m_H(N)$  dichotomies. If each group of  $N$  points can only produce up to  $m_H(N)$  dichotomies, then combining them gives us one set of  $2N$  points that can produce a max number  $2 \times m_H(N)$  dichotomies. Thus,  $2N$  points produce at most  $m_H(N) \times m_H(N) = m_H(N)^2$ .

$$\therefore m_H(2N) \leq m_H(N)^2$$

VC Generalization bound:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_H(2N)}{\delta}\right)}$$

With our results, combining it with the VC generalization bound gives us:

$$\begin{aligned}
E_{\text{out}}(g) &\leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_H(2N)}{\delta}\right)} \leq E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_H(N)^2}{\delta}\right)} \\
E_{\text{out}}(g) &\leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_H(N)^2}{\delta}\right)}
\end{aligned}$$

#### 5. (100) LFD Problem 2.12

VC Generalization bound:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_H(2N)}{\delta}\right)}$$

$$d_{vc} = 10$$

Need 95% confidence

$$\delta = 5\% = 0.05$$

VC Generalization bound:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4((2N)^{10} + 1)}{0.05}\right)} \leq 0.05$$

With 10,000 training examples,

$$N = 4.52957 \times 10^5$$

Thus, a sample size of  $4.52957 \times 10^5$  or greater is needed to receive such confidence.