

1.



提交您的作業

截止時間 12月1日 22:59 PST 答題次數 3/8 hours

再試



收到成績

通過條件 75% 或更高

成績

100%

查看反饋

我們會保留您的最高分數

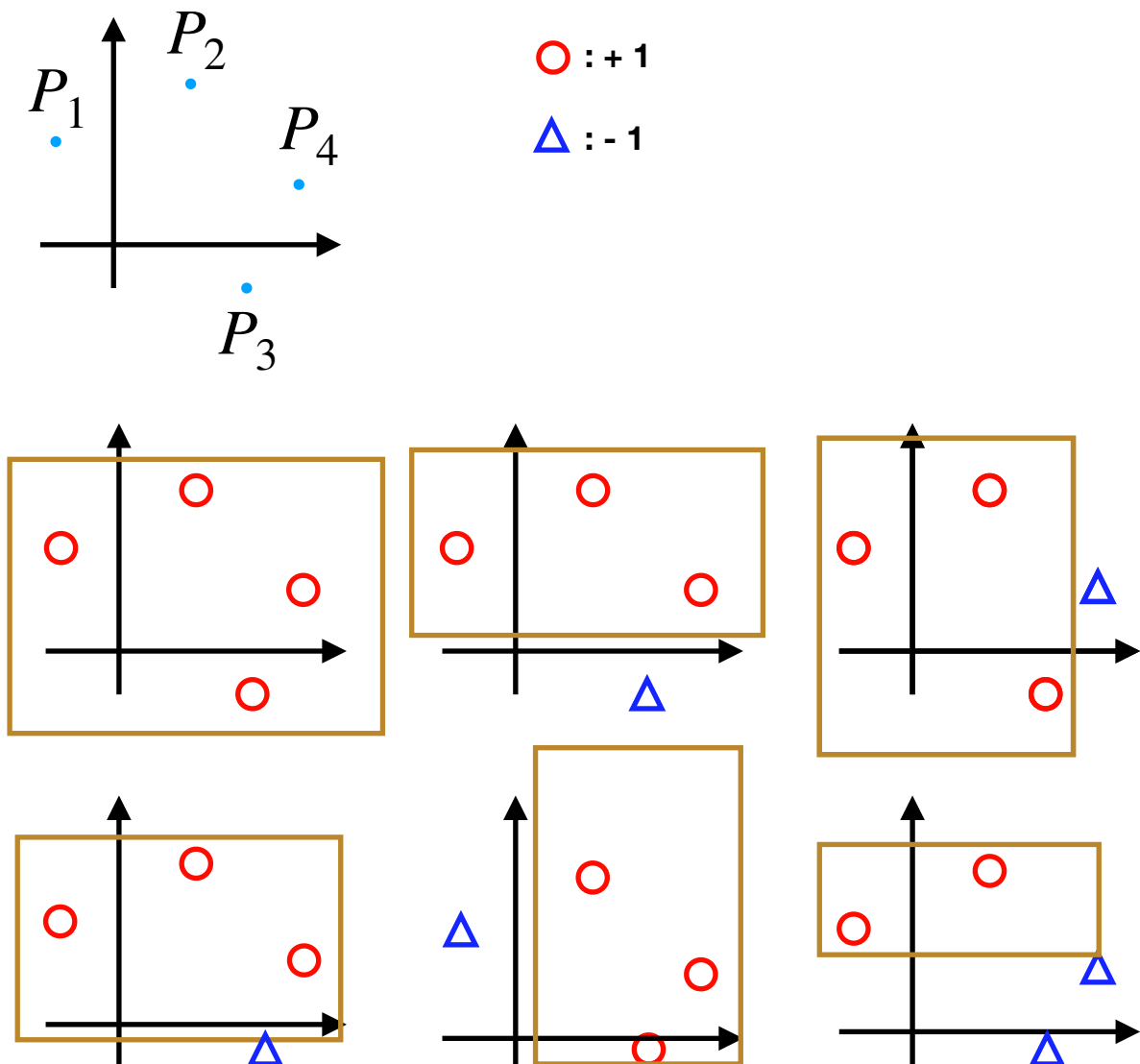


