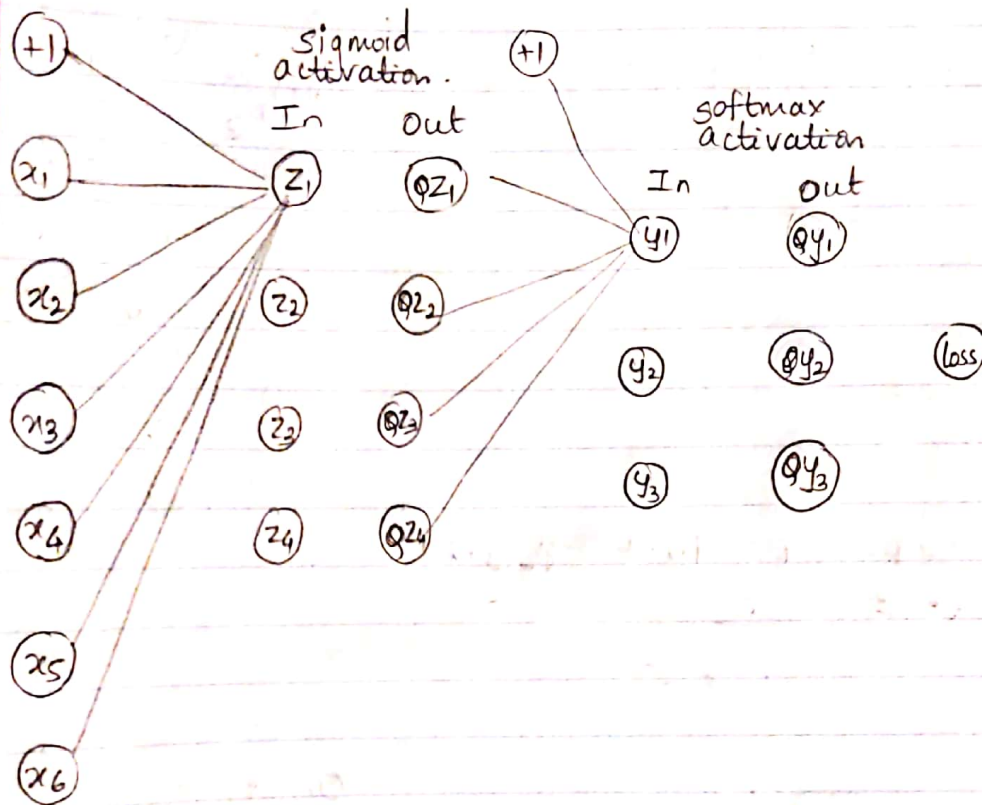


Problem 4 : Feed forward and Back Propagation.



a) Forward feed:
Linear combination at first layer

$$a_j = a_0 + \sum_{i=1}^n a_{ji} x_i$$

$$\begin{aligned} z_1 &= b_0 + x_1 w_{11} + x_2 w_{12} + x_3 w_{13} + x_4 w_{14} + x_5 w_{15} + x_6 w_{16} \\ &= 1 + (1 \times 1) + (1 \times 2) + (0 \times -3) + (0 \times 0) + (1 \times 1) + (1 \times -3) \\ &= 2 \end{aligned}$$

Similarly z_2 to z_4 can be calculated using the weight matrix α given.

$$\begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 & 2 & -3 & 0 & 1 & -3 \\ 3 & 1 & 2 & 1 & 0 & 2 \\ 2 & 2 & 2 & 2 & 2 & 1 \\ 1 & 0 & 2 & 1 & -2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 1 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 2 \\ 7 \\ 8 \\ 2 \end{bmatrix}$$

Activation of first layer using sigmoid
 $z_i = \frac{1}{1 + \exp(-z_i)}$

$$\phi z_1 = \frac{1}{1 + \exp(2)} = 0.880797, \quad \phi z_2 = \frac{1}{1 + \exp(7)} = 0.999088$$

$$\phi z_3 = \frac{1}{1 + \exp(8)} = 0.999664, \quad \phi z_4 = \frac{1}{1 + \exp(2)} = 0.880797$$

