

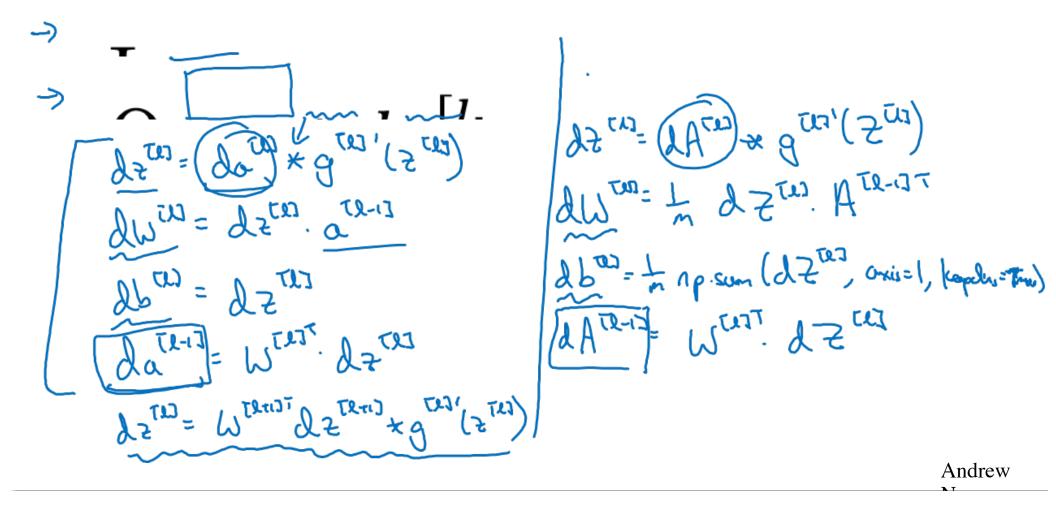
## Deep Neural Networks

Forward and backward propagation

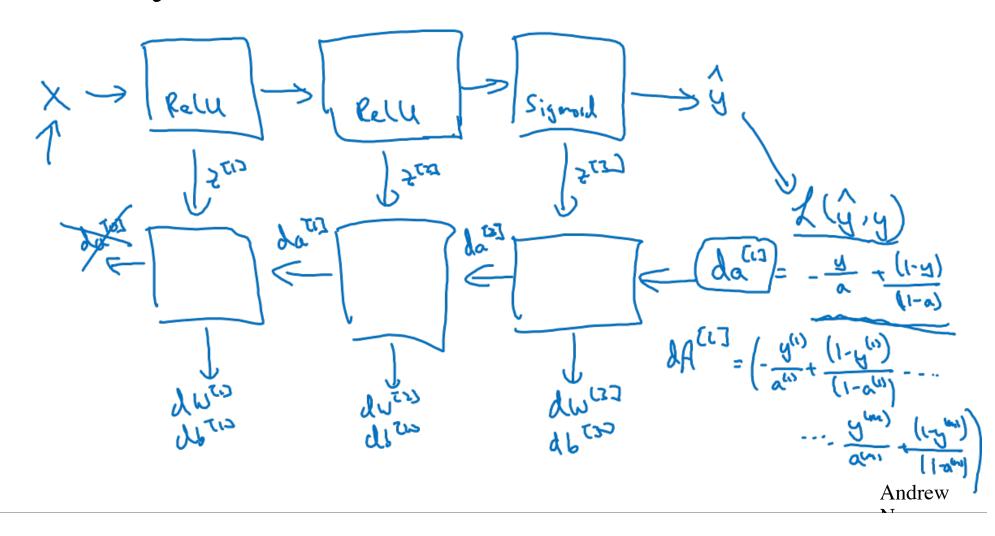
#### Forward propagation for layer *l*

Andrew

#### Backward propagation for layer l



#### Summary

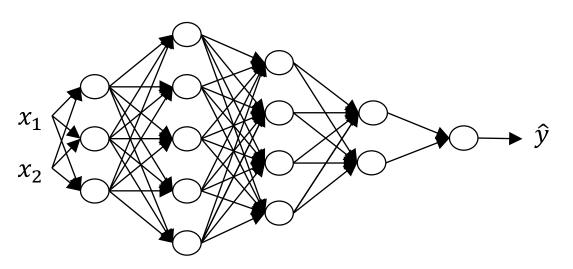




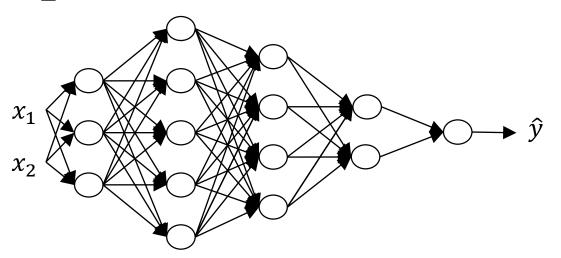
# Deep Neural Networks

Getting your matrix dimensions right

