

3 layer Network

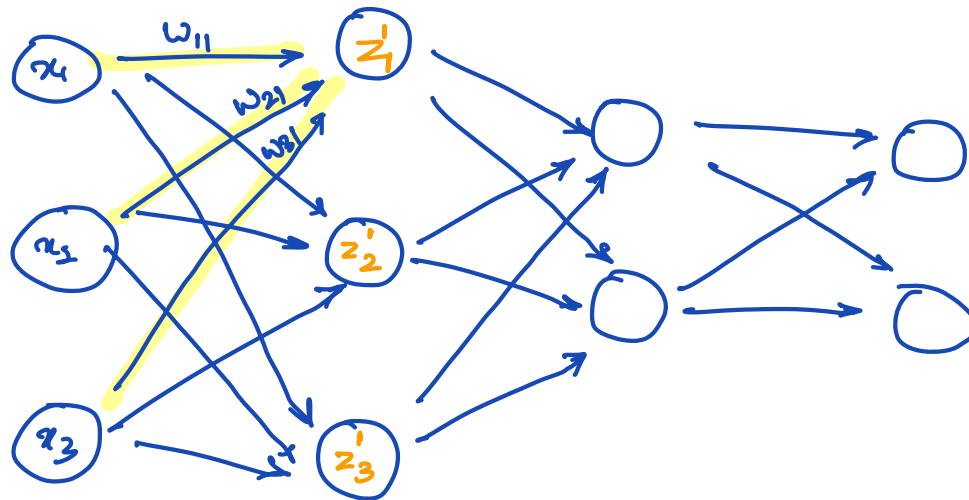
3 features
2 classes

layer 0
Input layer

layer 1
HL1

layer 2
HL2

layer 3
Output layer



$n = \text{features}$

$$Z_{j'}^L = \sum_{i=1}^n w_{i,j'} x_i + b \Rightarrow Z_{j'}^1 = \sum_{i=1}^n w_{i,j'} x_i + b$$

