

MNIST From Scratch using MNIST

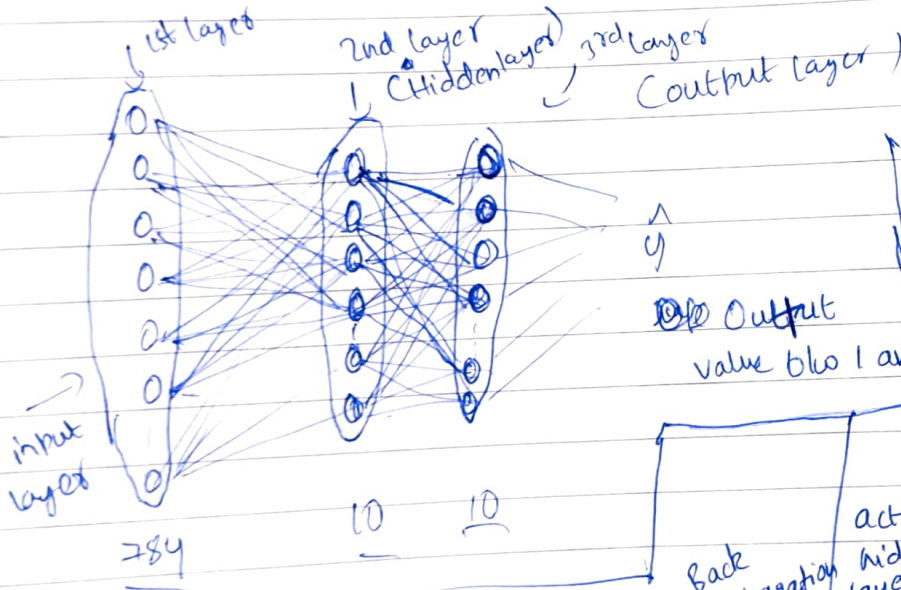
784
28x28

say we have m training images

each pixel is a value in $[0, 1]$

0, 1, ..., 255
completely Black completely white

$$X = \begin{bmatrix} x^{(1)} \\ x^{(2)} \\ \vdots \\ x^{(m)} \end{bmatrix}^T = \begin{bmatrix} | & | & | & | & | & | \\ x^{(1)} & x^{(2)} & x^{(3)} & x^{(4)} & \dots & x^{(m)} \\ | & | & | & | & | & | \end{bmatrix} \begin{matrix} 784 \\ \text{long} \end{matrix}$$



input layer value

$$A^{[0]} = X$$

$$Z^{[1]} = W^{[1]} A^{[0]} + b^{[1]}$$

$\uparrow 10 \times m$ unactivated 10×784 $784 \times m$ $10 \times m$

hidden layer values

Forward propagation

$$A^{[1]} = \text{ReLU}(Z^{[1]})$$

activated hidden layer

unactivated \uparrow output layer

$$\text{ReLU}(x) = \begin{cases} x & \text{if } x > 0 \\ 0 & \text{if } x \leq 0 \end{cases}$$

Back propagation

$$dZ^{[2]} = A^{[2]} - y_{\text{onehot encoded}}$$

$$dW^{[2]} = \frac{1}{m} dZ^{[2]} A^{[1]T}$$

$$db^{[2]} = \frac{1}{m} \sum dZ^{[2]}$$

$$dZ^{[1]} = W^{[2]T} dZ^{[2]} * g'(Z^{[1]})$$

$$dW^{[1]} = \frac{1}{m} dZ^{[1]} A^{[0]T}$$

$$db^{[1]} = \frac{1}{m} \sum dZ^{[1]}$$

activated output layer

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

$$A^{[2]} = \text{softmax}(Z^{[2]})$$

Output layer

Softmax activation function

$$\frac{e^{z_i}}{\sum_{j=1}^K e^{z_j}}$$

$w^{(1)} := w^{(1)} - \alpha dw^{(1)}$
 $b^{(1)} := b^{(1)} - \alpha db^{(1)}$
 $w^{(2)} := w^{(2)} - \alpha dw^{(2)}$
 $b^{(2)} := b^{(2)} - \alpha db^{(2)}$

learning rate

update params

