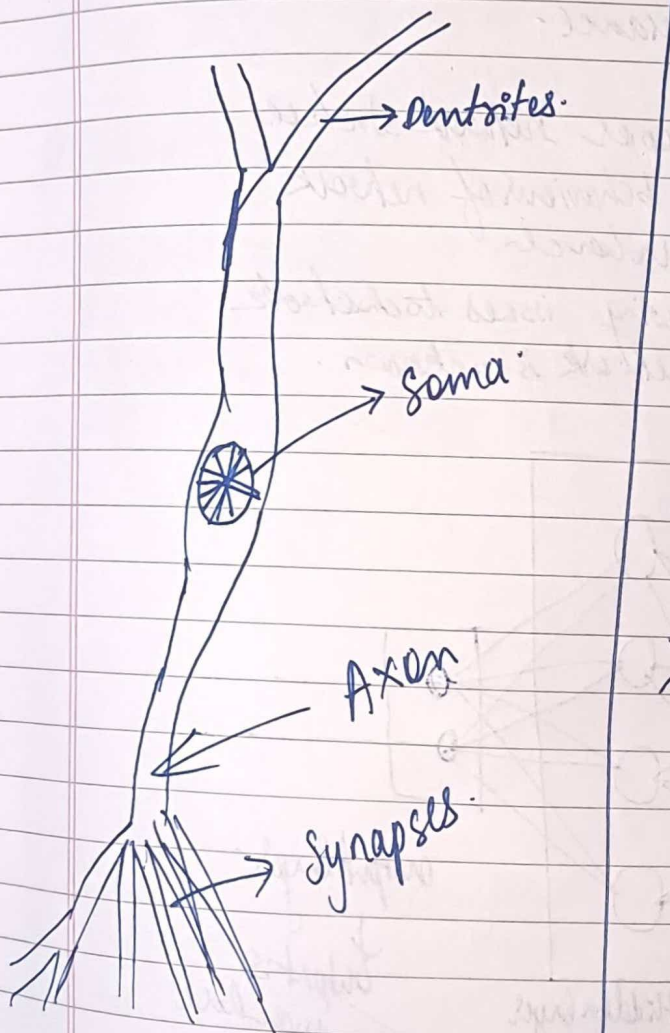
U-6
(ML)

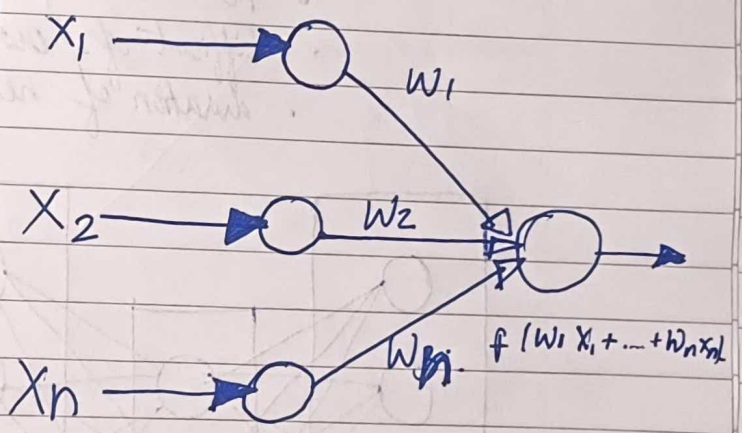
Biological Neuron Model

- Dendarites → accepts inputs.
- Soma → process input
- Axon → Turns processed inputs into output
- Synapses → electrochemical contact between neurons



Artifical Neuron Model.

