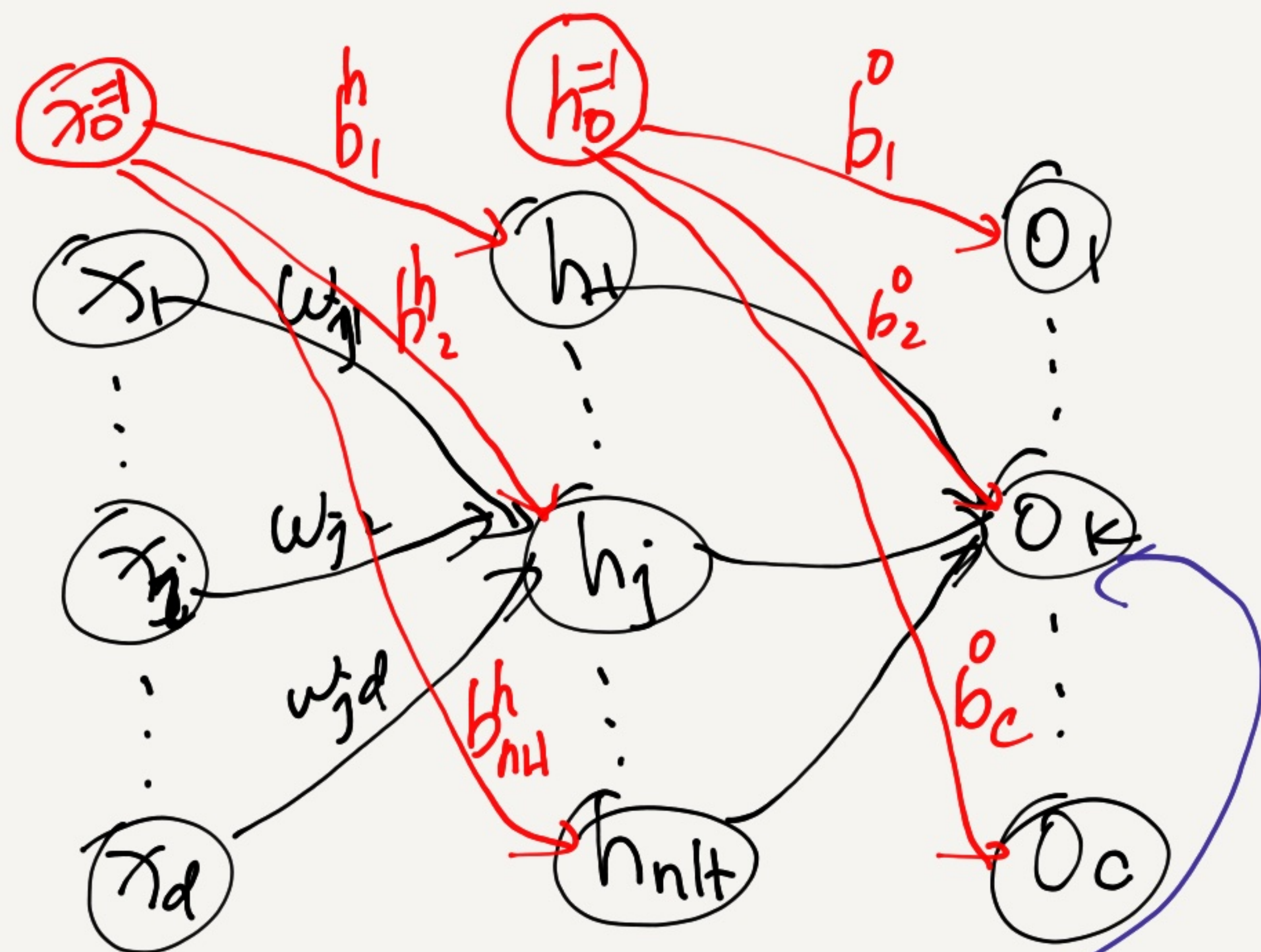


# Lecture 11 : NN Implementation

## 1. Bias nodes in NNs

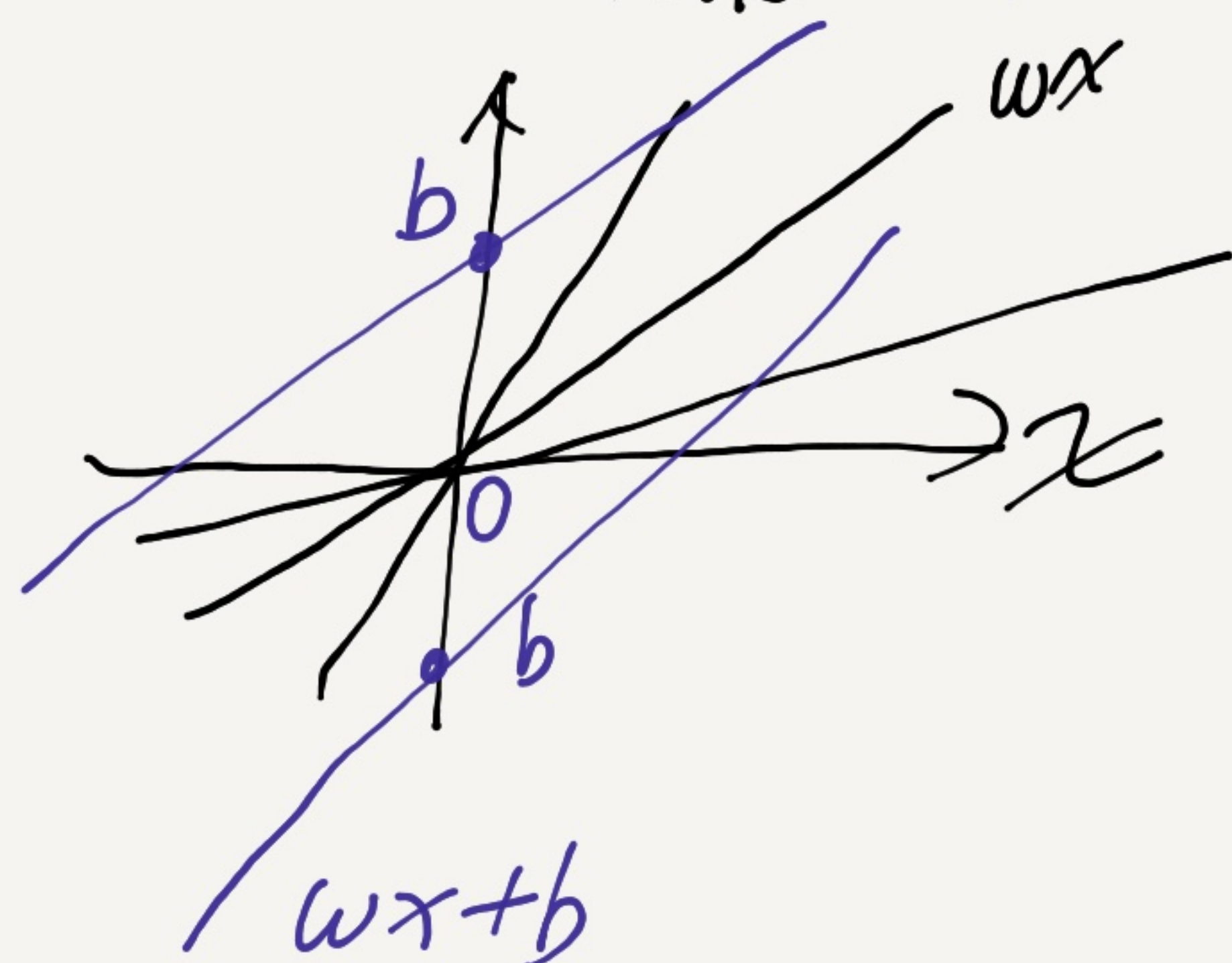


$$\nabla_{w_{kj}} L = \delta_k \cdot h_j \rightarrow \nabla_{b_k^o} L = \delta_k \leftarrow (h_j = h_0 = 1)$$

$$\nabla_{w_{ji}} L = \delta_j \cdot x_i \rightarrow \nabla_{b_j^h} L = \delta_j \leftarrow (x_i = x_0 = 1)$$

Net input :  $Net_j = w_{j1} \cdot x_1 + w_{j2} \cdot x_2 + \dots + w_{jd} \cdot x_d + b_j$

$$Net_k = w_{k1} \cdot h_1 + w_{k2} \cdot h_2 + \dots + w_{knh} \cdot h_{nh} + b_k$$



without bias nodes:

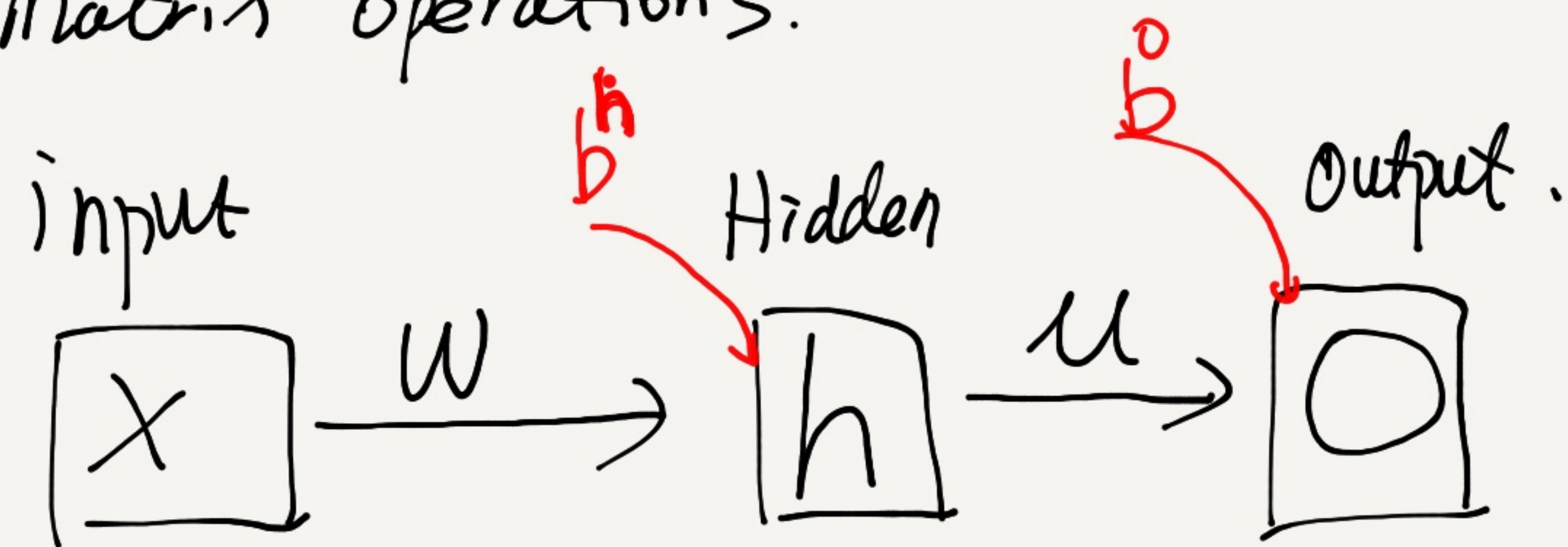
$$d \times n_H + n_H \times C$$

with the bias nodes:

$$(d+1) \times n_H + (n_H+1) \times C$$



2 Use Matrix operations.



$$X = \begin{bmatrix} x_{11} & x_{12} & \dots & x_{1d} \\ \vdots & \vdots & & \vdots \\ x_{n1} & x_{n2} & \dots & x_{nd} \end{bmatrix} n \times d$$

$n$ : # of data samples

$d$ : # of features.  $\leftarrow d$

$$W = \begin{bmatrix} w_{11} & w_{12} & \dots & w_{1d} \\ \vdots & \vdots & & \vdots \\ w_{nH1} & w_{nH2} & \dots & w_{nHd} \end{bmatrix} n_H \times d$$

$n_H$ : # of hidden nodes.

$$b^h = \begin{bmatrix} b_1^h \\ b_2^h \\ \vdots \\ b_{nH}^h \end{bmatrix}$$

$$b^o = \begin{bmatrix} b_1^o \\ b_2^o \\ \vdots \\ b_c^o \end{bmatrix}$$

Output of the hidden layer:  $h = g_h(\text{Net}_h)$   $\text{Net}_h = X \cdot W^T + (b^h)^T$

Output of the last layer:  $O = g_o(\text{Net}_o)$   $\text{Net}_o = h \cdot u^T + (b^o)^T$



### 3. Two settings

1) regression problem.

$$L = \frac{1}{2} \sum_{k=1}^C (y_k - o_k)^2$$

universal.  
↓

$$\begin{aligned} \nabla_{u_{kj}} L &= \delta_k \cdot h_j \\ \nabla_{u_{ji}} L &= \delta_j \cdot x_i \end{aligned}$$

↙ determined by the loss function

$$\begin{aligned} \delta_k &= (o_k - y_k) \cdot g'_o \\ \delta_j &= \sum_{k=1}^C \delta_k \cdot u_{kj} \cdot g'_h \end{aligned}$$

2) classification problem

$$L = - \sum_{k=1}^C y_k \cdot \log o_k \quad (\text{cross-entropy})$$

$g_o$ : softmax

$$\delta_k = (o_k - y_k)$$

$$\delta_j = \left( \sum_{k=1}^C \delta_k \cdot u_{kj} \right) \cdot g'_h$$