Machine Learning from Data CSCI 4100 Assignment 4 Jae Park 661994900

#### Exercise 2.4

(a) X can be represented as

$$X = \begin{bmatrix} x_{10} & x_{11} & & x_{1d} \\ x_{20} & x_{21} & \dots & x_{2d} \\ \vdots & \vdots & & \vdots \\ x_{d+1 \ 0} & x_{d+1 \ 1} & & x_{d+1 \ d} \end{bmatrix}$$

and w & y are given as

$$w = \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_d \end{bmatrix}, y = \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_d \end{bmatrix}$$

X is an invertible, nonsingular matrix of dimension  $(d+1) \times (d+1)$  where each column is an individual  $\vec{x}$ .

y is given as  $y = [y_0, y_1, ..., y_d]^T$  where each element of y is composed of +1's and -1's

Given  $X \cdot w = y$ , if w exists that can satisfy sign(Xw) = y, then the perceptron shatters the (d+1) points. In such cases, now  $w = X^{-1}y$  holds true because X is an invertible matrix, as stated above.

Because a valid solution matrix of w exists, it concludes the inequality  $d_{vc} \geq d+1$ .

(b)Let each point of X as  $[\vec{x}_0, \vec{x}_1, ..., \vec{x}_d]$  like the construction earlier in (a). Then new vector  $\vec{x}_{d+2}$  should be the linear combination of the previous (d+1) vectors in such form

$$\vec{x}_{d+2} = c_0 \vec{x}_0 + c_1 \vec{x}_1 + c_2 \vec{x}_2 + \dots + c_{d+1} \vec{x}_{d+1}$$

Now, we assume that  $w^T c_i x_i < 0$  for all  $c_i$ . If this is true, then for some dichotomy, it is always true that  $sign(w^T \cdot x_{d+2}) = -1$ . This means that there is some dichotomy, more specifically [+1], that cannot be implemented.

## Problem 2.3

- (a) With N data points, you can make N+1 regions. And for each positive and negative ray, you get  $m_H(N)$  is 2N+2. From this, you subtract two cases where points are either all +1's or -1's. So  $m_H(N)=2N$ .
- $m_{\mathcal{H}}(1) = 2 = 2^1$
- $m_{\mathcal{H}}(2) = 4 = 2^2$
- $m_{\mathcal{H}}(3) = 6 < 2^3$
- So  $d_{vc} = 2$
- (b) (referring to the graph in LFD pg.44) In the positive intervals, the maximum number of dichotomies is  $\frac{1}{2}N^2 + \frac{1}{2}N + 1$ .

Now, the negative intervals has the same case in positive intervals, and the only additional case arises when both end points are set to be +1. In such case, the number of dichotomies is  $\binom{N+1-2}{2}$ , which is  $\frac{(N-1)!}{(2)!(N-3)!} = \frac{1}{2}N^2 - \frac{3}{2}N + 1$ .

$$m_{\mathcal{H}}(N) = (\frac{1}{2}N^2 + \frac{1}{2}N + 1) + (\frac{1}{2}N^2 - \frac{3}{2}N + 1) = N^2 - N + 2$$

- $m_{\mathcal{H}}(1) = 2 = 2^1$
- $m_{\mathcal{H}}(2) = 4 = 2^2$
- $m_{\mathcal{H}}(3) = 8 = 2^3$
- $m_{\mathcal{H}}(3) = 14 < 2^4$
- So  $d_{vc} = 3$
- (c) For the values  $x_1, x_2, ... x_m$ , y is given as  $y = \sqrt{x_1^2 + x_2^2 + ... + x_n^2}$ . This is essentially same as finding the interval, where  $x_i$  mapped onto a line, and the interval is split by the points into at most N+1 regions. According to LFD pg 44,  $m_{\mathcal{H}}(N) = \binom{N+1}{2} + 1 =$

$$\frac{1}{2}N^2 + \frac{1}{2}N + 1. \ m_{\mathcal{H}}(2) = 2 = 2^1$$

- $m_{\mathcal{H}}(2) = 4 = 2^2$
- $m_{\mathcal{H}}(3) = 7 < 2^3$
- So  $d_{vc} = 2$

