

# Deep Learning | EX01

Feed Forward Neural Networks (FFNN)

800

**Computer Vision | Zahra Amini**

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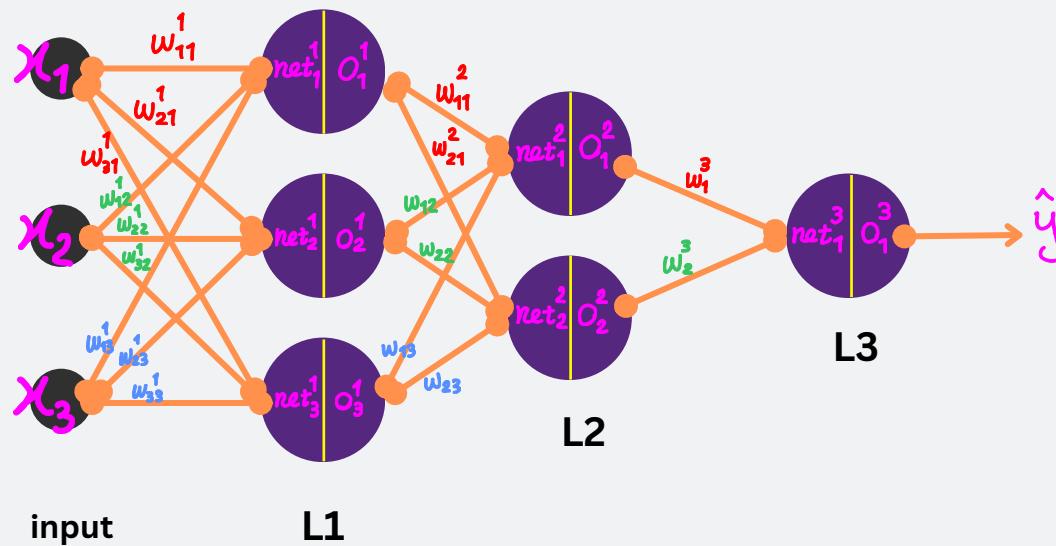
<https://zil.ink/zahraamini>

1. For the below feed-forward neural network architecture obtain the following variables:

$$O_1^1 = ?$$

$$O_2^2 = ?$$

$$O_1^3 = y_{predicted} = ?$$



inputs

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0.3 \\ 0.5 \\ 0.8 \end{bmatrix}$$

L1-Weights

$$\begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & 0.1 & 0.2 \\ w_{31} & 0.4 & 0.1 \end{bmatrix}$$

L1-Biases

$$\begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} 0.2 \\ 0.2 \\ 0.1 \end{bmatrix}$$

L2-Weights

$$\begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & 0.1 & 0.2 \\ w_{31} & 0.4 & 0.1 \end{bmatrix}$$

L2-Biases

$$\begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} 0.1 \\ 0.4 \end{bmatrix}$$

L3-Weights

$$\begin{bmatrix} w_1 & w_2 \\ w_3 & 0.5 \end{bmatrix}$$

L3-Biases

$$b_3 = 0.5$$

Activation Functions:

L1: Relu  
L2: Tanh  
L3: Sigmoid

Diagram:  $O_1^1 = ? 0.49 \quad O_2^1 = ? 0.621 \quad \hat{y} = ? 0.673$

Layer #1:

$$\text{net}_1^1 = \sum Wx + b = W_{11}^1 x_1 + W_{21}^1 x_2 + W_{31}^1 x_3 + b_1^1 = (0.1 \times 0.3) + (0.2 \times 0.5) + (0.2 \times 0.8) + 0.2 = 0.49$$

$$\begin{bmatrix} \text{net}_1^1 \\ \text{net}_2^1 \\ \text{net}_3^1 \end{bmatrix} = WX + b = \begin{bmatrix} 0.1 & 0.2 & 0.2 \\ 0.4 & 0.1 & 0.2 \\ 0.2 & 0.3 & 0.1 \end{bmatrix}_{3 \times 3} \begin{bmatrix} 0.3 \\ 0.5 \\ 0.8 \end{bmatrix}_{3 \times 1} + \begin{bmatrix} 0.2 \\ 0.2 \\ 0.1 \end{bmatrix} = \begin{bmatrix} 0.1 \times 0.3 + 0.2 \times 0.5 + 0.2 \times 0.8 \\ 0.4 \times 0.3 + 0.1 \times 0.5 + 0.2 \times 0.8 \\ 0.2 \times 0.3 + 0.3 \times 0.5 + 0.1 \times 0.8 \end{bmatrix} + \begin{bmatrix} 0.2 \\ 0.2 \\ 0.1 \end{bmatrix}$$