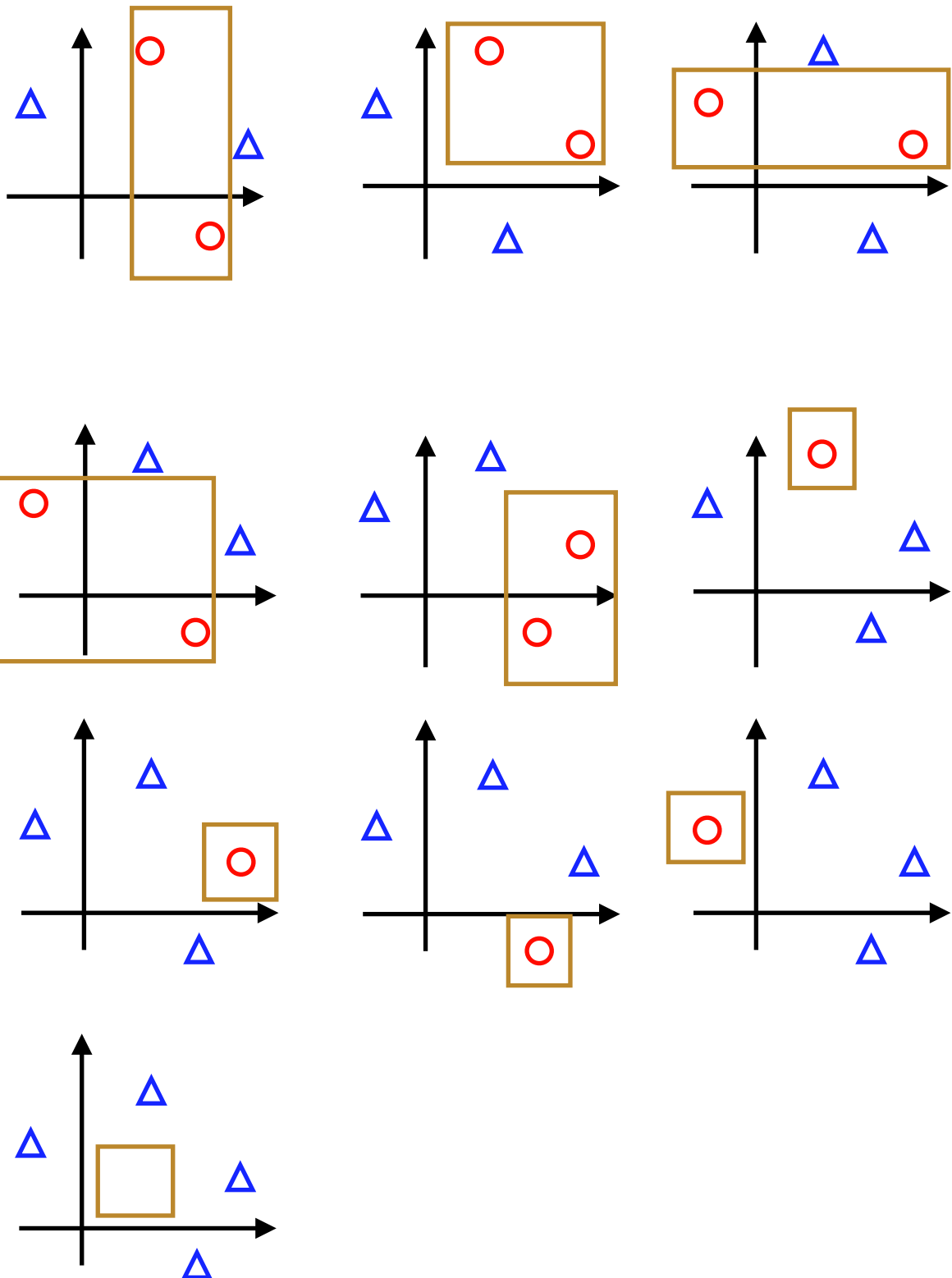
2. Show that the VC-Dimension of the hypothesis set is no less than 4.

