

## Problem Set 5

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## 1. Answer to problem 1

(a) 1,  $\mathbf{W} = [-1, 1]^T$   
 $\theta = 0$

2,  $\mathbf{W} = [-0.5, 0.25]^T$   
 $\theta = 0$

3, For SVM method in this problem, we need to find three lines that

$$w_1x_1 + w_2x_2 + \theta = 0$$

$$w_1x_1 + w_2x_2 + \theta = -1$$

$$w_1x_1 + w_2x_2 + \theta = 1$$

The first step is to find the two support vectors. Considering all the points from the table, it is easy to conclude that **(-1.2, 1.6, (+1))** and **(2, 0, (-1))** has the least distance. So, we can derive:

$$-1.2w_1 + 1.6w_2 + \theta = 1$$

$$2w_1 + 0w_2 + \theta = -1$$

Through these two questions, we can get that:  $\theta = -2w_1 - 1$  and  $w_2 = 2w_1 + 1.25$ . So, we can derive that

$$w_1^2 + w_2^2 = 5w_1^2 + 5w_1 + 1.25^2$$

so, in order to get the least value for  $w_1^2 + w_2^2$ , we need  $\mathbf{w}_1 = -0.5$ , so  $\mathbf{w}_2 = 0.25$  and  $\theta = 0$ .

So, we need that:  $\mathbf{W} = [-0.5, 0.25]^T$ ,  $\theta = 0$ .

(b) 1,  $\mathbf{I} = \{1, 6\}$  or  $\mathbf{I} = \{(-1.2, 1.6), (2, 0)\}$

2,  $\alpha = \{\frac{5}{32}, \frac{5}{32}\}$

3,  $objectivefunctionvalue = \frac{5}{32}$

- (c) C controls the tradeoff between large margin (small  $\|w\|$ ) and small loss. When C is larger, we concerned more on the mistakes. When C is smaller, we concerned more on larger margin.

- 1, When  $C = \infty$ , we need  $\xi_i$  to be 0, so, when  $C = \infty$ , we get the same answer in (a)-2.
- 2, When  $C = 1$ , we will make more misclassifications, but we can also have a larger margin.
- 3, When  $C = 0$ , we will concerned more on the larger margin. So, we have larger margin, but we will also make more misclassifications.

## 2. Answer to problem 2

### (a) Dual Perceptron:

- 1, Initialize  $\alpha$  and  $\theta$  to zero vectors
- 2, For each example  $(x_m, y_m)$ :  
if  $y_m \sum_{i=1}^m \alpha_i x_i x_m < 0$ ,  $\alpha_m = \alpha_m + 1$
- 3, Output the final result  $w$

- (b) Suppose that:  $\vec{x} = (x_1, x_2)^T$  and  $\vec{z} = (z_1, z_2)^T$ , so

$$K(\vec{x}, \vec{z}) = (\vec{x}^T \vec{z})^3 + 400(\vec{x}^T \vec{z})^2 + 100(\vec{x}^T \vec{z})$$

can be written as:

$$K(\vec{x}, \vec{z}) = (x_1 z_1 + x_2 z_2)^3 + 400(x_1 z_1 + x_2 z_2)^2 + 100(x_1 z_1 + x_2 z_2)$$

**First**, let's prove that  $K_1 = (x_1 z_1 + x_2 z_2)^3$  is a valid kernel.

$$K_1 = x_1^3 z_1^3 + 3x_1^2 x_2 z_1^2 z_2 + 3x_1 x_2^2 z_1 z_2^2 + x_2^3 z_2^3$$

$$K_1 = (x_1^3, \sqrt{3}x_1^2 x_2, \sqrt{3}x_1 x_2^2, x_2^3)(z_1^3, \sqrt{3}z_1^2 z_2, \sqrt{3}z_1 z_2^2, z_2^3)^T = \psi_1(\vec{x})^T \psi_1(\vec{z})$$

