

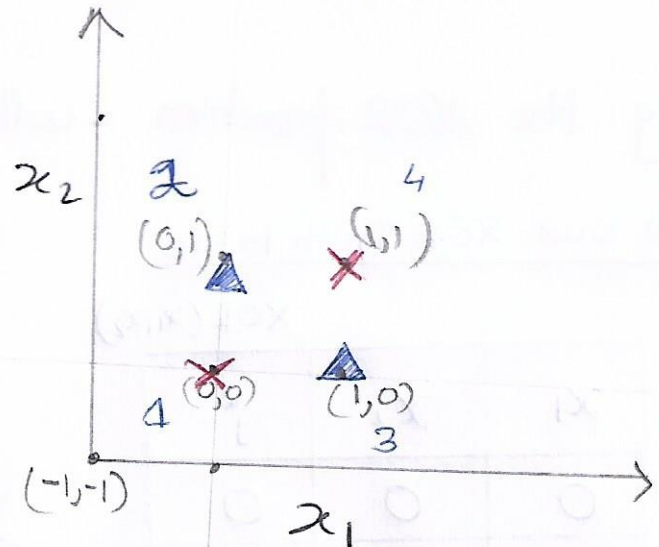
# Introduction to Feed Forward Networks.

①

## The XOR operation

$x_1$	$x_2$	$XOR(x_1, x_2)$
0	0	0
0	1	1
1	0	1
1	1	0

Let us see how the plot looks!



We have the following input:

1] (0,0)

2] (0,1)

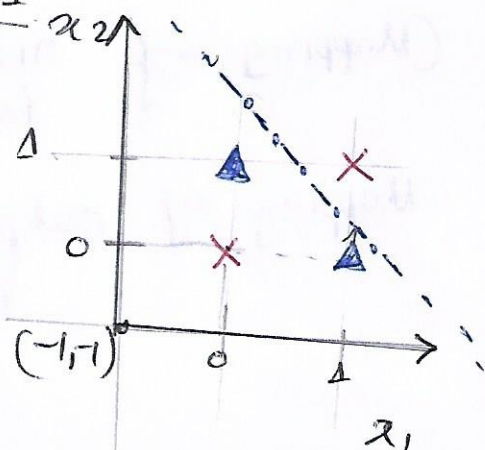
3] (1,0)

4] (1,1).

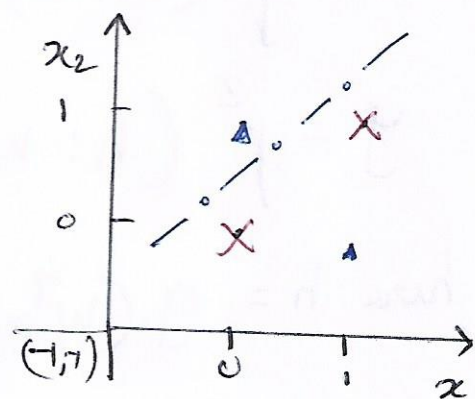
Can we draw a straight line to separate the X from the  $\Delta$ 's?

Let us try

Trial 1

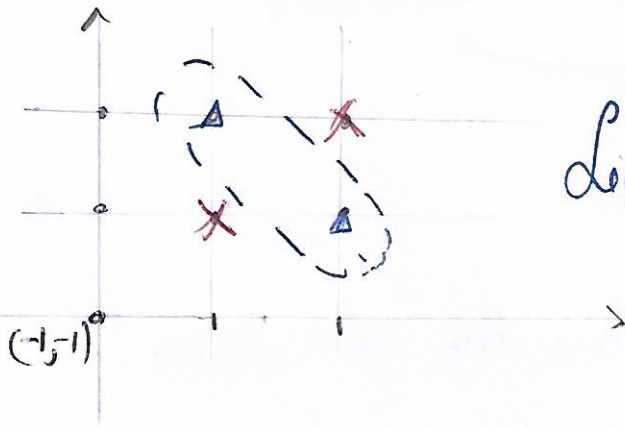


Trial 2



(2)

How can we separate the 1's from the 0's.