Proof : show that we could shatter

$point(i) = P_i = (x_i, y_i), i = 1, \dots, 4$ , where  $x_1 < x_2 < x_3 < x_4$ , and  $y_1 \neq y_2 \neq y_3 \neq y_4$

Ex :





Since we could find  $2^4 = 16$  different dichotomies  
 $\Rightarrow$  We could shatter 4 inputs  
 $\Rightarrow$  break point  $\geq 5$   
 $\Rightarrow$   $dvc(\text{positive rectangle}) \geq 4$

# QED

3.

考慮  $h_\alpha(x) = -1$  的情況  $\iff |\alpha x \bmod 4 - 2| - 1 < 0$

$$\iff 1 < \alpha x \bmod 4 < 3$$

令  $x = (x_1, x_2, \dots, x_k)$ , 其中  $x_i = 4^i$

$$\alpha = \left(\frac{1}{4}\right)^1 a_1 + \left(\frac{1}{4}\right)^2 a_2 + \dots + \left(\frac{1}{4}\right)^i a_i + \dots + \left(\frac{1}{4}\right)^k a_k$$

$$\begin{aligned} \rightarrow \alpha x_i &= \left[\left(\frac{1}{4}\right)^1 a_1 + \left(\frac{1}{4}\right)^2 a_2 + \dots + \left(\frac{1}{4}\right)^i a_i + \dots + \left(\frac{1}{4}\right)^k a_k\right] 4^i \\ &= \left(\frac{1}{4}\right)^{1-i} a_1 + \dots + a_i + \left(\frac{1}{4}\right) a_{i+1} \dots + \left(\frac{1}{4}\right)^{k-i} a_k \end{aligned}$$

$$\text{其中, } \left(\frac{1}{4}\right) a_{i+1} \dots + \left(\frac{1}{4}\right)^{i-k} a_k \leq 2 * \left(\frac{\frac{1}{4}}{1 - \frac{1}{4}}\right) = \frac{2}{3}, \text{ when } k \rightarrow \infty$$

(assume  $a_{i+1}, \dots, a_k = 2$  here, since we could choose  $\alpha$ )

$$\rightarrow \alpha x_i \bmod 4 \leq a_i + \frac{2}{3}$$

$$\rightarrow \text{let } a_i = 0, \alpha x_i \bmod 4 \leq \frac{2}{3} \rightarrow |\alpha x_i \bmod 4 - 2| - 1 > 0 \rightarrow h_\alpha(x_i) = 1$$

$$\text{or } 2, \alpha x_i \bmod 4 \leq 2 + \frac{2}{3} \rightarrow |\alpha x_i \bmod 4 - 2| - 1 < 0 \rightarrow h_\alpha(x_i) = -1$$

$\therefore$  透過此  $\alpha$  之設計, 可以 shatter 任意數量的 input  $x$

$$\therefore dvc(H) = \infty$$

4. Show that *if*  $(H_1 \cap H_2) \neq \text{empty}$  , *then*  $dvc(H_1 \cap H_2) \leq dvc(H_1)$  .

Proof :

已知  $(H_1 \cap H_2) \subseteq H_1$

case 1 :  $dvc(H_1 \cap H_2)$  is finite

assume  $dvc(H_1) = n_1$ ,  $dvc(H_1 \cap H_2) = n_2$ , claim  $n_1 \geq n_2$

$\therefore \exists h_1, \dots, h_{n_2} \in (H_1 \cap H_2)$ , could shatter some of  $n_2$  inputs

