

# Multi Layered Perceptron Model [Artificial Neural Net]

① Forward Propagation

② Backward Propagation → Geoffrey Hinton →

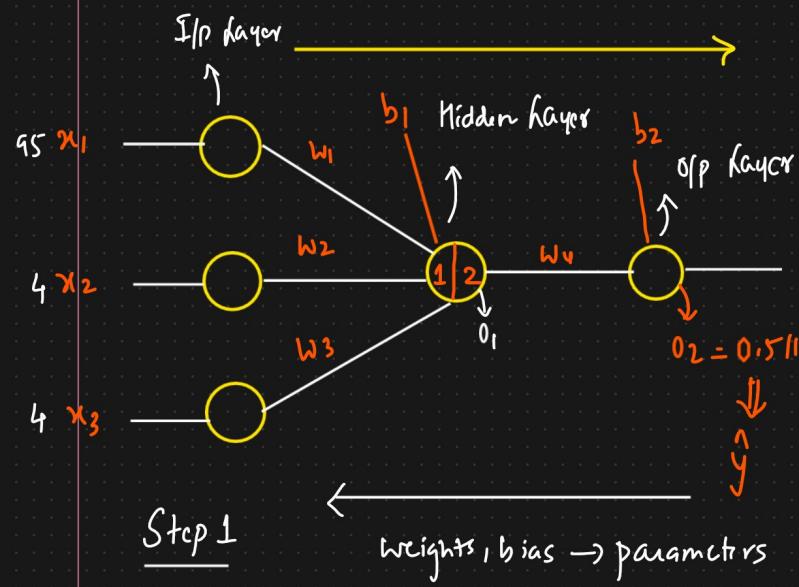
③ Activation Function

④ Optimizers ↗

⑤ Loss functions ↗



## ① Forward Propagation In Multi Layered NN



$$I_1 = x_1 w_1 + x_2 w_2 + x_3 w_3 + b_1$$

$$\boxed{I_1 = \sum_{i=1}^w w_i^T x_i + b}$$

$$I_1 = 95 \times 0.01 + 4 \times 0.02 + 4 \times 0.03 + 0.01$$

$$I_1 = 1.151$$

## Dataset

|  | $x_1$ | $x_2$       | $x_3$      | O/P |
|--|-------|-------------|------------|-----|
|  | IQ    | Study Hours | Play Hours | P/F |
|  | → 95  | 4           | 4          | 1   |
|  | → 100 | 5           | 2          | 1   |
|  | 95    | 2           | 7          | 0   |

