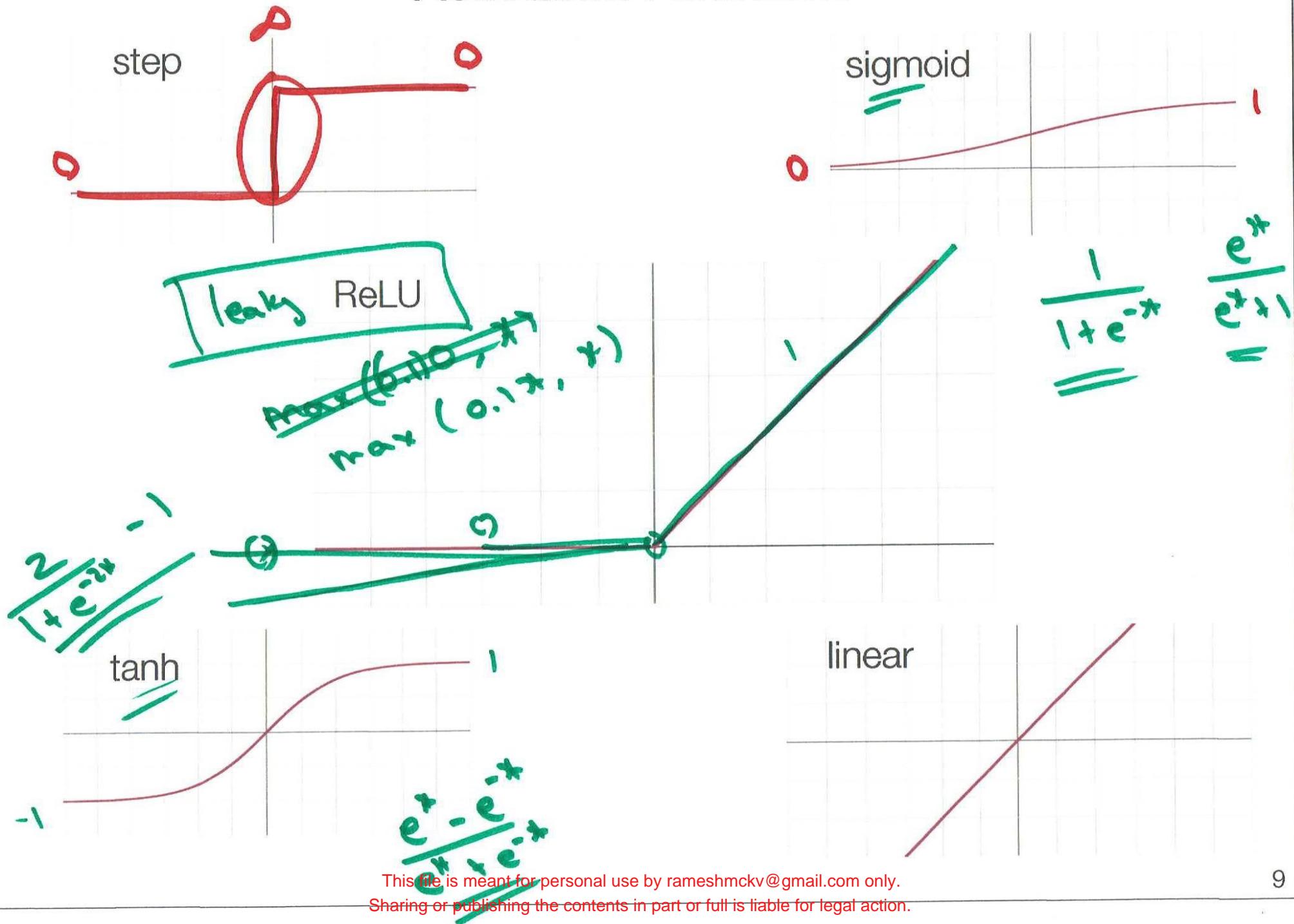
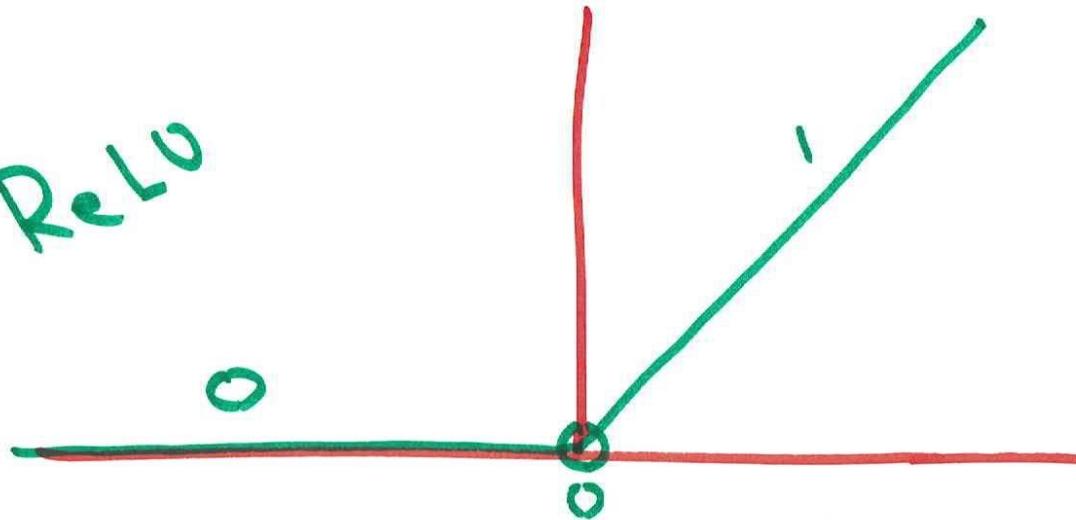