**So,  $K_1$  is a valid kernel.**

**Now**, let's prove that  $K_2 = 400(x_1 z_1 + x_2 z_2)^2$  is a valid kernel.

$$K_2 = 400(x_1^2 z_1^2 + 2x_1 x_2 z_1 z_2 + x_2^2 z_2^2)$$

$$K_2 = 20(x_1^2, \sqrt{2}x_1 x_2, x_2^2)(z_1^2, \sqrt{2}z_1 z_2, z_2^2)^T = \psi_2(\vec{x})^T \psi_2(\vec{z})$$

**So,  $K_2$  is valid kernel.**

**Finally**,  $K_3 = 100(x_1 z_1 + x_2 z_2) = 10(x_1, x_2)10(z_1, z_2)^T = \psi_3(\vec{x})^T \psi_3(\vec{z})$  is also a kernel.

Since,  $K_1$ ,  $K_2$  and  $K_3$  are all valid kernels, so

$$K(\vec{x}, \vec{z}) = K_1 + K_2 + K_3 = \psi_1(\vec{x})^T \psi_1(\vec{z}) + \psi_2(\vec{x})^T \psi_2(\vec{z}) + \psi_3(\vec{x})^T \psi_3(\vec{z})$$

$$K(\vec{x}, \vec{z}) = [\psi_1(\vec{x}) \psi_2(\vec{x}) \psi_3(\vec{x})][\psi_1(\vec{z}) \psi_2(\vec{z}) \psi_3(\vec{z})]^T$$

So,  $\mathbf{K}(\tilde{\mathbf{x}}, \tilde{\mathbf{z}})$  is also valid kernels.

### 3. Answer to problem 3

$i$	Label	Hypothesis 1				Hypothesis 2			
		$D_0$	$x_1 \equiv [\mathbf{x} > \mathbf{5}]$	$x_2 \equiv [\mathbf{y} > \mathbf{6}]$	$h_1 \equiv [\mathbf{x}_1]$	$D_1$	$x_1 \equiv [\mathbf{x} > \mathbf{3}]$	$x_2 \equiv [\mathbf{y} > \mathbf{8}]$	$h_2 \equiv [\mathbf{x}_2]$
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
1	—	1/10	—	+	—	1/16	—	+	+
2	—	1/10	—	—	—	1/16	+	—	—
3	+	1/10	+	+	+	1/16	+	—	—
4	—	1/10	—	—	—	1/16	+	—	—
5	—	1/10	—	+	—	1/16	—	+	+
6	+	1/10	+	+	+	1/16	+	—	—
7	+	1/10	+	+	+	1/16	+	+	+
8	—	1/10	—	—	—	1/16	+	—	—
9	+	1/10	—	+	—	1/4	+	+	+
10	—	1/10	+	+	+	1/4	+	—	—

Table 1: Table for Boosting results

- (a) There are 10 samples in total, so  $m = 10$ , in this way,  $D_0 = 1/m = 0.1$ .  
In order to reduce the error rate, through comparison, finally we find that for  $x$  and  $y$ , when we choose  $x > 5$  and  $y > 6$ , the error rate will be smallest, which are 0.2 and 0.3. So,  $\mathbf{x}_1 = [\mathbf{x} > \mathbf{5}]$  and  $\mathbf{x}_2 = [\mathbf{y} > \mathbf{6}]$ .
- (b) Since error rate is 0.2 and 0.3, we should choose  $\mathbf{x}_1 = [\mathbf{x} > \mathbf{5}]$  as the first hypothesis  $\mathbf{h}_1 = [\mathbf{x}_1]$ . The prediction are shown in tables.
- (c) From (a), we know that error rate  $\epsilon = 0.2$ . According to

$$\alpha = 0.5 * \ln[(1 - \epsilon)/\epsilon]$$

we can get that  $\alpha = \ln 2 = 0.693$ . According to

$$z_t = \sum D_t \exp(-\alpha_t y h(x))$$

When  $y_i = h(x_i)$ ,  $D_t \exp(-\alpha_t y h(x)) = 1/10 * \exp(-\ln 2) = 0.05$ ,

When  $y_i \neq h(x_i)$ ,  $D_t \exp(-\alpha_t y h(x)) = 1/10 * \exp(\ln 2) = 0.2$ .

