

## Question 1

### Using Kernel 0

```
C:\Users\nigel\OneDrive\Desktop\Code\libsvm-3.24\windows>svm-predict.exe test.txt training.txt.model training.out
Accuracy = 84.375% (27/32) (classification)
```

### Using Kernel 1

```
C:\Users\nigel\OneDrive\Desktop\Code\libsvm-3.24\windows>svm-predict.exe test.txt training.txt.model training.out
Accuracy = 81.25% (26/32) (classification)
```

### Using Kernel 2

```
C:\Users\nigel\OneDrive\Desktop\Code\libsvm-3.24\windows>svm-predict.exe test.txt training.txt.model training.out
Accuracy = 90.625% (29/32) (classification)
```

### Using Kernel 3

```
C:\Users\nigel\OneDrive\Desktop\Code\libsvm-3.24\windows>svm-predict.exe test.txt training.txt.model training.out
Accuracy = 43.75% (14/32) (classification)
```

## Question 2

a)

Referring to the equation, we want to minimize  $0.5 \|w\|^2$ , which implies we should minimize  $w$ . Using the constraint that it is subjected to, to minimize  $w$  we have:

$$y(w \cdot x + w_0) = 1$$

as bias  $b = 0$ ,

$$y(w \cdot x) = 1$$

Using the points given we have and setting our  $w$  as  $[w_1 \ w_2]$ :

For  $x_1 = [1 \ 1]$  transpose and  $y_1 = 1$ :

$$1 * [w_1 \ w_2] \cdot [1 \ 1]^T = 1$$

$$w_1 + w_2 = 1$$

For  $x_2 = [1 \ 0]$  transpose and  $y_2 = -1$ :

$$(-1) * [w_1 \ w_2] \cdot [1 \ 0].T = 1$$

$$w_1 = -1$$

Therefore we get  $w_1 = -1$ ,  $w_2 = 2$

Thus our  $w$  is  **$[-1 \ 2]$**

To get  $\gamma$ , we use  $\gamma(w \cdot x + w_0) / \|w\|$

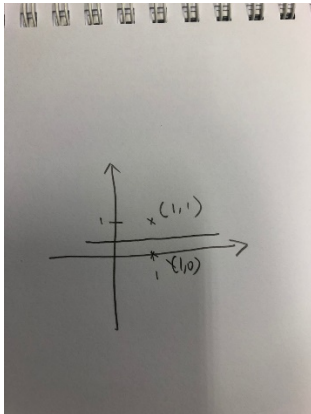
For  $x_1 = [1 \ 1]$  transpose and  $y_1 = 1$ :

$$1 * ([ -1 \ 2] \cdot [1 \ 1].T + 0) / (-1^2 + 2^2)^{0.5}$$

$$\text{Hence, } \gamma = 1/5^{0.5}$$

b)

If bias  $b$  is allowed to be non-zero, this would allow the classifier to be optimized as it is no longer restricted and results in optimal  $\gamma$  and  $w$  values. Given this, since the support vectors  $x_1$  and  $x_2$  are exactly 1 unit apart as seen in the following drawing (sorry I was rushing and could not plot a better one so I drew it out PS), we can find the best fit line as shown:



The margin  $\gamma$  can hence be deduced to be  $\frac{1}{2}$

Hence,  $\gamma=0.5$

Using the above graph, to calculate the gradient of the line we can take 2 points A(0, 0.5) and B(1, 0.5). in this case,  $dy/dx$  would be 0/1.

For every change in  $dx$  by 1,  $dy$  would change by 0. Resulting in the tangent vector being [1 0]. To get  $w$ , we can find the normal of this vector which would be [0 1], hence

$w = [0 \ 1]$

### Question 3

## Kernel Methods

#### • Properties of kernel functions

1.  $K(x, x') = 1$  is a kernel function.
2. Let  $f : \mathcal{R}^d \rightarrow \mathcal{R}$  be any real valued function of  $x$ . Then, if  $K(x, x')$  is a kernel function, then so is  $\tilde{K}(x, x') = f(x)K(x, x')f(x')$
3. If  $K_1(x, x')$  and  $K_2(x, x')$  are kernels, then so is their sum. In other words,  $K(x, x') = K_1(x, x') + K_2(x, x')$  is a kernel.
4. If  $K_1(x, x')$  and  $K_2(x, x')$  are kernels, then so is their product  $K(x, x') = K_1(x, x')K_2(x, x')$

The above is reference for answering the following questions.

1.  $K(x, z) = K_1(x, z)K_2(x, z)$  is a Kernel

#### **Proof:**

Using the 4<sup>th</sup> property from the above and given that:

$K_1(x, z)$  and  $K_2(x, z)$  are kernels over  $\mathbb{R}^n \times \mathbb{R}^n$ .

since  $K_1(x, z)$  and  $K_2(x, z)$  are Kernels, their product  $K(x, z)$  will also be a Kernel.

2.  $K(x; z) = aK_1(x; z) + bK_2(x; z)$ , where  $a; b > 0$  are real numbers, is also considered a Kernel.

**Proof:**

Using the 3<sup>rd</sup> property from above and given that:

$K_1(x; z)$  and  $K_2(x; z)$  are kernels over  $R_n \times R_n$ .

As  $aK_1(x, z)$  and  $bK_2(x, z)$  are Kernels and  $a, b$  are real numbers greater than 0, the summation of these two Kernels which results in  $K(x, z)$  is a Kernel as well.

3.  $K(x; z) = aK_1(x; z) - bK_2(x; z)$ , where  $a; b > 0$  are real numbers

In this case,  $K(x, z)$  is not a kernel as there is the potential for it to be less than 0. Given that **Kernels have the property of being non-negative**, the next part will be to prove that this is indeed not a Kernel

**Proof:**

As  $a$  and  $b$  are given to be real numbers that are greater than 0, we can take any non negative values. In this case lets take  **$a = 9$**  and  **$b = 17$** .

Also given that

$K_1(x; z)$  and  $K_2(x; z)$  are kernels over  $R_n \times R_n$ .

Using the 1<sup>st</sup> property in the properties list above,  $K_1(x, z)$  and  $K_2(x, z)$  can be taken to be equal to 1.

Plugging these values into the given equation of:

$$K(x, z) = aK_1(x, z) - bK_2(x, z)$$

We get

$$K(x, z) = (9 * 1) - (17 * 1)$$

This would result in  $K(x, z) = -8$  which is a value lesser than 0.

As this does not follow the property of non-negativity, we can prove that in this case,  $K$  is not a Kernel.

4.