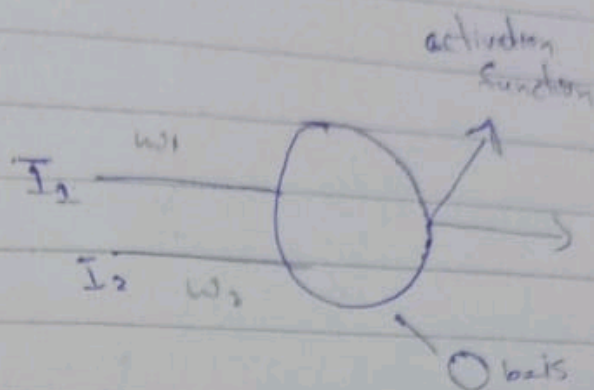


- data should be numeric
- data should be normalized  $(-1, 1)$
- other than these data neural network will not work

## ANN



- give weights to every input.
- multiple value of input and weights and then sum all values
- weight is given by connectors
- Activation Function is ~~the~~ non-linear regression

(Activation function do this)

- data should be normalized between  $(-1 \text{ to } 1)$  to  $(0 \text{ to } 1)$

## Simplest Neuron: / Single Perceptron

- only ~~one~~ one neuron is use
- It has capability to give same output.

- input layer :
- internal layer/hidden layer :
- output layer :

Deep learning :

- no. of hidden layer use -
- increase time complexity
- computing
- no of hidden layer is the dept of deep learning.

Drawbacks :

- interpretation <sup>of working</sup> is ~~hard~~ is not understandable.

Advantages :

- How to ~~data~~ give input to model should be known
- And interpretation of output is needed.



→ These iteration will run until  
the model give error less the  
decide threshold.

threshold:-

Any decide value,  
the error rate which we can bear  
in our model.

→ because nothing is 100%  
efficient.

→ when comparing threshold with  
error (we neglect negative sign) <sup>but we send</sup> actual value of

$$\text{Err}_j = O_j (1 - O_j) (T_j - O_j)$$

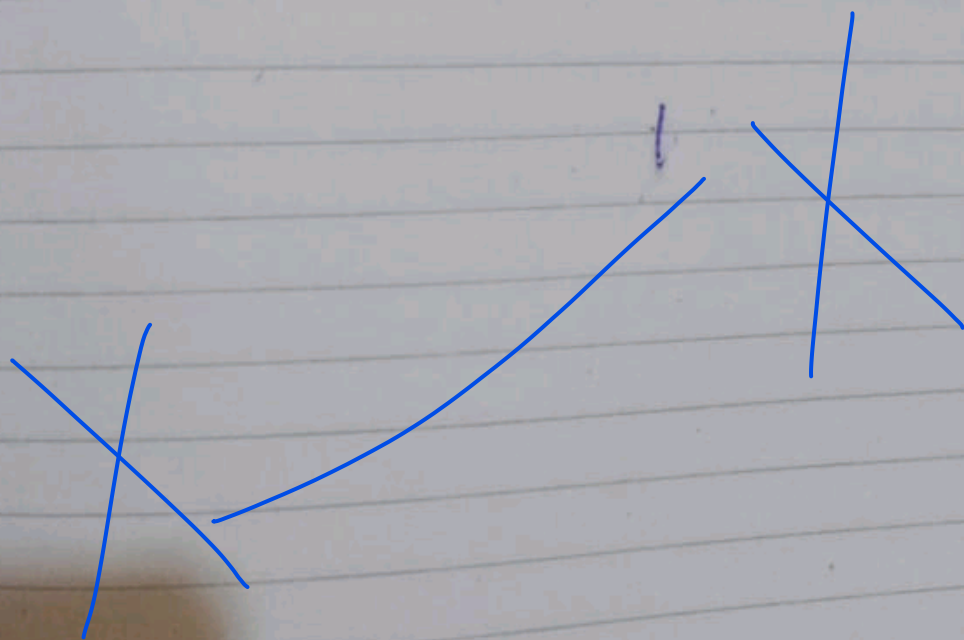
sign  
for back  
propagation

according to our examples:-

$$\text{Err}_6 = O_6 (1 - O_6) (T_6 - O_6)$$

$\bar{I}_4 =$	$-0.4083$	$O_4 =$	$0.399$
$\bar{I}_5 =$	$0.4077$	$O_5 =$	$0.6005$
$\bar{I}_6 =$	$0.5805$	$O_6 =$	$0.6411$

