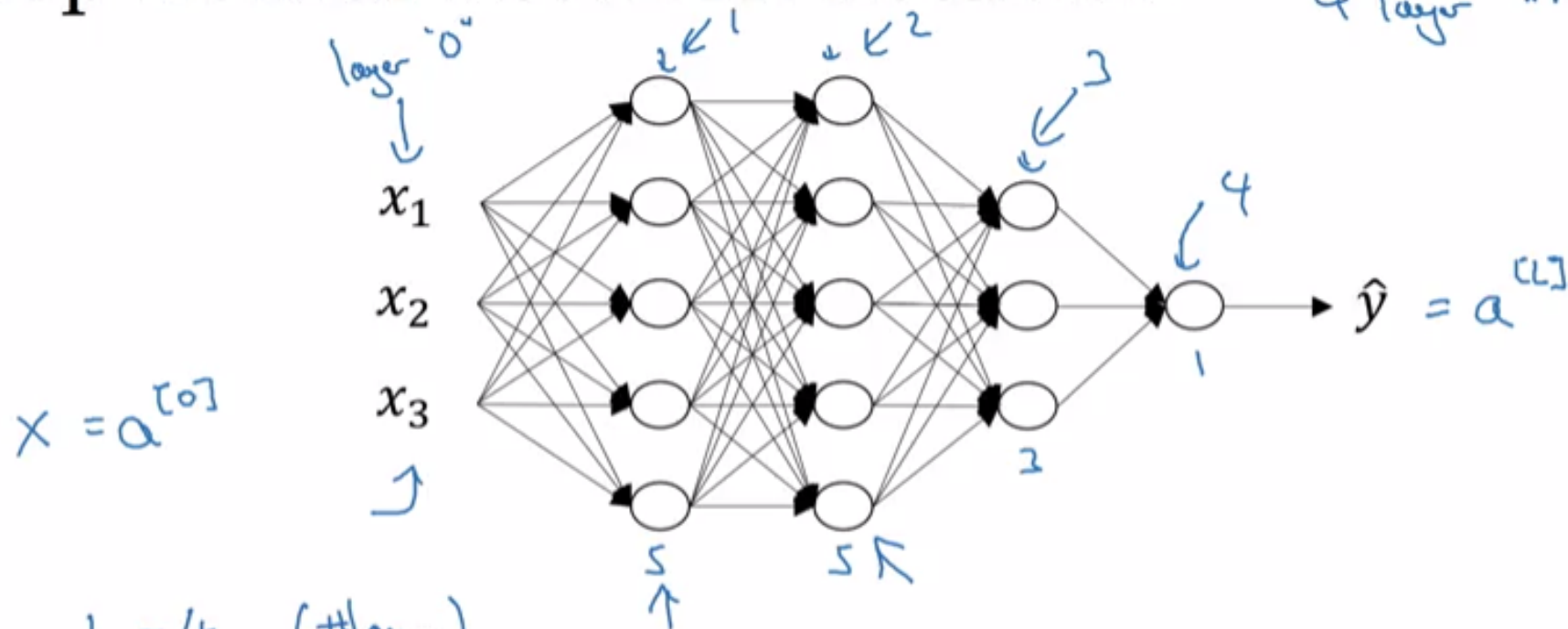


Deep neural network notation

4 layer NN



$L = 4$ (#layers)

$n^{[l]} = \# \text{units in layer } l$

$a^{[l]} = \text{activations in layer } l$

$a^{[l]} = g(z^{[l]})$, $w_{j,i}^{[l]} = \text{weights for } \underline{z^{[l]}}$

$n^{[1]} = 5, n^{[2]} = 5, n^{[3]} = 3, n^{[4]} = n^{[L]} = 1$

$n^{[0]} = n_x = 3$

Forward propagation for layer l

→ Input $a^{[l-1]}$ ←

→ Output $a^{[l]}$, cache $(z^{[l]})$ ← $w^{[l]}, b^{[l]}$

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

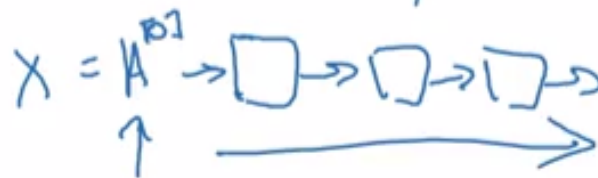
$$a^{[l]} = g^{[l]}(z^{[l]})$$

Verknüpfung:

$$z^{[l]} = w^{[l]} \cdot A^{[l-1]} + b^{[l]}$$

$$A^{[l]} = g^{[l]}(z^{[l]})$$

$a^{[0]}$
 $A^{[0]}$



Backward propagation for layer l

→ Input $da^{[l]}$

→ Output $da^{[l-1]}$, $dW^{[l]}$, $db^{[l]}$

$$dz^{[l]} = da^{[l]} * g^{[l]'}(z^{[l]})$$

$$dW^{[l]} = dz^{[l]} \cdot a^{[l-1]}$$

$$db^{[l]} = dz^{[l]}$$

$$da^{[l-1]} = W^{[l]T} \cdot dz^{[l]}$$

$$dz^{[l+1]} = W^{[l+1]T} dz^{[l]} * g^{[l+1]'}(z^{[l+1]})$$

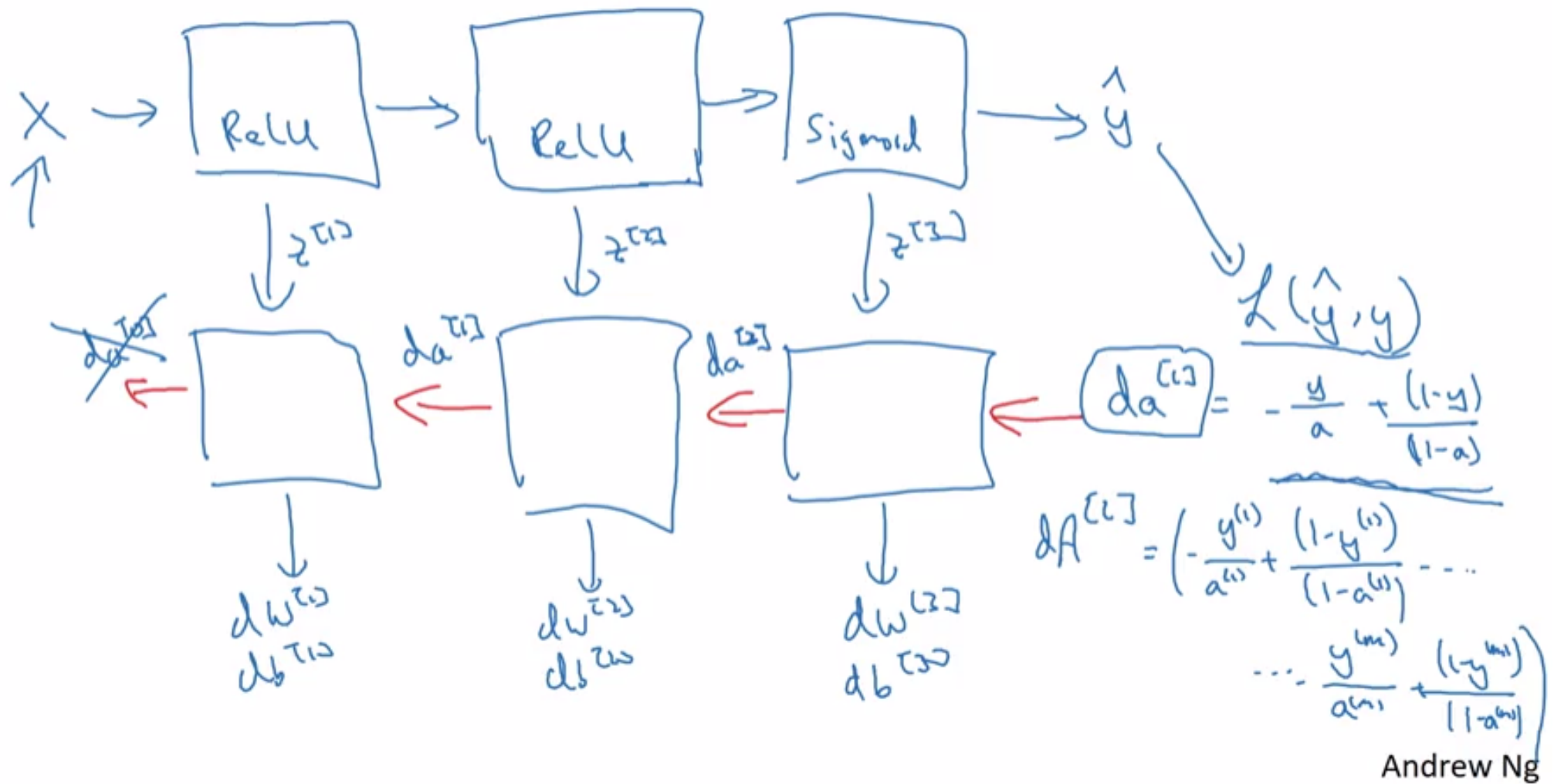
$$dz^{[l]} = dA^{[l]} * g^{[l]'}(z^{[l]})$$