- ANN
- Edge or connection or link.
- Weight or connection strength

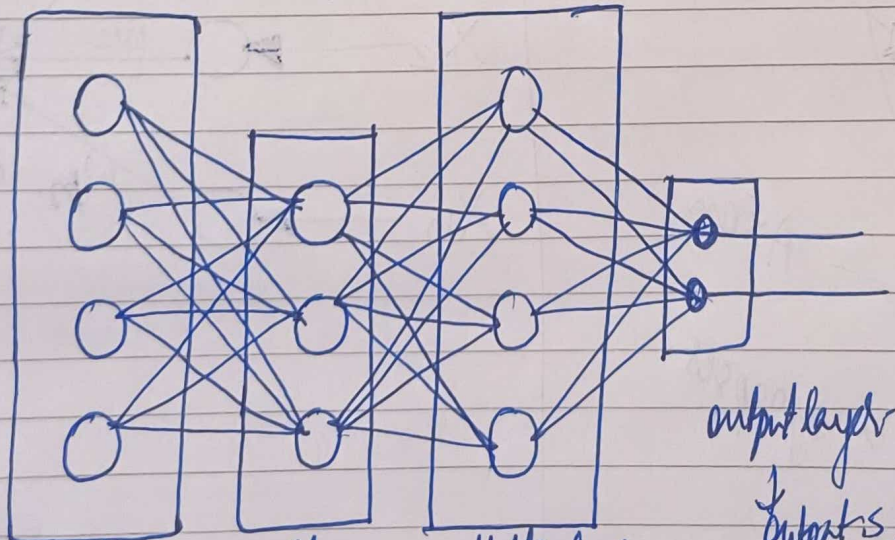


• Artifical Neural network

- inspired by biological neural network.
- mimic network of neurons like human brain so computer can be make decision like human like manner.

Adv → • parallel processing capability
- storing data on entire network.
- capability to work with incomplete knowledge
- Having memory distribution.
- Having fault tolerance.

Disadv → • assurance of proper network structure
- unrecognizable behaviour of network
- Hardware dependence.
- difficult of showing issues to the network
- duration of network is unknown.



Types of ANN

1) Feed Forward ANN
- atleast one layer of neurons
- network intensity → output & input layer
- output → network's output in context of its input.
- benefit → learns to assess & identify input patterns.

2) Feedback ANN
- output loops back into network to achieve best internally involved results
- addressing optimization problems.
- feed info into themselves.
- internal system errors repairs.

Appln of ANN

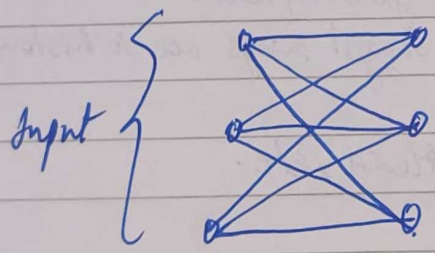
1) Social Media → Insta → People you may know.
2) Sales & marketing → ecommerce → suggest things acc to history
3) Healthcare → facial analysis
4) Personal Assistants → Siri Alexa, etc.

Single Layer Neural Network

- Perceptron is most basic form of neural network.
- consists of set of inputs nodes connected to output nodes using weighted connections.
- Neurons in perceptron are independent of each other.
- No. of input may not be same as number of neurons (output)

Mathematically -

1) Perceptron → 'm' inputs 'n' neurons.
2) ~~each~~ weights are labelled w_{ij} where $1 \le i \le m$.
3) each neuron has its own specific weight denoted as w_{ij}, where i is input node no. & j is output node number.

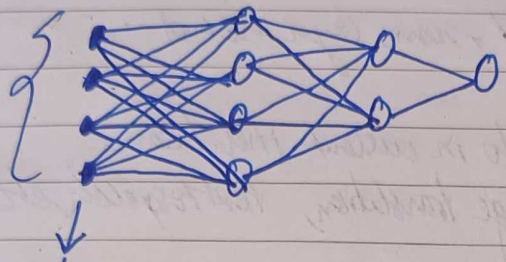


Limitations -

- uses just binary activation function
- only applicable to linear network
- provides an optimal soln due to supervised learning. more training time.
- unable to tackle linear inseperable problems

* Multilayer Perceptron-

- If neural network requires complex decision making we can create multiple layers of perceptron network.



Layer 1 process inputs its output is input for layer 2. along with output, weights are also given.