$$E_{0.6} = |-0.0246|$$



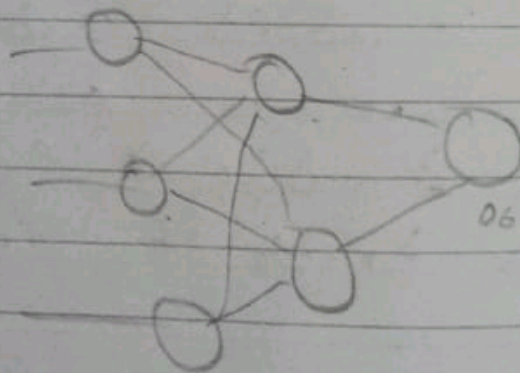


then model take the difference  
of  $O_6$  &  $T_6$

→ this is actually calculating  
error. according to difference.

↳ weight are now updated  
according to error.

This is known as back  
propagation (learning is done on basis  
of this)



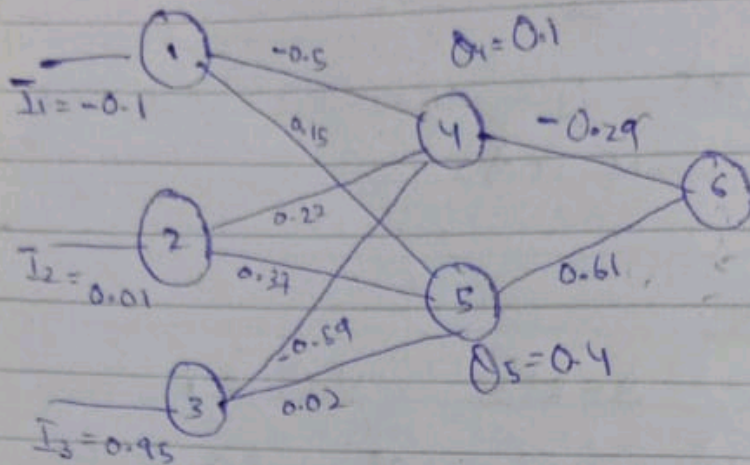
The steps taken from  
actual starting point of inputs  
to the final output.

Feed forward.

Back propagation

→ Every time when back  
propagation occur, weights and  
bias will be updated.

$\odot 1 \dots$



$$\bar{T} = 0.533$$

$$O_1 = -0.1$$

$$O_2 = 0.01$$

$$O_3 = 0.95$$

$$\begin{aligned} \bar{I}_4 &= -0.1 \times -0.5 + 0.01 \times 0.22 + 0.95 \times -0.59 \\ &\quad + 0.1 \\ &= 0.05 + 0.0022 + (-0.5605) + 0.1 \\ &= -0.4082 \end{aligned}$$

$$O_4 = \frac{1}{1 + e^{-\bar{I}_4}} = \frac{1}{1 + e^{-(-0.4082)}} = 0.664845$$

$$\begin{aligned} \bar{I}_5 &= -0.1 \times 0.15 + 0.01 \times 0.37 + 0.95 \times 0.02 \\ &\quad + 0.4 \end{aligned}$$