# Activation Functions

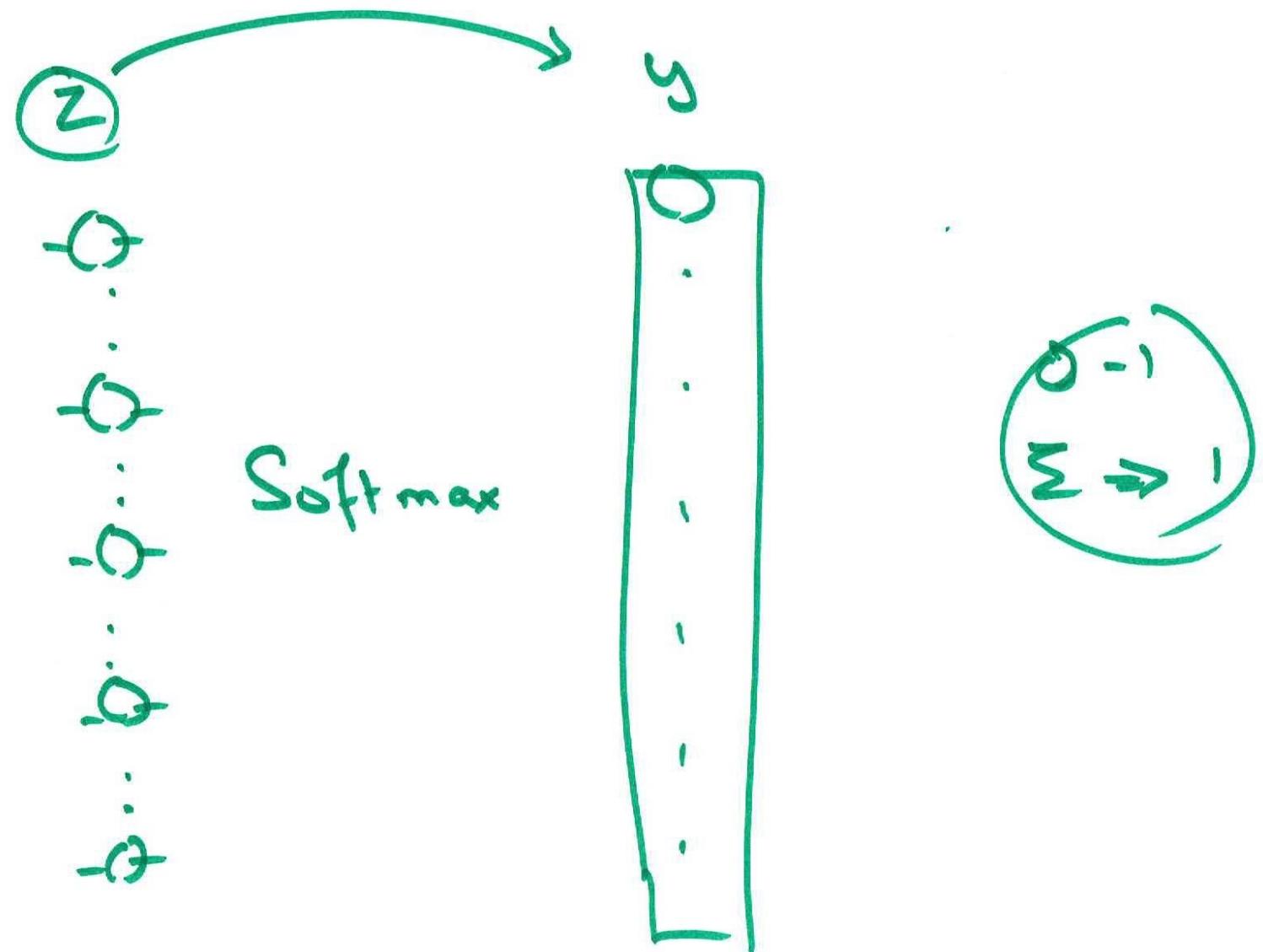


ReLU



$$f(x) = \begin{cases} x & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

$$f(x) = \max(0, x)$$



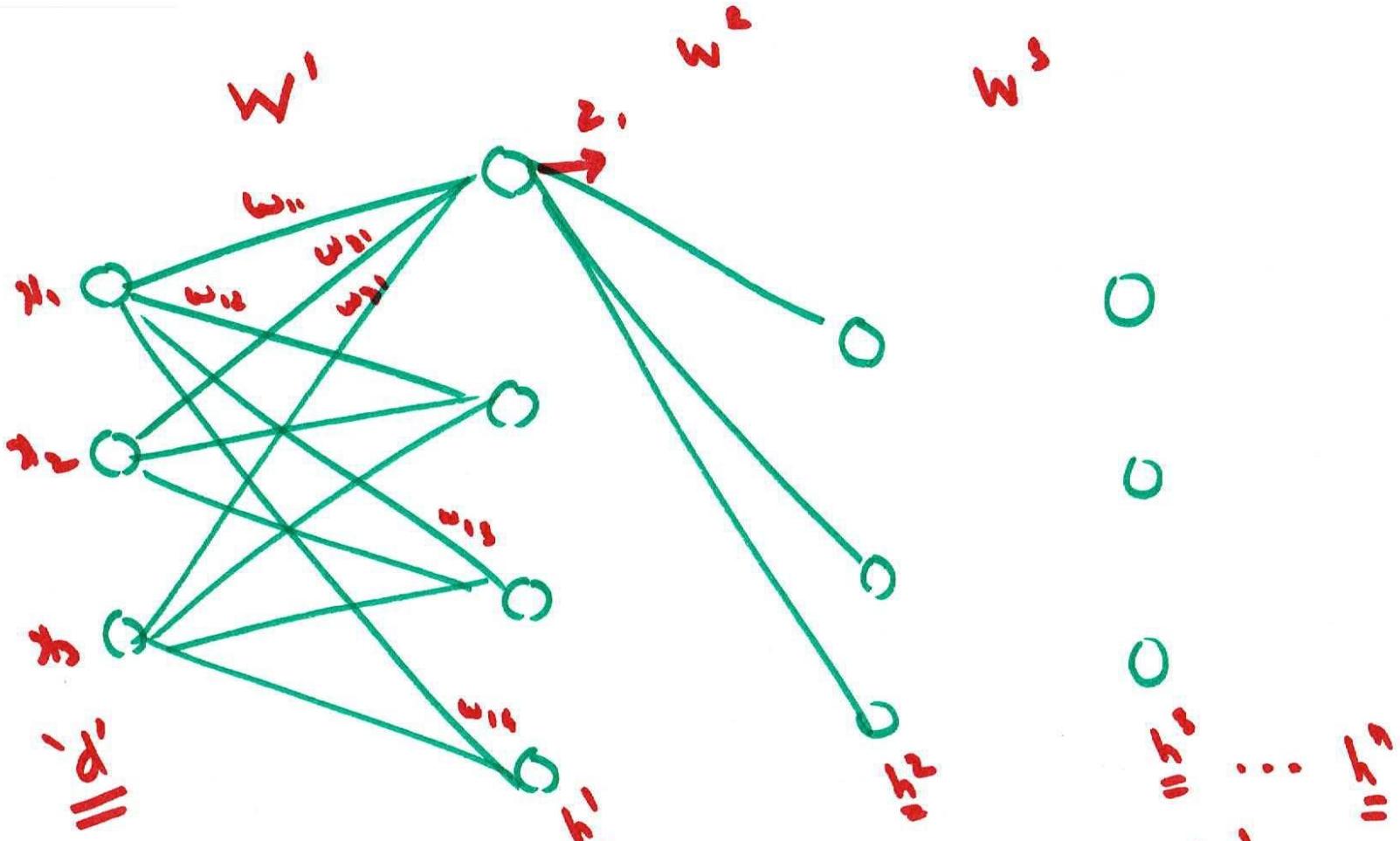
$z$

$y$

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

$$\omega^{\text{new}} = \omega^{\text{old}} - \eta \nabla_{\omega} l(\omega)$$

$$= \omega^{\text{old}} - \frac{1}{N} \eta \sum \nabla_{\omega} l_i(\omega)$$



$$z_7 = \text{f}(w_{03}x_0 + w_{13}x_1 + w_{23}x_2 + b)$$

$$\boxed{z_j = \text{f}(\sum_i w_{ij}x_i + b_j)}$$

$$W' = \begin{pmatrix} w_{00} & w_{01} & w_{02} & \dots & w_{0d} \\ w_{10} & w_{11} & w_{12} & \dots & w_{1d} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w_{h0} & w_{h1} & w_{h2} & \dots & w_{hd} \\ \dots & \dots & \dots & \dots & \dots \\ w_{dh} & \dots & \dots & \dots & w_{dd} \end{pmatrix}$$