$$\begin{bmatrix} \text{net}_1^1 \\ \text{net}_2^1 \\ \text{net}_3^1 \end{bmatrix} = \begin{bmatrix} 0.49 \\ 0.53 \\ 0.39 \end{bmatrix}_{3 \times 1} \xrightarrow{\text{ReLU}} \begin{bmatrix} O_1^1 \\ O_2^1 \\ O_3^1 \end{bmatrix} = \begin{bmatrix} 0.49 \\ 0.53 \\ 0.39 \end{bmatrix}$$

✓  $\text{ReLU}(z) = \max(0, z)$

Layer #2:

$$\text{net}_1^2 = \sum Wx + b = W_{11}^2 O_1^1 + W_{12}^2 O_2^1 + W_{13}^2 O_3^1 + b_1^2 = (0.1 \times 0.49) + (0.2 \times 0.53) + (0.2 \times 0.39) + 0.1$$

$$\text{net}_1^2 = 0.333 \xrightarrow{\tanh} O_1^2 = \tanh(\text{net}_1^2) = 0.321$$

✓  $\tanh(x) = \frac{\sinh(x)}{\cosh(x)} = \frac{e^x - e^{-x}}{e^x + e^{-x}}$

$$\text{net}_2^2 = \sum Wx + b = W_{21}^2 O_1^1 + W_{22}^2 O_2^1 + W_{23}^2 O_3^1 + b_2^2 = (0.4 \times 0.49) + (0.1 \times 0.53) + (0.2 \times 0.39) + 0.4$$

$$\text{net}_2^2 = 0.727 \xrightarrow{\tanh} O_2^2 = \tanh(\text{net}_2^2) = 0.621$$

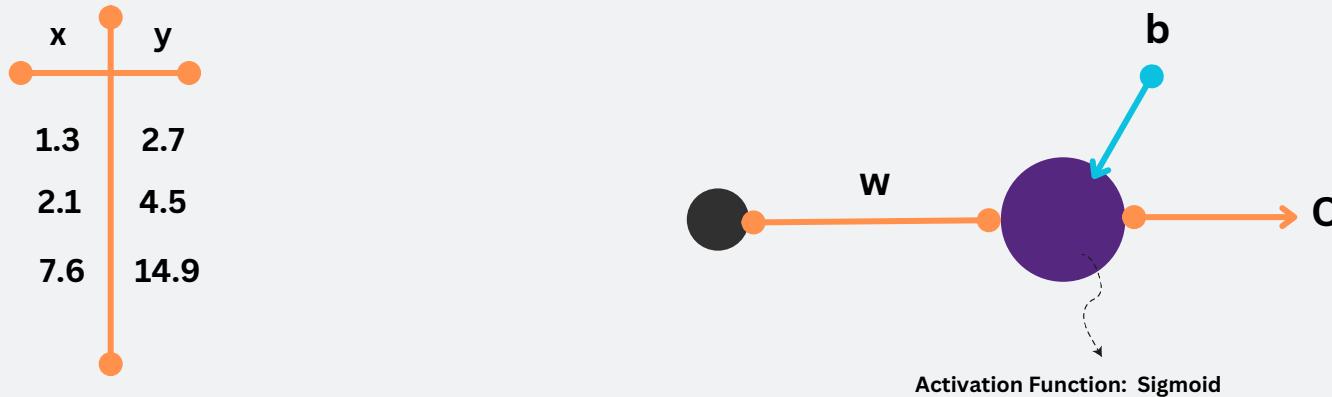
Layer #3:

$$net_1^3 = \sum Wx + b = W_1^3 O_1^2 + W_2^3 O_2^2 + b^3 = (0.5 \times 0.321) + (0.1 \times 0.627) + 0.5 =$$

