

# Session-4

# NEURAL NETWORK - 4

# N LAYER NN

Feb 12, 2024



## AGENDA

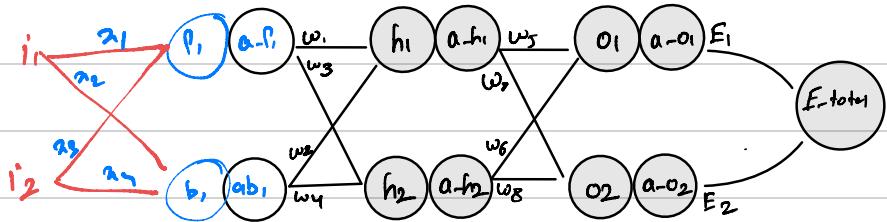
- ① Recap → Backprop  
→ CCE  
→ Linear activation function.

- ② Activation function → Sigmoid  
→ Tanh  
→ Any problems  
→ Relu

- ③ Notation & Tensorflow

# CAN WE GENERALIZE

$$\frac{\partial L}{\partial w_1}$$



$$\frac{\partial E_{\text{total}}}{\partial w_1}$$

$$\frac{\partial E_{\text{total}}}{\partial w_1} \Rightarrow \frac{\partial E}{\partial w_1} \times \frac{\partial w_1}{\partial a_{f1}} \times \frac{\partial a_{f1}}{\partial P_1} \times \frac{\partial P_1}{\partial w_1}$$

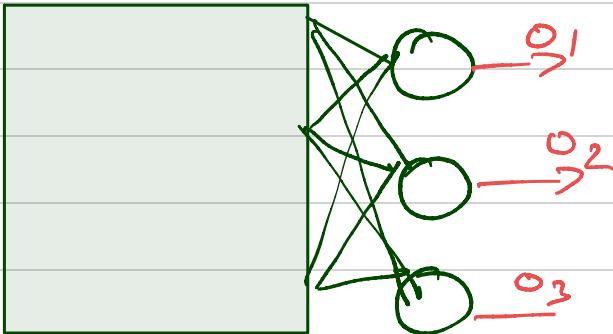
Incomplete

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

Sigmoid  
tanh  
ReLU

Leaky ReLU

Elkh Dog Cat  
 [1 0 0]



Actual Predict

1st row

$$\begin{bmatrix} 0 & 1 & 0 \end{bmatrix} \quad \begin{bmatrix} 0.9 & 0.05 & 0.05 \end{bmatrix}$$

2nd row

$$\begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0.1 & 0.8 & 0.1 \end{bmatrix}$$

3rd row

$$\begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 0.2 & 0.2 & 0.6 \end{bmatrix}$$

$CCE = \sum_i^{actual} p_i \times \log(\hat{y}_i)$

	Actual	Predict
1st row	$[0 \ 1 \ 0]$	$[0.9, 0.05, 0.05]$
$\Rightarrow -$	$-\left[ 0 \times \log(0.9) + 1 \times \log(0.05) + 0 \times \log(0.05) \right]$	
$\Rightarrow -$	<u>1.3</u>	

Row 2 & Row 3

$$-\log(0.1) - \log(0.6) - \log(0.05)$$

$$\Rightarrow -\underline{2.52}$$

$\rightarrow$  Linear activation

$$n \rightarrow f(n) = 2n + 1 \xrightarrow{\text{fines}} g(f(n)) = 3n + 2 \xrightarrow{\text{fines}} g(f(n))$$

