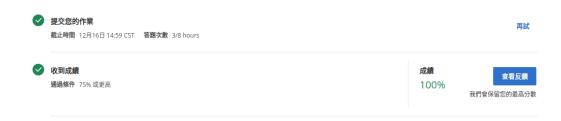
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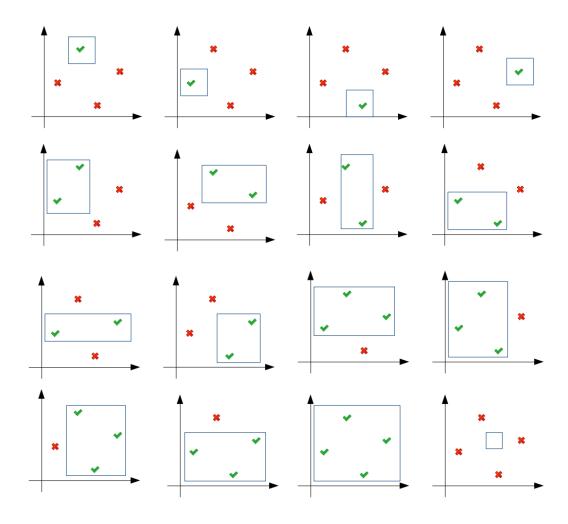
# 作業二



 $\mathbf{2}$ 

we need to proof that  $d_v c \ge 4$ , which means we must find a set of 4 inputs that can be shattered.

Consider 4 points are in the first quadrant in 2D space, which arranged in four corners of a rotated square.



3

#### Step1

First, we analyze the formula  $h_{\alpha}(x) = sign(|\alpha x \mod 4 - 2| - 1)$  in the question.

For simplicity, we define sign(0) = -1. And we can get that:  $h_{\alpha}(x) = \begin{cases} -1, & \text{when } 1 \leq \alpha x \mod 4 \leq 3 \\ 1, & \text{otherwise} \end{cases}$ 

#### Step2

Now we need to come up with the configurations of  $\alpha$  and x. Suppose there are N points,  $x_i = 4^i$   $(1 \le i \le N)$ . So we need to construct  $2^N$  kinds of  $\alpha$  to satisfy all  $\{+1, -1\}^N$  combinations.

We construct one of the  $\alpha$  as follows:

$$\alpha_k$$
 is the kth alpha,  $1 \leq k \leq 2^N$  
$$\alpha_k = C_1 4^{-1} + C_2 4^{-2} + \cdots + C_N 4^{-N}$$
 
$$C_i = \{0,1\}, 1 \leq i \leq N$$

Let  $T = (x_i \alpha_k) \mod 4$ .

$$T = 4^{i}(C_{1}4^{-1} + C_{2}4^{-2} + \dots + C_{N}4^{-N}) \mod 4$$
$$= (C_{1}4^{i-1} + C_{2}4^{i-2} + \dots + C_{N}4^{i-N}) \mod 4$$
$$= C_{i} + C_{i+1}4^{-1} + \dots + C_{N}4^{i-N}$$

then we can reduce the formula about  $h_{\alpha}(x)$  as follows,

$$h_{\alpha_k}(x_i) = sign(|\alpha_k x_i \mod 4 - 2| - 1)$$
$$= sign(|T - 2| - 1)$$

and we will get 
$$h_{\alpha_k}(x_i) = \begin{cases} -1, & \text{when } 1 \leq T \leq 3 \\ 1, & \text{otherwise} \end{cases}$$

By the properties of the sum of geometric sequence, we know  $\begin{cases} \text{ if } C_i = 1 \text{ then } 1 \leq T \leq \frac{4}{3} & \Rightarrow h_{\alpha_k}(x_i) = -1 \\ \text{ if } C_i = 0 \text{ then } 0 \leq T \leq \frac{1}{3} & \Rightarrow h_{\alpha_k}(x_i) = 1 \end{cases}$ 

# Step3

For any finite N, we must find N inputs that we can shatter.

Let 
$$X = \{x_i = 4^i\}$$
 and  $Y \in \{0, 1\}^N$  for  $1 \le i \le N$ , then  $\alpha = \sum_{i=1}^N C_i 4^{-i}$ , 
$$\begin{cases} C_i = 1, & \text{if } y_i = h_\alpha(x_i) = -1 \\ C_i = 0, & \text{if } y_i = h_\alpha(x_i) = 1 \end{cases}$$
Thus  $d_{VC} = \infty$ .

# 4

First, we assume that  $d_{vc}(H_1 \cap H_2) = n$  and  $d_{vc}(H_1) = m$ , which means n inputs can be shattered by  $H_1 \cap H_2$  and m inputs can be shattered by  $H_1$ . Proof that  $n \leq m$ .

#### Prove by Contradiction.

Suppose that n > m.

Since n > m, by the definition of VC-Dimension, we know that n inputs can be shattered by  $H_1 \cap H_2$  but cannot be shattered by  $H_1$ .

However,  $H_1 \cap H_2 \subseteq H_1$ , the inputs shattered by  $H_1 \cap H_2$  must also be shattered by  $H_1$ , contradiction to the assumption. Therefore, we have proven that  $n \leq m$ ,  $d_{vc}(H_1 \cap H_2) \leq d_{vc}(H_1)$ .

