

## Deep learning techniques

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B.Tech - AI - 'A' section

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NAME: JAMILSELVAN M STD: \_\_\_\_\_ SEC: \_\_\_\_\_ ROLL NO.: \_\_\_\_\_ SUB \_\_\_\_\_

## Ex-Note Study Of Activation Functions And

Its Role In Machine Learning.

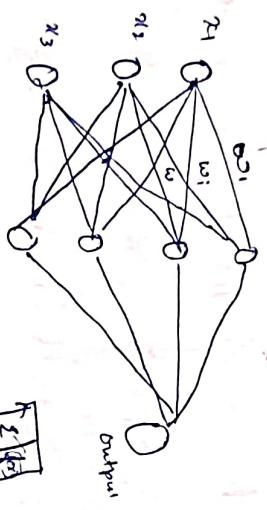
AIM: To understand their role.

→ What are the units in a neural network?

→ Why we need them?

→ filter each input

→ Non-linearity



Neuron Computation

$$z = w_1x_1 + w_2x_2 + \dots + w_nx_n + b$$

Input Symbols

→ To observe how activation functions affect the training of Perceptrons or models.

Pseudocode :

1. Import libraries
2. Define activation function

(i) Sigmoid ( $x$ )

(ii) tanh ( $x$ )

(iii) ReLU

(iv) Softmax

(v) Logit

(vi) Tanh

(vii) Sigmoid

(viii) ReLU

(ix) Softmax

(x) Logit

(xi) Tanh

(xii) Sigmoid

(xiii) ReLU

(xiv) Softmax

(xv) Logit

(xvi) Tanh

(xvii) Sigmoid

(xviii) ReLU

(xix) Softmax

(xx) Logit

(xxi) Tanh

(xxii) Sigmoid

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(xxxvi) Tanh

(xxxvii) Sigmoid

(xxxviii) ReLU

(xxxix) Softmax

(xxxx) Logit

(xxxxi) Tanh

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## Sigmoid (Not Zero-Centred)

- Graph has a characteristic S-shaped curve.

Sigmoid function

- Graph has a characteristic S-shaped curve.
- Vanishing Gradient problem

formula

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

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

No gradient at zero, gradient goes to infinity

Exponential gradient vanishes

gradient

$$\sigma'(x) = \sigma(x)(1-\sigma(x))$$

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$$\tanh(x) = \frac{\sinh(x)}{\cosh(x)}$$

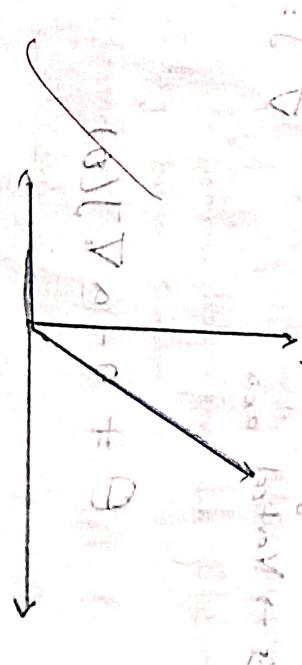
$$\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

graph



Softmax ReLU (Rectified Linear Unit)

$$f(x) = \max(0, x)$$



Result

Learned Activation function and its

different types [tanh, sigmoid, ReLU].

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

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

$$\sigma'(x) = \sigma(x)(1-\sigma(x))$$

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