

Answer: 1~5 c d c b b
 6~10 d d d c c
 11~15 d b b d b
 16~20 d b e c a

ML foundation HW 2

Electrical Engineering
 B06504016 林家宏

Solution:

1 Let's write $\{(1, 1, 1, 3), (1, 7, 8, 9), (1, 15, 16, 17), (1, 21, 23, 25)\}$ to the matrix form.

$$\begin{matrix} \begin{bmatrix} 1 & 1 & 1 & 3 \\ 1 & 7 & 8 & 9 \\ 1 & 15 & 16 & 17 \\ 1 & 21 & 23 & 25 \end{bmatrix} & \begin{bmatrix} w_0 \\ w_1 \\ w_2 \\ w_3 \end{bmatrix} & = & \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} \\ X & W & & Y \end{matrix} \Rightarrow \text{The 4 points can be shattered if column vectors of } X \text{ span } \mathbb{R}^4.$$

The reduced row echelon form of X is $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$, whose

rank is 4. Hence, any $Y \in \mathbb{R}^4$ can be spanned by column vectors of X .

\Rightarrow The 4 points can be shattered by the 3D perceptron hypothesis set. \Rightarrow Choose (c)

2

Horizontal line (Excluding all 1 or all -1 case): $(N-1) \times 2$

Vertical line (Excluding all 1 or all -1 case): $(N-1) \times 2$

All y are 1: 1

All y are -1: 1

$$2(N-1) + 2(N-1) + 1 + 1 = 4N - 2 \Rightarrow \text{Choose (d)}$$

3

$y = \text{sign}(w_0 + w_1 x_1 + w_2 x_2)$, if $w_0 > 0$, it means the point $O(0, 0)$ should be always classified to 1. Hence, it just like adding 1 point $O(0, 0)$ to run perceptron and the point should be always assigned 1 as consequence.

In class, we are taught that 4 points cannot shattered by 2D perceptron hypothesis set. In this case, 3 normal points + 1 origin point $(0, 0)$ cannot be shattered, too. Therefore, the VC dimension = $3 - 1 = 2 \Rightarrow$ Choose (c)

$$4 \quad h(x) = \begin{cases} +1 & \text{if } \sqrt{a} \leq |x| \leq \sqrt{b} \\ -1 & \text{otherwise} \end{cases}$$

N points mean that we have $n+1$ location to choose two from it. It can determine the start and end of the ring. But there is still one case excluded, which is the all -1 case. Therefore, the growth function is $C_2^{n+1} + 1 \Rightarrow$ Choose (b).

$$5 \quad \text{The growth function } m_h(n) = \frac{n(n+1)}{2} + 1$$

$$n=2 \Rightarrow m_h(2) = \frac{2 \times 3}{2} + 1 = 4 = 2^2$$

$$n=3 \Rightarrow m_h(3) = \frac{3 \times 4}{2} + 1 = 7 \leq 2^3$$

3 is the break point, so 2 is VC dimension.

6 From the slide, we obtain

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

$$E_{in}(g^*) - \sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)} \leq E_{out}(g^*) \leq E_{in}(g^*) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)}$$

$$\Rightarrow E_{out}(g) \leq E_{in}(g) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)} \leq E_{in}(g^*) + \sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)} \leq E_{out}(g^*) + 2\sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)}$$

$$\Rightarrow E_{out}(g) \leq E_{out}(g^*) + 2\sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)}$$

$$\Rightarrow E_{out}(g) - E_{out}(g^*) \leq 2\sqrt{\frac{8}{N} \ln\left(\frac{4m_h(2N)}{\delta}\right)} \Rightarrow \text{choose (d)}$$

$E_{in}(g) \leq E_{in}(g^*)$ since g is the hypothesis minimizing E_{in} .

7 N points have 2^N combination. If the hypothesis set can shatter N points, $M \geq 2^N$.

$N = \lfloor \log_2 M \rfloor \Rightarrow d_{VC}(H) = \lfloor \log_2 M \rfloor$ because H at most shattered

$N = \lfloor \log_2 M \rfloor$ points.

8 $\{-1, 1\}^k$: The number of 1 range from 0 to k, k+1 possible combination in total. If the number of data exceeds k+1, the outcomes will always be the same since they have same number of 1 and -1 so that k+2 data cannot be shattered. Therefore, VC dimension is k+1.