- at least one layer b/w input & output layer.
- hidden layer b/w i/o layers.
- offers best soln to every categorization issue
- input → non linear multilayer network with linear discriminants us

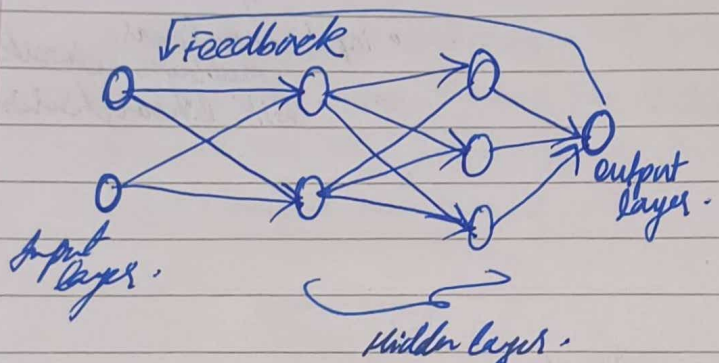
* need for Multilayer networks-

- to solve complex problems
- for huge input output data
- intricate info is too complex for single layer to handle
- multilayer go beyond limitation of single layer.

Recurrent Neural Network (RNN)-

- Feed Forward Neural Network.
- Perceptrons are arranged in layers. Hidden layers are not connected with outside world.
- All nodes fully connected, some layer already.

- need to access previous info in current iteration.
- commonly used in language translation, text to speech, etc



Adv → • remember each info through time
• useful in time series prediction → feature to remember input (LSTM)
• extend effective pixel neighbourhood.

Disadv → • gradient vanishing & exploding problems.
• Training is an difficult task.
• can't process very long sequences if using tanh or relu as activation func.

Types of RNN → one to one
one to many
many to one
many to many.

FLANN →

- Functional Link Artificial Neural Network.
- High order neural network with low computational complexity
- No hidden layers.
- generate additional synthetic inputs for a feed forward NN. by appln fucn to original, raw input

Two choices for input data –

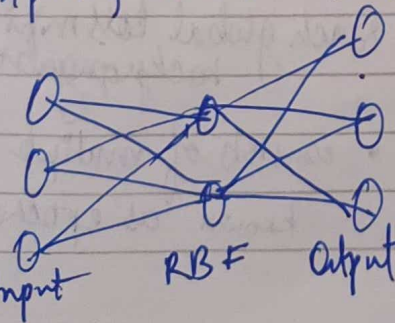
1) Outerproduct or Tensor model –
   - product of input with each other
   - inputs = {x0, x1, x2} · after tensor model input will be inputs = {x0, x1, x2, x0x1, x1x2, x2x0}.

2) Fuctional expansion model –
   - apply one or more univariant fucn to each input.
   eg → Input → x
   extra inputs → x^2, x^3, x^n ——

RBF →

- Radial Basis Functions.
  → kernal trick used to classify non linear data using SVM.
  given as $K(x,y) = \exp\left[-\frac{(x-y)^2}{2\sigma^2}\right]$.

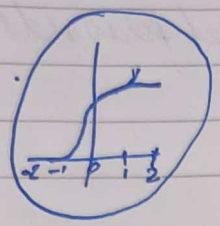
RBFN → consists of input nodes connected by weights to set. RBF neurons which fine proportional to distance b/w input & neuron is weight space



- they never have ≥1 layer of non-linear neurons.

Activation Function –

decides whether a biological neural world fire for a given set of inputs.



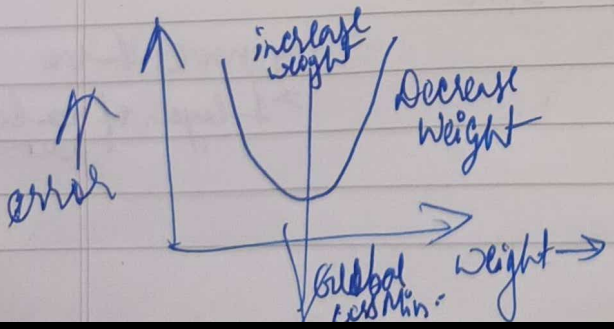
It's crucial in determining accuracy & computational efficiency of model.