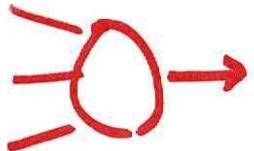
$\text{h' } \times d$   
matrix

$$x = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_d \end{pmatrix}$$

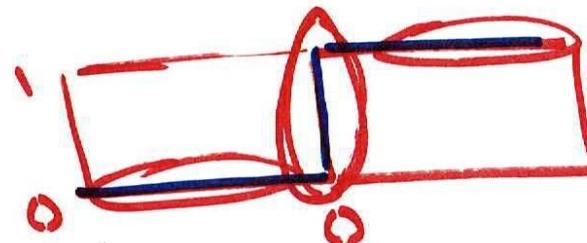
$$\hat{y} = f(-f(f(W^1)^\top W^2)^\top f(W' x + b') + b^2) + b^3 \dots$$

Diagram illustrating the dimensions of the layers:

- $W'$  is  $h' \times d$ .
- $x$  is  $d \times 1$ .
- The result of  $W' x + b'$  is  $h' \times 1$ .
- The result of  $f(W' x + b')$  is  $h' \times 1$ .
- The result of  $f(f(W^1)^\top W^2)^\top f(W' x + b')$  is  $h^2 \times h^1$ .
- The result of  $-f(f(W^1)^\top W^2)^\top f(W' x + b')$  is  $h^2 \times h^1$ .
- The final result  $\hat{y}$  is  $h^2 \times 1$ .

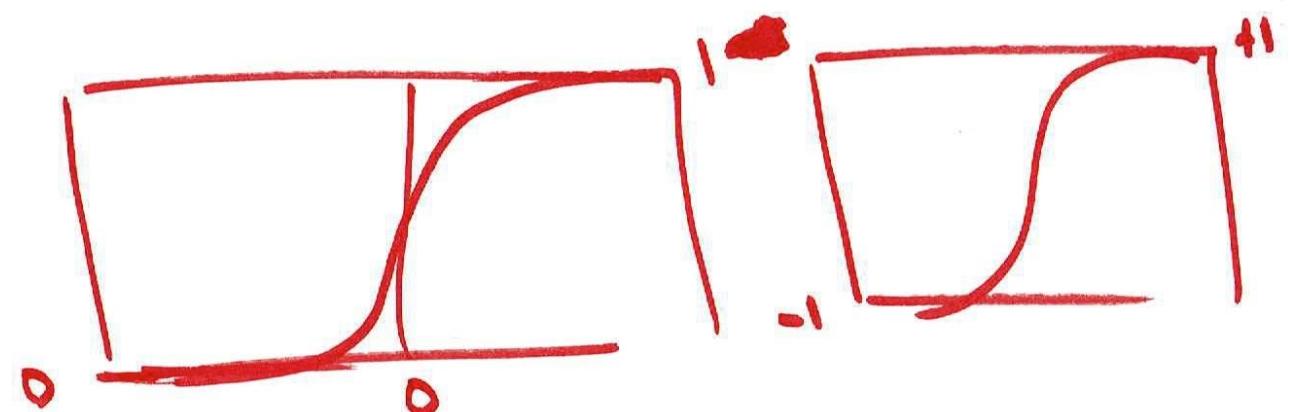


Step ( $w_1x_1 + w_2x_2 + \dots + w_nx_n + b$ )