$$net_1^3 = 0.7232 \xrightarrow{\text{Sigmoid}} O^3 = \tanh(net_1^3) = 0.673$$

✓ Sigmoid( $x$ ) =  $\sigma(x) = \frac{1}{1 + e^{-x}}$

2. Obtain new value of weight and bias after Gradient descent first iteration, use MSE as a loss function.

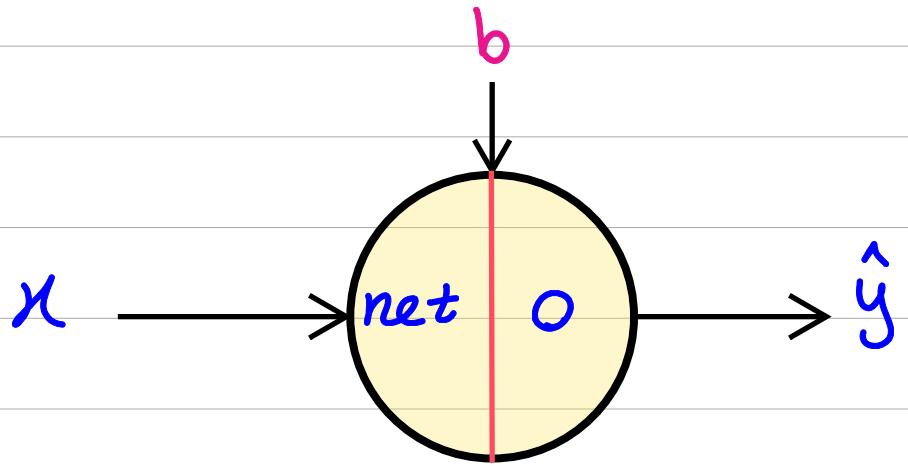


Initial weight value (weight have been initialized randomly) : 1.5

Initial bias value (bias have been initialized randomly) : 0.5

learning rate (alpha) = 0.1

# Gradient Descent in NN

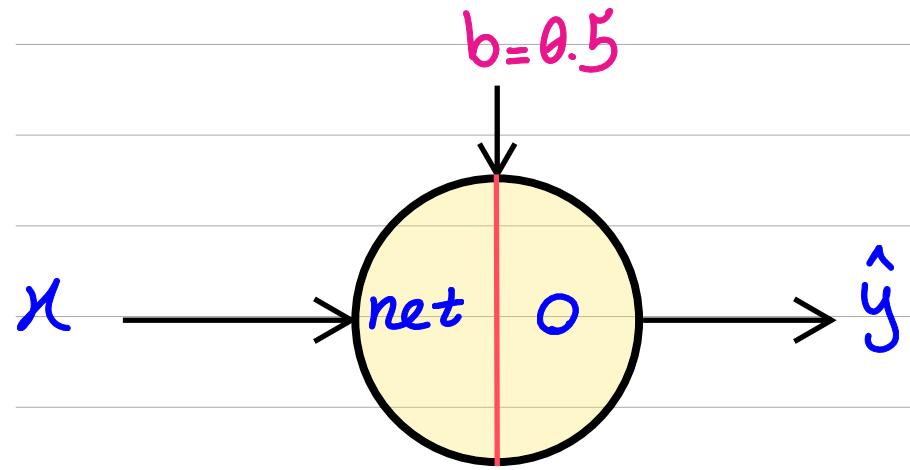


1.  $\hat{y} = w x + b \rightarrow$  خرد جی مربوط به درودی را محاسبه کنید (FF)

2.  $Loss = \frac{1}{m} (\hat{y} - y)^2 \rightarrow$  تابع هنزینه را محاسبه کنید  $\frac{\delta Loss}{\delta w}$ ,  $\frac{\delta Loss}{\delta b} = ?$

3.  $w_{new} = w_{old} - \alpha \frac{\delta Loss}{\delta w} \rightarrow$  وزن خارا آپدیت کنید (BF)

4. این مرحله را تا زمانیله به حداقل Loss بررسیه کنید.