So,

$$z_t = \sum D_t \exp(-\alpha_t y h(x)) = 8 * 0.05 + 2 * 0.2 = 0.8$$

Then, according to  $D_{t+1} = D_t / z_t * \exp(-\alpha_t y h(x))$ ,

**when**  $y_i = h(x_i)$ ,

$$D_1 = 0.1 / 0.8 * \exp(-\ln 2) = 1/16$$

**when**  $y_i \neq h(x_i)$ ,

$$D_1 = 0.1 / 0.8 * \exp(\ln 2) = 1/4$$

So, we can fill in the Table as shown above.

- (d) First, let's calculate  $\alpha_2$ . For  $h_2 = x_2 = [y > 8]$ , error rate  $\epsilon = 4/16 = 1/4$ , so, according to

$$\alpha = 0.5 * \ln[(1 - \epsilon)/\epsilon]$$

we can get that

$$\alpha_2 = 0.5 * \ln[(1 - 0.25)/0.25] = 1/2 * \ln 3 = 0.549$$

So,  $\alpha_1 = \mathbf{0.693}$ ,  $\alpha_2 = \mathbf{0.549}$

According to the above analysis, we can write the final hypothesis as follows:

$$\mathbf{h} = \text{sign}(\sum \alpha_t \mathbf{h}_t) = \text{sign}[\mathbf{0.693}(\mathbf{x} > \mathbf{5}) + \mathbf{0.549}(\mathbf{y} > \mathbf{8})]$$

#### 4. Answer to problem 4

- (a) i. The expected number of children in towns A is 1 and 2 in towns B

**Proof:**

In **town A**, each family has just one child, so the expected number of children in towns A is 1.

In **town B**, the probability of a family has k children is:

$$P(k) = \left(\frac{1}{2}\right)^{k-1} \frac{1}{2} = \left(\frac{1}{2}\right)^k$$

so the expected number of children will be:

$$N = \sum_1^{\infty} k \left(\frac{1}{2}\right)^k = 2$$

So, the expected number of children in towns B is **2**.

ii. The boy to girl ratio in towns A is 1 : 1 and 1 : 1 in towns B

**Proof:**

In **town A**, at first the boy to girl ratio is 1:1. So, the probability of boy and girl is 0.5 and 0.5. Now, when they have children, they will have only one child, with the probability of 0.5 to be a boy and 0.5 to be a girl. So, at the end of the first generation, the probability of boy will be:  $0.5 * 0.5 * 2 = 0.5$  and the probability of girl will be:  $0.5 * 0.5 * 2 = 0.5$ . So, the boy to girl ratio in towns A is **1 : 1**.

In **town B**, at first the boy to girl ratio is 1:1. So, the probability of boy and girl is 0.5 and 0.5. Now, for each family, there be definitely one boy, so the number of boy for each family will be: 1. For girls, the expected number of girls will be:

$$\sum_1^{\infty} k \left(\frac{1}{2}\right)^{k+1} = 1$$

So, after one generation, the expected number of boys will be:  $0.5 * 1 = 0.5$ . and the expected number of girls will be:  $0.5 * 1 = 0.5$ . So the boy to girl ratio in towns B is **1 : 1**.