$$dW^{[l]} = \frac{1}{n} dz^{[l]} \cdot A^{[l-1]T}$$

$$db^{[l]} = \frac{1}{n} \text{np.sum}(dz^{[l]}, \text{axis}=1, \text{keepdims}=True)$$

$$dA^{[l-1]} = W^{[l]T} \cdot dz^{[l]}$$

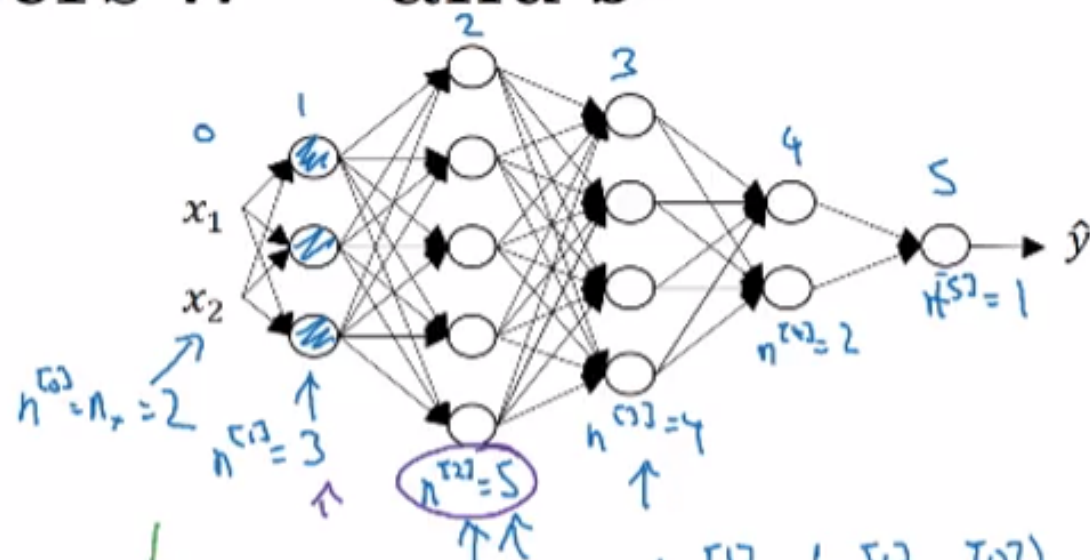
Summary



Andrew Ng

Parameters $W^{[l]}$ and $b^{[l]}$

$$z^{[L]} = g^{[L]}(a^{[L]})$$



$L=5$

$$\begin{cases} \rightarrow W^{[L]}: (n^{[L]}, n^{[L-1]}) \\ \rightarrow b^{[L]}: (n^{[L]}, 1) \\ \rightarrow \Delta W^{[L]}: (n^{[L]}, n^{[L-1]}) \\ \rightarrow \Delta b^{[L]}: (n^{[L]}, 1) \end{cases}$$

$$z^{[1]} = \underbrace{W^{[1]} \cdot x}_{(3,1) \leftarrow (3,2) \quad (2,1)} + \underbrace{b^{[1]}}_{(3,1)}$$

$(n^{[1]}, 1) \quad (n^{[1]}, n^{[0]}) \quad (n^{[0]}, 1)$

$$\begin{bmatrix} \vdots \\ \vdots \end{bmatrix} = \begin{bmatrix} \vdots \\ \vdots \end{bmatrix} \begin{bmatrix} \vdots \end{bmatrix}$$