$\therefore (H_1 \cap H_2) \subseteq H_1$

$\therefore h_1, \dots, h_{n_2} \in H_1$

$\therefore h_1, \dots, h_{n_2} \in H_1$ , could shatter some of  $n_2$  inputs

$\therefore dvc(H_1) = n_1 \geq n_2 = dvc(H_1 \cap H_2)$

case 2 :  $dvc(H_1 \cap H_2)$  is infinite

$\forall n$  input,  $\exists h_1, \dots, h_{2^n}$  could shatter them

$\therefore (H_1 \cap H_2) \subseteq H_1$

$\therefore h_1, \dots, h_{2^n} \in H_1$

$\therefore h_1, \dots, h_{2^n} \in H_1$ , could shatter some of  $2^n$  inputs

$\therefore H_1$  could shatter all number of inputs

$$5. m_{H_1 \cup H_2}(N) = 2N, \text{ dcv}(H_1 \cup H_2) = 2$$

One of solutions :

$\therefore H_1$  is positive ray and  $H_2$  is negative ray

$\therefore H_1 \cup H_2$  is one dimension perceptron

$$\therefore m_{H_1 \cup H_2}(N) = 2N$$

$$\therefore \text{dvc}(H_1 \cup H_2) = d + 1 = 2$$

6.

Derivation :

According to question 1 form Coursera, we know the formula which could calculate the probability of error that hypothesis makes in approximating the noisy target  $y$

$$E_{out} = \lambda\mu + (1 - \lambda)(1 - \mu) \quad - \quad (1)$$

,where  $(1 - \mu)$  is the probability that hypothesis predict the same value as target function do, and  $(1 - \lambda)$  is the noise rate of input.

So, the problem now is how could we use represent  $\lambda$  and  $\mu$  by  $\theta$  and  $s$

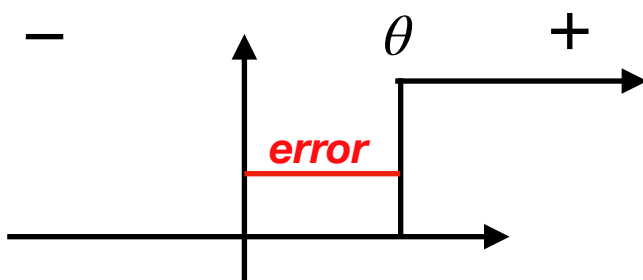
Now we try to give  $\theta$  and  $s$  some mathematical meanings

$\theta$  determines the threshold of this hypothesis, and

$s$  determines positive(+1) or negative(-1) ray it is

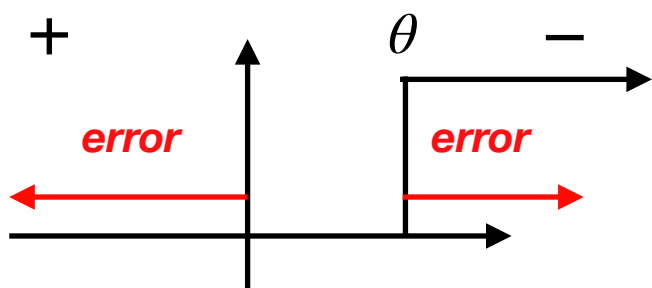
Discuss  $\lambda$  and  $\mu$  from 4 case:

case 1 :  $\theta > 0, s = 1$



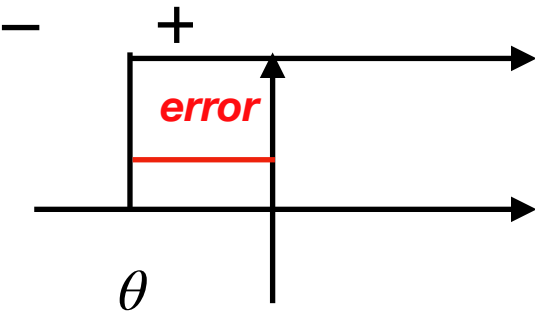
$$\text{error} = \theta/2$$

case 2 :  $\theta > 0, s = -1$



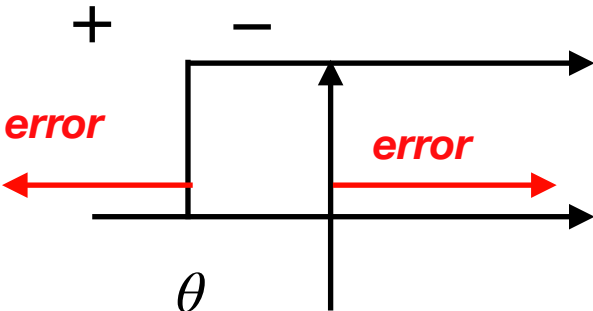
$$error = 1 - \theta/2$$

case 3 :  $\theta < 0, s = 1$



$$error = \theta/2$$

case 4 :  $\theta < 0, s = -1$



$$error = 1 - \theta/2$$

So, we could get the formula :

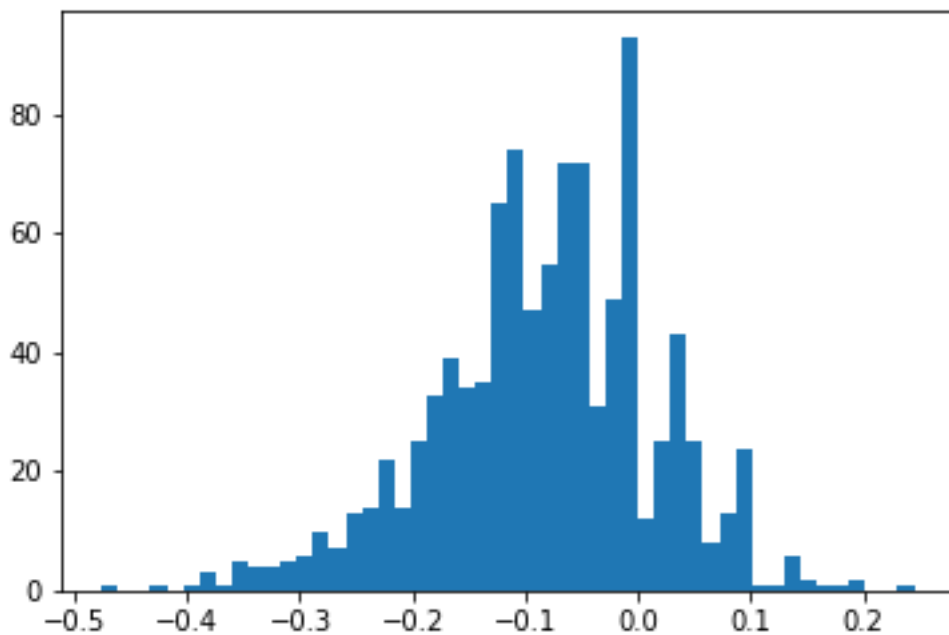
$$\mu = \left(\frac{1-s}{2}\right) + \left(s * \frac{\theta}{2}\right)$$

And  $(1 - \lambda) = 0.2$

Then we can modify formula(1)

$$\begin{aligned} E_{out} &= (1 - \mu)(1 - \lambda) + \mu\lambda \\ &= \left(\frac{1+s}{2} - \left(s * \frac{\theta}{2}\right)\right) * 0.2 + \left(\frac{1-s}{2} + \left(s * \frac{\theta}{2}\right)\right) * 0.8 \\ &= \frac{1}{2} + \frac{-0.6 * s}{2} + \frac{0.6 * s * \theta}{2} \\ &= \frac{1}{2} + \frac{0.6 * s}{2}(-1 + \theta) = 0.5 + 0.3s(\theta - 1) \end{aligned}$$

7.



(1.)  $E_{in} - E_{out} < 0$

Explain: 因為我們取hypothesis時是依照 $E_{in}$ 而不是 $E_{out}$ 挑選，因此挑出的hypothesis，平均多次之後有很大機率  $E_{in} < E_{out}$

(2.)