Sigmoid

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

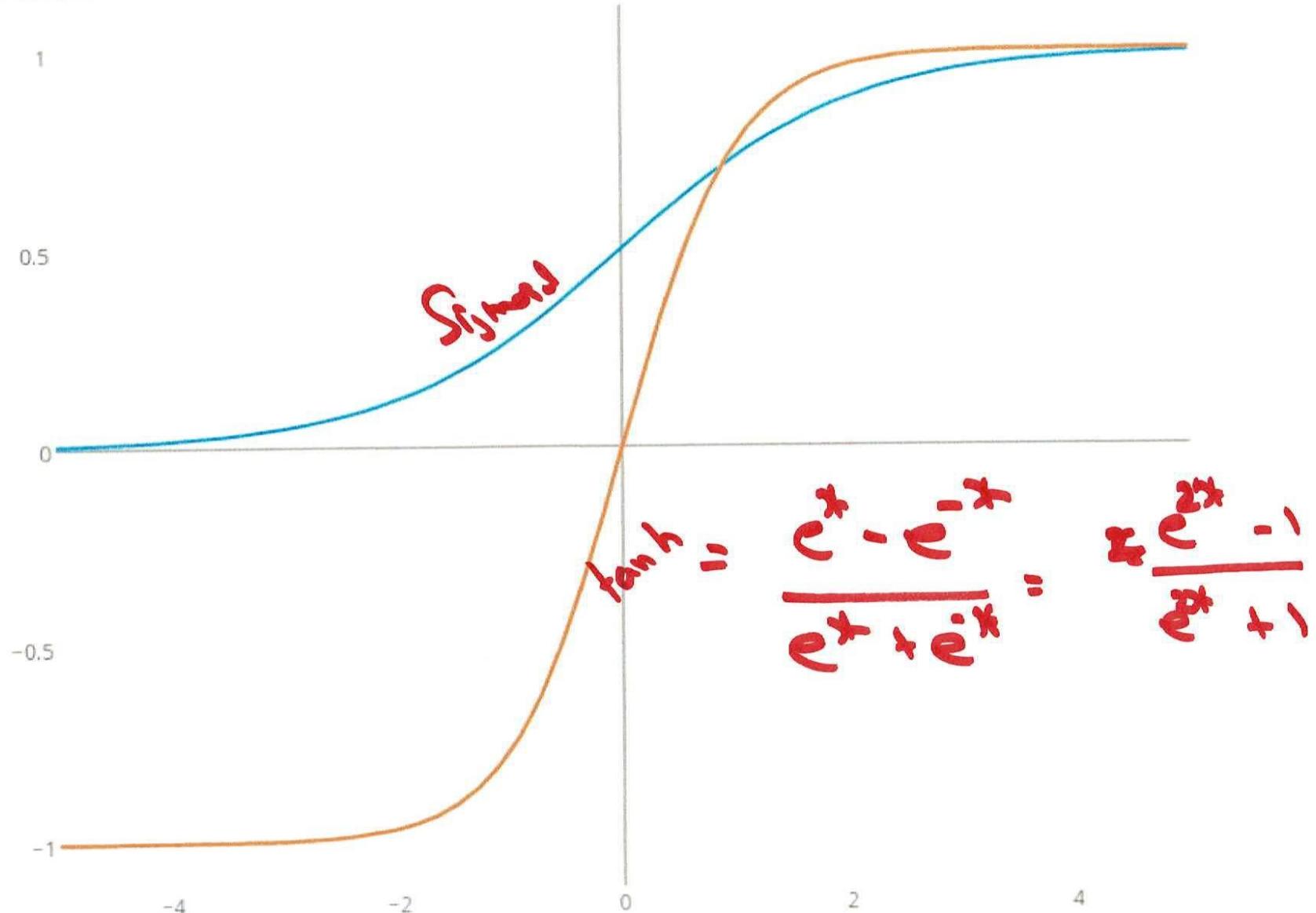


$$2\sigma(x) - 1$$

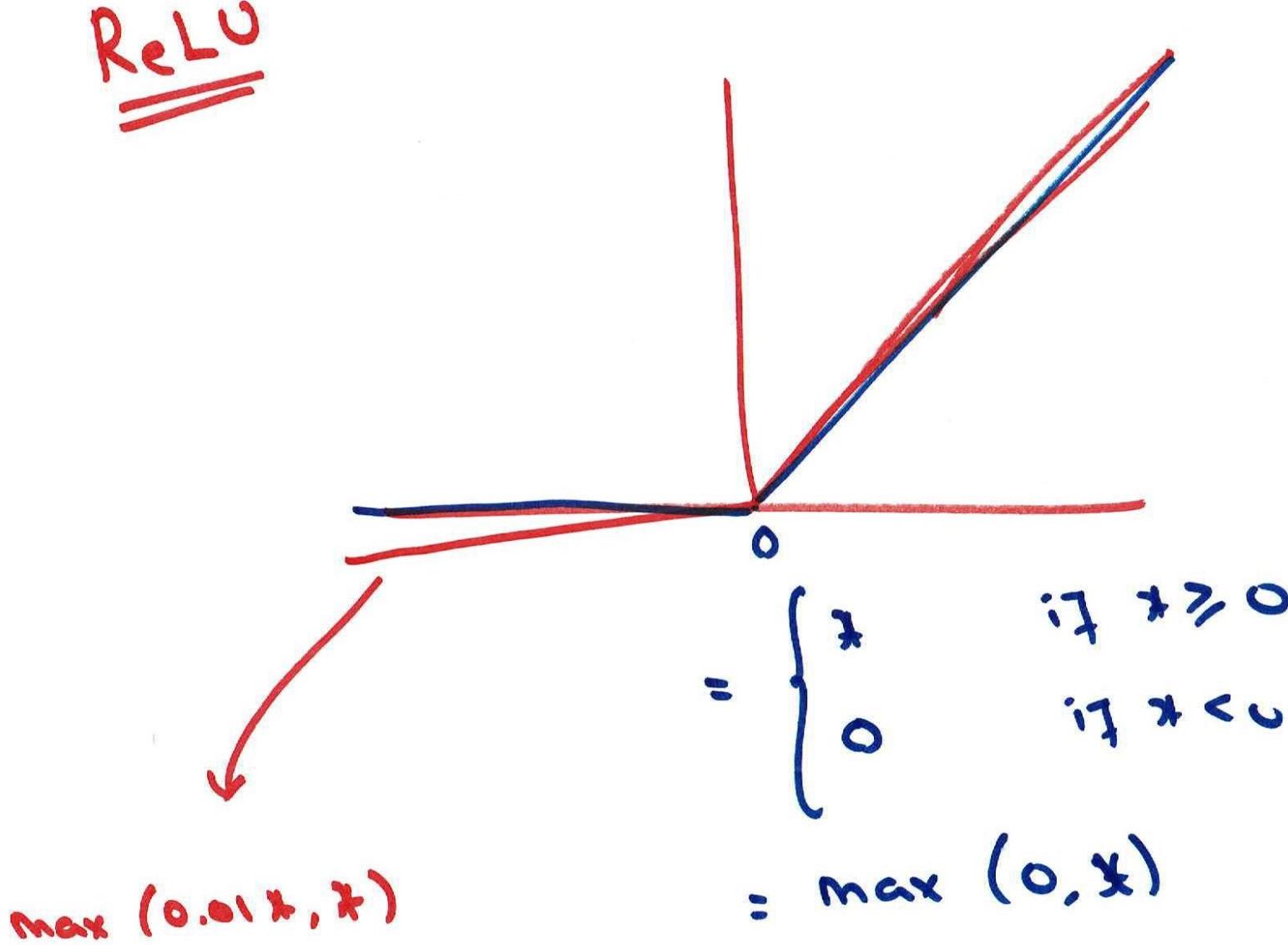
$$\tanh = 2\sigma(2x) - 1$$



Sigmoid function  
Tanh function