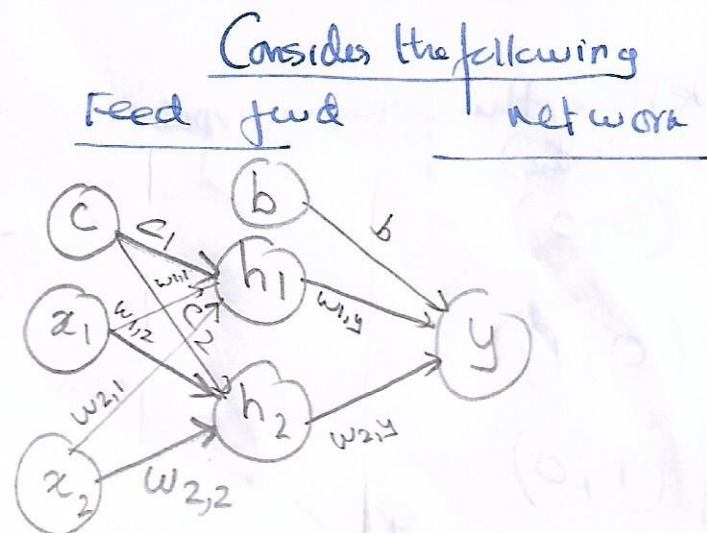


Linear Separation is not possible!

Learning the XOR function with Feed Forward Networks

Revisit our XOR Truth table.

XOR( $x_1, x_2$ )		
$x_1$	$x_2$	$y^*$
0	0	0
0	1	1
1	0	1
1	1	0



Consider a mapping as follows

$$f(x; w, b) = x^T w + b$$

Now

$$h = f'(x; w, c)$$

(Mapping of hidden to input layer)

$$y = f^2(h; w, b)$$

Mapping of output to hidden.

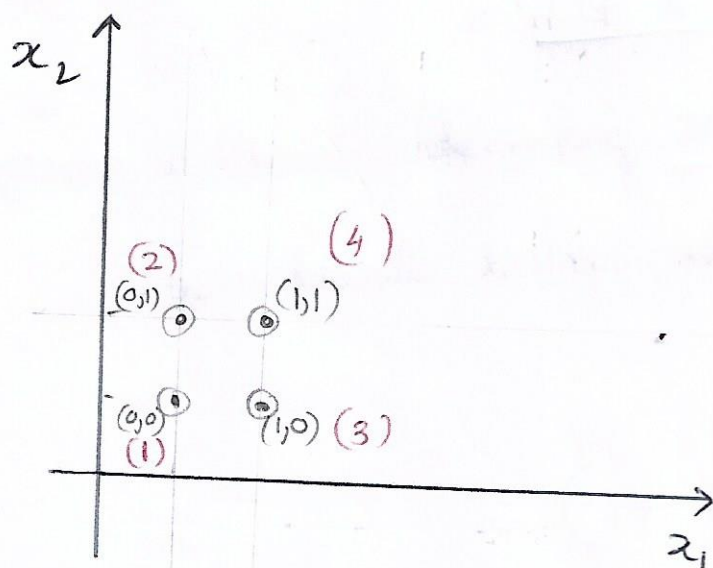
$$\text{here } h = g(w^T x + c)$$

$g$  is the activation function to be used.



# Solving XOR by Linear Functions.

Consider original 2 space as shown below.



$x_1$	0	0	1	1
$x_2$	0	1	0	1
$y$	0	1	1	0

Let us try Fitting Linear Functions with weights  $w_1, w_2$ .  
Thus For the four points in question, we have the following.

$$w_1 \times 0 + w_2 \times 0 + b = 0 \quad \text{--- (1)}$$

$$w_1 \times 0 + w_2 \times 1 + b = 0 \quad \text{--- (2)}$$

$$w_1 \times 1 + w_2 \times 0 + b = 0 \quad \text{--- (3)}$$

$$w_1 \times 1 + w_2 \times 1 + b = 0 \quad \text{--- (4)}$$

if from (1) we get  $b = 0$

and from (2), (3) if  $w_1 = 1$  or  $w_2 = 1$

Then (4) does NOT hold!