9 ① Some set of d distinct inputs is shattered by H

\Rightarrow It means that $d_{VC}(H) \geq d$ (Definition)

② Some set of d+1 distinct inputs is not shattered by H.

\Rightarrow If $d_{VC}(H) = d$, any set of d+1 distinct is not shattered by H. It's definition. Hence, some set of d+1 distinct inputs is not shattered by H.

③ Any set of d+1 distinct inputs is not shattered by H.

\Rightarrow It means that $d_{VC}(H) < d+1$ (Definition)

3 conditions in total. \Rightarrow Choose (C).

10 Let's look at (C)

Let $\alpha = \pi(1 + \sum_{i=1}^m 2^i \frac{1-y_i}{2})$ over a set of points (x_1, x_2, \dots, x_m) with arbitrary labels $(y_1, y_2, \dots, y_m) \in \{-1, 1\}^n$

$$x_i = 2^{-i}$$

$$\alpha x_i = \pi(2^{-i} + \sum_{j=1}^m 2^{i-j} \frac{1-y_j}{2}) = \pi(2^{-i} + \sum_{j=1}^i 2^{i-j} \frac{1-y_j}{2} + \sum_{j=i+1}^m 2^{i-j} \frac{1-y_j}{2})$$

$$x_i = 2^{-i}$$

$$\Rightarrow \alpha x_i = \pi(2^{-i} + \sum_{j=1}^m 2^{i-j} \frac{1-y_j}{2}) \leq \pi(\sum_{j=0}^{i-1} 2^{i-j} + \frac{1-y_i}{2})$$

$$< \pi(1 + \frac{1-y_i}{2})$$

$$\alpha x_i > \pi \frac{1-y_i}{2} \Rightarrow \pi \frac{1-y_i}{2} < \alpha x_i < \pi(1 + \frac{1-y_i}{2})$$

\downarrow
drop since it only contributes multiples of 2π

for $y_i = -1 \Rightarrow \pi < \alpha x_i < 2\pi$, $y_i = +1 \Rightarrow 0 < \alpha x_i < \pi \Rightarrow$ Choose (C).

$$11 \quad E_{\text{out}}(h, z) = (1-\tau)E_{\text{out}}(h, 0) + \tau(1-E_{\text{out}}(h, 0)) = (1-2\tau)E_{\text{out}}(h, 0) + \tau$$

$$\Rightarrow E_{\text{out}}(h, 0) = \frac{E_{\text{out}}(h, z) - \tau}{1-2\tau}$$

$$12 \quad f(\vec{x}) = 1$$

$$(1-1)^2 \times 0.7 + (1-2)^2 \times 0.1 + (1-3)^2 \times 0.2 = 0.9$$

$$f(\vec{x}) = 2$$

$$(2-2)^2 \times 0.7 + (2-3)^2 \times 0.1 + (2-1)^2 \times 0.2 = 0.3$$

$$f(\vec{x}) = 3$$

$$(3-3)^2 \times 0.7 + (3-1)^2 \times 0.1 + (3-2)^2 \times 0.2 = 0.6 \quad E_{\text{out}}(f) = \frac{1}{3} \times 0.9 + \frac{1}{3} \times 0.3 + \frac{1}{3} \times 0.6 = 0.6$$

$$13 \quad f(\vec{x}) = 1$$

$$f_*(\vec{x}) = 0.7 + 2 \times 0.1 + 3 \times 0.2 = 1.5$$

$$f(\vec{x}) = 2$$

$$f_*(\vec{x}) = 0.2 + 2 \times 0.7 + 3 \times 0.1 = 1.9$$

$$f(\vec{x}) = 3$$

$$f_*(\vec{x}) = 0.1 + 2 \times 0.2 + 3 \times 0.7 = 2.6$$

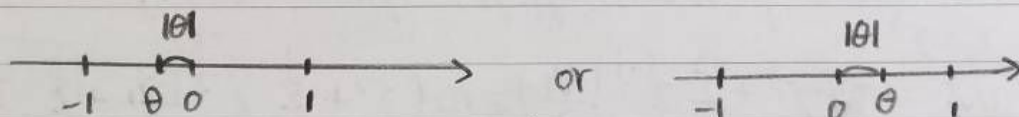
$$\Delta(f, f_*) = \frac{1}{3}(1-1.5)^2 + \frac{1}{3}(2-1.9)^2 + \frac{1}{3}(3-2.6)^2 = 0.14$$

