

Machine learning

Assignment - 3

- 1) For the following function f , create the computation graph.

$$f(x, y, z) = \frac{\sin(xz)}{2} + \frac{e^{xz}}{xy}$$

Let,

$$a = x \cdot y$$

$$b = x \cdot z$$

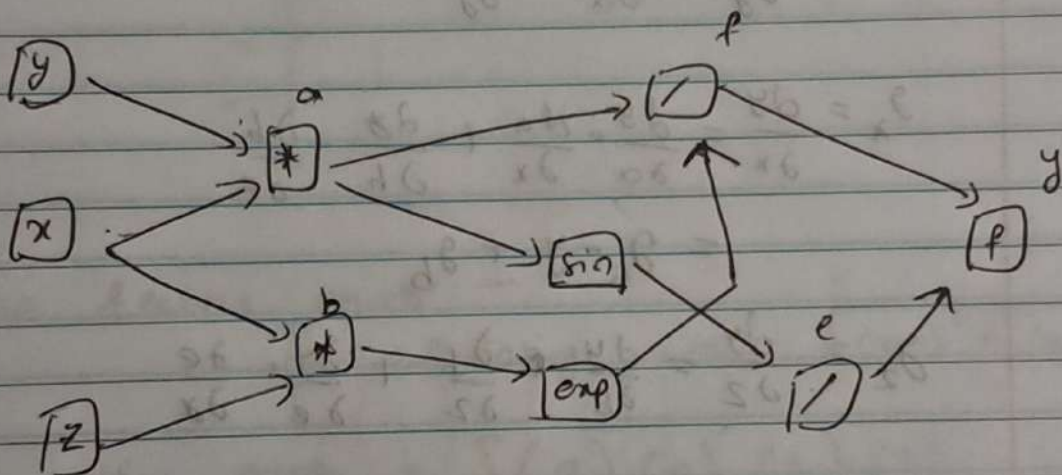
$$c = \sin(b)$$

$$d = e^b$$

$$f = d/2$$

$$e = c/2$$

$$y = f + e$$



- 2) Based on computation graph and chain rule, show the derivative $\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}$

please note that derivative

$$\frac{\partial \sin(x)}{\partial x} = \cos(x), \text{ derivative } \frac{\partial (1/x)}{\partial (x)} = -\frac{1}{x^2},$$

$$\text{the derivative of } \frac{\partial e^x}{\partial x} = e^x$$

$$\frac{\partial y}{\partial y} = \frac{\partial f}{\partial y} = 1$$

$$g_c = g_f = 1$$

$$g_c = \frac{\partial y}{\partial c} = \frac{\partial y}{\partial c} \cdot \frac{\partial e}{\partial c} = g_e \cdot \frac{1}{2} = \frac{1}{2}$$

$$g_d = \frac{\partial y}{\partial d} = \frac{\partial y}{\partial f} \cdot \frac{\partial f}{\partial d} = g_f \cdot \frac{1}{a} = \frac{1}{a}$$

$$g_a = \frac{\partial y}{\partial a} = \frac{\partial y}{\partial c} \cdot \frac{\partial c}{\partial a} + \frac{\partial y}{\partial f} \cdot \frac{\partial f}{\partial a} = g_c$$

$$\cos x + g_f \cdot \frac{-d}{a^2}$$

$$g_a = \frac{\cos a}{2} - \frac{d}{a^2}$$

$$g_b = \frac{\partial y}{\partial b} = \frac{\partial y}{\partial d} \cdot \frac{\partial d}{\partial b} = g_d \cdot e^b = \frac{e^b}{a}$$

$$g_y = \frac{\partial y}{\partial y} = \frac{\partial y}{\partial a} \cdot \frac{\partial a}{\partial y} = g_a \cdot x$$

$$g_x = \frac{\partial y}{\partial x} = \frac{\partial y}{\partial a} \cdot \frac{\partial a}{\partial x} + \frac{\partial y}{\partial b} \cdot \frac{\partial b}{\partial x}$$