# Problem 2.8

A growth function is either bounded by some polynomial function of N or the function of  $2^N$ . If the growth function has a break point, then it is bounded by a polynomial function; otherwise,  $2^N$ .

$$m_{\mathcal{H}}(N) = 1 + N$$
:  $m_{\mathcal{H}}(1) = 2 = 2^1 m_{\mathcal{H}}(2) = 3 < 2^2$   
 $d_{vc} = 1$  so a possible growth function.

$$m_{\mathcal{H}}(N) = 1 + N + \frac{N(N-1)}{2}$$
:  $m_{\mathcal{H}}(2) = 4 = 2^2 m_{\mathcal{H}}(3) = 7 < 2^3$   $d_{vc} = 2$  so a possible growth function.

 $m_{\mathcal{H}}(N) = 2^N$ :  $d_{vc} = \infty$  so a possible growth function.

$$m_{\mathcal{H}}(N) = 2^{\left\lfloor \sqrt{N} \right\rfloor}$$
:  $m_{\mathcal{H}}(0) = 1 = 2^0 m_{\mathcal{H}}(1) = 2 = 2^1 m_{\mathcal{H}}(2) = 2 < 2^2$   $d_{vc} = 1$  and this does not satisfy the inequality  $m_{\mathcal{H}}(N) \leq \sum_{i=0}^{d_{vc}} \binom{N}{i}$ 

$$m_{\mathcal{H}}(N) = 2^{\lfloor N/2 \rfloor} : m_{\mathcal{H}}(0) = 1 = 2^0 \, m_{\mathcal{H}}(1) = 1 < 2^1$$
  
 $d_{vc} = 0$  and this does not satisfy the inequality  $m_{\mathcal{H}}(N) \leq \sum_{i=0}^{d_{vc}} {N \choose i}$ 

$$m_{\mathcal{H}}(N) = 1 + N + \frac{N(N-1)(N-2)}{6}$$
:  $m_{\mathcal{H}}(0) = 1 = 2^0 m_{\mathcal{H}}(1) = 2 = 2^1 m_{\mathcal{H}}(2) = 3 < 2^2$   $d_{vc} = 1$  and this does not satisfy the inequality  $m_{\mathcal{H}}(N) \leq \sum_{i=0}^{d_{vc}} {N \choose i}$ 

Therefore,  $m_{\mathcal{H}}(N) = 2^{\left\lfloor \sqrt{N} \right\rfloor}, 2^{\left\lfloor N/2 \right\rfloor}, 1 + N + \frac{N(N-1)(N-2)}{6}$  cannot be a possible growth function.

### Problem 2.10

 $m_{\mathcal{H}}(2N)$  means the greatest number of dichotomies obtainable by two independent groups of N points, or in other words, maximum possible combination of two different groups of dichotomies. The case for this maximum possible combination is given as  $m_{\mathcal{H}}(N) \times m_{\mathcal{H}}(N) = m_{\mathcal{H}}(N)^2$ . Therefore,  $m_{\mathcal{H}}(2N) \leq m_{\mathcal{H}}(N)^2$ 

A generalization bound which involves only  $m_{\mathcal{H}}(N)$  is given as

$$E_{out}(g) \le E_{in}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(N)^2}{\delta}} \quad \dots \text{LFD (2.12)}$$
  
which is the most conservative bound possible with probability  $\ge 1 - \delta$ .

### Problem 2.12

$$N \ge \frac{8}{\epsilon^2} \ln \left( \frac{4((2N)^{d_{vc}} + 1)}{\delta} \right)$$

...LFD (2.13) on pg. 57

where,  $d_{vc} = 10$ ,  $\epsilon = 0.05$ ,  $\delta = 0.05$ .

$$N \ge \frac{8}{0.05^2} \ln \left( \frac{4((2N)^{10} + 1)}{0.05} \right)$$

Trying an initial guess of N = 1,000 in the RHS, we get

$$N \ge \frac{8}{0.05^2} \ln \frac{4((2 \times 1000)^{10} + 1)}{0.05} = 257251$$

After an iterative process, N rapidly converges to an estimate of  $N \approx 452957$  Therefore, the sample size needs to be greater than or equal to 452957.