$$14 \quad P[\text{BAD}] \leq 4m_n(2N) \exp(-\frac{1}{8}\epsilon^2 N) \quad m_n(2N) = 4N \quad \epsilon = 0.1$$

$$N = 11543 \quad P[\text{BAD}] = 0.10002 > 0.1 = \delta$$

$$N = 11544 \quad P[\text{BAD}] = 0.0999 < 0.1 = \delta$$

15 Hence, 12000 is the smallest N which can let $P[\text{BAD}] < 0.1$



In both cases, $E_{\text{out}}(h_{s,0}, 0) = \frac{|\theta|}{2}$ since $\text{sign}(x - \theta) \neq \text{sign } x$ between 0 and θ .

```

import math
import random
import numpy as np
def h(s, theta, x):
    if s * (x - theta) > 0:
        return 1
    else:
        return -1
def Ein(size, s, theta, data, data2):
    err = 0
    for i in range(size):
        if h(s, theta, data[i][0]) != data[i][1]:
            err = err + 1
    return err/size
data_size = 20
tou = 0
Eoutin = []
for j in range(10000):
    data = []
    data1 = []
    data2 = [-1]
    for i in range(data_size):
        x = random.uniform(-1, 1)
        data.append((x, h(1, 0, x)))
        data1.append(x)
    data.sort(key = lambda s: s[0])
    data1.sort()
    for i in range(data_size-1):
        data2.append((data1[i]+data1[i+1])/2)
    min_Ein = 1
    cur_s = 1
    cur_theta = -1
    for i in data2:
        s = -1
        if Ein(data_size, s, i, data, data2)<min_Ein:
            cur_s = s

```

```

        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
        s = 1
        if Ein(data_size, s, i, data, data2) < min_Ein:
            cur_s = s
            cur_theta = i
            min_Ein = Ein(data_size, s, i, data, data2)
        if cur_s == 1:
            E = abs(cur_theta)/2
        else:
            E = (2-abs(cur_theta))/2
        Eoutin.append(E - min_Ein)
average = 0
for i in Eoutin:
    average = average + i
average = average/10000
print('Eout-Ein = {}'.format(average))

```

17

```

import math
import random
import numpy as np
def h(s, theta, x):
    if s * (x - theta) > 0:
        return 1
    else:
        return -1
def Ein(size, s, theta, data, data2):
    err = 0
    for i in range(size):
        if h(s, theta, data[i][0]) != data[i][1]:
            err = err + 1
    return err/size
data_size = 20
tou = 0
Eoutin = []
for j in range(10000):
    data = []
    data1 = []

```

```

data2 = [-1]
for i in range(data_size):
    x = random.uniform(-1, 1)
    data.append((x, h(1, 0, x)))
    data1.append(x)
data.sort(key = lambda s: s[0])
data1.sort()
for i in range(data_size-1):
    data2.append((data1[i]+data1[i+1])/2)
min_Ein = 1
cur_s = 1
cur_theta = -1
for i in data2:
    s = -1
    if Ein(data_size, s, i, data, data2)<min_Ein:
        cur_s = s
        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
    s = 1
    if Ein(data_size, s, i, data, data2)<min_Ein:
        cur_s = s
        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
if cur_s == 1:
    E = abs(cur_theta)/2
else:
    E = (2-abs(cur_theta))/2
Eoutin.append(E - min_Ein)
average = 0
for i in Eoutin:
    average = average + i
average = average/10000
print('Eout-Ein = {}'.format(average))

```

18

```

import math
import random
import numpy as np
def h(s, theta, x):