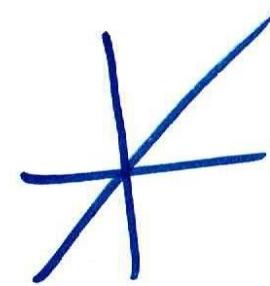


ReLU

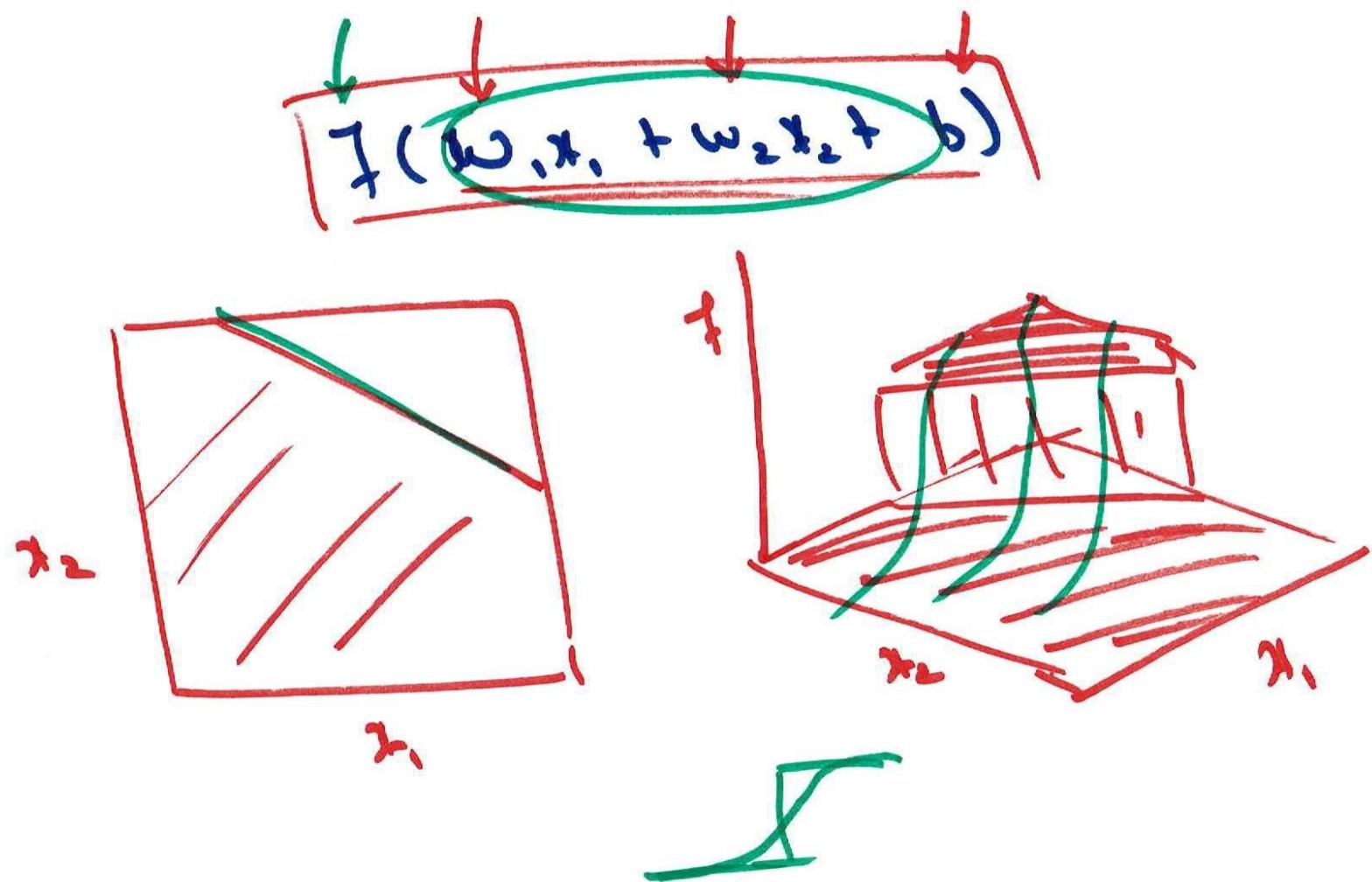


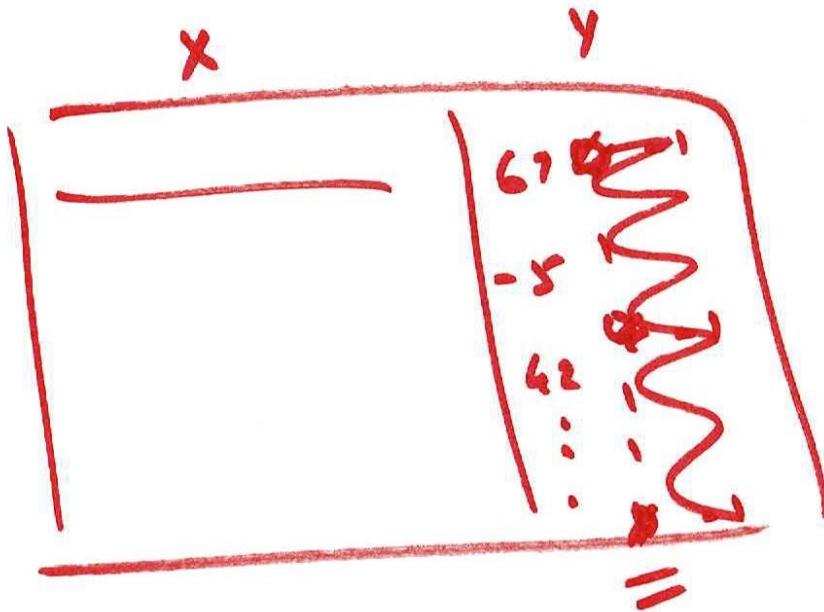
linear

= \*



$$\text{Step } = f(x) = \begin{cases} 1 & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$





