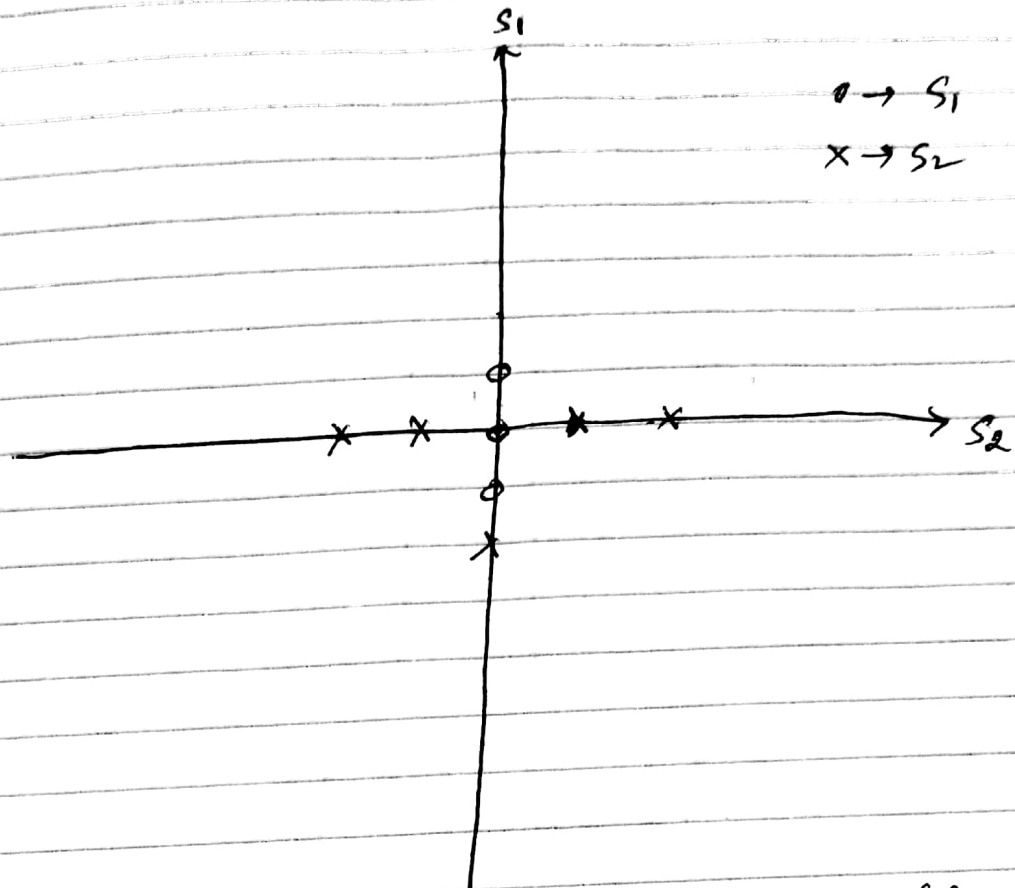


## Problem-2

a)



No, they are not linearly separable.

$$\underline{u}_0 = \phi(\underline{x}) = \begin{bmatrix} 1 \\ x_1 \\ x_2 \\ x_1 x_2 \\ x_1^2 \\ x_2^2 \end{bmatrix} \quad \begin{bmatrix} (D^2+1)x_1 \\ (D^2)x_1 \end{bmatrix}$$

$$\begin{aligned} \text{Total no. of terms} &= \frac{D^2 + 3D}{2} + 1 = \frac{2^2 + 2 \times 3}{2} + 1 \\ &= \frac{4 + 6}{2} + 1 = 6 \end{aligned}$$

b) Applying the  $\phi(x)$  function,

datapoint  
(0,0)  $u_1 = \phi(x_1) = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$

datapoint  
(0,1)  $u_2 = \phi(x_2) = \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$

datapoint  
(0,-1)  $u_3 = \phi(x_3) = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$

datapoint  
(-2,0)  $u_4 = \phi(x_4) = \begin{bmatrix} 1 \\ -2 \\ 0 \\ 4 \\ 0 \\ 0 \end{bmatrix}$

datapoint  
(-1, 0)

$$u_5 = \phi(\underline{x}_5) = \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

datapoint  
(0, 1)

$$u_6 = \phi(\underline{x}_6) = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 2 \\ 0 \\ 4 \end{bmatrix}$$

datapoint  
(0, -2)

$$u_7 = \phi(\underline{x}_7) = \begin{bmatrix} 1 \\ 0 \\ -2 \\ 0 \\ 0 \\ 4 \end{bmatrix}$$