### **5**

Since the intersection of  $H_1$  and  $H_2$  is all positive or all negetive. And we know  $m_{H_1}(N) = m_{H_2}(N) = N + 1$ .

$$m_{H_1 \cup H_2}(N) = m_{H_1}(N) + m_{H_2}(N) - m_{H_1 \cap H_2}(N)$$
$$= 2(N+1) - 2$$
$$= 2N.$$

when 
$$N=2$$
,  $m_{H_1\cup H_2}(N)=2N=4=2^2$ .  
when  $N=3$ ,  $m_{H_1\cup H_2}(N)=2N=6<2^3$ . Thus,  $d_{vc}(H_1\cup H_2)=2$ 

we know  $h_{s,\theta}(x) = s \cdot sign(x - \theta)$  and f(x) = sign(x) + noise.

 $\mu = \text{average error rate } h(x) \neq f(x) = \text{average false accept and false reject}$ .

$$\mu = \frac{\textit{green part}}{\textit{yellow part} + \textit{green part}}$$

when s=1  $h_{s,\theta}(x) = sign(x-\theta)$  f'(x) = sign(x)  $h_{s,\theta}(x) = -sign(x-\theta)$  f'(x) = sign(x)  $\frac{\theta}{-1}$  -1 0 1  $\mu = \frac{|\theta|}{2}$   $\mu = \frac{(2-|\theta|)}{2} = 1 - \frac{|\theta|}{2}$ 

we can make a small conclusion by eliminating the noise factor (20 percent flip) first.

$$\mu = \begin{cases} \frac{|\theta|}{2}, & \text{when } s = 1\\ 1 - \frac{|\theta|}{2}, & \text{when } s = -1 \end{cases}$$

Combine the two cases above, we have  $\mu = \frac{s(|\theta|-1)+1}{2}$ .

By Problime 1 in coursera, we know  $E_{out}(h_s;\theta) = \lambda \mu + (1-\lambda)(1-\mu)$ .  $\lambda$  is rate of case with no noise.

$$E_{out}(h_s; \theta) = P[\text{no flip}]P[h(x) \neq f(x)] + P[\text{ flip }]P[h(x) = f(x)]$$

$$= \lambda \mu + (1 - \lambda)(1 - \mu)$$

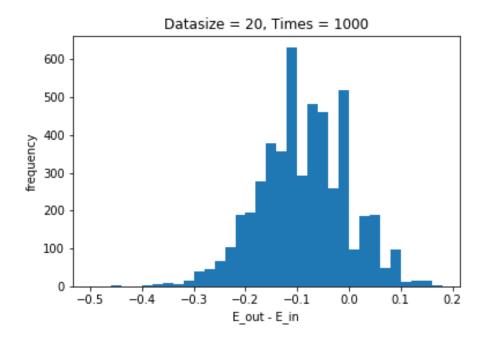
$$= 0.8\mu + 0.2(1 - \mu)$$

$$= 0.2 + 0.6\mu$$

$$= 0.2 + 0.6(\frac{s(|\theta| - 1) + 1}{2})$$

$$= 0.5 + 0.3s(|\theta| - 1)$$

The average of  $E_{in}-E_{out}$  falls around -0.089, and the range of  $E_{in}-E_{out}$  is actually -0.5 to 0.2.

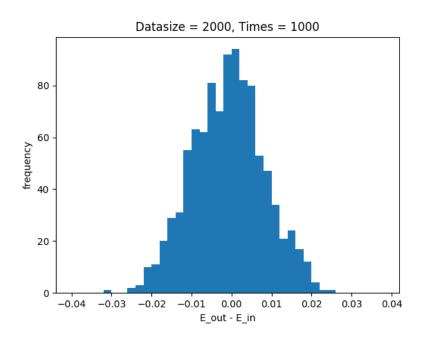


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The average of  $E_{in}-E_{out}$  falls around -0.0008, and the range of  $E_{in}-E_{out}$  is actually -0.03 to 0.03.

Compared with the previous case,  $|E_{in} - E_{out}|$  becomes smaller with bigger datasize.

And the distribution of 1000 times experiment with 2000 data is more like a normal distribution whose mean is closer to 0.



Consider the "simplified decision trees" hypothesis set on  $\mathbb{R}^d$ , which is given by

$$H = \{h_{t,s} | h_{t,s}(x) = 2[[v \in S]] - 1, \text{ where } v_i = [[x_i > t_i]],$$

$$\mathbf{S} \text{ a collection of vectors in } \{0, 1\}^d, \mathbf{t} \in \mathbb{R}^d\}$$

By the definition of H, we know that  $v_i = \begin{cases} 1, & \text{when } x_i > t_i \\ 0, & \text{when } x_i \le t_i \end{cases}$  Each  $t_i$  can divide the space into two part.

Therefore if there are d dimension in the space, we can divide the space into  $2^d$  different regoins.

If there are more than  $2^d$  points in the space, it must exist two points in the space belong to the same separated region, we can't assign these two points to different regions.

For example when d = 2. There are 5 points in the space. Always more than 1 points in one of the regions. Suppose  $p_4$  and  $p_5$  in the same region, then the following cases can't appear at the same time.

$$(p_1, p_2, p_3, p_4, p_5)$$
  
 $(+, -, -, +, +)$   
 $(+, -, -, +, -)$ 

Thus, we can know that the VC-dimension of the "simplified decision trees" hypothesis set is  $2^d$ .