$$= g_a \cdot y + g_b$$

$$g_z = \frac{\partial y}{\partial z} = \frac{\partial y}{\partial b} \cdot \frac{\partial b}{\partial z} + \frac{\partial y}{\partial e} \cdot \frac{\partial e}{\partial z}$$

$$= g_b \cdot x - \frac{e}{2}$$

$$g_x = \frac{y \cos y}{2} + \frac{e^{yz} \cdot z}{2} - \frac{e^{xz}}{2yz}$$

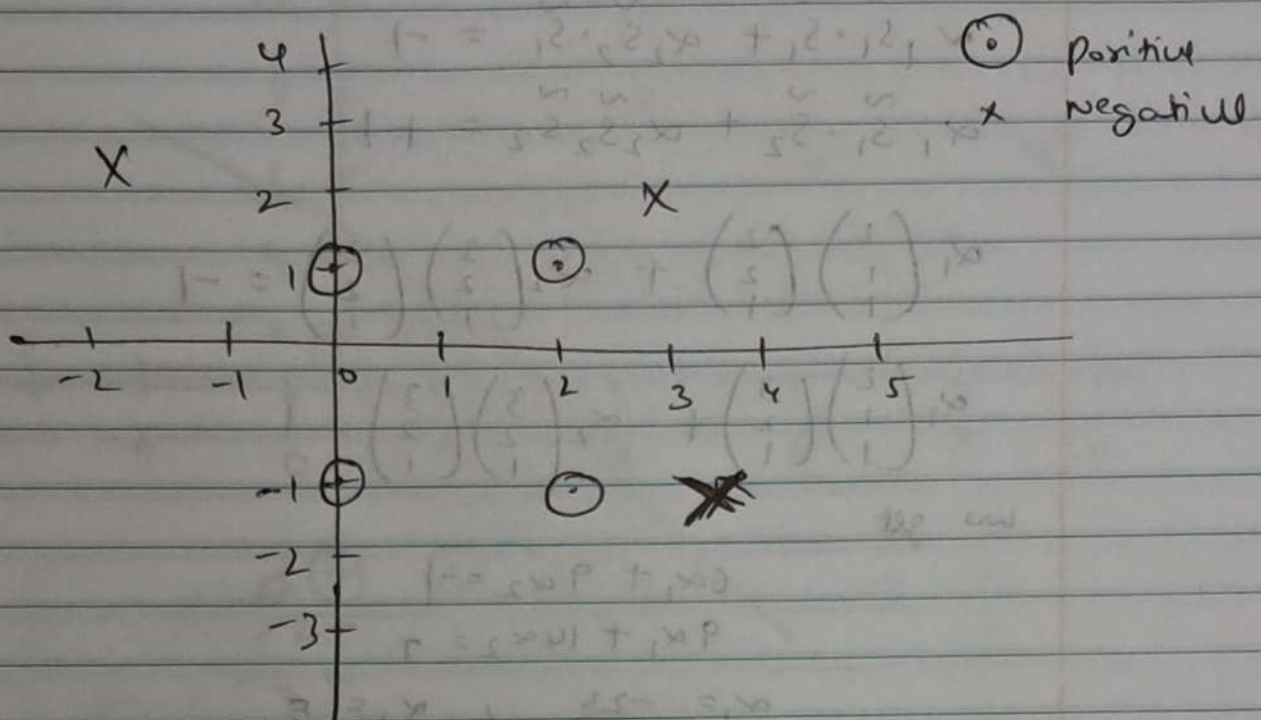
$$g_y = \frac{\cos y}{2} \cdot x - \frac{e^{xz}}{y^2}$$

$$g_2 = \frac{-\sin x_3}{z^2} + \frac{e^{x_2}}{y}$$

✱

SVM

Positive points: $(0, -1)$ $(0, 1)$ $(2, -1)$ $(2, 1)$
 Negative points: $(-1, 1)$ $(-1, -2)$ $(3, 2)$ $(3, -1)$