Type of activation fun –

1) Identity fun = $g(x) = x$.
2) Binary step function = $g(x) = 1$ when $x > 0$ otherwise 0.
3) Logistic / Sigmoid = $S(x) = \frac{1}{1+e^{-x}}$
4) Tan H = $TanH(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$.
5) Rectified Linear Unit (ReLU):
$R(x) = \max(0, x)$

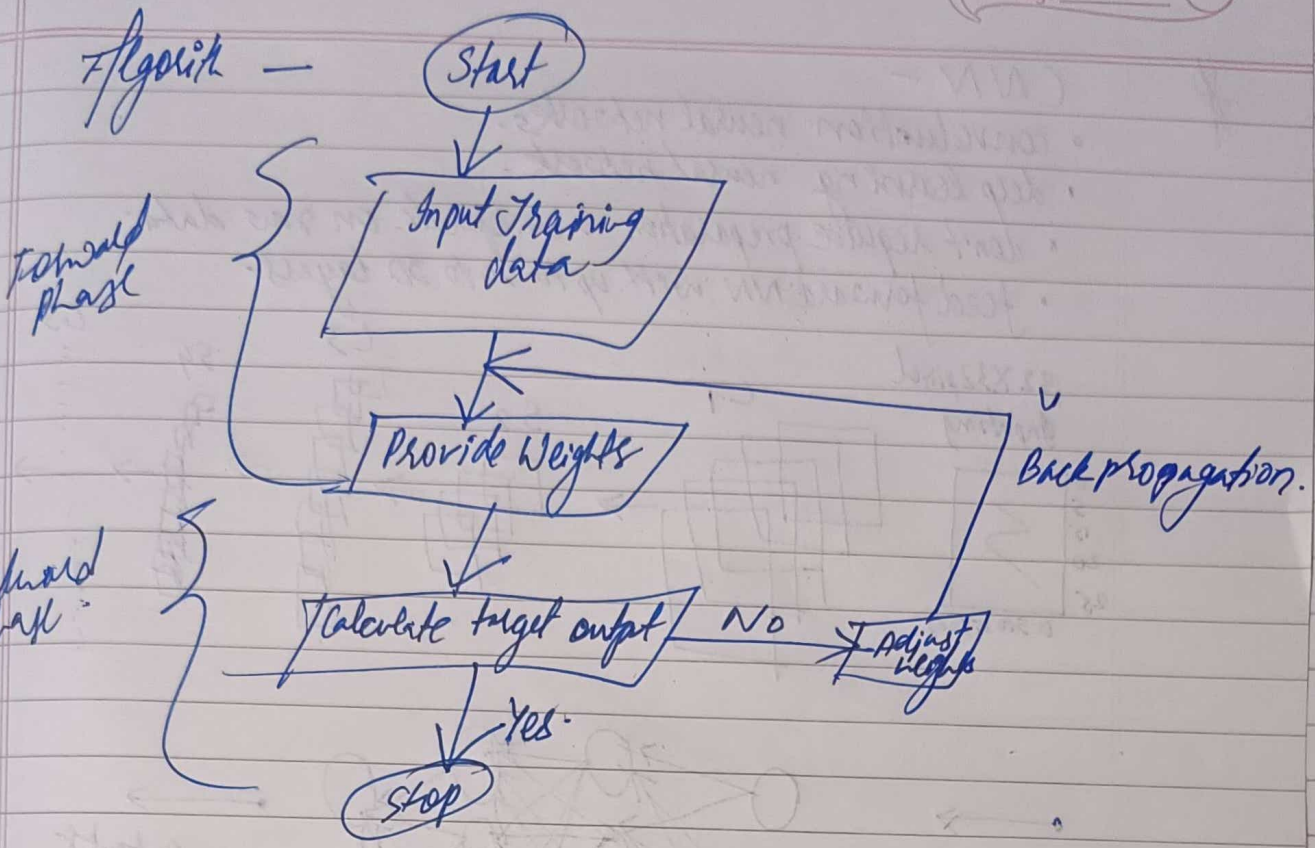
Back propagation –

- algo for supervised learning for ANN
- keeps adjusting weights of connected neurons with a goal to reduce deviation of output signal with target output.