loss  $[y - \hat{y}] \approx 0$

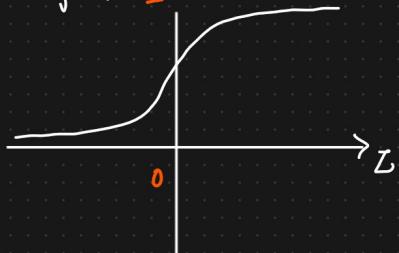
$w_1, w_2, w_3, b_1, b_2$

$[0.01, 0.02, 0.03] [0.001]$

## Step 2 [Activation Function] → 0 to 1

$$f(z) = 1$$

$$\boxed{f(z) = \frac{1}{1+e^{-z}}}$$



$$f(z) = \frac{1}{1+e^{-1.151}} = 0.759 = 0_1$$

## O/P Layer

$$w_4 = [0.02] \quad b_2 = 0.03$$

$$\text{Step 1} \Rightarrow I_2 = o_1 * w_4 + b_2$$

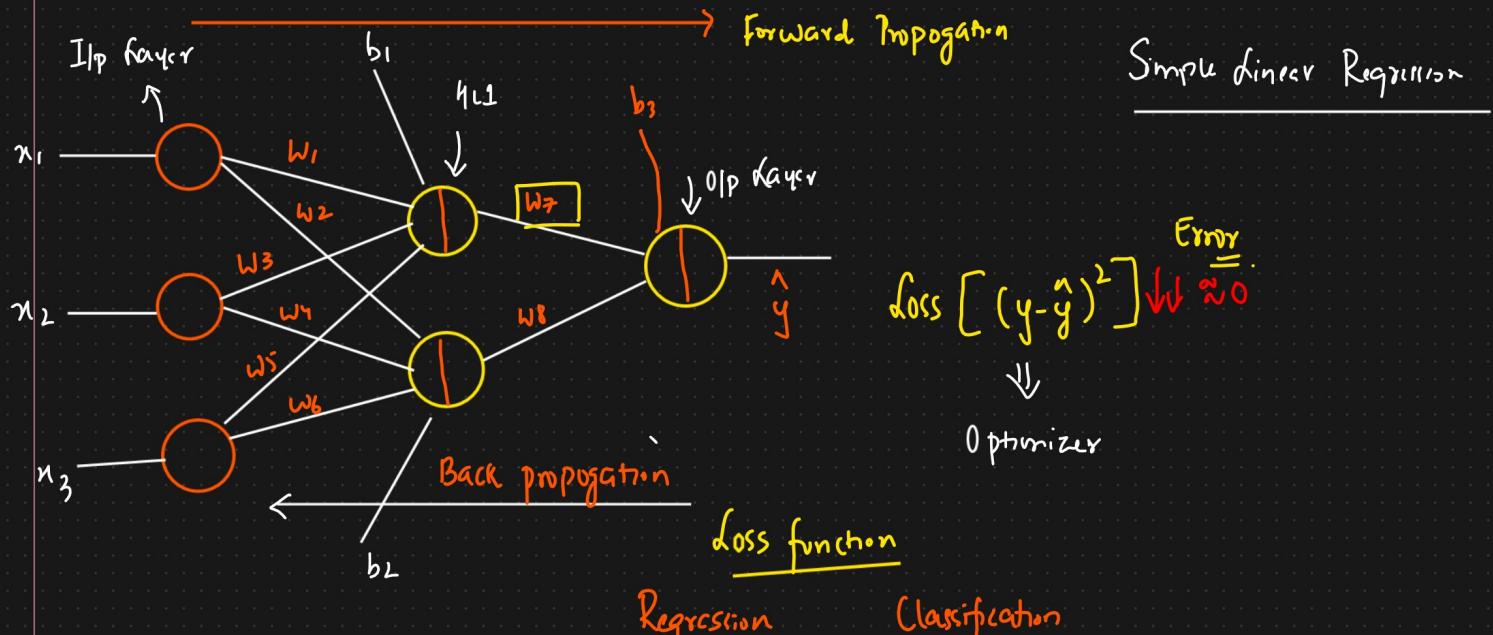
$$= 0.759 + 0.02 + 0.03$$

$$= 0.04518$$

Step 2 Activation( $z$ )  $\Rightarrow \frac{1}{1+e^{-(0.04518)}} = 0.511$

$$\theta_2 = 0.511$$

## ② Backward Propagation And Weight Updation Formula



① MSE

② MAE

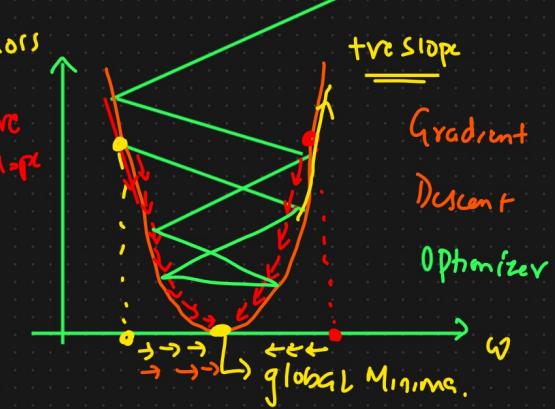
③ Huber loss

① Binary Cross Entropy

② Categorical Cross Entropy

## Weight Updation Formula - Back Propagation

$$W_{\text{new}} = W_{\text{old}} - \eta \frac{\partial L}{\partial W_{\text{old}}} \quad \begin{array}{l} \text{Slope} \\ \text{Learning Rate} \end{array}$$