```

```

    if s * (x - theta) > 0:
        return 1
    else:
        return -1
def Ein(size, s, theta, data, data2):
    err = 0
    for i in range(size):
        if h(s, theta, data[i][0]) != data[i][1]:
            err = err + 1
    return err/size
data_size = 2
tou = 0.1
Eoutin = []
for j in range(10000):
    data = []
    data2 = [-1]
    for i in range(data_size):
        x = random.uniform(-1, 1)
        y = random.uniform(0, 1)
        if y < tou:
            data.append((x, -h(1, 0, x)))
        else:
            data.append((x, h(1, 0, x)))
    data.sort(key = lambda s: s[0])
    for i in range(data_size-1):
        data2.append((data[i][0]+data[i+1][0])/2)
    min_Ein = 1
    cur_s = 1
    cur_theta = -1
    for i in data2:
        s = -1
        if Ein(data_size, s, i, data, data2)<min_Ein:
            cur_s = s
            cur_theta = i
            min_Ein = Ein(data_size, s, i, data, data2)
        s = 1
        if Ein(data_size, s, i, data, data2)<min_Ein:
            cur_s = s

```



```

        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
    if cur_s == 1:
        E = abs(cur_theta)/2
    else:
        E = (2-abs(cur_theta))/2
    Eoutin.append(E*(1-2*tou) + tou - min_Ein)
average = 0
for i in Eoutin:
    average = average + i
average = average/10000
print('Eout-Ein = {}'.format(average))

```

19

```

import math
import random
import numpy as np
def h(s, theta, x):
    if s * (x - theta) > 0:
        return 1
    else:
        return -1
def Ein(size, s, theta, data, data2):
    err = 0
    for i in range(size):
        if h(s, theta, data[i][0]) != data[i][1]:
            err = err + 1
    return err/size
data_size = 20
tou = 0.1
Eoutin = []
for j in range(10000):
    data = []
    data1 = []
    data2 = [-1]
    for i in range(data_size):
        x = random.uniform(-1, 1)
        y = random.random()
        if y < tou:

```

```

        data.append((x, -h(1, 0, x)))
    else:
        data.append((x, h(1, 0, x)))
    data1.append(x)
data.sort(key = lambda s: s[0])
data1.sort()
for i in range(data_size-1):
    data2.append((data1[i]+data1[i+1])/2)
min_Ein = 1
cur_s = 1
cur_theta = -1
for i in data2:
    s = -1
    if Ein(data_size, s, i, data, data2)<min_Ein:
        cur_s = s
        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
    s = 1
    if Ein(data_size, s, i, data, data2)<min_Ein:
        cur_s = s
        cur_theta = i
        min_Ein = Ein(data_size, s, i, data, data2)
if cur_s == 1:
    E = abs(cur_theta)/2
else:
    E = (2-abs(cur_theta))/2
Eoutin.append(E*(1-2*tou) + tou - min_Ein)
average = 0
for i in Eoutin:
    average = average + i
average = average/10000
print('Eout-Ein = {}'.format(average))

```

20

```

import math
import random
import numpy as np
def h(s, theta, x):
    if s * (x - theta) > 0:

```

```

        return 1
    else:
        return -1
def Ein(size, s, theta, data, data2):
    err = 0
    for i in range(size):
        if h(s, theta, data[i][0]) != data[i][1]:
            err = err + 1
    return err/size
data_size = 200
tou = 0.1
Eoutin = []
for j in range(10000):
    data = []
    data2 = [-1]
    for i in range(data_size):
        x = random.uniform(-1, 1)
        y = random.random()
        if y < tou:
            data.append((x, -h(1, 0, x)))
        else:
            data.append((x, h(1, 0, x)))
    data.sort(key = lambda s: s[0])
    for i in range(data_size-1):
        data2.append((data[i][0]+data[i+1][0])/2)
    min_Ein = 1
    cur_s = 1
    cur_theta = -1
    for i in data2:
        s = -1
        if Ein(data_size, s, i, data, data2)<min_Ein:
            cur_s = s
            cur_theta = i
            min_Ein = Ein(data_size, s, i, data, data2)
        s = 1
        if Ein(data_size, s, i, data, data2)<min_Ein:
            cur_s = s
            cur_theta = i

```

```
        min_Ein = Ein(data_size, s, i, data, data2)
    if cur_s == 1:
        E = abs(cur_theta)/2
    else:
        E = (2-abs(cur_theta))/2
    Eoutin.append(E*(1-2*tou) + tou - min_Ein)
average = 0
for i in Eoutin:
    average = average + i
average = average/10000
print('Eout-Ein = {}'.format(average))
```