Linear combination at second layer

$$y_i = b_0 + z_1 w_{21} + z_2 w_{22} + z_3 w_{23}$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 & 2 & -2 & 1 \\ 1 & -1 & 1 & 2 \\ 3 & 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 0.880797 \\ 0.999088 \\ 0.999664 \\ 0.880797 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 2.760442 \\ 3.642966 \\ 4.5226126 \end{bmatrix}$$

Activation of second layer using softmax

$$ay_i = \frac{e^{y_i}}{\sum e^{y_i}}$$

$$\begin{bmatrix} ay_1 \\ ay_2 \\ ay_3 \end{bmatrix} = \begin{bmatrix} 0.108201 \\ 0.261521 \\ 0.630277 \end{bmatrix}$$

$$\begin{aligned} \text{Loss } \therefore L &= - \sum_{i=1}^3 \hat{y}_i \times \log(ay_i) \quad \text{log is to base } e \\ &= -0 \times \log(0.108201) - 1 \times \log(0.261521) \\ &\quad - 0 \times \log(0.630277) \\ &= 1.3412 \end{aligned}$$

(b) Back propagation
 (A) Back propagation for second layer
 Updating weights,

$$\begin{aligned} \frac{\partial L}{\partial w_{21}} &= \frac{\partial L}{\partial ay_1} \times \frac{\partial ay_1}{\partial y_1} \times \frac{\partial y_1}{\partial w_{21}} + \frac{\partial L}{\partial ay_2} \times \frac{\partial ay_2}{\partial y_1} \times \frac{\partial y_1}{\partial w_{21}} \\ &\quad + \frac{\partial L}{\partial ay_3} \times \frac{\partial ay_3}{\partial y_1} \times \frac{\partial y_1}{\partial w_{21}} \end{aligned}$$

$$= \frac{-\hat{y}_1}{ay_1} \times ay_1 \times (1 - ay_1) \times \phi z_1 + \frac{(-\hat{y}_2)}{ay_1 \times ay_2} \times (ay_2 \times ay_1) \phi z_1$$

$$+ \frac{(-\hat{y}_3)}{ay_3} \times (ay_3 \times ay_1) \phi z_1$$

$$\begin{aligned}
 &= -\hat{y}_1 (1 - \phi y_1) \times \phi z_1 + \hat{y}_2 \times \phi y_1 \times \phi z_1 + \hat{y}_3 \times \phi y_1 \times \phi z_1 \\
 &= -\hat{y}_1 \times \phi z_1 + \hat{y}_1 \phi z_1 \phi y_1 + \hat{y}_2 \times \phi z_1 \times \phi y_1 + \hat{y}_3 \phi y_1 \times \phi z_1 \\
 &= -\hat{y}_1 \times \phi z_1 + \phi z_1 \phi y_1
 \end{aligned}$$

$$w_{21}^{\text{new}} = w_{21} - \frac{\alpha \partial L}{\partial w_{21}} = 1 - 1 \times (\phi y_1 - \hat{y}_1) \times \phi z_1$$

$$w_{21}^{\text{new}} = 1 - 1(0.1082 - 0) \times 0.8808 = 0.9047$$

We can generalize $\frac{\partial L}{\partial w_{2i}} = (-\hat{y}_i \times \phi z_i + \phi z_i \times \phi y_i)$

There updated weights are as follows.

$$W_2^{\text{new}} = B^{\text{new}} = \begin{bmatrix} 0.9047 & 1.8919 & -2.1082 & 0.9047 \\ 1.6505 & -0.2622 & 1.7382 & 2.6505 \\ 2.4449 & 0.3703 & -1.6301 & 0.4449 \end{bmatrix}$$