$$w_{new} = w_{old} - \eta (+v_c)$$

$$\boxed{w_{new} << w_{old}}$$

$$w_{new} = w_{old} - \eta (-v_c)$$

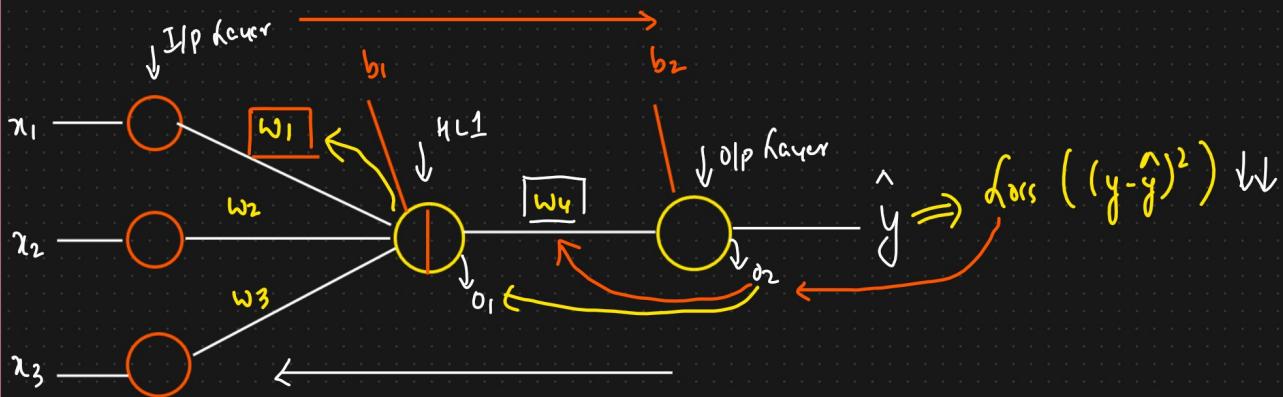
$$w_{new} = w_{old} + (+v_c)$$

$$\boxed{w_{new} >> w_{old}}$$

When  $w$  reaches global Minima

$$\boxed{w_{new} = w_{old}} \Rightarrow \text{loss} \downarrow \downarrow \text{minimal}$$

### ③ Chain Rule of Derivative



$$w_{new} = w_{old} - \eta \frac{\partial L}{\partial w_{old}}$$

$$b_{new} = b_{old} - \eta \frac{\partial L}{\partial b_{old}}$$

$$w_{4,new} = w_{4,old} - \eta \left[ \frac{\partial L}{\partial w_{4,old}} \right]$$

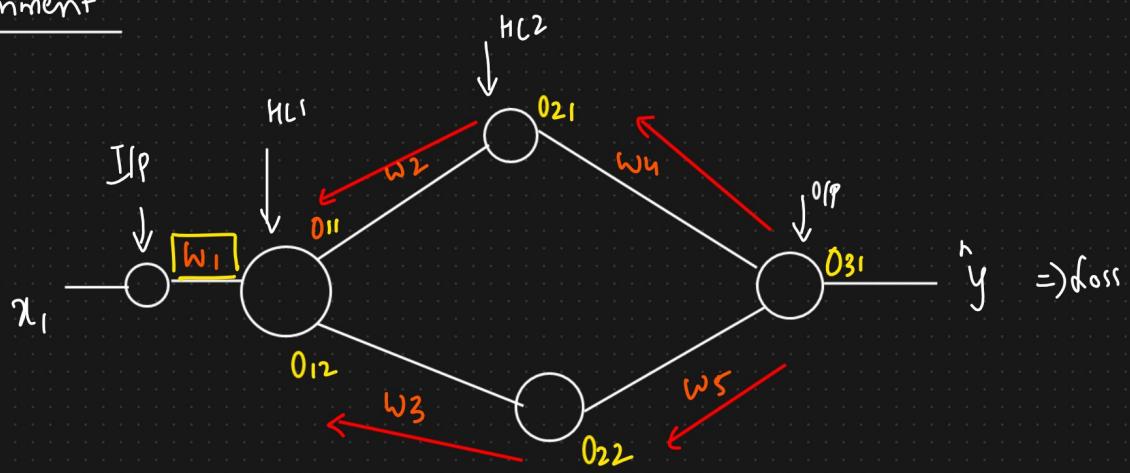
$$b_{2,new} = b_{2,old} - \eta \left[ \frac{\partial L}{\partial b_{2,old}} \right]$$

$$\frac{\partial L}{\partial w_{4,old}} = \frac{\partial L}{\partial o_2} * \frac{\partial o_2}{\partial w_{4,old}} \quad \left\{ \text{Chain Rule} \right\}$$

$$w_{1\text{new}} = w_{1\text{old}} - \eta \left[ \frac{\partial L}{\partial w_{1\text{old}}} \right]$$

$$\frac{\partial L}{\partial w_{1\text{old}}} = \frac{\partial L}{\partial o_2} * \frac{\partial o_2}{\partial o_1} * \frac{\partial o_1}{\partial w_{1\text{old}}}$$