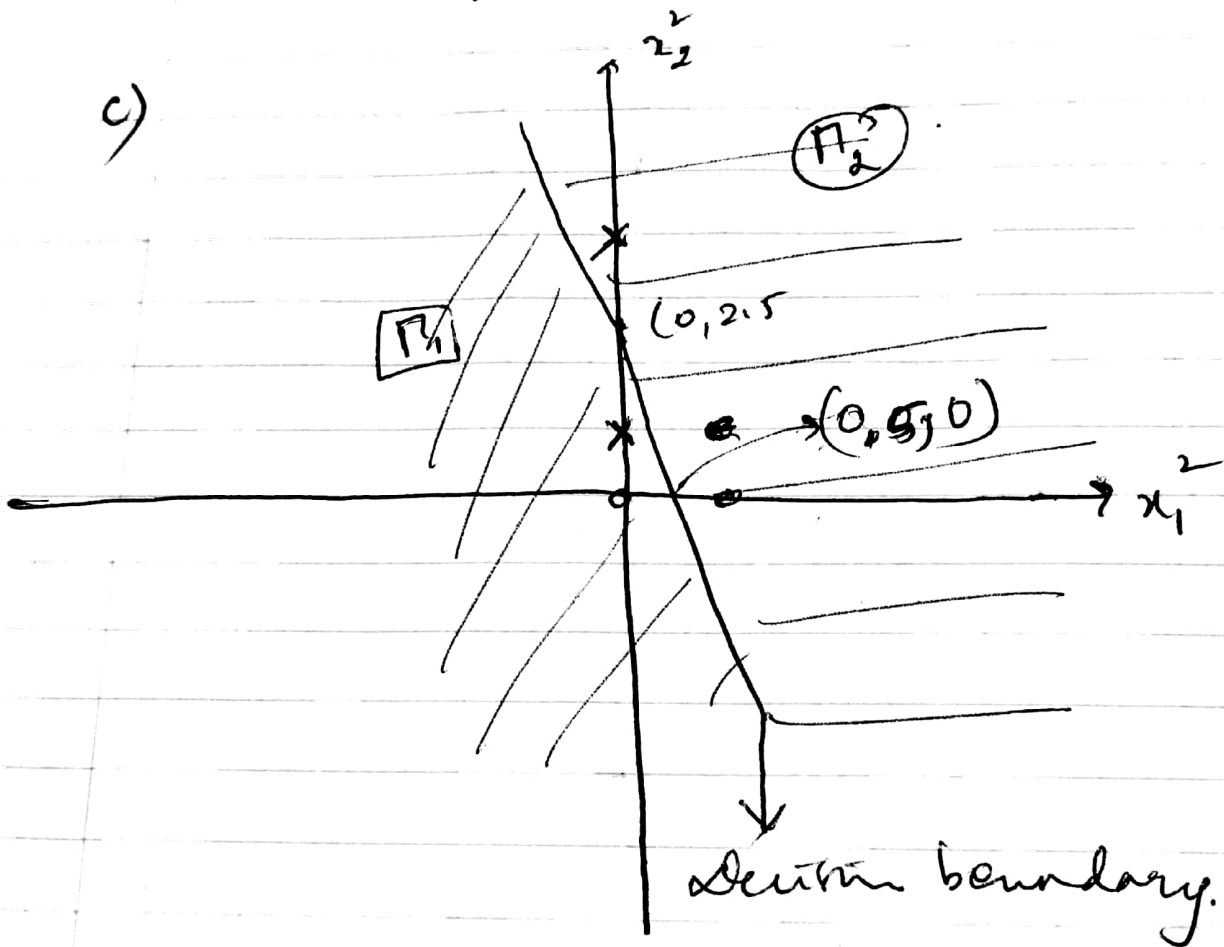
datapoint  
(1, 0)

$$u_8 = \phi(\underline{x}_8) = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