$$I_5 = 0.4077$$



8/11/21

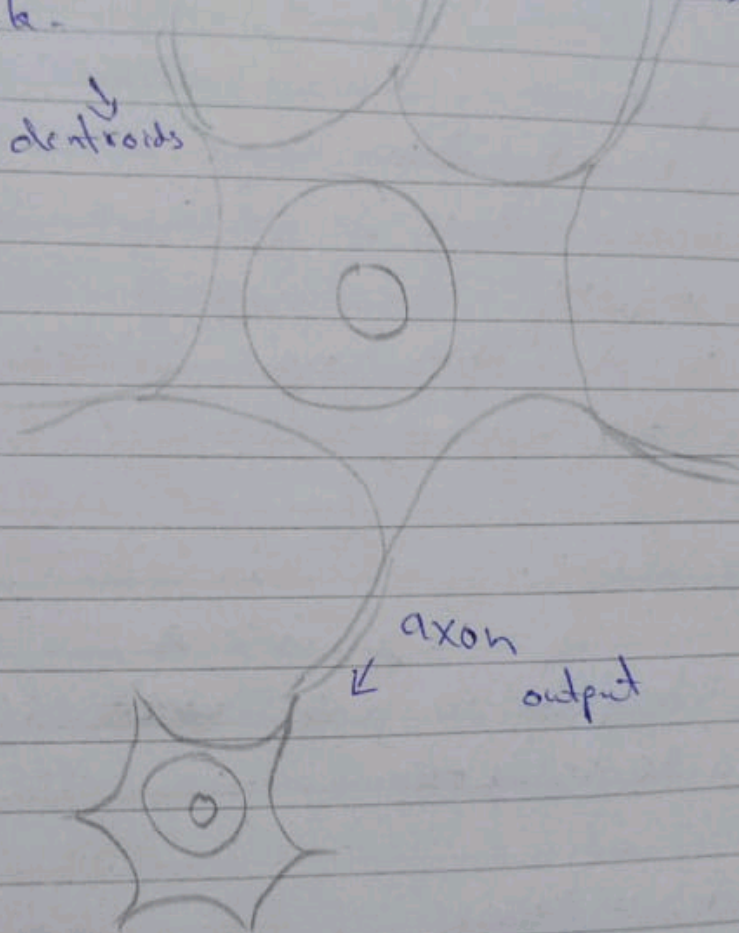
# Artificial Neural Network (ANN)

## Neural Network

mimic human neurons  
network.

input

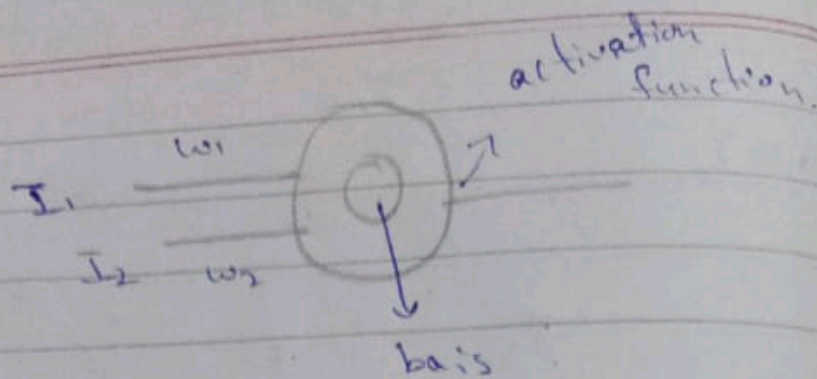
dendroids



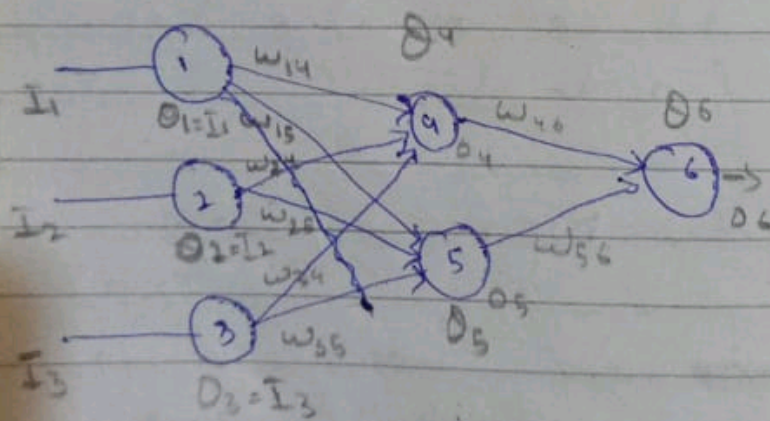
axon

output

→ Take inputs from sensors (input may be multiple) process it and give output (output may be give to multiple neuron).



- weight is given according to neuron.
- initially weight is provided randomly
- we can only manipulate weights and bias.



- $O \rightarrow$  output
- $w \rightarrow$  weight b/w nodes
- $\theta \rightarrow$  bias



# MLP

## Multi-layer perceptron

if atleast 3 ~~hidden~~ layers are present in model.

Initially, simple input is forward to ~~the~~ output of its node

then weighted sum of every node is sum up and value of bias is added and the pass it through activation function then ~~a~~ value is forward to next layers as input.

$$I_4 = I_1 \times w_{14} + I_2 \times w_{24} + I_3 \times w_{34} + b_4$$

$$O_4 = \frac{1}{1 + e^{-I_4}}$$

suppose  $O_4$  (final output) is 0.33  
and  $I_4$  (Target/actual output) is 0.42