1) New feature points: -

after calculating with $\sqrt{(x_1-1)^2 + (x_2)^2} > 2$

positive points = $\left\{ \begin{pmatrix} 0 \\ -1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 2 \\ -1 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} \right\}$

negative points = $\left\{ \begin{pmatrix} 5 \\ 8 \end{pmatrix} \begin{pmatrix} 7 \\ 6 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix} \begin{pmatrix} 9 \\ 5 \end{pmatrix} \right\}$

2) Support vectors =

$$\left\{ s_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix} \quad s_2 = \begin{pmatrix} 3 \\ 2 \end{pmatrix} \right\}$$

each arg 1 with α as bias $1/p$

$$\tilde{s}_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix} \quad \tilde{s}_2 = \begin{pmatrix} 3 \\ 2 \end{pmatrix}$$

$$\alpha_1 \tilde{s}_1 \cdot \tilde{s}_1 + \alpha_2 \tilde{s}_2 \cdot \tilde{s}_1 = -1$$

$$\alpha_1 \tilde{s}_1 \cdot \tilde{s}_2 + \alpha_2 \tilde{s}_2 \cdot \tilde{s}_2 = +1$$

$$\alpha_1 \begin{pmatrix} 2 \\ 1 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \alpha_2 \begin{pmatrix} 3 \\ 2 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = -1$$

$$\alpha_1 \begin{pmatrix} 2 \\ 1 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix} + \alpha_2 \begin{pmatrix} 3 \\ 2 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix} = +1$$

we get

$$6\alpha_1 + 9\alpha_2 = -1$$

$$9\alpha_1 + 14\alpha_2 = 1$$

$$\alpha_1 = -\frac{23}{3}, \quad \alpha_2 = 5$$

$$\bar{w} = \sum \alpha_i \tilde{s}_i$$

$$-\frac{23}{3} \begin{pmatrix} 2 \\ 1 \end{pmatrix} + 5 \begin{pmatrix} 3 \\ 2 \end{pmatrix}$$

$$= \begin{pmatrix} -4\frac{2}{3} \\ -23\frac{1}{3} \end{pmatrix} + \begin{pmatrix} 15 \\ 10 \end{pmatrix}$$