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



### Vectorized implementation





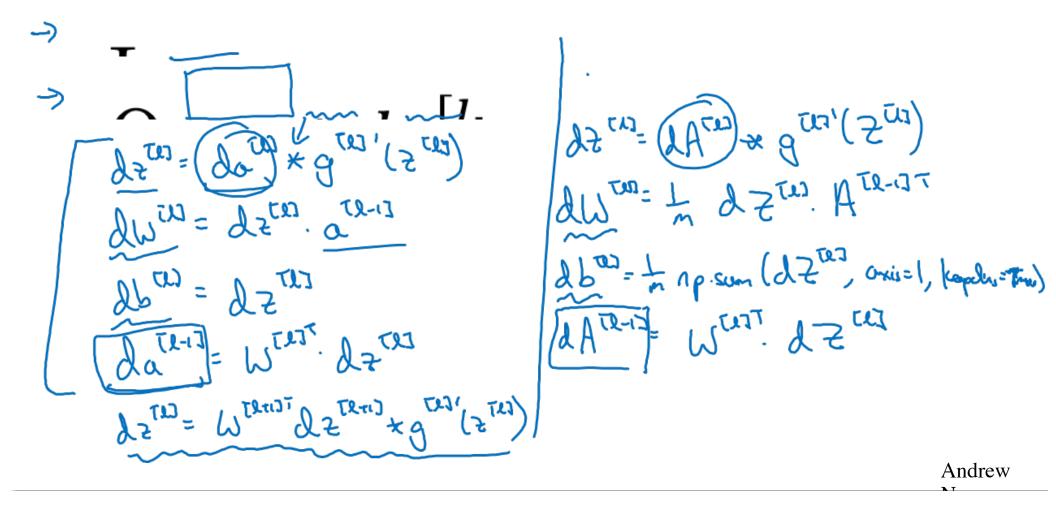
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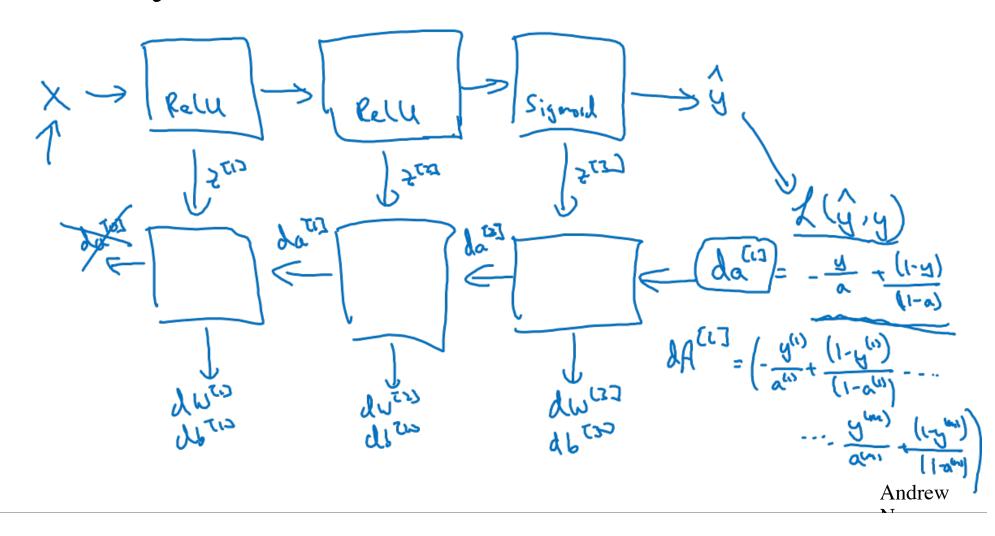
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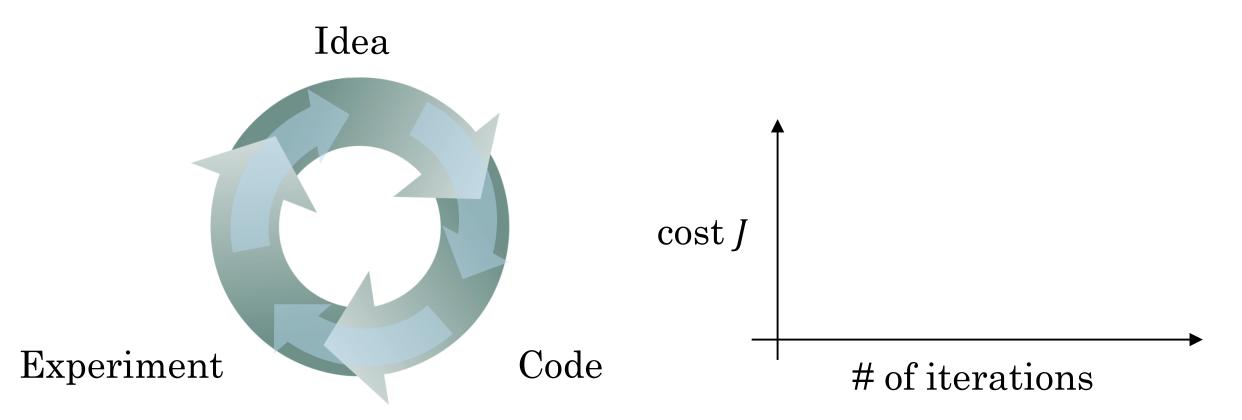
# Deep Neural Networks