Thus a Linear Fit is not directly possible!

Drawbacks of Linear Model.

Vector of weights + Scalar bias parameters

Affine transformation from  $\vec{x}$  to  $\vec{h}$ .

- ① Entire Vector of bias parameters would be needed!
- ② Activation Function to be applied elementwise.

Consider the following.

$$h_i = g(x^T w_i + c_i)$$

Relu  $\rightarrow$  Rectified Linear Unit.

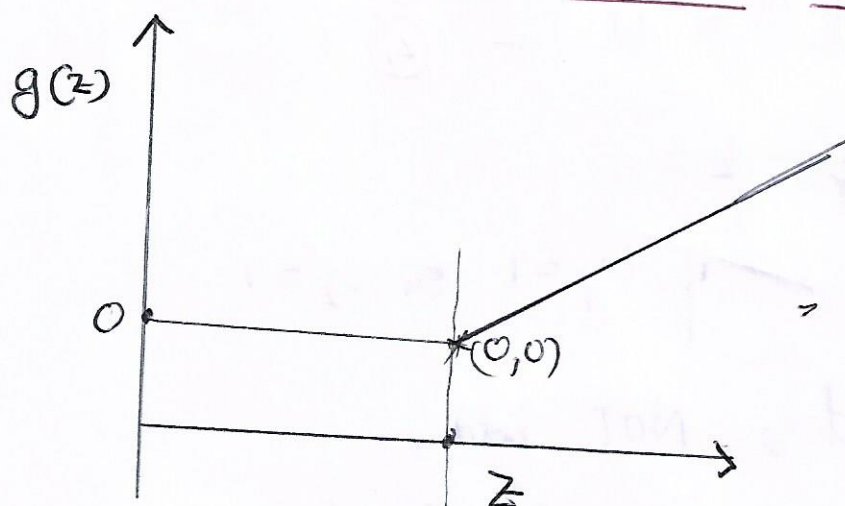
$\hookrightarrow$  how do we model this?

$$g(z) = \max(0, z)$$

Thus our final output maps to input in the following way:

$$f(x; w, c, w, b) = w^T \max\{0, (w^T x + c)\} + b$$

ReLU  $\rightarrow$  Activation Function

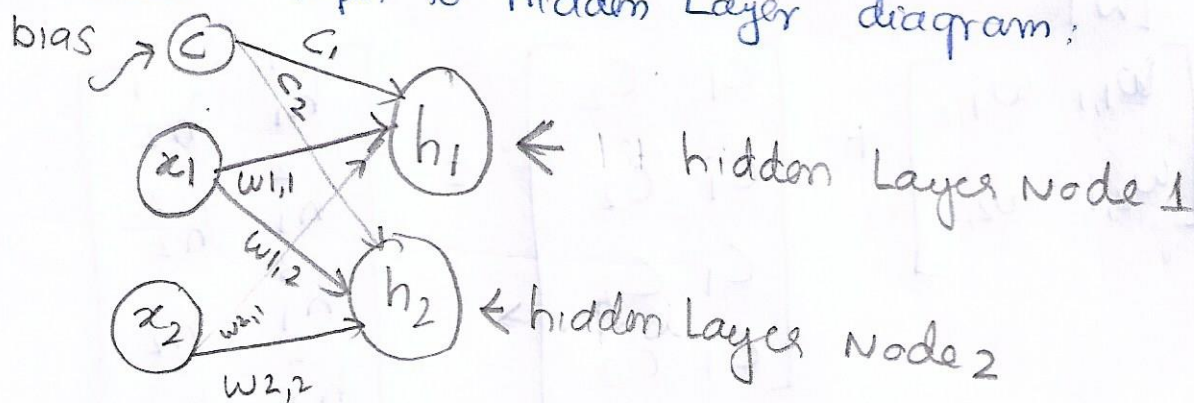




# Let us try Linear Activation fn. with Feed Forward Network.

(5)

Consider input to hidden Layer diagram;