$x$	$y$	$\sigma(x) = \frac{1}{1 + e^{-x}}$
1.3	2.7	
2.1	4.5	
7.6	14.9	

$$\textcircled{1} \quad \hat{y} = Wx + b \rightsquigarrow \text{net}_1 = 1.5 \times 1.3 + 0.5 = 2.45 \xrightarrow{\text{Sigmoid}} \hat{y}_1 = 0.921$$

$$\text{net}_2 = 1.5 \times 2.1 + 0.5 = 3.65 \xrightarrow{\text{Sigmoid}} \hat{y}_2 = 0.975$$

$$\text{net}_3 = 1.5 \times 7.6 + 0.5 = 11.9 \xrightarrow{\text{Sigmoid}} \hat{y}_3 = 0.999$$

$$\textcircled{2} \quad \text{Loss} = \text{MSE} = \frac{1}{m} \sum_{i=1}^m (y_i - \hat{y}_i)^2 = \frac{1}{3} \left[ (2.7 - 0.92)^2 + (4.5 - 0.97)^2 + (14.9 - 0.99)^2 \right] = 69.7$$

$$\text{Loss} = \frac{1}{m} \sum_{i=1}^m (\hat{y}_i - y_i)^2 = \frac{1}{m} (Wx + b - y)^2 \xrightarrow{\frac{\partial \text{loss}}{\partial W} = ?}$$

$$\xrightarrow{\frac{\partial \text{loss}}{\partial b} = ?}$$

$$② \text{MSE} = \frac{1}{m} \sum_{i=1}^m (\hat{y}_i - y_i)^2$$

$$\frac{\partial \text{MSE}}{\partial w} = \frac{\partial \text{MSE}}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial z} \frac{\partial z}{\partial w} = \frac{2}{m} \sum (\hat{y} - y) \hat{y} (1 - \hat{y}) x \quad Z_i = w x_i + b$$

$$\frac{\partial z}{\partial w} = (w x + b)' = x$$

$$\frac{\partial \hat{y}}{\partial z} = \sigma'(z) (1 - \sigma(z)) \quad * \hat{y} = \sigma(z_i) = \frac{1}{1 + e^{-z_i}}$$

\*  $\frac{\partial \hat{y}}{\partial z} = \hat{y} (1 - \hat{y})$

$$\frac{\partial \text{MSE}}{\partial \hat{y}} = \frac{\partial}{\partial \hat{y}} \left( \frac{1}{m} \sum (\hat{y} - y)^2 \right) = \frac{2}{m} \sum (\hat{y} - y) (1)$$

$$③ W_{\text{new}} = W_{\text{old}} - \alpha \frac{\partial \text{loss}}{\partial w} =$$

$$\frac{2}{m} \sum (\hat{y} - y) \hat{y} (1 - \hat{y}) x$$

$2 (0.921 - 2.7) \times 0.921 \times (1 - 0.921) \times 1.3 = -0.337$

$2 (0.975 - 4.5) \times 0.975 \times (1 - 0.975) \times 2.1 = -0.356$

$2 (0.999 - 14.9) \times 0.999 \times (1 - 0.999) \times 7.6 = -0.211$

AVG  $\frac{1}{3} [(-0.337) + (-0.356) + (-0.211)] = -0.301$

$$W_{\text{new}} = W_{\text{old}} - \alpha \frac{\partial L}{\partial w} \Rightarrow W_{\text{new}} = 1.5 - (0.1 \times (-0.301)) = 1.5301$$

$$② MSE = \frac{1}{m} \sum_{i=1}^m (\hat{y}_i - y_i)^2$$

$$\frac{\partial MSE}{\partial b} = \frac{\partial MSE}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial z} \frac{\partial z}{\partial b} = \frac{2}{m} (\hat{y} - y) \hat{y} (1 - \hat{y})$$

$$\frac{\partial MSE}{\partial \hat{y}} = \frac{2}{m} (\hat{y} - y) (1)$$

$$\frac{\partial \hat{y}}{\partial z} = \hat{y} (1 - \hat{y})$$

$$\frac{\partial z}{\partial b} = 1$$

$$③ b_{new} = b_{old} - \alpha \frac{\partial loss}{\partial b} =$$

$$\frac{2}{m} \sum (\hat{y} - y) \hat{y} (1 - \hat{y}) \rightarrow \left. \begin{array}{l} 2(0.921 - 2.7) \times 0.921 \times (1 - 0.921) = -0.258 \\ 2(0.975 - 4.5) \times 0.975 \times (1 - 0.975) = -0.172 \\ 2(0.999 - 14.5) \times 0.999 \times (1 - 0.999) = -0.026 \end{array} \right\} \text{AVG} \rightarrow \frac{1}{3} \times (-0.456) = -0.152$$

$$b_{new} = b_{old} - \alpha \frac{\partial L}{\partial b} \Rightarrow b_{new} = 0.5 - (0.1 \times (-0.152)) = 0.5152$$



