

Neural network

what are
the neurons?

how are they connected?



Thing that holds a number
Generally betⁿ 0 & 1

number inside
neuron is

black
pixel

white pixel

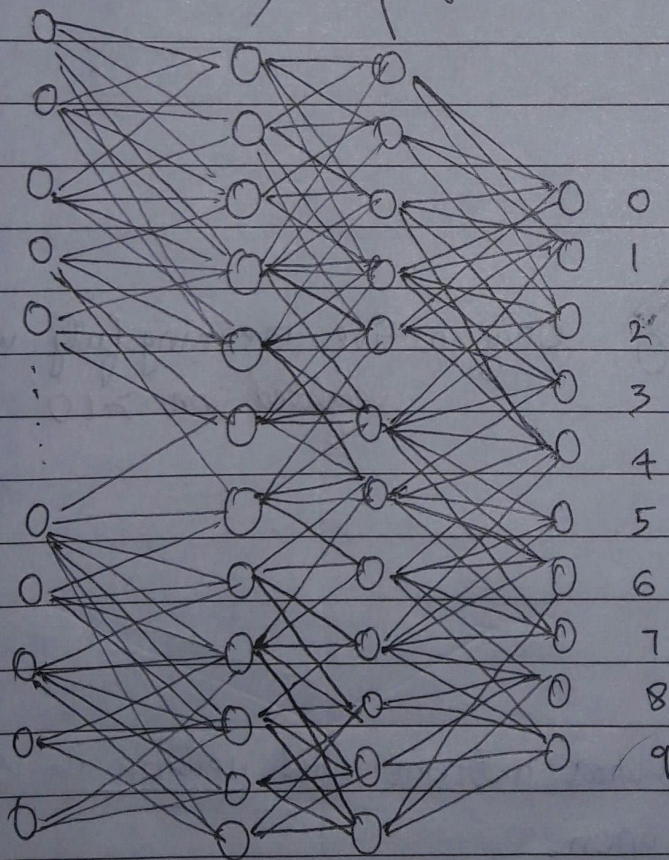
called activation

represent grey
scale value of pixel.

hidden layers

*

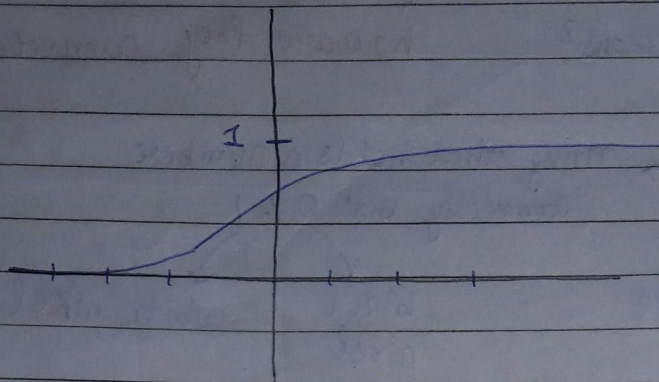
let's say
784
neurons
in
1st layer



* Sigmoid

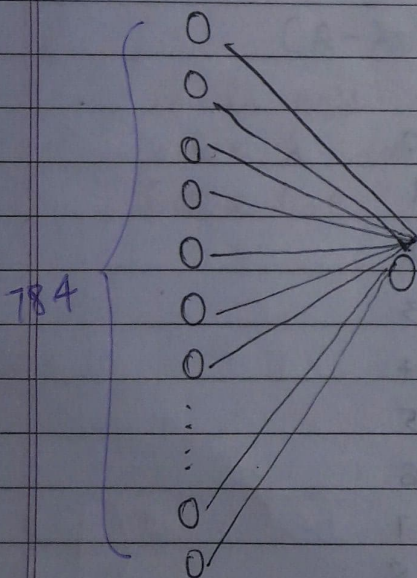
$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

[We use this function as we need range of 0 to 1 for activation values]



$$\sigma(w_1 a_1 + w_2 a_2 + w_3 a_3 + \dots + w_n a_n + \text{bias})$$

here w_1, w_2, \dots are weights of neurons, a is activation



Only active meaningfully when weighted sum > 10

weights tells what pattern this neuron in second layer is picking upon.

Bias tells how high the weighted sum needs to be before neuron becomes active.

This is for one neuron.

$\begin{array}{ccc} \text{1st \& 2nd} & \text{2nd \& 3rd} & \text{3rd \& 4th} \\ \text{layers} & & \end{array}$

$$\boxed{\begin{array}{c} 784 \times 16 + 16 \times 16 + 16 \times 10 \\ \text{weights} \\ 16 + 16 + 10 \\ \text{biases} \end{array}}$$

Total 13,002 weights & biases.

$$\alpha \left(\begin{bmatrix} w_{0,0} & w_{0,1} & \dots & w_{0,n} \\ w_{1,0} & w_{1,1} & \dots & w_{1,n} \\ \vdots & \vdots & \ddots & \vdots \\ w_{k,0} & w_{k,1} & \dots & w_{k,n} \end{bmatrix} \begin{bmatrix} a_0^{(0)} \\ a_1^{(0)} \\ \vdots \\ a_n^{(0)} \end{bmatrix} + \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_n \end{bmatrix} \right)$$

Each row represent connection betⁿ one layer & particular neuron in next layer.

$$\overset{\substack{\text{1st} \\ \text{layer}}}{a^{(1)}} = \alpha (W a^{(0)} + b)$$

Initial layer

Cost = $(0.43 - 0)^2 +$
 (Should be as small as possible) $(0.28 - 0)^2 + (0.8 - 1)^2 + \dots$ actual values
 Wrong values that we get

we consider average cost of all data's.

