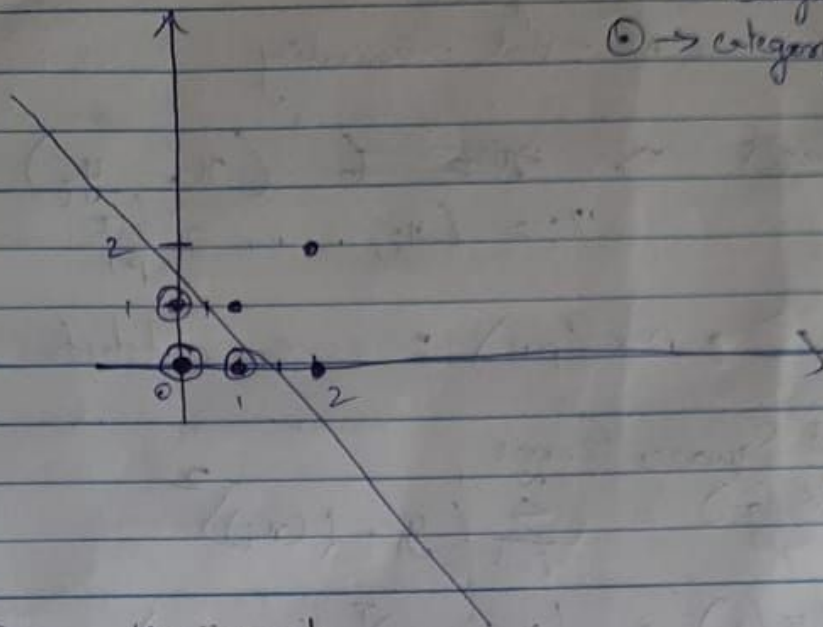


Tutorial 7

1. a)

•  $\rightarrow$  category 1⊙  $\rightarrow$  category 2

From the graph,  
hyperplane passes through  $(0.5, 1), (1.5, 0)$

$$w = [1, 1] \quad \text{bias} = -1.5$$

b) Support vectors  $\Rightarrow (0, 1), (1, 1), (1, 0), (2, 0)$

c) Separable hyperplane depends only on the support vectors. As  $D$  is not a support vector, deleting it won't change solution.

2. a) TRUE

As Gaussian kernel can be expanded using Taylor series expansion. It can be seen as basis exp with as dimensions as Gaussian using Taylor series has as terms.  
Hence due to  $\infty$  dimensions it can model any complex separating hyperplane.

2. b) FALSE

For Gaussian kernel, there is no closed form basis expansion.

2. c) TRUE

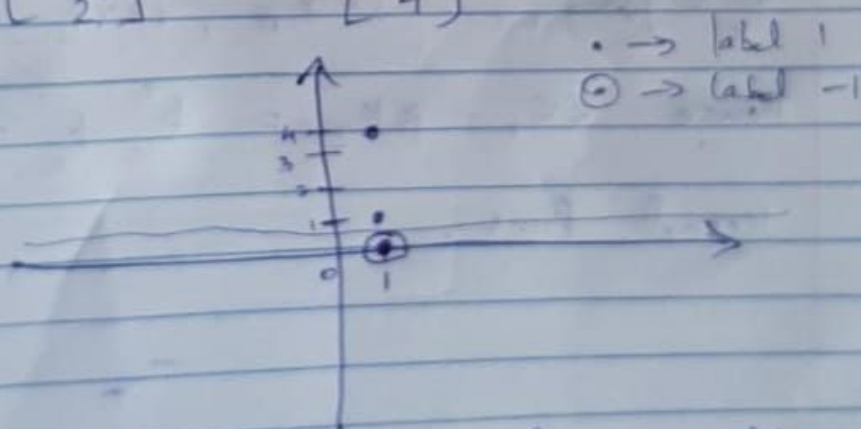
As separator depends on support vectors, more the num supp vecs, more overfitting.

$$3. a) K = \Phi^T \Phi = \begin{bmatrix} 1 & x^2 \end{bmatrix} \begin{bmatrix} 1 \\ x^2 \end{bmatrix} = 1 + x^4$$

3. b) Using  $\phi(x) = \begin{bmatrix} 1 & x^2 \end{bmatrix}$ ,  
the pts become,

$$x_1' = \begin{bmatrix} 1 \\ (-1)^2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad x_2' = \begin{bmatrix} 1 \\ 0^2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$x_3' = \begin{bmatrix} 1 \\ 2^2 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \end{bmatrix}$$



From graph, it is linearly separable.



3.c) Let  $K_1 = \phi_1^T \phi_1$  and  $K_2 = \phi_2^T \phi_2$

$$\therefore \alpha K_1 = (\sqrt{\alpha} \phi_1^T) (\sqrt{\alpha} \phi_1)$$

$$\beta K_2 = (\sqrt{\beta} \phi_2^T) (\sqrt{\beta} \phi_2)$$

$$\therefore \alpha K_1 + \beta K_2 = (\sqrt{\alpha} \phi_1^T) (\sqrt{\alpha} \phi_1) + (\sqrt{\beta} \phi_2^T) (\sqrt{\beta} \phi_2)$$

$$= \phi_K^T \phi_K$$

This is possible by forming  $\phi_K$  by

Simply putting the elements of  $\sqrt{\alpha} \phi_1$  and  $\sqrt{\beta} \phi_2$  together.

If elements of  $\phi_1$  are  $[\phi_{11}, \phi_{12}, \dots, \phi_{1n}]$   
and  $\phi_2$  are  $[\phi_{21}, \phi_{22}, \dots, \phi_{2m}]$ ,

$$\phi_K = [\sqrt{\alpha} \phi_{11}, \dots, \sqrt{\alpha} \phi_{1n}, \sqrt{\beta} \phi_{21}, \dots, \sqrt{\beta} \phi_{2m}]$$

Hence  $\alpha K_1 + \beta K_2 = \phi_K^T \phi_K \Rightarrow$  another

kernel func.