Assignment



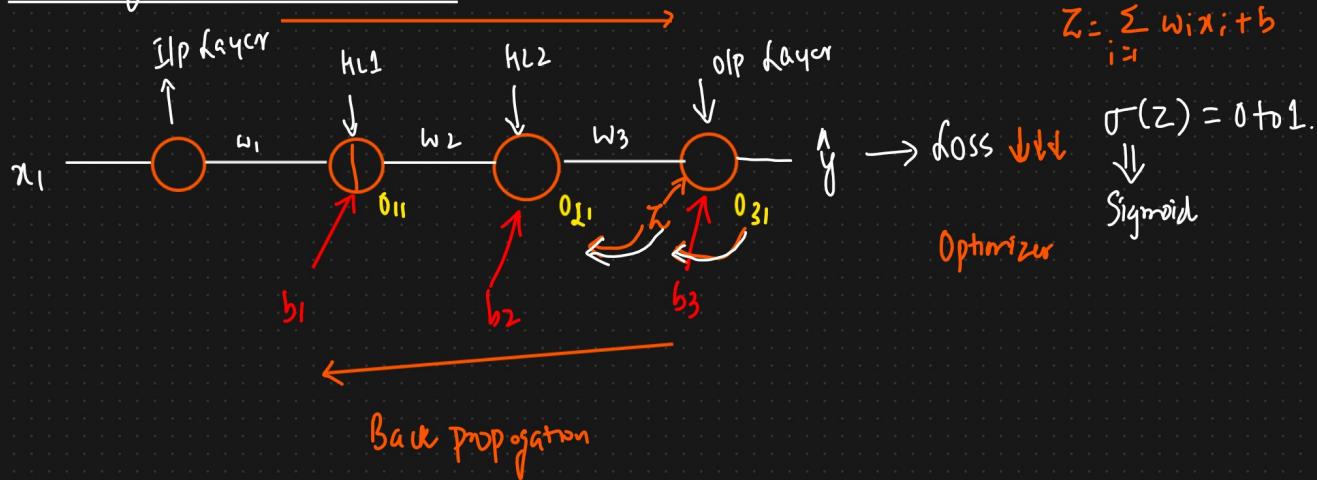
$$w_{1\text{new}} = w_{1\text{old}} - \eta \left[ \frac{\partial L}{\partial w_{1\text{old}}} \right]$$

$$\frac{\partial L}{\partial w_{1\text{old}}} = \left[ \frac{\partial L}{\partial o_{31}} * \frac{\partial o_{31}}{\partial o_{21}} * \frac{\partial o_{21}}{\partial o_{11}} * \frac{\partial o_{11}}{\partial w_{1\text{old}}} \right]$$

+

$$\left[ \frac{\partial L}{\partial o_{31}} * \frac{\partial o_{31}}{\partial o_{22}} * \frac{\partial o_{22}}{\partial o_{12}} * \frac{\partial o_{12}}{\partial w_{1\text{old}}} \right]$$

#### ④ Vanishing Gradient Problem



$$z = \sum_{i=1}^n w_i x_i + b$$

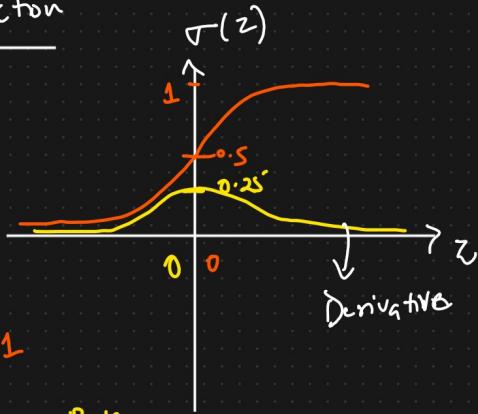
$$\sigma(z) = 0 \text{ to } 1.$$

↓  
Sigmoid

Optimizer

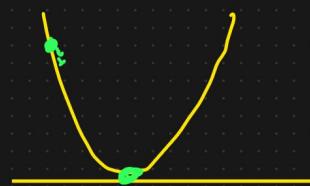
#### ⑤ Sigmoid Activation Function

$$\sigma(z) = \frac{1}{1 + e^{-z}}$$



$$0 \leq \sigma(z) \leq 1$$

$$\partial(\sigma(z)) = 0 \text{ to } 0.25$$



$$w_{\text{new}} = w_{\text{old}} - \eta \left[ \frac{\partial L}{\partial w_{\text{old}}} \right] \Rightarrow [w_{\text{new}} \approx w_{\text{old}}]$$

$0.25 * 0.05 + 0.05 * 0.0025 \approx 0.0025 \approx$  Small value

$$\frac{\partial L}{\partial w_{\text{old}}} = \frac{\partial L}{\partial o_{31}} + \left[ \frac{\partial o_{31}}{\partial o_{21}} * \frac{\partial o_{21}}{\partial o_{11}} + \frac{\partial o_{11}}{\partial w_{11}} \right] \Rightarrow \text{Chain Rule}$$

$$o_{31} = \sigma(w_3 * o_{21} + b_3)$$

$$o_{31} = \sigma(z)$$

$\Rightarrow$  Sigmoid

$$\frac{\partial o_{31}}{\partial o_{21}} = \frac{\partial(\sigma(z))}{\partial z} + \frac{\partial z}{\partial o_{21}} \quad \{ \text{Chain Rule} \}$$

$$= \frac{0 \text{ to } 0.25}{\overbrace{\dots}^{\partial o_{21}}} * \frac{\partial((w_3 * o_{21}) + b_3)}{\partial o_{21}}$$

$$\boxed{\frac{\partial o_{31}}{\partial o_{21}} = 0 \text{ to } 0.25 * w_3} \Rightarrow \text{Small values}$$

## Different Activation

- ① Tanh
- ② ReLU
- ③ Parametric ReLU