

# Weight, Bias

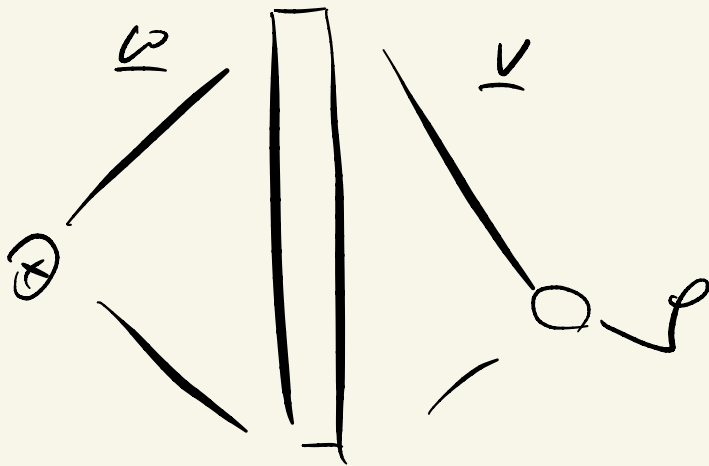
$$\begin{bmatrix} \underline{w}_1 & \underline{w}_2 & \dots & \underline{w}_n \\ | & | & & | \end{bmatrix} + \begin{bmatrix} \underline{b} & \underline{b} & \dots & \underline{b} \end{bmatrix}$$



$$\sigma \left( \begin{bmatrix} \underline{w}_1 + \underline{b} & \underline{w}_2 + \underline{b} & \dots & \underline{w}_n + \underline{b} \\ | & | & & | \end{bmatrix} \right)$$



$$\Rightarrow v \begin{bmatrix} \sigma(\underline{w}_1 + \underline{b}) & \dots & \sigma(\underline{w}_n + \underline{b}) \\ | & & | \end{bmatrix}$$



$$I \mathcal{J}_1 \mathcal{J}_2 \dots \mathcal{J}_k I$$

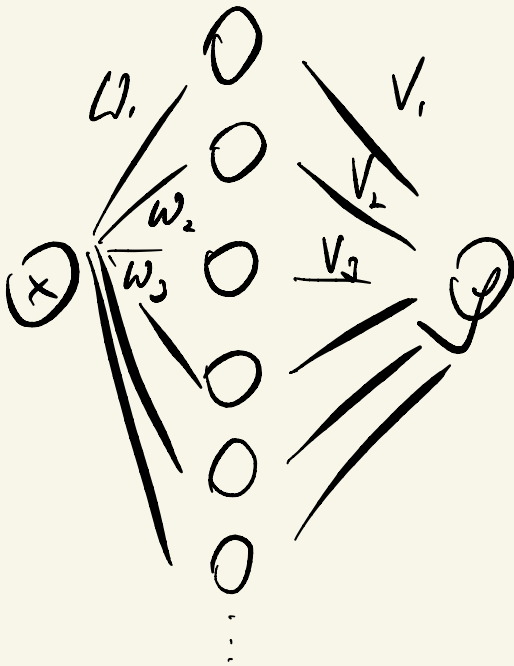
$$= V^T \begin{bmatrix} \delta(\omega_{d_1+b}) & \delta(\omega_{d_2+b}) & \dots \\ \vdots & \vdots & \\ \dots & \delta(\omega_{d_k+b}) & \end{bmatrix}$$

$$f^T = v^T \sigma(\omega d + \underline{b})$$

$$d \longrightarrow \sigma(\omega d + \underline{b})$$

$$\left\{ \begin{array}{l} \sigma_{\text{ReLU}} = \max(d, 0) \end{array} \right.$$

$$\sigma_{\text{sigmoid}} = \frac{1}{1 + e^{-d}}$$



$$\sigma \left( \begin{bmatrix} \omega_1 d \\ \omega_2 d \\ \vdots \\ \omega_n d \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \right)$$

$$\sigma(\underline{v}) = \begin{bmatrix} \sigma(v_1) \\ \sigma(v_2) \\ \vdots \\ \sigma(v_n) \end{bmatrix}$$