由觀察得知，此時大部分的 $E_{in} - E_{out}$ 都落在 $-0.2$ 到 $0.04$ 之間(平均值約落在 $-0.083$ )，根據上課提到的公式

$$P_D[|E_{in}(g) - E_{out}(g)| > \epsilon] \leq 4(2N)^{dc} \exp\left(-\frac{1}{8}\epsilon^2 N\right)$$

已知 $N = 20$ ,  $dvc = 2$

$$\text{觀察 } \epsilon = 0.05, 4(2N)^{dc} \exp\left(-\frac{1}{8}\epsilon^2 N\right) = 6360.1$$

$$\text{觀察 } \epsilon = 0.12, 4(2N)^{dc} \exp\left(-\frac{1}{8}\epsilon^2 N\right) = 6173.7$$

但實際上 $E_{out}$ 與 $E_{in}$ 已經相當靠近，發現理論數值與實際數值的差距，但若是根據老師上課所述，將 $N$ 放大1000倍(相當於需要的input size量減少1000倍)，再利用同樣的公式去計算

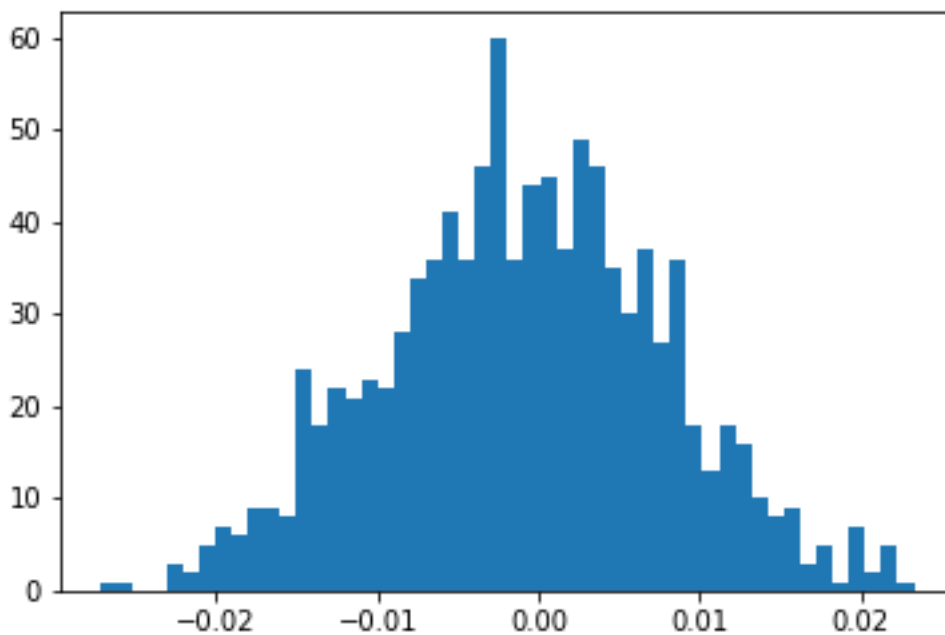


$$\epsilon = 0.05, 4(2N)^{dc} \exp\left(-\frac{1}{8}\epsilon^2 N\right) = 1.2355e + 07$$

$$\epsilon = 0.12, 4(2N)^{dc} \exp\left(-\frac{1}{8}\epsilon^2 N\right) = 1.4845e - 06$$

即當 $\epsilon = 0.05$ , 壞事情( $|E_{in}(g) - E_{out}(g)| > \epsilon$ )發生的機率已經非常大, 當 $\epsilon = 0.12$ , 壞事情( $|E_{in}(g) - E_{out}(g)| > \epsilon$ )發生的機率已經非常小, 雖然和真實數據仍有一定的差距, 但相對原本的公式已經合理許多。

8.