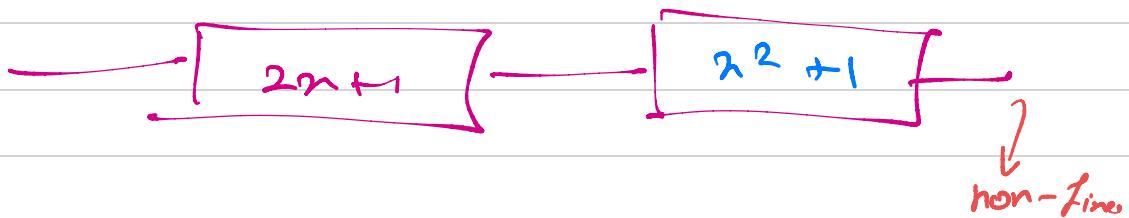
$$g(f(n)) = 3(2n+1) + 2$$

$$\Rightarrow 6n + 3 + 2$$

$$- \quad \text{6n} + 5$$

Line

$$\boxed{n(m) = 6m + 5}$$



## ACTIVATION FUNCTION

① Sigmoid function

→ Domain  $(-\infty, \infty)$

→ Range  $(0 - 1)$

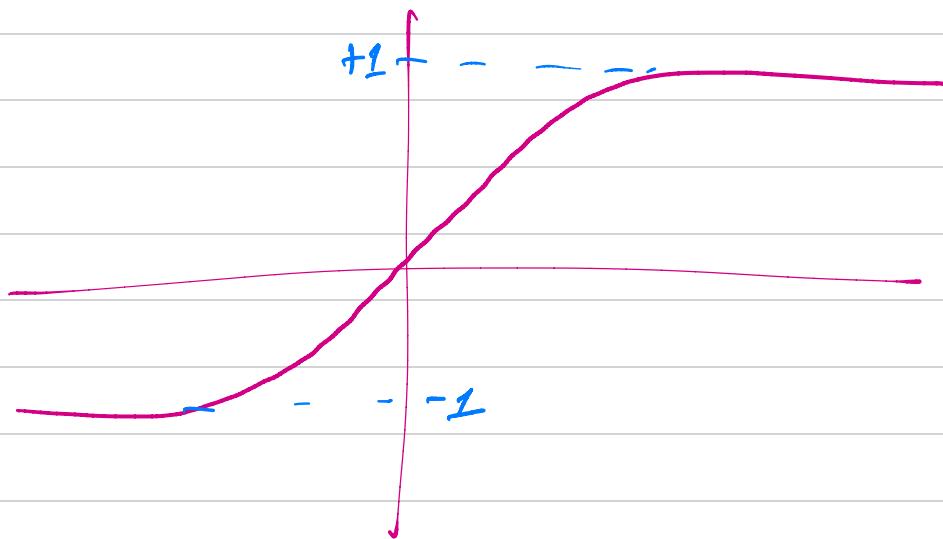
$$\Rightarrow f(m) = \sigma(m) (1 - \sigma(m)) \rightarrow \underline{(0 - 1)}$$

2

Tan - h activation func

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

Domain =  $(-\infty, \infty)$   
Range =  $(-1, 1)$



Always works better than sigmoid

→ Derivative of  $\tanh \rightarrow$   $1 - \overbrace{\tanh^2 x}$   
 $\overbrace{\text{Range} \rightarrow (0, 1)}$

→ Possible problem with tanh / Sigmoid

① Comp. expensive

② Range D.