(B) Back propagation for 1st layer

$$\begin{aligned}
 \frac{\partial L}{\partial w_{11}} &= \left[\frac{\partial L}{\partial w_{21}} \frac{\partial w_{21}}{\partial \phi z_1} \frac{\partial \phi z_1}{\partial z_1} \frac{\partial z_1}{\partial w_{11}} + \frac{\partial L}{\partial w_{22}} \frac{\partial w_{22}}{\partial \phi z_1} \frac{\partial \phi z_1}{\partial z_1} \frac{\partial z_1}{\partial w_{11}} \right. \\
 &\quad \left. + \frac{\partial L}{\partial w_{31}} \frac{\partial w_{31}}{\partial \phi z_1} \frac{\partial \phi z_1}{\partial z_1} \frac{\partial z_1}{\partial w_{11}} \right]
 \end{aligned}$$

$$\begin{aligned}
 &= \left[(\phi y_1 - \hat{y}_1) w_{21} + (\phi y_2 - \hat{y}_2) w_{22} + (\phi y_3 - \hat{y}_3) w_{23} \right] \\
 &\quad \times \phi z_1 (1 - \phi z_1) x_1
 \end{aligned}$$

$$\therefore \frac{\partial L}{\partial w_{11}} = \left[(0.1082 - 0) \times 1 + (0.2615 - 1) \times 1 + (0.6303 - 0) \times 3 \right] \times (0.8808) \times (1 - 0.8808) \times 1$$

$$\frac{\partial L}{\partial w_{11}} = 0.1324$$

$$\begin{aligned} \therefore w_{11 \text{ update}} &= w_{11} - \alpha \frac{\partial L}{\partial w_{11}} \\ &= 1 - 1 \times 0.1324 \\ &= 0.8676 \end{aligned}$$

The matrix of updated weights is as follows.

$$W_{\text{new}} = \alpha\text{-update} = \begin{bmatrix} 0.8676 & 1.8676 & -3 & 0 & 0.8676 & -3.1324 \\ 2.9986 & 0.9986 & 2 & 1 & -0.0014 & 1.9986 \\ 2.0005 & 2.0005 & 2 & 2 & 2.0005 & 1.0005 \\ 1.0775 & 0.0775 & 2 & 1 & -1.9224 & 2.0775 \end{bmatrix}$$

Updating 2nd layer bias.

$$\frac{\partial L}{\partial B_{21}} = \frac{\partial L}{\partial y_1} \times \frac{\partial y_1}{\partial B_{21}} + \frac{\partial L}{\partial y_2} \times \frac{\partial y_2}{\partial B_{21}}$$

$$+ \frac{\partial L}{\partial y_3} \times \frac{\partial y_3}{\partial B_{21}}$$

$$= (y_1 - \hat{y}_1) B_{21}$$

$$\begin{aligned}\therefore B_{21\text{new}} &= B_{21} - \alpha [0y_1 - \hat{y}_1] B_{21} \\ &= 1 - (0.1082 - 0) \times 1 \\ &= 0.8918\end{aligned}$$

B_{22}, B_{23} can be calculated as follows

$$B_{22} = 1.7385$$

$$B_{23} = 0.3697$$

Updating 1st layer Bias

$$\begin{aligned}\frac{\partial L}{\partial B_{11}} &= \left(\sum_{i=1}^3 \frac{\partial L}{\partial w_{1i}} \right) \times \frac{\partial z_1}{\partial z_1} \times \frac{\partial z_1}{\partial B_{11}} \\ &= 0.1324\end{aligned}$$

$$\text{Therefore, } B_{11\text{-new}} = B_{11} - \alpha (0.1324) = 0.8676$$

Similarly we can calculate updated values as below,

$$B_{12} = 0.9986$$

$$B_{13} = 1.0005$$

$$B_{14} = 1.0775$$

Answers are as follows:

1(a) $a_1^{(1)} = z \Rightarrow z_1^{(1)} = 0.88079$

1(b) $a_3^{(1)} = 8 \Rightarrow z_3^{(1)} = 0.99966$

1(c) $b_2^{(1)} = 3.6430$

1(d) $\hat{y}_2^{(1)} = 0.2615$

1(e) class 3

1(f) loss = 1.3412

2(a) $B_{2,1} = 1.6505$

(b) Bias = 0.8918

(c) $\alpha_{3,4} = 2$

(d) Bias = 0.9986

(e) class 2