- reach global loss minimum using backpropagation
- consists of multiple iterations known as epochs

Algorithm —



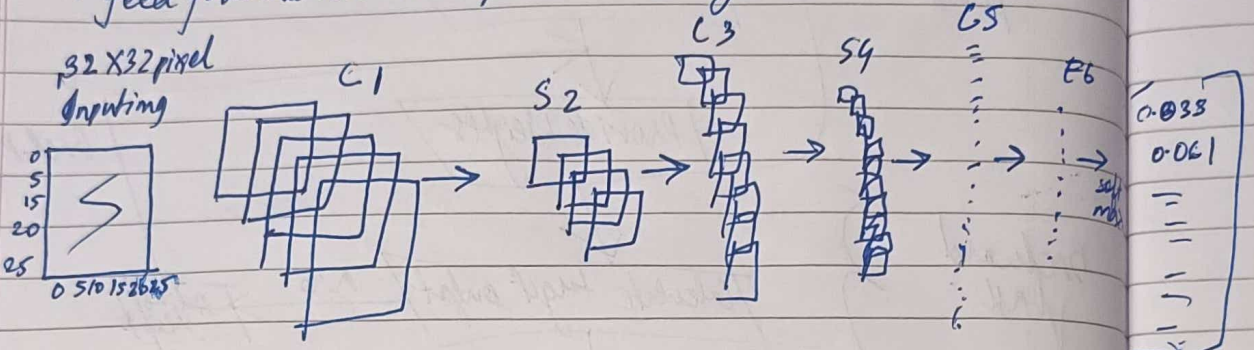
Features → gradient descent method → case of simple perceptron network with diff unit
→ weights are calculate in learning period of network
→ • feed forward of input training pattern
• calculation & backpropagation of error
• updation of weight.

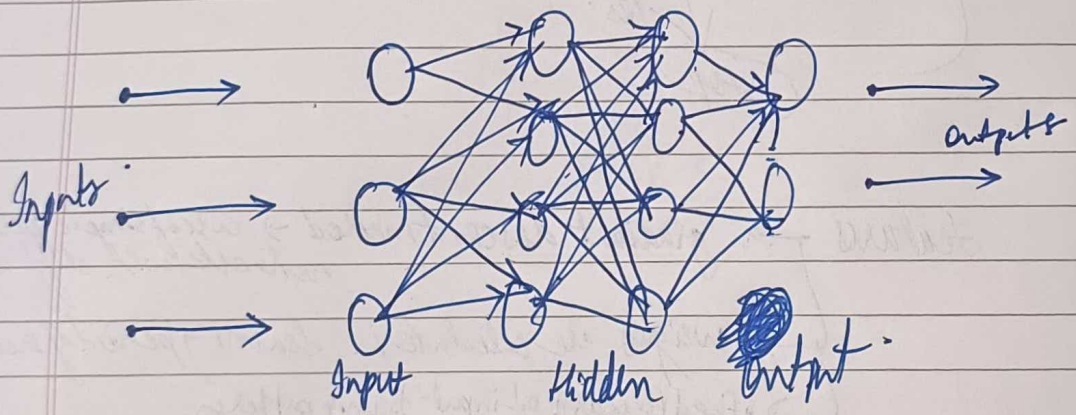
Adv → • simple, fast & easy
• only no. of input alternated, not any other parameter
• flexible & efficient
• No need for user to learn any special function.

Disadv → • Sensitive to noisy data & irregularities.
• performance → depended → data
• too much time taking
• matrix based approach preffed wel mini batch

CNN -

- convolution neural networks.
- deep learning neural network.
- don't require preparation → can operate on raw data.
- feed forward NN with up to 20 to 30 layers.





Adv → • detecting patterns & feature in img, video & audio signals
- Robust to translation, rotation, & scaling invariance
- end-to end training, no need for manual feature extraction
- can handle large amt of data & high accuracy

Dis → • computationally expensive & need lot of memory
- prone to overfitting
- large amount of labeled data
- Interpretability is limited, hard to understand what network has learned.