Range  $\rightarrow (-1, 1) \rightarrow \text{tanh}$

Sigmoid  $\rightarrow (0, 1)$

\* After 2-3 layers of NN training  
gradients  $\rightarrow 0$

$$w_m = w_{old} - \alpha \frac{\partial L}{\partial w_m}$$

s. no.  $\downarrow$   
 $U \cdot U \cdot U \cdot \text{small}$

When  $\alpha$   $\ll$  1  
 $10^{-5}$   $10^{-6}$

Vanishing Gradients

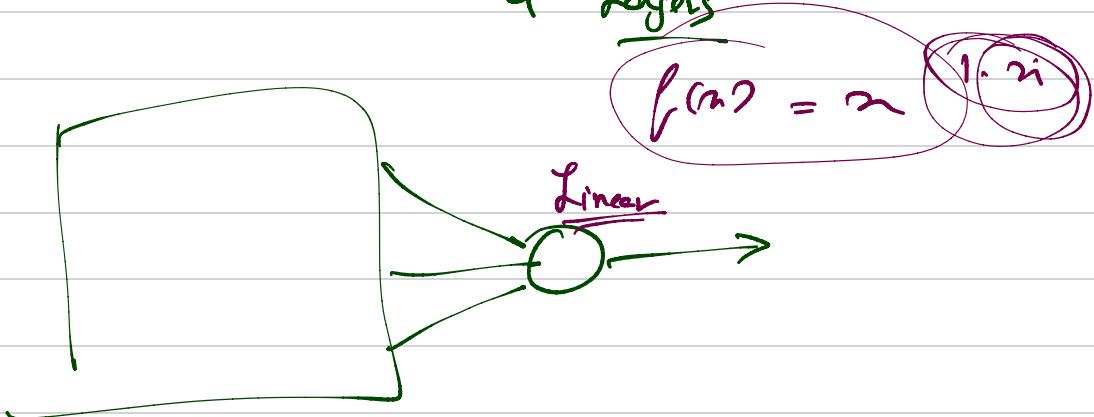
I'm not able to optimize

Q. Why weights of Neuron randomly initialized, why not zero?

$\frac{\partial L}{\partial w}$  also zero

$w$

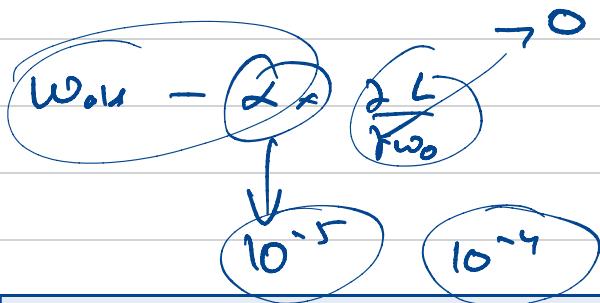
→ For almost 2 decades we couldn't train network for more than 4 layers



How do we deal with Vanishing  
gradient

→ Relu

$$f(x) = \begin{cases} \max(0, x) & x > 0 \\ 0 & x \leq 0 \end{cases}$$



gradient  $\rightarrow \frac{\partial L}{\partial w / \text{Parameter}}$

$$f(x) = \underbrace{\max(0, x)}_{= \begin{cases} x & x > 0 \\ 0 & x \leq 0 \end{cases}}$$