datapoint  
(2, 0)

$$u_9 = \phi(\underline{x}_9) = \begin{bmatrix} 1 \\ 2 \\ 0 \\ 4 \\ 0 \\ 0 \end{bmatrix}$$

$$u = \begin{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \\ \begin{bmatrix} 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \\ \begin{bmatrix} 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \\ \begin{bmatrix} 1 & -2 & 0 & 4 & 0 & 0 \end{bmatrix} \\ \begin{bmatrix} 1 & -1 & 0 & 1 & 0 & 0 \end{bmatrix} \\ \begin{bmatrix} 1 & 0 & 2 & 0 & 0 & 4 \end{bmatrix} \\ \begin{bmatrix} 1 & 0 & -2 & 0 & 0 & 4 \end{bmatrix} \\ \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \end{bmatrix} \\ \begin{bmatrix} 1 & 2 & 0 & 4 & 0 & 0 \end{bmatrix} \end{bmatrix}$$



Equation of →

$$\frac{x_1^2}{0.5} + \frac{x_2^2}{2.5} = 1$$

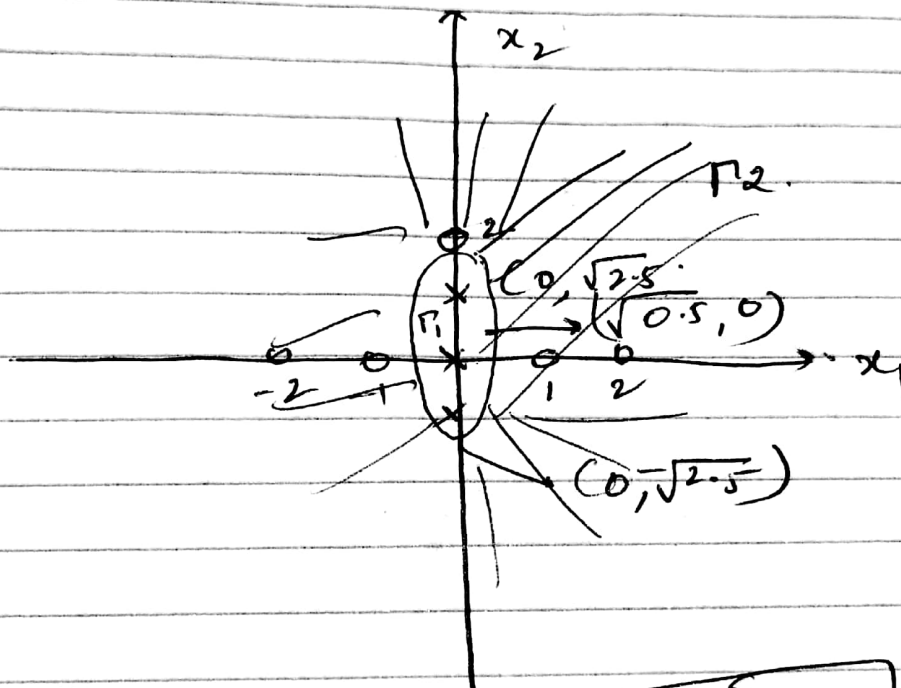
$$2x_1^2 + 0.4x_2^2 - 1 = 0$$

$$\underline{w} = [-1 \ 0 \ 0 \ 0 \ 2 \ 0 \ 0 \ 0.4]$$

$$g(\underline{u}_i) = \underline{w}^T \underline{u}_i \sum_{s_1}^{s_2} 0$$

$$g(\underline{u}_i) = 2x_1^2 + 0.4x_2^2 - 1 \sum_{s_1}^{s_2} 0$$

a)



$$g(\underline{x}) = 2x_1^2 + 0.4x_2^2 - 1 \sum_{s_1}^{s_2} 0$$

as mentioned  
above

1) c) (i) According to the doc, the initial weight vector (by default) is  $w = 1$  for all classes.

1) a) Python. ~~matplotlib~~, sklearn-learn.

1) b)

→ The testing is not available during model building,  
∴ To get a good estimate of the quality of the model, we use normalisation parameters on training data

→ Avoiding small model weights in order to get numerical stability.

→ To achieve quick convergence of optimization algorithms.

~~Ques~~  
(ii)