$$W^{[1]}: (n^{[1]}, n^{[0]})$$

$$W^{[2]}: (5, 3) \quad (n^{[2]}, n^{[1]})$$

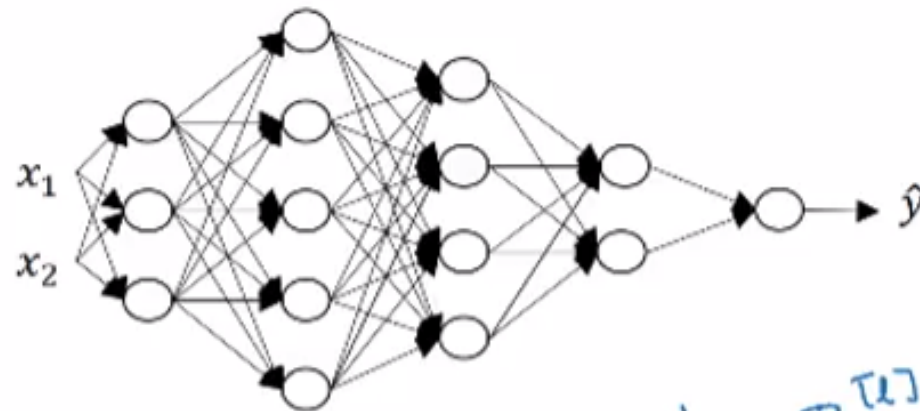
$$z^{[2]} = \underbrace{W^{[2]} \cdot a^{[1]}}_{(5,1) \leftarrow (5,3) \quad (2,1)} + \underbrace{b^{[2]}}_{(5,1)}$$

