

## 1) Notations:

$X$  - Input

$y$  - True Output

$W[i], B[i]$  - Weights & bias for layer  $i$

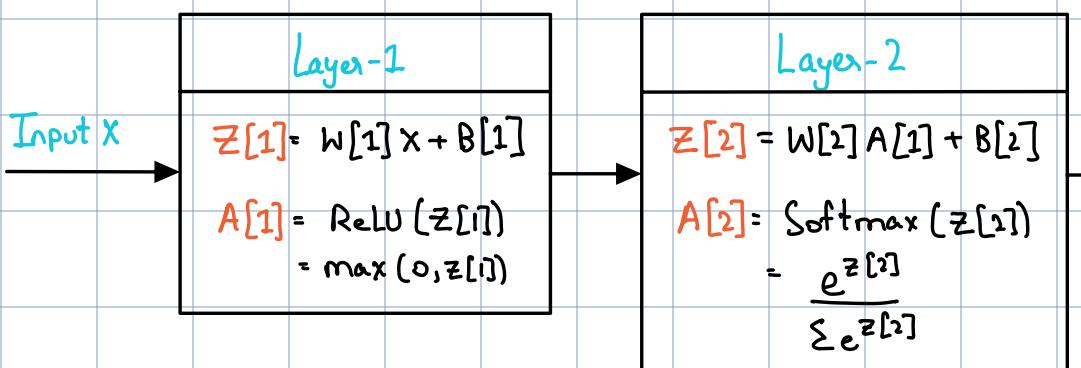
$z[i], A[i]$  - Linear & Activation for layer  $i$

$\alpha$  - Learning rate

$m$  - Number of Examples

## ANN from Scratch

## 2) Forward Propagation



## 4) Gradient Descent

$$W[i] = W[i] - \alpha \cdot \partial W[i]$$

$$B[i] = B[i] - \alpha \cdot \partial B[i]$$

## 3) Backward Propagation

$$\text{Cost function } J = \frac{1}{m} \sum_{i=1}^m (\hat{y} - y)$$

$$\text{Chain Rule } \frac{\partial J}{\partial w} = \frac{\partial J}{\partial z} \cdot \frac{\partial z}{\partial w}$$

$$\text{Starting point } \partial z[2] = A[2] - y$$

### ① Layer-2:

Weight  $\partial W[2]$ :

$$z[2] = W[2]A[1] + B[2] \quad [\text{from Chain rule}]$$

Chain Rule:

$$\partial W[2] = \frac{\partial J}{\partial z[2]} \cdot \frac{\partial z[2]}{\partial W[2]} \rightarrow A[1]$$

Bias  $\partial B[2]$ :

$$z[2] = W[2]A[1] + B[2]$$

Chain Rule:

$$\partial B[2] = \frac{\partial J}{\partial z[2]} \cdot \frac{\partial z[2]}{\partial B[2]} \rightarrow 1$$

$$\partial W[2] = \frac{1}{m} \partial z[2] A[1]^T$$

$$\partial B[2] = \frac{1}{m} \sum_{i=1}^m \partial z[2]$$

### ② Layer-1:

$$\partial z[1] = \frac{\partial J}{\partial z[2]} \cdot \frac{\partial z[2]}{\partial A[1]} \cdot \frac{\partial A[1]}{\partial z[1]} \rightarrow g'(z[1])$$

$$\partial z[1] = W[2]^T \partial z[2] \circ g'(z[1])$$

Weight  $\partial W[1]$ :

$$z[1] = W[1]A[0] + B[1]$$

Chain Rule:

$$\partial W[1] = \frac{\partial J}{\partial z[2]} \cdot \frac{\partial z[2]}{\partial W[2]} \rightarrow A[0]$$

Bias  $\partial B[1]$ :

$$z[1] = W[1]A[0] + B[1]$$

Chain Rule:

$$\partial B[1] = \frac{\partial J}{\partial z[2]} \cdot \frac{\partial z[2]}{\partial B[2]} \rightarrow 1$$

by

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$$\partial W[1] = \frac{1}{m} \partial z[1] A[0]^T$$

$$\partial B[1] = \frac{1}{m} \sum_{i=1}^m \partial z[1]$$