$$\begin{pmatrix} -0.33 \\ 2.33 \\ 8 \end{pmatrix}$$

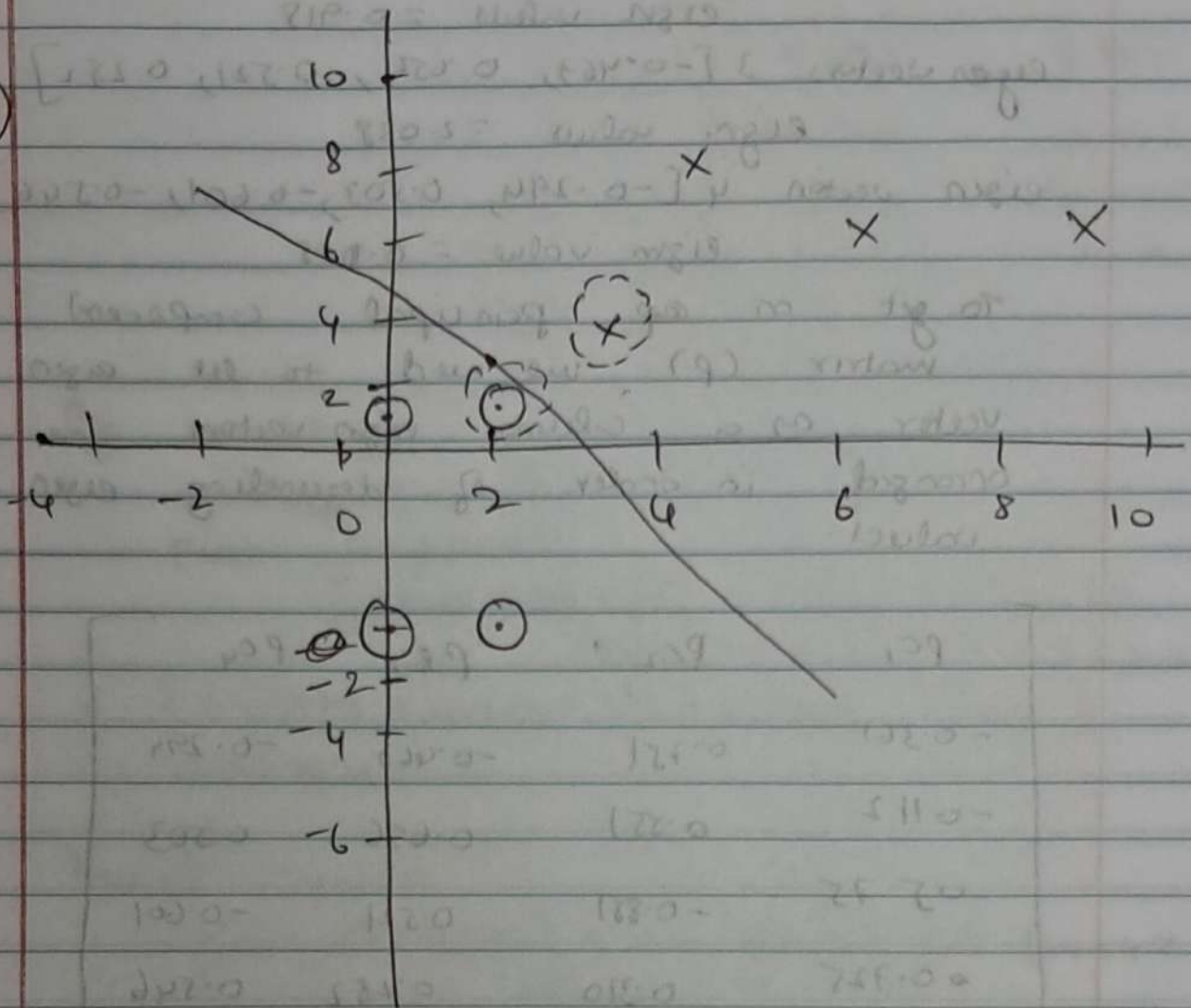
PCA

(R)

3)

$$w = \begin{pmatrix} -0.33 \\ 2.33 \end{pmatrix} \quad b = 8$$

4)



*)

PCA

Given eigen vectors A eigen values

1)

eigen vectors 1 $[-0.361, -0.112, -0.575, 0.725]$

eigen value $= 1.756$

eigen vectors 2 $[0.751, 0.551, -0.18, 0.310]$

eigen value $= 0.918$

eigen vectors 3 $[-0.467, 0.656, 0.521, 0.282]$

eigen value $= 2.038$

eigen vectors 4 $[-0.294, 0.503, -0.601, -0.546]$

eigen value $= 0.088$

To get an principal component matrix (P) we need to let eigen vector as a column. eigen vectors are arranged in order of descending eigen values.

PC ₁	PC ₂	PC ₃	PC ₄
-0.361	0.751	-0.467	-0.294
-0.112	0.551	0.656	0.503
-0.575	-0.881	0.521	-0.601
0.725	0.310	0.282	0.546

Given,

-0.467	0.751
0.656	0.551
0.521	-0.188
0.282	0.310

From data ,
First sample

$$X_{\text{first sample}} = [7.49, 19.01, 14.64, 11.97]$$

Transformation formulae $X_{\text{PCA}} = X P$

$$X_{\text{PCA}} = [7.49, 19.01, 14.64, 11.97] \times$$

$$\begin{bmatrix} -0.467 & 0.751 \\ 0.656 & 0.551 \\ 0.521 & -0.188 \\ 0.282 & 0.310 \end{bmatrix}$$

2)