$\rightarrow (5,1) \quad (5,3) \quad (2,1) \quad (n^{[2]}, 1)$

$$W^{[3]}: (4, 5)$$

$$W^{[4]}: (2, 4) \quad , \quad W^{[5]}: (1, 2)$$

Vectorized implementation



$$z^{[1]} = W^{[0]} \cdot X + b^{[1]}$$

$(n^{[0]}, 1)$ $(n^{[0]}, n^{[0]})$ $(n^{[0]}, 1)$ $(n^{[1]}, 1)$

$[z^{[0]}, z^{[1]}, \dots, z^{[L-1]}]$

$$Z^{[1]} = W^{[0]} \cdot X + b^{[1]}$$

$(n^{[1]}, m)$ $(n^{[0]}, n^{[0]})$ $(n^{[0]}, m)$ $(n^{[1]}, 1)$

$(n^{[0]}, m)$

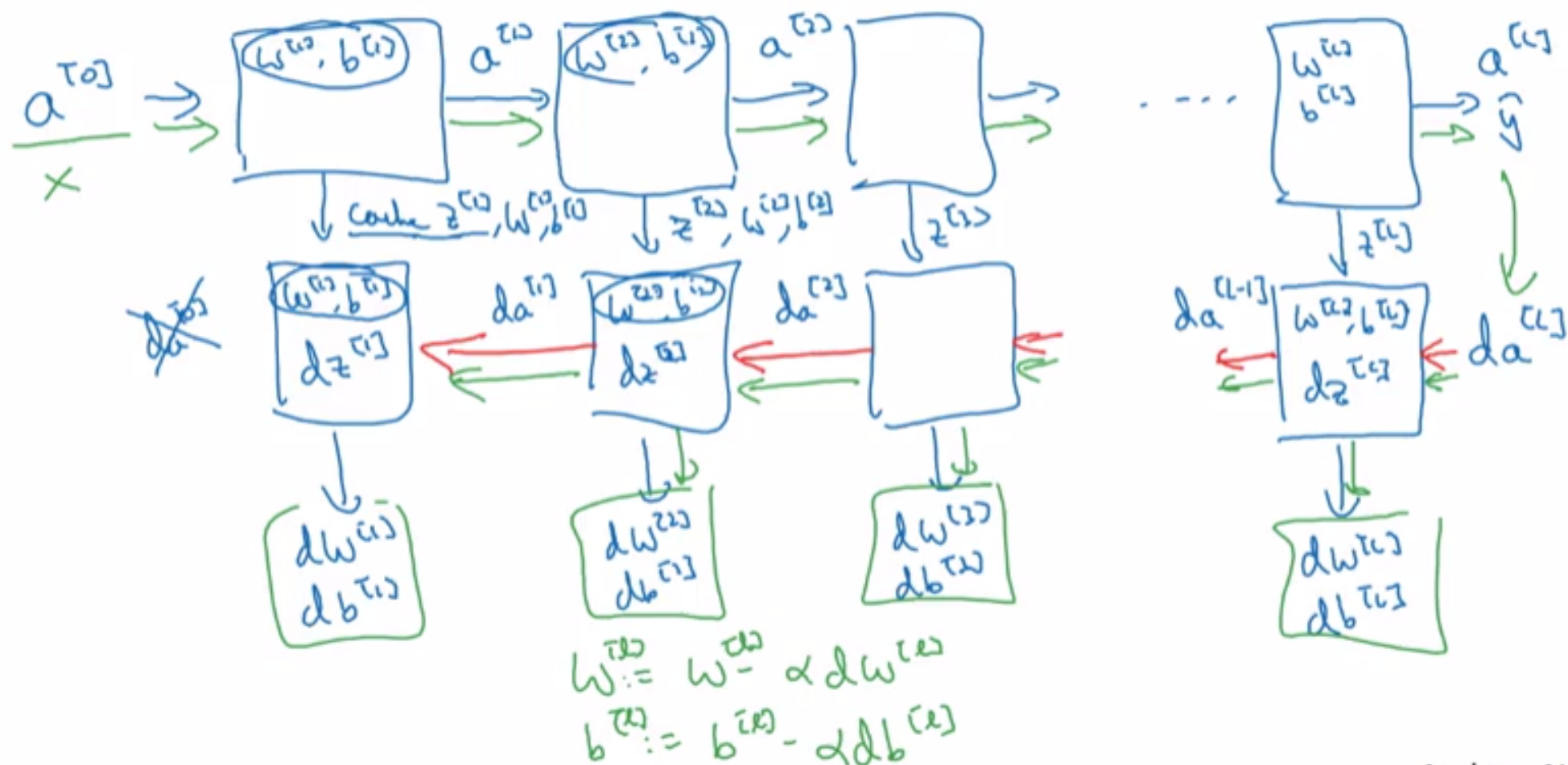
$$z^{[1]}, a^{[1]} : (n^{[1]}, 1)$$

$$z^{[2]}, A^{[2]} : (n^{[2]}, m)$$

$l=0 \quad A^{[0]} = X = (n^{[0]}, m)$

$$dz^{[1]}, dA^{[1]} : (n^{[1]}, m)$$

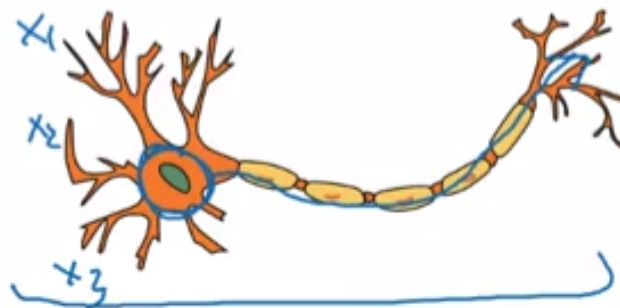
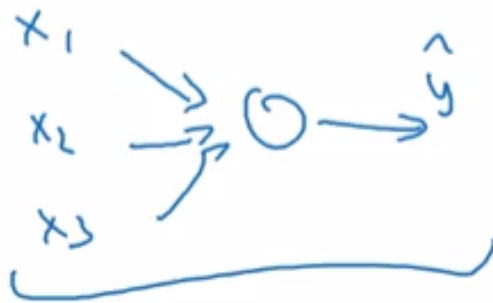
Forward and backward functions



Forward and backward propagation

$$\begin{aligned} Z^{[1]} &= W^{[1]}X + b^{[1]} \\ A^{[1]} &= g^{[1]}(Z^{[1]}) \\ Z^{[2]} &= W^{[2]}A^{[1]} + b^{[2]} \\ A^{[2]} &= g^{[2]}(Z^{[2]}) \\ &\vdots \\ A^{[L]} &= g^{[L]}(Z^{[L]}) = \hat{Y} \end{aligned}$$

"It's like the brain."



$$\begin{aligned} dZ^{[L]} &= A^{[L]} - Y \\ dW^{[L]} &= \frac{1}{m} dZ^{[L]} A^{[L]T} \\ db^{[L]} &= \frac{1}{m} \text{np.sum}(dZ^{[L]}, \text{axis} = 1, \text{keepdims} = \text{True}) \\ dZ^{[L-1]} &= dW^{[L]T} dZ^{[L]} g'^{[L]}(Z^{[L-1]}) \\ &\vdots \\ dZ^{[1]} &= dW^{[L]T} dZ^{[2]} g'^{[1]}(Z^{[1]}) \\ dW^{[1]} &= \frac{1}{m} dZ^{[1]} A^{[1]T} \\ db^{[1]} &= \frac{1}{m} \text{np.sum}(dZ^{[1]}, \text{axis} = 1, \text{keepdims} = \text{True}) \end{aligned}$$