Output nodes

Classification

Sigmoid, tanh  
Softmax

hidden layer

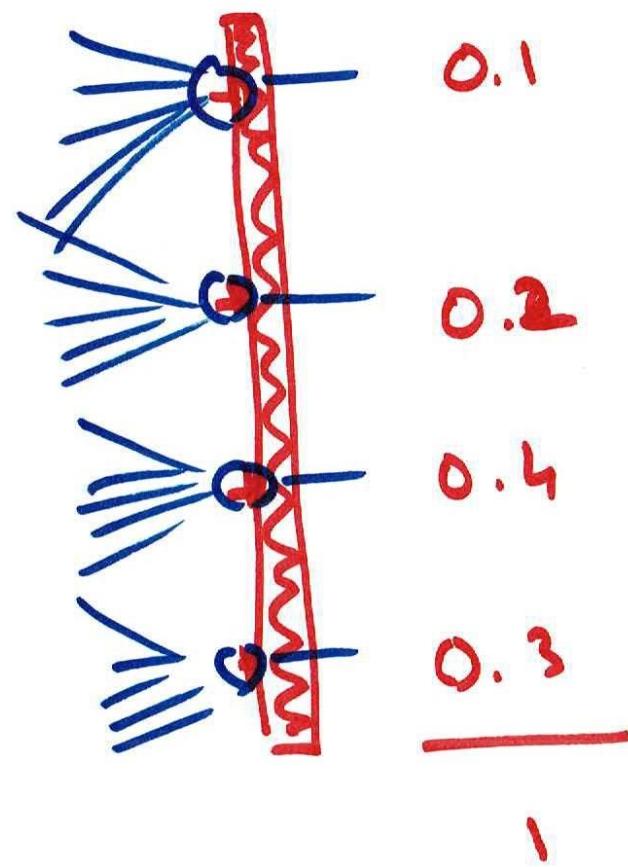
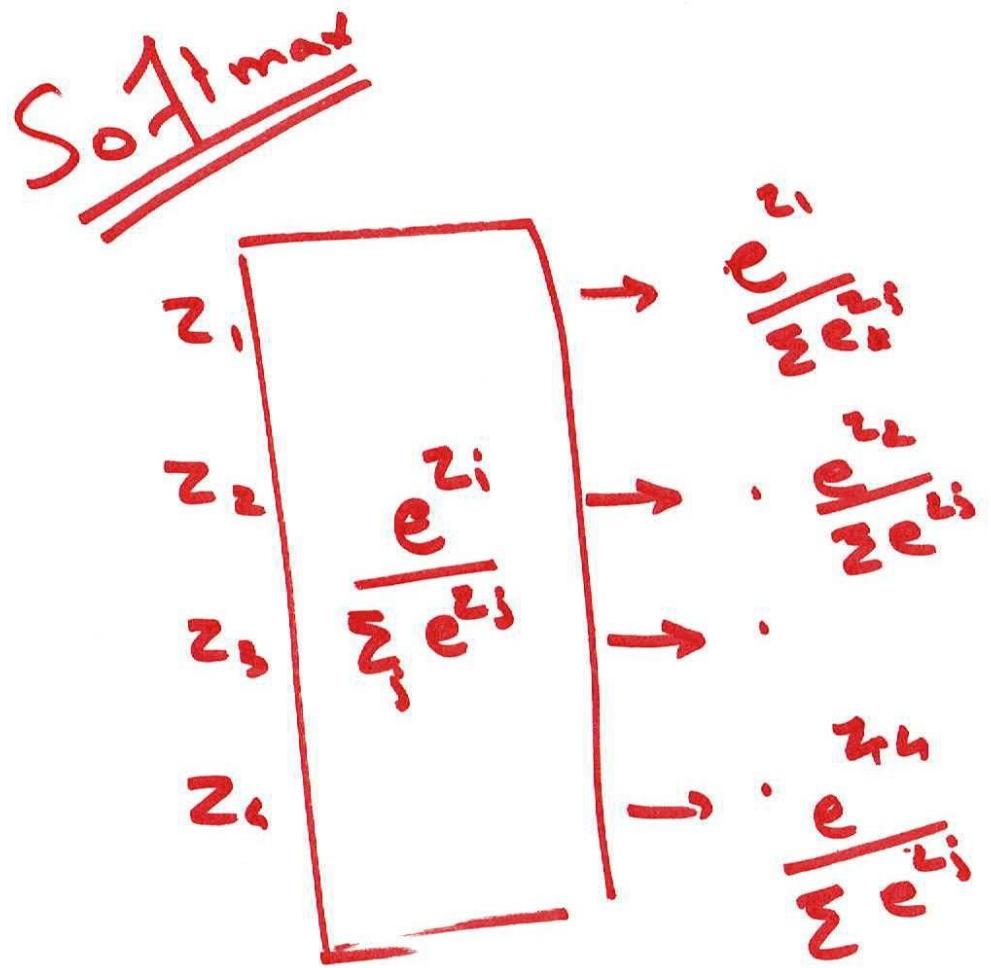
Sigmoid ✓  
tanh ✓

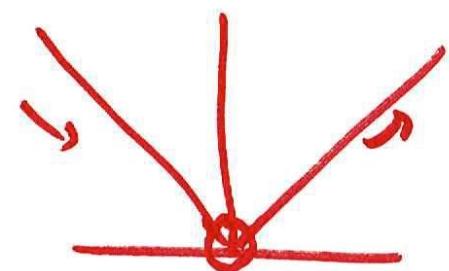
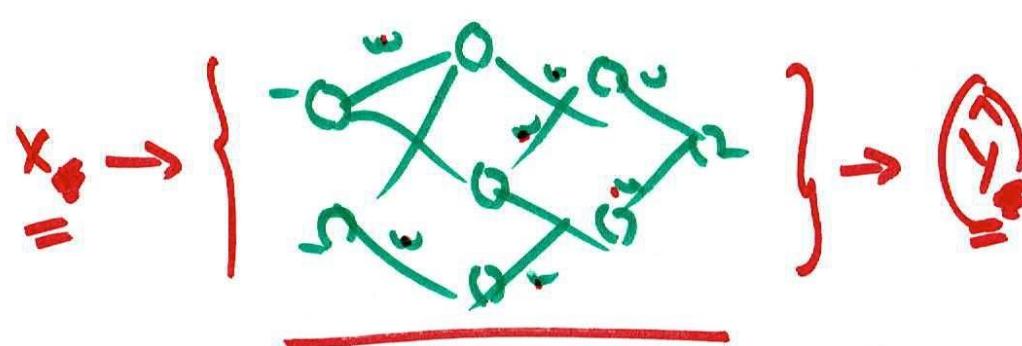
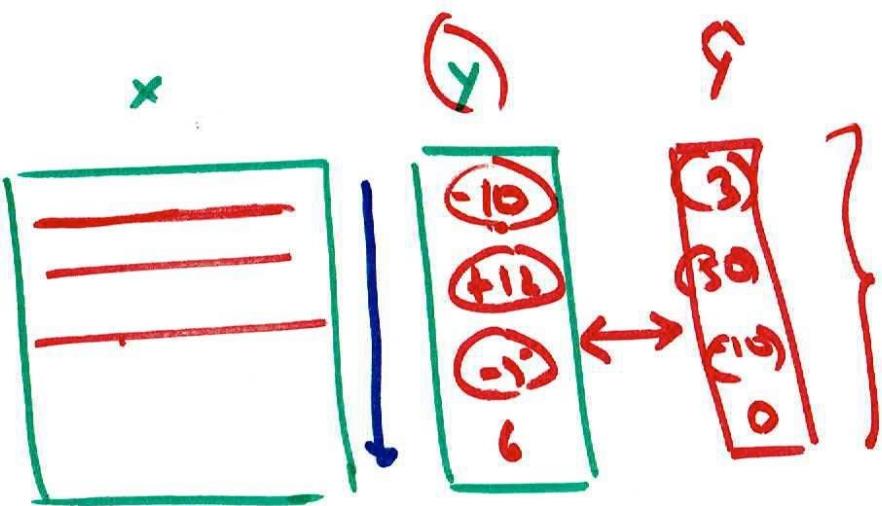
ReLU ✓

