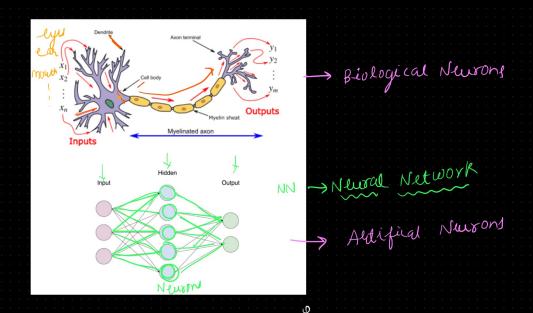
Artificial Newal Network

Artifical Neural Network (ANN)



Neural Network -> computational model

Neural -> Neurons + layers

Neurone: Building Block of Neural Network

O Each Neurong ruceires inputs

Lager: Input layer -> receive the input

Hidden layer -> Neural Network -> operation

Output layer -> final output

ANN Preception Single layered Neural Network Hidden layer J/P layer OIP layer > 0/P Neuron ≥ signal X3 procus Simplest form of Neural Network is preciptron
Backward Propogation Hidden layer #/P layer 7 Olp layer Benary classification 2 Activation Function WI Activation Act (Zxi wi) + big (2) (1) → 0/P $\mathcal{X}_{\mathcal{L}}$ v_3

Neuron

x + W3 + B procus

= x1W1 + x2W2 +

X3

≥ segnal

torward

propogation

Simple livear Regression $y = x_i \omega_i + \beta$ y = mx + C $y = \beta_0 + \beta_1 x$ xi = data points W = WeightSlope, Interept 13 = Bi ay [B, Constant W -> strength of connection eg - You're leaving to ride a sigule paddly, balany practise - perfect Strong + Neurony -> Weight is a parameter that determine the strength of connections blw neurons in defferent layers Bias - It allows to neurons activate even when lyuts on zero. $y = \beta$ $y = x_i \omega_i + \beta$ input = 0

Mathematical Intution

ANN -> Conapt of Perceptson

A simple model of neurons

- (2) Inputs \times weight \Rightarrow $x_i \omega_i$, $x_i \omega_i$, $x_i \omega_i$, $x_i \omega_i$, $x_i \omega_i$
- $3) \qquad \chi_1 w_1 + \chi_2 w_2 + \chi_3 w_3 + b$
- 4) The output of the perceptron of Apply Activation Function

$$y = \sqrt{(\omega_1 x_1 + \omega_2 x_2 + \omega_3 x_3 + \cdots + \omega_n x_n + 6)}$$
sigmoid
$$y \to \text{truth value}$$

$$\hat{y} \to \text{predicted value}$$

$$(y - \hat{y}) \to \text{minimise}$$

$$[y - \hat{y}] \rightarrow \underset{=}{\text{minimist}}$$