By computing

$$X_{\text{PCA}} =$$

First component

$$\begin{aligned} & [(7.49 \times -0.467) + (19.01 \times 0.656) + \\ & \quad (14.64 \times 0.521) + (11.97 \times 0.282)] \\ & = [-3.49983 + 12.47056 + 7.62984 + 3.375] \\ & = 19.96761 \end{aligned}$$

Second component

$$\begin{aligned} & = [(7.49 \times 0.751) + (19.01 \times 0.551) + (14.64 \times -0.188) \\ & \quad + (11.97 \times 0.310)] \\ & = [5.62599 + 10.47151 - 2.75232 + 3.711] \\ & = 17.05518 \end{aligned}$$

PCA transformed, Result of First Sampled
using third component.

$$X_{\text{PCA}} \text{ first sample} = [19.96711 \quad 17.05618]$$

Clustering :-

	P_1	P_2	P_3	P_4	P_5	P_6
P_1	0					
P_2	0.23	0				
P_3	0.22	0.15	0			
P_4	0.37	0.20	0.15	0		
P_5	0.14	0.14	0.18	0.29	0	
P_6	0.23	0.25	0.11	0.22	0.39	0

Given to merge (only $P_3 \times P_6$)

After merging both $P_3 \times P_6$

we consider both as one entity
for further calculations.

1) Single Link (min distance)

To calculate minimum distance b/w the new cluster (P_3, P_6) and all other points

To distance b/w matrix

$$\begin{aligned} \text{MIN} (\text{dist}(P_3, P_1), (P_6, P_1)) \\ = \min(0.22, 0.23) \\ = 0.22 \end{aligned}$$

To update distance matrix $\text{MIN} [\text{dist}(P_3, P_6), P_2]$

$$\begin{aligned} \text{MIN} (\text{dist}(P_3, P_2), (P_6, P_2)) \\ = \min(0.15, 0.25) = 0.15 \end{aligned}$$

To update the distance matrix

$$\begin{aligned} \text{MIN} [\text{dist}(P_3, P_6), P_4] \\ = \min(0.37, 0.22) \\ = 0.15 \end{aligned}$$

To update the distance matrix

$$\begin{aligned} \text{MIN} [\text{dist}(P_3, P_6), P_5] \\ = \min(0.28, 0.39) = 0.28 \end{aligned}$$

To update distance matrix for P_3, P_6

	P_1	P_2	P_3, P_6	P_4	P_5
P_1	0				
P_2	0.23	0			
P_3, P_6	0.22	0.15	0		
P_4	0.37	0.20	0.15	0	
P_5	0.34	0.14	0.28	0.24	0

2) Complete Link (max distance)

To update the distance matrix

$$= \text{MAX} [\text{dist} (P_3, P_6) / P_1]$$

$$= \text{MAX} [\text{dist} (P_3, P_1), (P_6, P_1)]$$

$$= \text{MAX} [0.22, 0.23] = 0.23$$

To update the distance matrix

$$\text{MAX} [\text{dist} (P_3, P_6), P_2]$$

$$\text{MAX} [0.15, 0.25] = 0.25$$

To update the distance matrix

$$\text{MAX} [\text{dist} (P_3, P_6), P_4]$$

$$= \text{MAX} [\text{dist} (P_3, P_4)]$$

$$= \text{MAX} [0.15, 0.22] = 0.22$$

To update the distance matrix

$$= \text{MAX} [\text{dist} (P_3, P_6), P_5]$$

$$= \text{MAX} [\text{dist} (P_3, P_5), (P_6, P_5)]$$

$$= \text{MAX} [0.28, 0.39] = 0.39$$

To update distance matrix for cluster (P_3, P_6)

	P_1	P_2	P_3, P_6	P_4	P_5
P_1	0				
P_2	0.23	0			
P_3, P_6	0.23	0.25	0		
P_4	0.37	0.20	0.22	0	
P_5	0.34	0.19	0.39	0.29	0