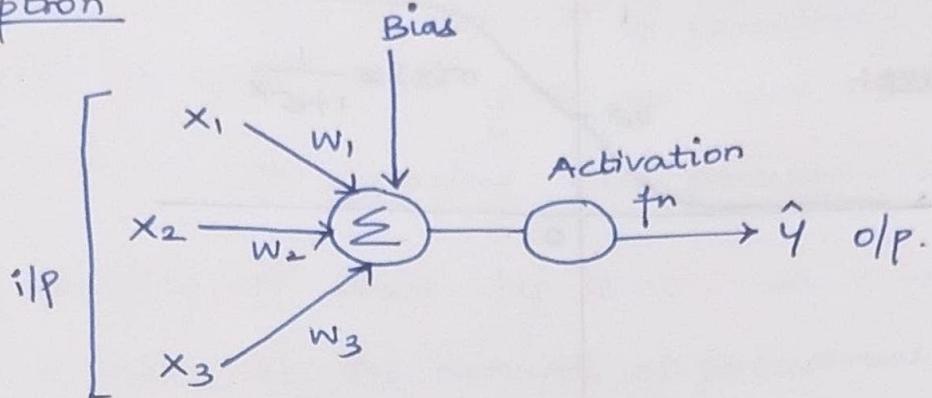


Perception



$$\text{i.e., } \hat{y} = \text{Activation-fn} \left(\sum_{i=1}^n w_i x_i + b \right)$$

so, it gives weight to each of its input values and computes the o/p based on its sum and threshold value.

In case of sigmoid neuron, the activation function is a sigmoid function:-

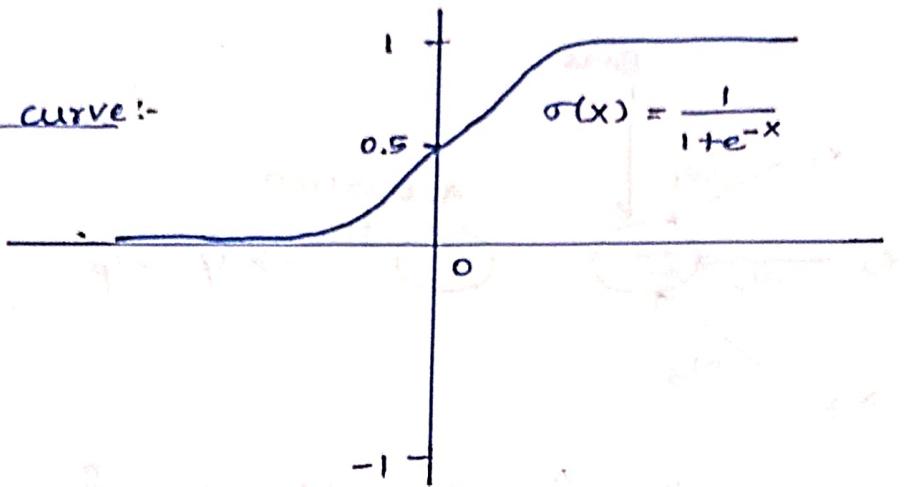
$$\text{i.e., } \sigma(z) = \frac{1}{1 + e^{-z}}$$

In case of perception, it is a step function

$$\text{i.e., } H(z) = \begin{cases} 0 & \text{if } z < a \\ 1 & \text{if } z \geq a \end{cases}$$

So, can we say sigmoid neuron as a perception with sigmoid fn as its activation fn? YES

Sigmoid curve:-



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

OPPONENT

thus, $\sigma(x) : \mathbb{R} \rightarrow [0, 1]$

so we define f to be $f(x) = w_0x + b$

but it's not differentiable at the points off the sigmoid base

$$\frac{\partial f}{\partial w} = x \quad \frac{\partial f}{\partial b} = 1$$

so if we want gradient descent for this, we can't use the sigmoid because it's not differentiable at the points off the sigmoid base

$$\Delta_w \approx \frac{\partial f}{\partial w} \Delta w + \frac{\partial f}{\partial b} \Delta b$$

so we can't use the sigmoid for gradient descent

Ex: 1

suppose we take all the w & b in a network of perceptrons, and multiply them by a +ve constant, show that the behaviour of the network doesn't change.

converting to the sigmoid example part three

Given is a network of perceptron

$$f(x) = g(\sum w_i x_i + b)$$

here,

$g(x)$ is a step function, i.e., $g(x) = \begin{cases} 0, & x < 0 \\ 1, & x \geq 0 \end{cases}$

so, when we multiply the result by a +ve constant, it does not change the result. Because, it is based only on the sign (not magnitude).

part II

→ Now, we have a network of perceptrons.

→ At perceptron, it is given as $w \cdot x + b \neq 0$

→ Now, we are replacing each perceptron with a sigmoid neuron.

→ multiply all web by $c \rightarrow 0$. as $c \rightarrow \infty$, the network behaviour is same as the network of perceptrons. (show)

$$z_c = c(w \cdot x + b) \quad \text{as } c \rightarrow \infty \Rightarrow z_c \rightarrow \infty$$

$$\sigma(z_c) = \frac{1}{1 + e^{-c(w \cdot x + b)}} \rightarrow 1 \quad \text{as } c \rightarrow \infty$$

taking $\lim_{c \rightarrow \infty}$

$$\lim_{c \rightarrow \infty} \sigma(z_c) = \sigma(\infty)$$

Then, case 1 $w \cdot c + b > 0 \Rightarrow -c(w \cdot c + b) \rightarrow +\infty$

$$\lim_{w \cdot c + b > 0} \sigma(z_c) \Rightarrow \sigma(\infty) = 1$$

same as perceptron o/p.

case 2 $w \cdot c + b < 0 \Rightarrow -c(w \cdot c + b) \rightarrow -\infty$

$$\Rightarrow \sigma(z_c) = 0$$

same as perceptron o/p.

But when $w \cdot c + b = 0$, sigmoid becomes not defined.

$$\sigma(z_c) = \frac{1}{1+1} = 0.5$$

Not the behavior of perceptron.

$$\boxed{\text{PP.0} - \times \text{d}}$$

Exercise #8 Find the set of weights & biases for the 4th layer.

Assumption

First 3 layers are correct with 3rd layer has an activation of ≥ 0.99 for correct o/p neuron & < 0.01 activation for incorrect neuron.

ilp layer is $a = (a_0, a_1, a_2, \dots, a_9)$.

o/p layer is $O = (O_0, O_1, O_2, O_3)$.

Thus, we can have $w^{[1]} = [0, 0, 0, 0, 0, 0, 0, 0, 1, 1]$

$w^{[2]} = [0, 0, 0, 0, 1, 1, 1, 1, 0, 0]$

$w^{[3]} = [0, 0, 1, 1, 0, 0, 1, 1, 0, 0]$

Binary representation

$w^{[4]} = [0, 1, 0, 1, 0, 1, 0, 1, 0, 1]$

since each neuron in 3rd layer has correct o/p, we can directly use these weights ($w^{[1]} \dots w^{[4]}$).

Bias

since for correct digits, o/p will be ≥ 0.99 .

Thus,

$$f = wx + b \geq 0$$

$\downarrow \downarrow$

(1) (0.99)

i.e., $b \geq -0.99$

for wrong digits, o/p will be < 0.01

Thus, $f = wx + b < 0$

(i.e.) $b < -0.01$

so, bias can be in anywhere in range $(-0.99, -0.01)$.