* Neural network function:-

Input:- 784 no's

Parameters:- 13,0002 weights/biases

Output:- 10 no's

Cost function

Input :- 13,002 weights / biases

output :- 1 no

Parameters :- multiple training examples.

"Gradient", the direction of steepest increase $\nabla C(x, y)$

$$\vec{W} = \begin{bmatrix} 2.43 \\ -1.12 \\ 1.47 \\ \vdots \\ \vdots \\ 1.21 \end{bmatrix}$$

13,002
weights & bias

$$-\nabla C(\vec{W}) =$$

(This tells
which changes
to which
weight
matters
most.)

$$\begin{bmatrix} 0.18 \\ 0.45 \\ -0.51 \\ \vdots \\ -0.32 \\ 0.82 \end{bmatrix}$$

How to nudge all
weights & biases

$$a = w_0 a_0 + w_1 a_1 + \dots + w_{n-1} a_{n-1} + b$$

let's say

$a = 0.2$, to increase a ,

i) increase b

ii) Increase w_i in proportion to a_i

[For more active neurons increase in w will have higher effect]

iii) change a_i [you cannot change a_i]
[in proportion to w_i]

Now, when you have desired effects that you want betⁿ last & 2nd last layers, same you can do for other previous layers, this is back propagation.

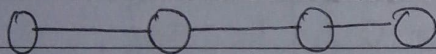
→ You do this for all training data,

	2	5	0	4	...	Average
w_0	-0.08	0.02	-0.02	0.11		-0.08
w_1	-0.11	0.11	0.07	0.05		0.04
w_2	-0.04	0.02	0.05	0.72		0.34
\vdots						
w_n	0.13	0.08	-0.06	0.09		0.76

This collection is

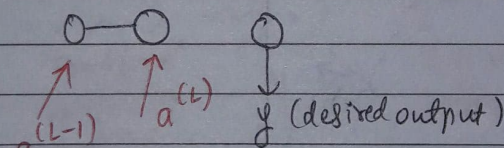
$$-\nabla C(w_1, w_2, \dots) = \begin{bmatrix} -0.08 \\ 0.04 \\ \dots \\ 0.76 \end{bmatrix}$$

* Consider, 4 neuron



For now,

consider last 2,

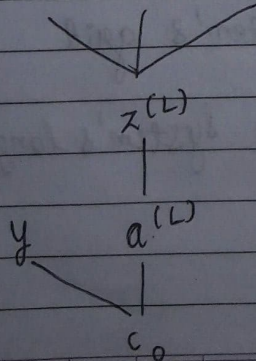


Cost $\rightarrow C_0 = (a^{(L)} - y)^2$

$$a^{(L)} = w^{(L)} \cdot a^{(L-1)} + b^{(L)} = z^{(L)}$$

$$a^{(L)} = \sigma(z^{(L)})$$

$w^{(L)} \quad a^{(L-1)} \quad b^{(L)}$



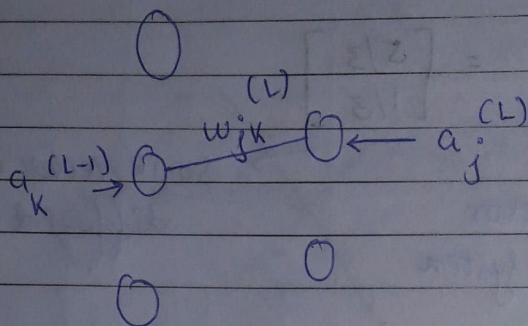
$$\frac{\partial C_0}{\partial w^{(L)}} = \frac{\partial z^{(L)}}{\partial w^{(L)}} \cdot \frac{\partial a^{(L)}}{\partial z^{(L)}} \cdot \frac{\partial C_0}{\partial a^{(L)}} = 2(a^{(L)} - y) \cdot \sigma'(z^{(L)}) \cdot a^{(L-1)}$$

$$\frac{\partial C_0}{\partial a^{(L-1)}} = \frac{\partial z^{(L)}}{\partial a^{(L-1)}} \cdot \frac{\partial a^{(L)}}{\partial z^{(L)}} \cdot \frac{\partial C_0}{\partial a^{(L)}} = w^{(L)} \cdot \sigma'(z^{(L)}) \cdot 2(a^{(L)} - y)$$

$$\frac{\partial C_0}{\partial b^{(L)}} = 1 \cdot \sigma'(z^{(L)}) \cdot 2(a^{(L)} - y)$$

Continued..
neural
networks.

Repeat this backward to see how sensitive cost f^n is to previous weights, biases, activations.



$$C_0 = \sum_{j=0}^{n_L-1} (a_j^{(L)} - y_j)^2$$

Same

$$a_j^{(L)} = \alpha(z_j^{(L)})$$