$= w_{11}x_1 + w_{21}x_2 + w_{31}x_3$

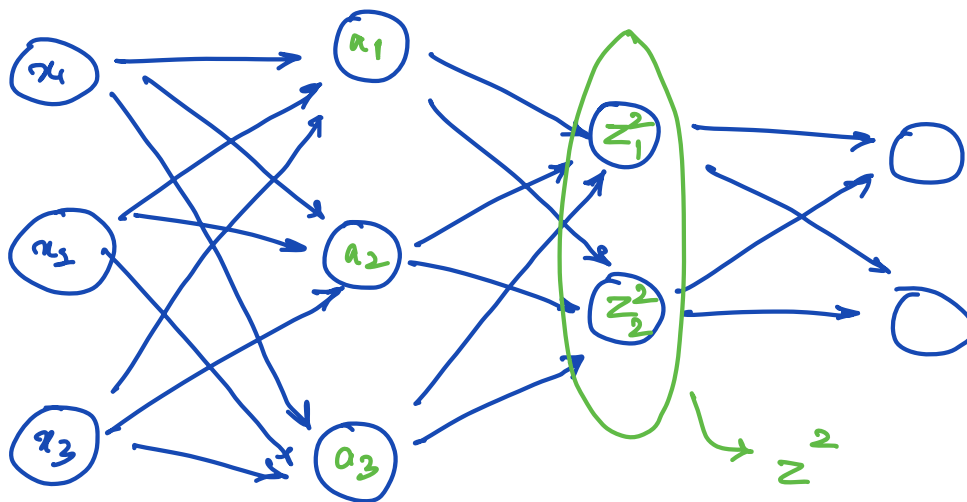
$j = \text{neuron no.}$

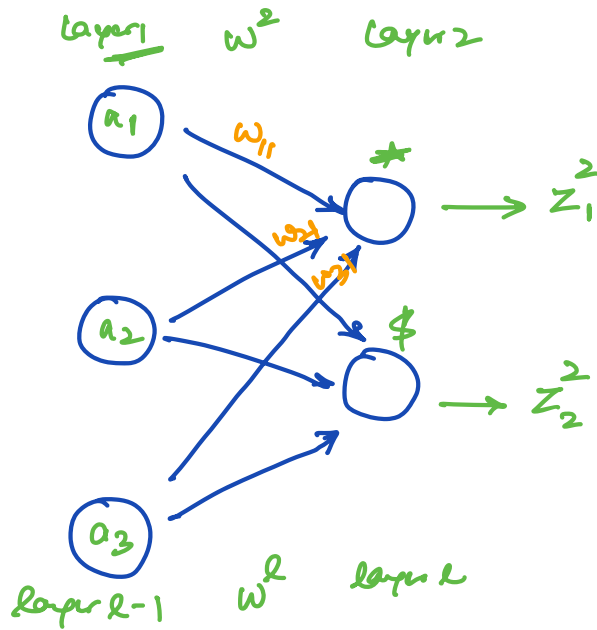
layer 1
HL1

layer 2
HL2

layer 3

w^1 w^2 w^3





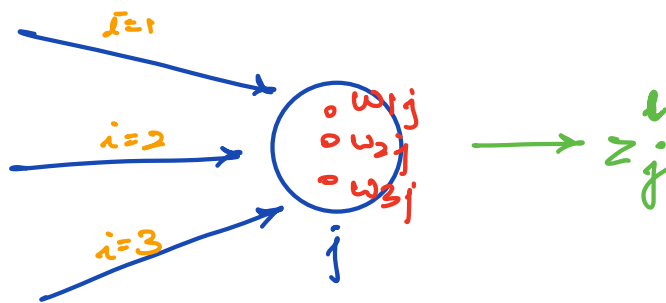
$$w^2 = \begin{bmatrix} \begin{matrix} w_{11} \\ w_{21} \\ w_{31} \end{matrix} & \begin{matrix} w_{12} \\ w_{22} \\ w_{32} \end{matrix} \end{bmatrix}$$

3x2

$$z_j^l = \sum_i w_{ij}^l a_i^{l-1} + b$$

no. of neurons in $l-1$ layer

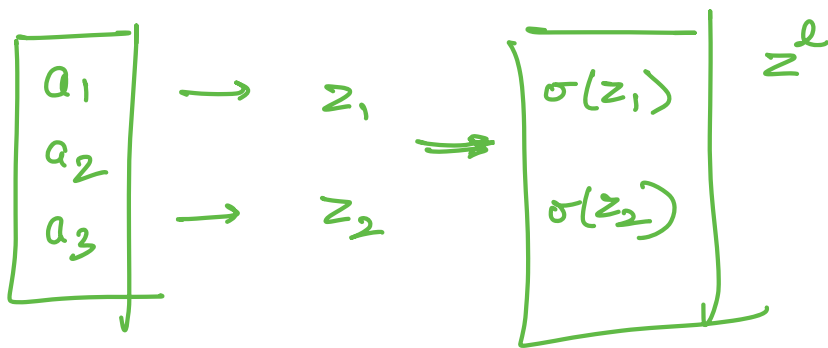
$j = \text{neurons}$
 $j = 1, 2$
 2 neurons in layer 2



w_{ij}^l is weight connecting the neuron in $l-1$ layer & j th neuron of l layer

$$w^l = \begin{bmatrix} \begin{matrix} w_{11}^l \\ w_{21}^l \\ w_{31}^l \end{matrix} & \begin{matrix} w_{12}^l \\ w_{22}^l \\ w_{32}^l \end{matrix} & \begin{matrix} w_{13}^l \\ w_{23}^l \\ w_{33}^l \end{matrix} \end{bmatrix}$$

weights for neuron 1 in l th layer



Vector Notation:

$$z_j^l = \sum_i w_{ij}^l a_i^{l-1} + b^l$$

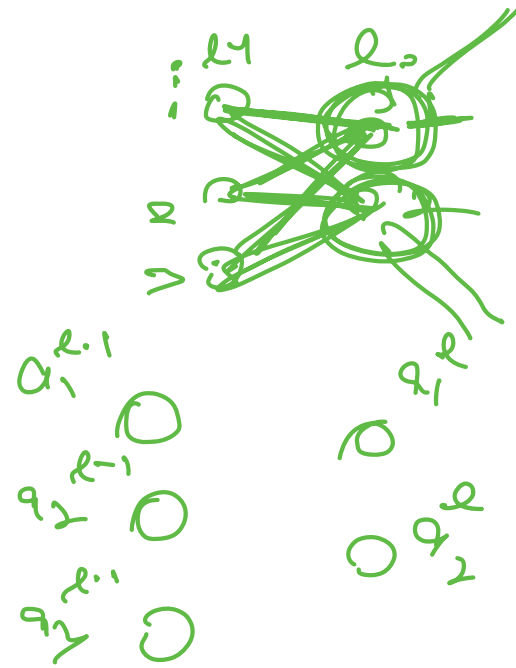
Diagram illustrating the vector notation for the equation above. Arrows point from the terms to their dimensions:

- z_j^l (every neuron)
- w_{ij}^l (3x2)
- a_i^{l-1} (3x1)

$$z^l = [w^l]^T a^{l-1} + b^l$$

$$w^l = \begin{bmatrix} w_{11}^l & w_{12}^l \\ w_{21}^l & w_{22}^l \\ w_{31}^l & w_{32}^l \end{bmatrix}$$

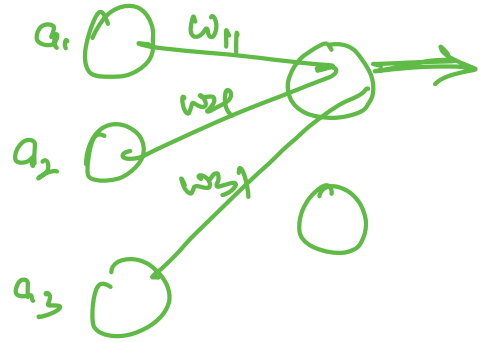
3x2



$$(w^l)^T = \begin{bmatrix} w_{11}^l & w_{21}^l & w_{31}^l \\ w_{12}^l & w_{22}^l & w_{32}^l \end{bmatrix} \begin{bmatrix} a_1^{l-1} \\ a_2^{l-1} \\ a_3^{l-1} \end{bmatrix}$$

2x3 3x1

$$w_{11}^l a_1^{l-1} + w_{21}^l a_2^{l-1} + w_{31}^l a_3^{l-1} = \textcircled{z_1^l}$$



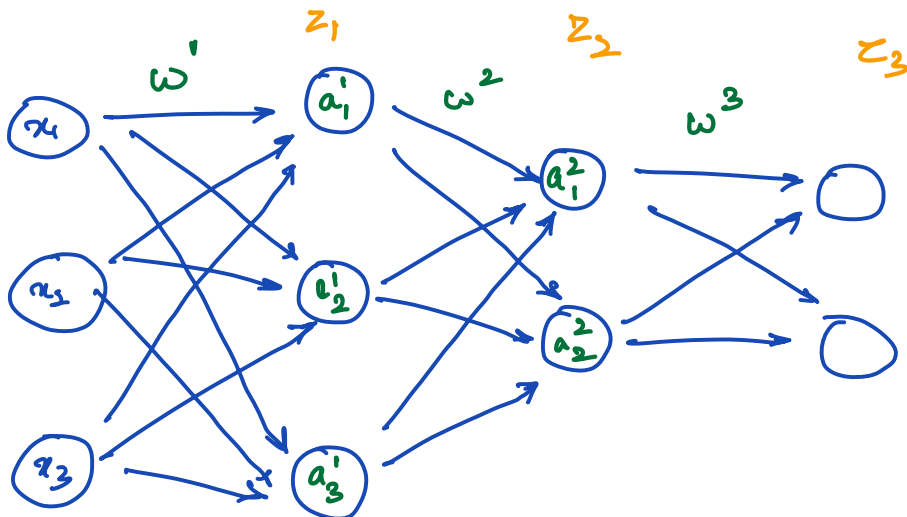
$$z^l = \begin{bmatrix} z_1^l \\ z_2^l \end{bmatrix}$$

$$\sigma(z^l) = \begin{bmatrix} \sigma(z_1^l) \\ \sigma(z_2^l) \end{bmatrix}$$

$$z^1 = (w^1)^T a + b^1$$

$$a^1 = \sigma(z^1) \rightarrow$$

$$z^2 = (w^2)^T a + b^2$$



$$z^3 = (w^3)^T a^2 + b^3$$

$$\hat{y} = \text{softmax}(z^3)$$

Exact Softmax:

$[1, 2, 3]$

$$\left[\frac{e^1}{e^1 + e^2 + e^3}, \frac{e^2}{e^1 + e^2 + e^3}, \frac{e^3}{e^1 + e^2 + e^3} \right]$$

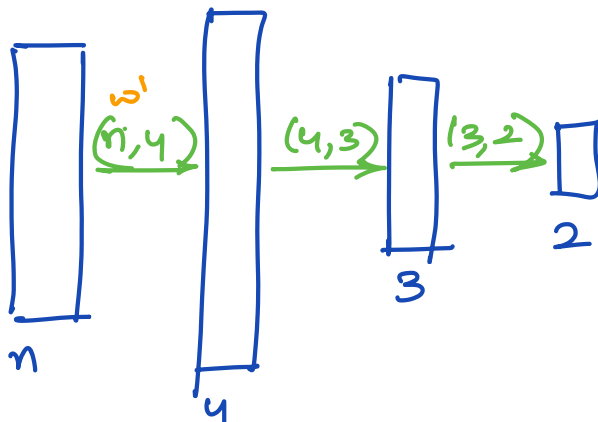
$$\frac{1}{1+2+3}, \frac{2}{1+2+3}, \frac{3}{1+2+3} \quad \times$$