(b) i. **Proof:**

$$P(A|B) = \frac{P(A, B)}{P(B)}$$

Since  $P(A, B) = P(A)P(B|A)$ , so

$$\frac{P(A, B)}{P(B)} = \frac{P(A)P(B|A)}{P(B)}$$

So,

$$\mathbf{P(A|B)} = \frac{\mathbf{P(B|A)P(A)}}{\mathbf{P(B)}}$$

ii.

$$P(A, B, C) = P(A|B, C)P(B, C)$$

Since

$$P(B, C) = P(B|C)P(C)$$

So:

$$P(A, B, C) = P(A|B, C)P(B, C) = P(A|B, C)P(B|C)P(C)$$

So:

$$\mathbf{P(A, B, C)} = \mathbf{P(A|B, C)P(B|C)P(C)}$$

**(c) Proof:**

First, let's calculate the probability distribution of X. Suppose that for n events, A events will occur for m times. So,

$$P(A) = \frac{m}{n}$$

And, it is clear that:

$$P(X = 1) = \frac{m}{n} \quad P(X = 0) = \frac{n - m}{n}$$

So,

$$E(X) = P(X = 1) * 1 + P(X = 0) * 0 = P(X = 1) = \frac{m}{n}$$

So,

$$\mathbf{E(X) = P(A)}$$

**(d)** First, let's calculate the distribution for (X, Y). The result are shown in following Table 2:

	$X = 0$	$X = 1$	
$Y = 0$	30/90	18/90	48/90
$Y = 1$	25/90	17/90	42/90
	55/90	35/90	1

Table 2: T(X, Y) Distribution

**i. X is not independent of Y.**

**Reason:** If X is independent of Y, we should have:  $P(X, Y) = P(X)P(Y)$ . So, for example, from Table 2, we can know that:  $P(X = 1, Y = 1) = 18/90$ , and  $P(X = 1) = 35/90$ ,  $P(Y = 1) = 42/90$ . It is clear that  $P(X, Y) \neq P(X)P(Y)$ .

So, X is not independent of Y.

**ii. X is conditionally independent of Y given Z**

**Reason:** According to Table 3:

	$Z = 0$		$Z = 1$	
	$X = 0$	$X = 1$	$X = 0$	$X = 1$
$Y = 0$	1/15	1/15	4/15	2/15
$Y = 1$	1/10	1/10	8/45	4/45

Table 3: Table for (X, Y) Distribution

We can get that:

$$P(Z = 0) = 30/90, P(X = 0|Z = 0) = 15/30, P(X = 1|Z = 0) = 15/30, \\ P(Y = 0|Z = 0) = 12/30, P(Y = 1|Z = 0) = 18/30,$$

So,

$$\mathbf{P(X, Y|Z = 0) = P(X|Z = 0)P(Y|Z = 0)}$$

Similarly,

$$P(Z = 1) = 60/90, P(X = 0|Z = 1) = 40/60, P(X = 1|Z = 1) = 20/60, \\ P(Y = 0|Z = 1) = 36/60, P(Y = 1|Z = 1) = 24/60,$$

So,

$$\mathbf{P(X, Y|Z = 1) = P(X|Z = 1)P(Y|Z = 1)}$$

In conclusion, **X is conditionally independent of Y given Z.**

iii. From Table 2, we can know that:

$$\{X + Y > 0\} = \{(X = 0, Y = 1), (X = 1, Y = 0), (X = 1, Y = 1)\}$$

So,

$$P(X = 0|X+Y > 0) = P(X = 0|\{(X = 0, Y = 1), (X = 1, Y = 0), (X = 1, Y = 1)\})$$

So

$$P(X = 0|X+Y > 0) = \frac{P(X = 0, X + Y > 0)}{P(\{(X = 0, Y = 1), (X = 1, Y = 0), (X = 1, Y = 1)\})}$$

And

$$P(X = 0, X + Y > 0) = P(X = 0, Y = 1) = 25/90$$

$$P(\{(X = 0, Y = 1), (X = 1, Y = 0), (X = 1, Y = 1)\}) = 25/90 + 18/90 + 17/90 = 60/90$$

So

$$P(X = 0|X + Y > 0) = \frac{25/90}{60/90} = 5/12$$

So,

$$\mathbf{P(X = 0|X + Y > 0) = 5/12}$$