$$a = \sigma(\underline{w}d + \underline{b})$$

$$J = V_1 a_1 + V_2 a_2 + \dots + V_n a_n \\ = V \cdot a$$

$$J = V^T a$$

$$J = V^T (\sigma(\underline{w}d + \underline{b}))$$

$$J_1 = V^T \overset{\uparrow}{\sigma}(\underline{w}d_1 + \underline{b})$$

$$J_2 = V^T \sigma(\underline{w}d_2 + \underline{b})$$

$$d = [1, 2, \dots, 100]^T$$

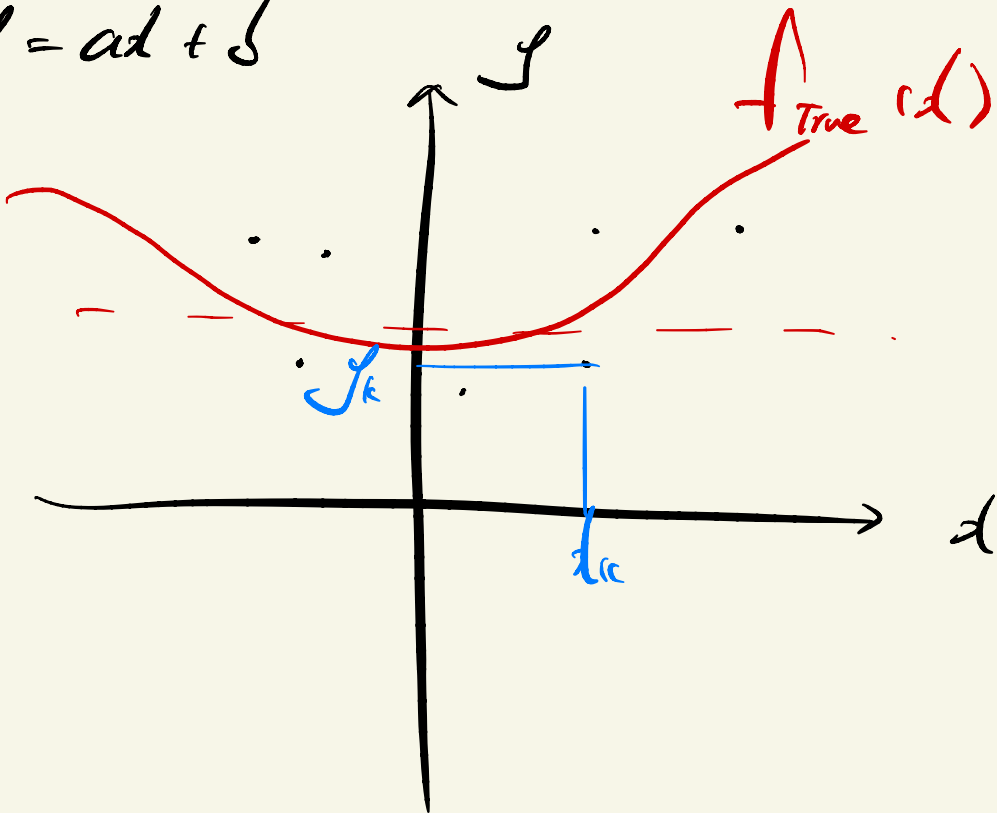
$$J = [0, \dots, 0]^T$$

$$J_i = \underline{V}^T (\underline{w}d_i + \underline{b})$$

$$\underline{W} \underline{d}^T \Rightarrow \begin{bmatrix} W_1 & d_1 & \dots & d_k \\ \vdots & & & \\ W_n & & & \end{bmatrix} \quad \begin{matrix} 1 \times k \\ k \times 1 \end{matrix}$$

**Linear Regression** : line of best fit

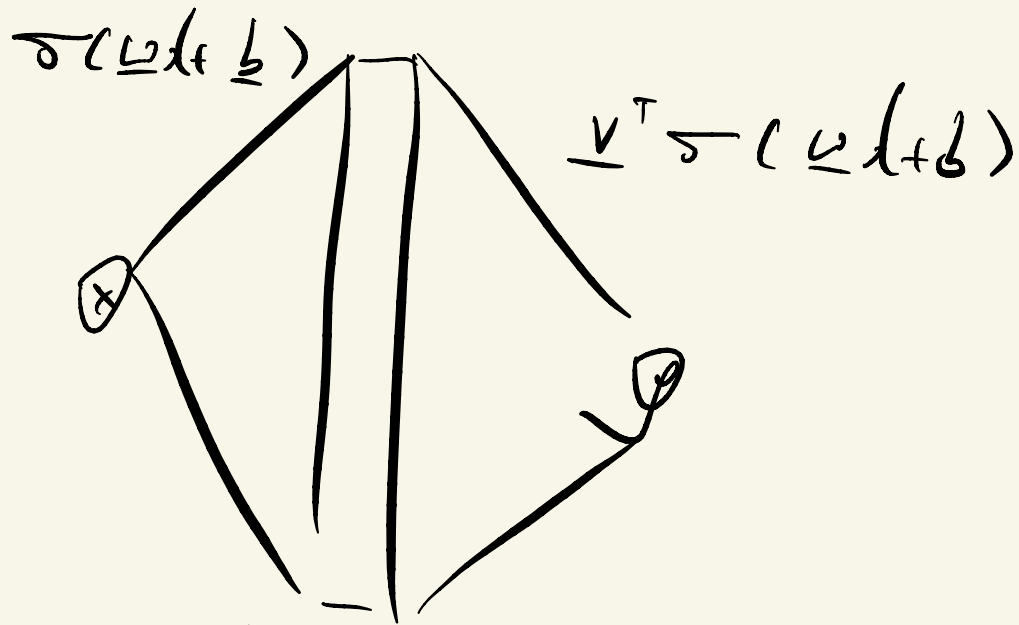
$$y = ad + b$$



# NN Regression : Best NN

$$f = f_{nn}(x_i; \theta)$$

$$\theta = w, b, v$$



$$\min_{w, b, v} \text{Loss} = \frac{1}{N} \sum_{i=1}^N |f_{nn}(x_i; \theta)|$$

-  $|f_i|^2$  should be small