linear

$$\text{Reg} \rightarrow \underline{\text{linear}}$$

$$\hat{a} + \hat{b} (a + b x)$$





Loss Function

Reg

$$L(y, \hat{y}) = \frac{1}{n} \sum_i (y_i - \hat{y}_i)^2$$

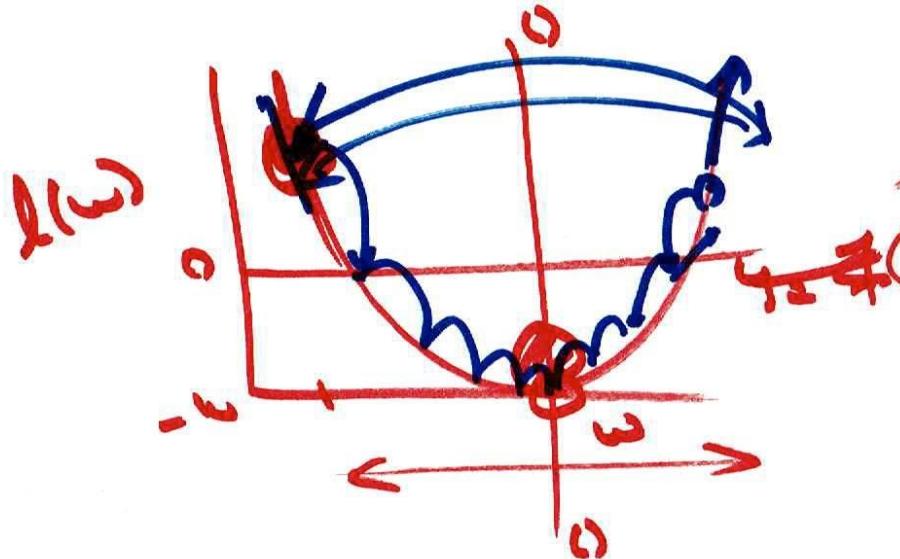
$$L(y, \hat{y}) = L(\omega)$$

$L_2$  loss  
MSE  
SSE

Classification  $L(y, \hat{y}) = -[y_i \log(\hat{y}_i) + (1-y_i) \log(1-\hat{y}_i)]$

Cross entropy loss

How  $\min_{\underline{w}} \underline{L}(\underline{y}, \underline{\hat{y}})$  by changing  $w^1, w^2, \dots, w^n$



$$y = x^2 - 10 = -10$$

$$\frac{dy}{dx} = 2x = 0$$

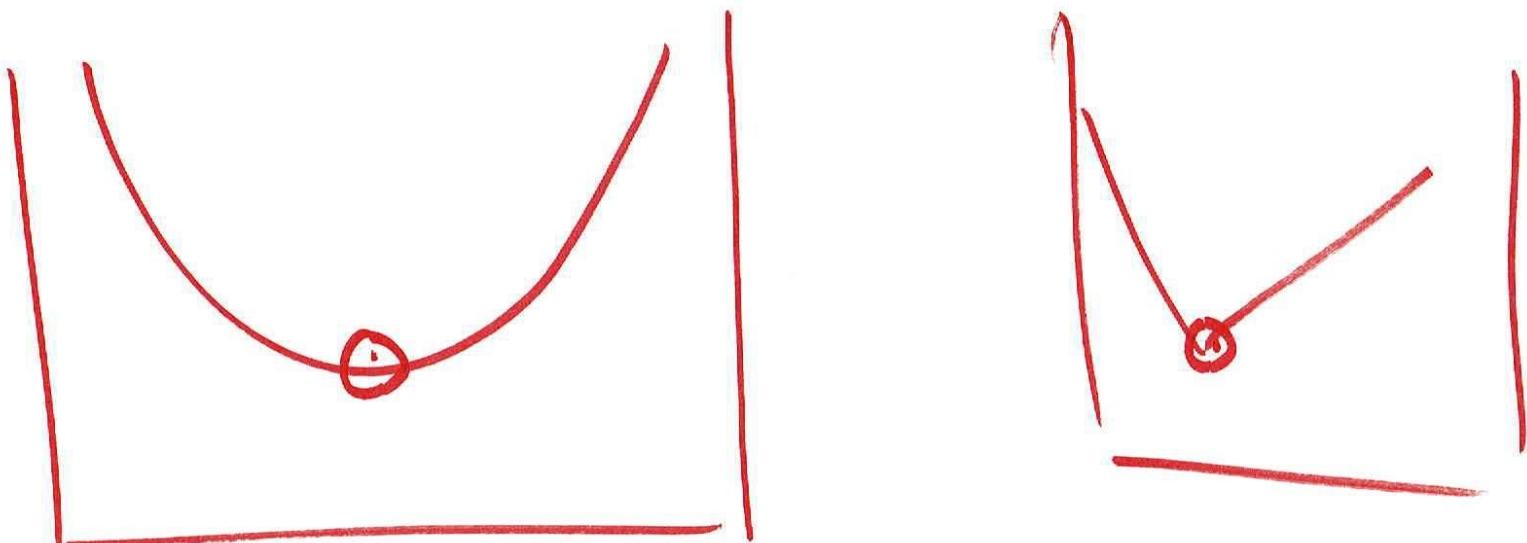
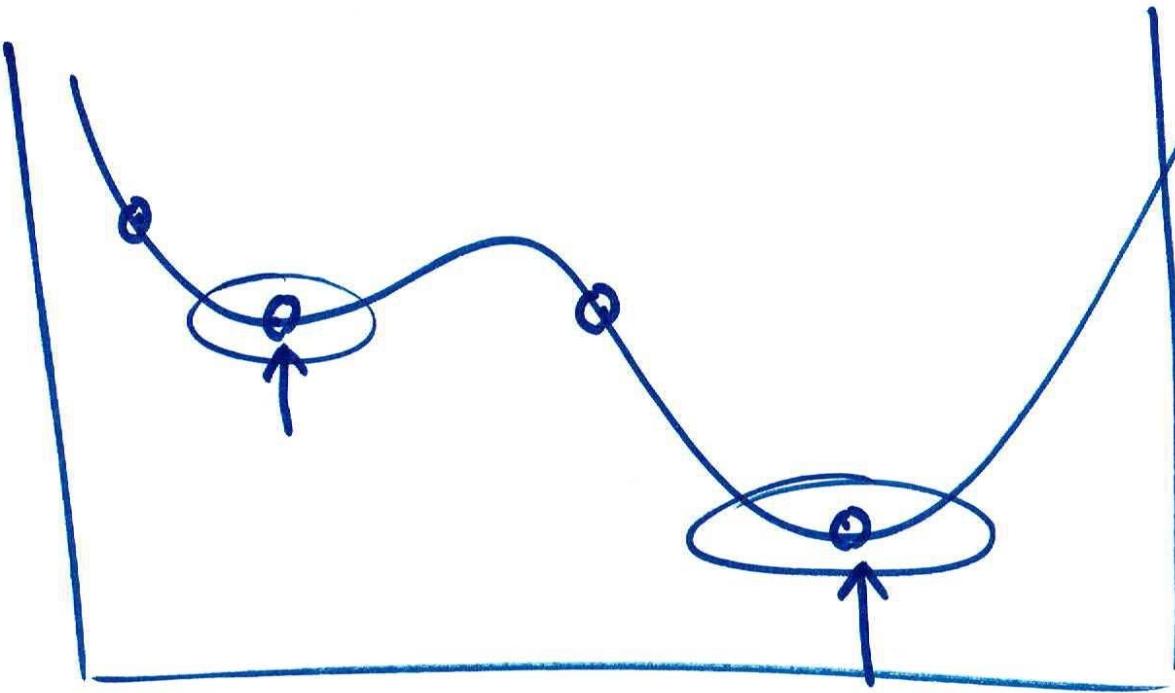
$$x = 0$$