Parameters vs Hyperparameters

### What are hyperparameters?

Parameters:  $W^{[1]}$  ,  $b^{[1]}$  ,  $W^{[2]}$  ,  $b^{[2]}$  ,  $W^{[3]}$  ,  $b^{[3]}$  ...

# Applied deep learning is a very empirical process





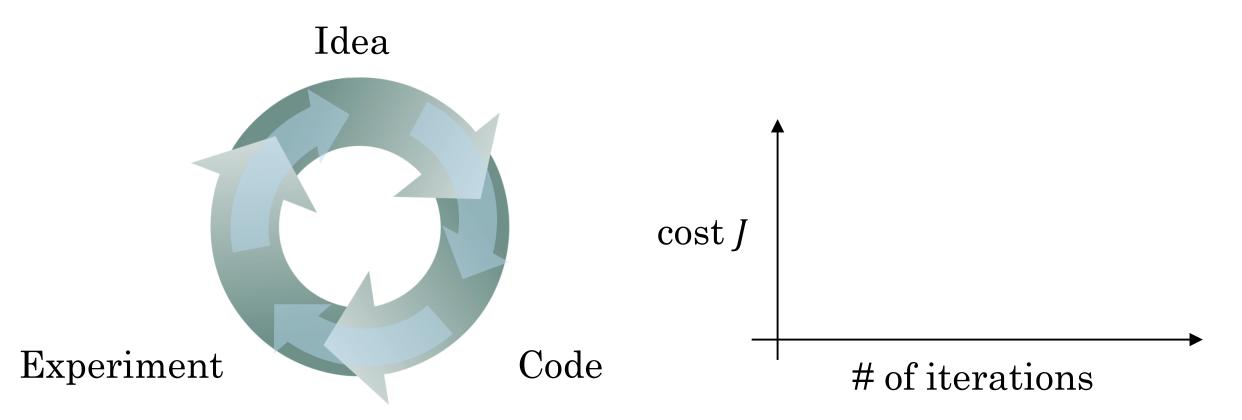
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# Deep Neural Networks

What does this have to do with the brain?

### Forward and backward propagation

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

$$A^{[1]} = g^{[1]}(Z^{[1]})$$

$$Z^{[2]} = W^{[2]}A^{[1]} + b^{[2]}$$

$$A^{[2]} = g^{[2]}(Z^{[2]})$$

$$\vdots$$

$$A^{[L]} = g^{[L]}(Z^{[L]}) = \hat{Y}$$

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