(1.)  $E_{in} - E_{out} < 0$

Explain: 因為我們取hypothesis時是依照 $E_{in}$ 而不是 $E_{out}$ 挑選, 因此挑出的hypothesis, 平均多次之後有很大機率  $E_{in} < E_{out}$

(2.) 可以發現當 $N$ 變大的時候, 在 $\epsilon$ 相同的情況下, 壞事情( $|E_{in}(g) - E_{out}(g)| > \epsilon$ )發生的機率變小, 此時大部分的 $E_{in} - E_{out}$ 都落在 $-0.02$ 到 $0.018$ 之間(平均值約落在 $-0.0008$ )

已知  $N = 2000$ ,  $d_{vc} = 2$  ,

$$\text{觀察 } \epsilon = 0.005, 4(2N)^{d_{vc}} \exp(-\frac{1}{8}\epsilon^2 N) = 6.3601e + 07$$

$$\text{觀察 } \epsilon = 0.012, 4(2N)^{d_{vc}} \exp(-\frac{1}{8}\epsilon^2 N) = 6.1737e + 07$$

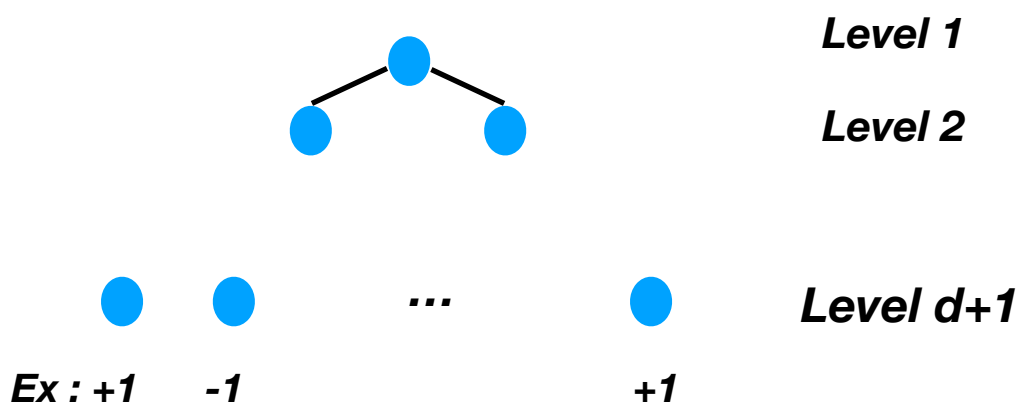
當  $N$  放大1000倍之後

$$\epsilon = 0.005, 4(2N)^{d_{vc}} \exp(-\frac{1}{8}\epsilon^2 N) = 1.2355e + 11$$

$$\epsilon = 0.012, 4(2N)^{d_{vc}} \exp(-\frac{1}{8}\epsilon^2 N) = 0.0148$$

結果合理許多，與Problem 7的結果相比較後，證明  $N$  的數量的確會影響壞事情發生的機率。

9. Decision tree looks like this



Since the dimension of  $S$  is  $d$ , the height of tree is  $d + 1$ , and this tree have  $2^d$  leaves. Each leaf could be or  $+1$  or  $-1$ , so at most we could have  $2^{2^d}$  dichotomies which is not influenced by input number.

對於decision tree的每一個level，我們可以看成在dimension  $i = t_i$ 上切一刀，令其分成兩個subregion， $x_i \leq t_i$ 代表decision tree的左子樹， $x_i > t_i$ ，代表decision tree右子樹

因為實數為countable infinity set，所以對於任意 $t_i \in R$ ，  
 $\exists t_i^l = t_i - 1 \in R$  and  $t_i^r = t_i + 1 \in R$

因此對於任意一個leave我們都可以根據decision tree的rule找到一個point  $\in$  該leave代表的subregion

$\therefore$  for each subregion  $s$ ，我們至少可以找到一個點 $\in s$   
 $\therefore$  我們可以找到 $2^{2^d}$ 種dichotomies

Then we try to find break point of  $H$ , which is the first number of input we could not shatter.

$$2^{2^d} < 2^N$$

$$\Rightarrow 2^d < N$$

$$\Rightarrow N = 2^d + 1$$

And we could calculate that  $dvc(H) = break\ point(H) - 1 = 2^d$