$$\frac{dL}{dw} = \boxed{\text{Diagram showing a trapezoid with zero area}} = 0$$

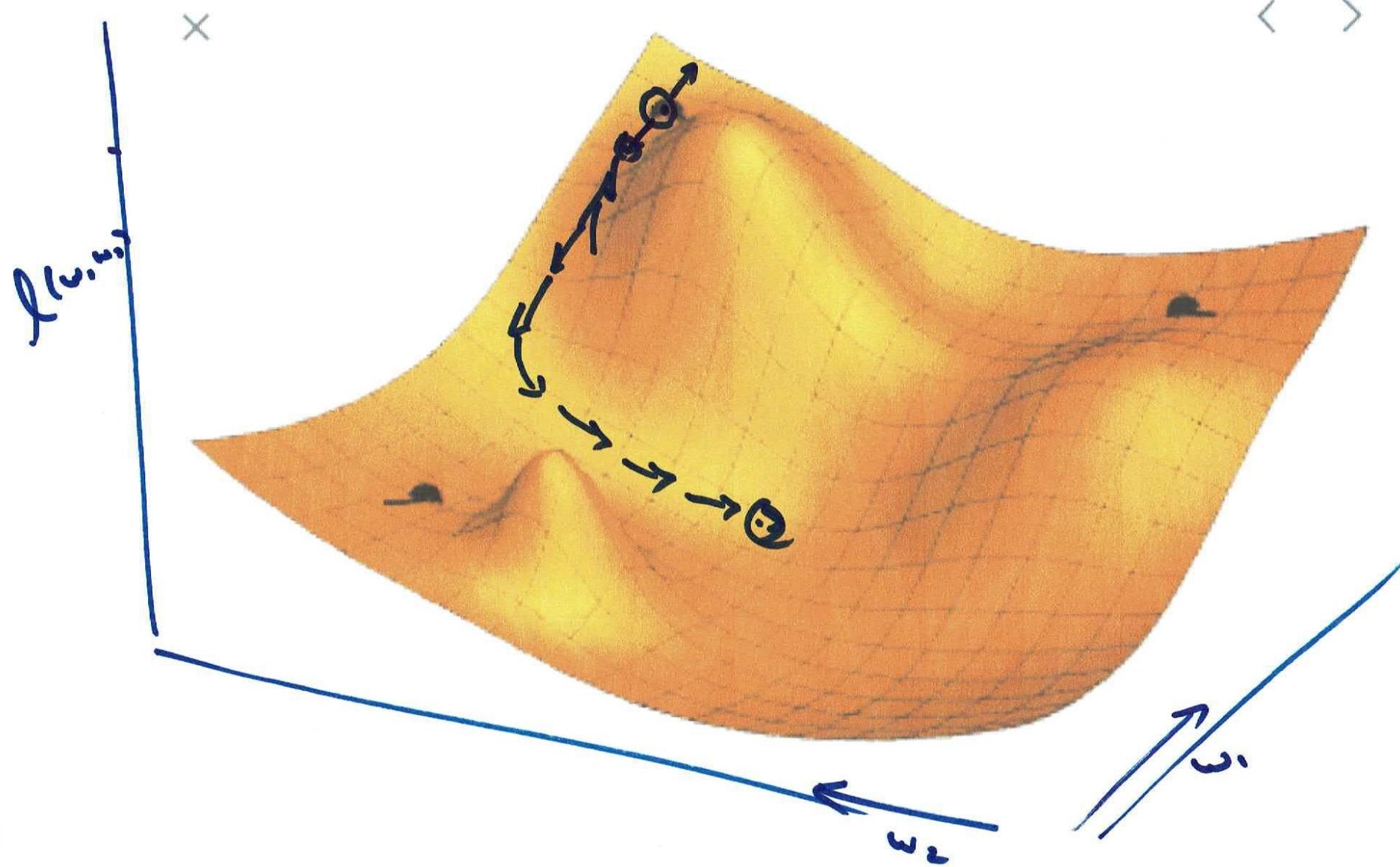
$$\frac{dl}{dw}$$

$$\underline{w}^{\text{new}} = \underline{w} - \eta \nabla_{\underline{w}} l$$

↑ learning rate



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$$\text{Eq} \quad \omega^{\text{new}} = \omega^{\text{old}} - \frac{\eta \nabla_{\omega} l(\omega)}{N}$$

$$= \omega^{\text{old}} - \frac{1}{N} \sum \nabla_{\omega} l_i(\omega) \leftarrow$$

(SGD)

$$\omega^{\text{new}} = \omega^{\text{old}} - \eta \nabla_{\omega} l_i(\omega) \leftarrow$$

$$\boxed{\omega^{\text{new}} = \omega^{\text{old}} - \frac{1}{N} \eta \sum \nabla_{\omega} l_i(\omega)}$$

↓  
over  
a min  
batch

Loss

function of  $\omega$  ( $l(\omega)$ )

$$L = \frac{1}{N} \sum \text{Loss} \left( \underbrace{y_i - f^{\circ}(\dots f(w^2 \cdot f'(w^1 x + b') + b^2) \dots)_i)}_2 \right)$$

Chain Rule

$$f(g(h(x)))$$

$$\frac{df}{dx} = \boxed{\frac{df}{dg} \cdot \frac{dg}{dh} \cdot \frac{dh}{dx}}$$