1 example $\rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

m examples

$$X = \begin{bmatrix} x'_1 & x'_2 & x'_3 & \dots & x'_n \\ \hline & x^{(i)} & & & \end{bmatrix}_{m \times n}$$

layer 0 layer 1 layer 2 layer 3

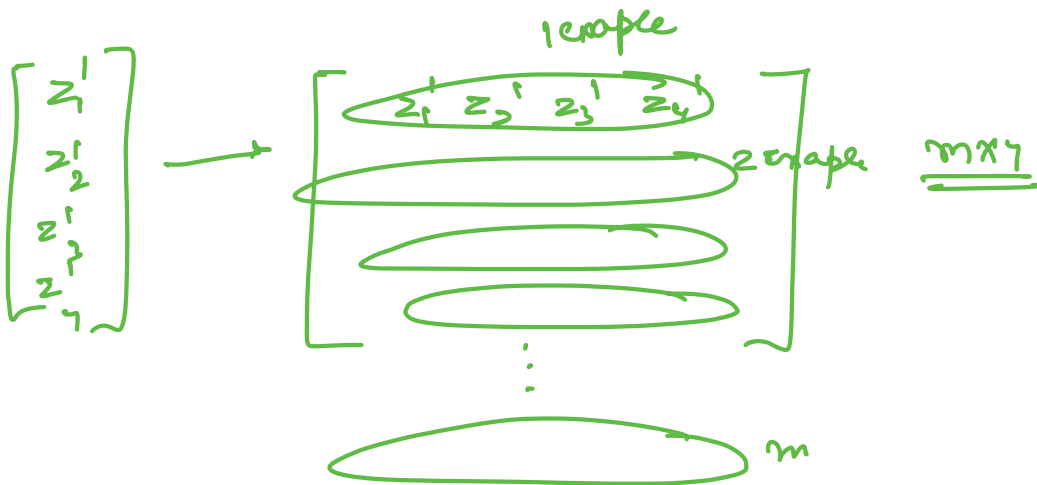


$$z' = \underbrace{(w')^T}_{4 \times 4} \cdot \underbrace{x}_{m \times 4} + b'$$

$$w' = \underline{m \times 4} \quad x = \underline{m \times n}$$

$$z' = \underbrace{x \cdot w'}_{m \times n \cdot n \times 4} + b'$$

$$\underline{m \times 4}$$



$$Z' = \begin{bmatrix} z_1^{11} & z_2^{11} & z_3^{11} & z_4^{11} \\ z_1^{12} & z_2^{12} & z_3^{12} & z_4^{12} \\ \vdots & \vdots & \vdots & \vdots \\ z_1^{1m} & z_2^{1m} & z_3^{1m} & z_4^{1m} \end{bmatrix}_{m \times 4}$$

z_j^{li} → i th example
 l layer j th neuron output

$$a' = \sigma \left[\frac{z'}{1} \right]$$

$$Z' = \begin{bmatrix} z_1^{1'} & z_2^{1'} & z_3^{1'} & z_4^{1'} \\ z_1^{2'} & z_2^{2'} & z_3^{2'} & z_4^{2'} \\ \vdots & \vdots & \vdots & \vdots \\ z_1^{m'} & z_2^{m'} & z_3^{m'} & z_4^{m'} \end{bmatrix}_{m \times 4} + \begin{bmatrix} b_1' & b_2' & b_3' & b_4' \end{bmatrix}_{1 \times 4}$$

$$Z' = \begin{bmatrix} z_1^{1'} + b_1' & z_2^{1'} + b_2' & z_3^{1'} + b_3' & z_4^{1'} + b_4' \\ z_1^{2'} + b_1' & z_2^{2'} + b_2' & z_3^{2'} + b_3' & z_4^{2'} + b_4' \\ \vdots & \vdots & \vdots & \vdots \\ z_1^{m'} + b_1' & z_2^{m'} + b_2' & z_3^{m'} + b_3' & z_4^{m'} + b_4' \end{bmatrix}_{m \times 4}$$

Backpropagation