$f'(x) = \{$   $x$  if  $x > 0$  Not differentiable

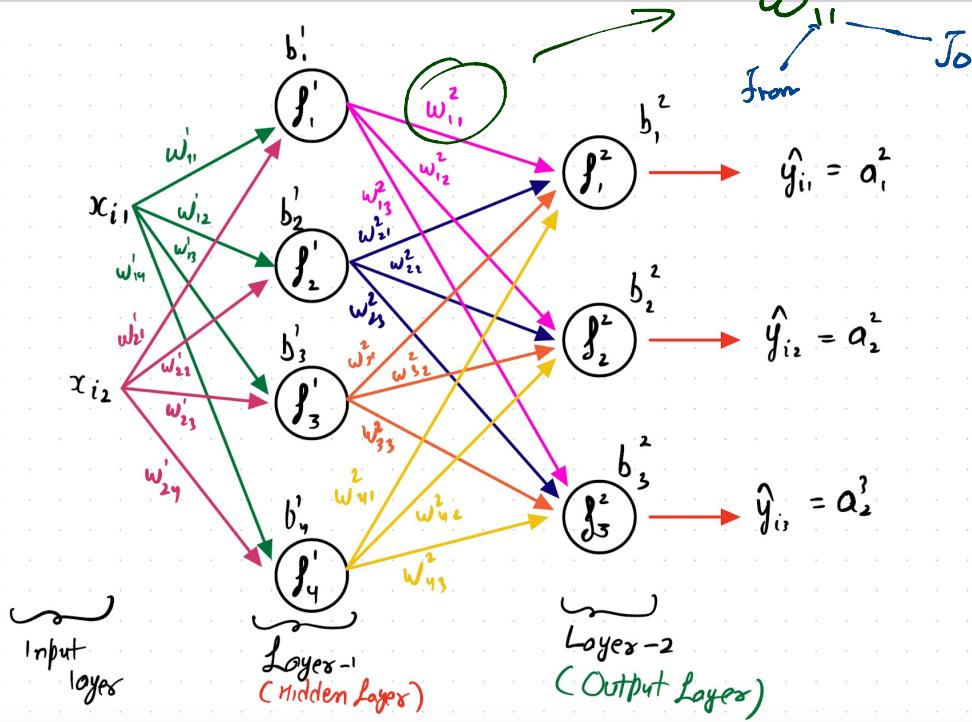
Leaky ReLU  $\rightarrow$

$$g(x) = \begin{cases} x & \text{if } x > 0 \\ 0.1x & \text{if } x \leq 0 \end{cases}$$

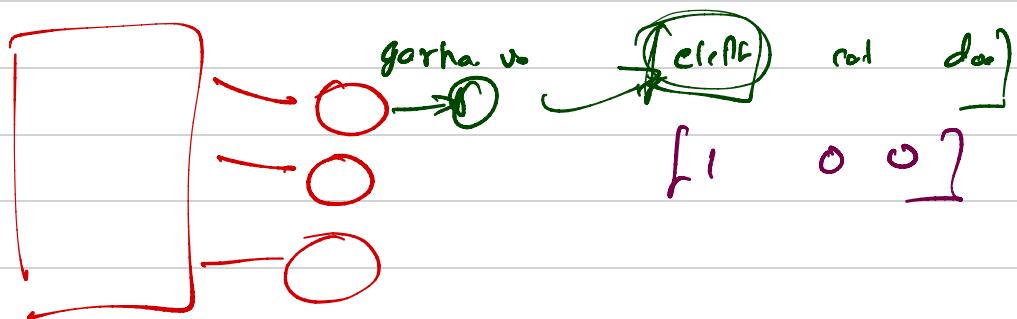
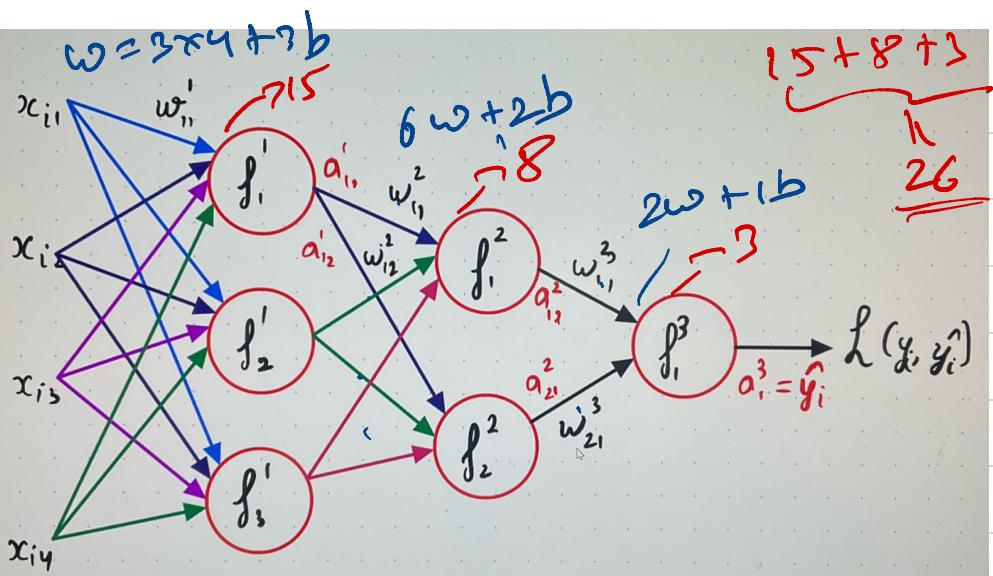
ReLU still performs activation

$$g(x) = \begin{cases} 1 & \text{if } x > 0 \\ x & \text{if } x \leq 0 \end{cases}$$

# NOTATIONS



$$w_{ij}^l = \text{?}$$



$SGD \rightarrow$  Gradient descent  
 $Adam \rightarrow$  Most commonly used  
 Optimizer