Thus we have; for the 1st hidden node

$$0 \times w_{1,1} + 0 \times w_{2,1} + C_1 = h_1^1$$

$$0 \times w_{1,1} + 1 \times w_{2,1} + C_1 = h_1^2$$

$$1 \times w_{1,1} + 0 \times w_{2,1} + C_1 = h_1^3$$

$$1 \times w_{1,1} + 1 \times w_{2,1} + C_1 = h_1^4$$

and similarly for the 2nd hidden node

$$0 \times w_{1,2} + 0 \times w_{2,2} + C_2 = h_2^1$$

$$0 \times w_{1,2} + 1 \times w_{2,2} + C_2 = h_2^2$$

$$1 \times w_{1,2} + 0 \times w_{2,2} + C_2 = h_2^3$$

$$1 \times w_{1,2} + 1 \times w_{2,2} + C_2 = h_2^4$$

MATRIX FOR

$$\begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} w_{1,1} \\ w_{2,1} \end{bmatrix} + \begin{bmatrix} C_1 \end{bmatrix} = \begin{bmatrix} h_1^1 \\ h_1^2 \\ h_1^3 \\ h_1^4 \end{bmatrix}$$

MATRIX FORM

$$\begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} w_{1,2} \\ w_{2,2} \end{bmatrix} + \begin{bmatrix} C_2 \end{bmatrix} = \begin{bmatrix} h_2^1 \\ h_2^2 \\ h_2^3 \\ h_2^4 \end{bmatrix}$$

New Combining together,

COMBINED MATRIX FORM

$$\begin{bmatrix} [X] \\ [W] \end{bmatrix} + \begin{bmatrix} C_1 & C_2 \\ C_1 & C_2 \\ C_1 & C_2 \\ C_1 & C_2 \end{bmatrix} = \begin{bmatrix} h_1^1 & h_2^1 \\ h_1^2 & h_2^2 \\ h_1^3 & h_2^3 \\ h_1^4 & h_2^4 \end{bmatrix}$$

Where we have

$$x = \begin{matrix} x_1 \\ x_2 \end{matrix} \begin{matrix} 1 & 2 & 3 & 4 \\ \begin{bmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{bmatrix} \end{matrix}$$

$$X = x^T = \begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}$$

Parameters Assumed for solution are as follows:

$$W = \begin{bmatrix} w_{1,1} & w_{1,2} \\ w_{2,1} & w_{2,2} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \leftarrow \text{Weights matrix (Input} \rightarrow \text{hidden)}$$

$$C = \begin{bmatrix} C_1 \\ C_2 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \leftarrow \text{bias vector (Input} \rightarrow \text{hidden)}$$

$$w = \begin{bmatrix} w_{1,y} \\ w_{2,y} \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix} \leftarrow \text{Weights vector (hidden} \rightarrow \text{output)}$$

$$b = [0] \leftarrow \text{bias (hidden} \rightarrow \text{output)}$$



# Computing the Solution with Assumed Parameters.

Step 1 : Multiply Input matrix with 1st Layer weights

$$[X][W] = \begin{matrix} X \\ \begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix} \\ [4 \times 2] \end{matrix} \cdot \begin{matrix} W \\ \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ [2 \times 2] \end{matrix} = \begin{matrix} XW \\ \begin{bmatrix} 0 & 0 \\ 1 & 1 \\ 1 & 1 \\ 2 & 2 \end{bmatrix} \\ [4 \times 2] \end{matrix}$$

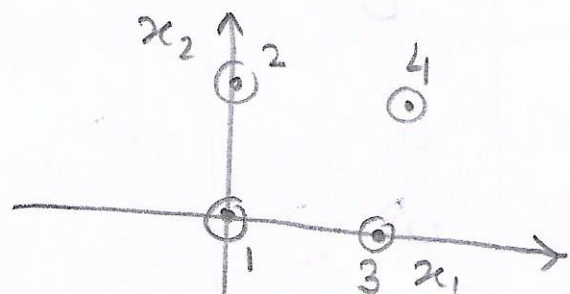
Step 2 : Add the bias vector

$$[X][W] + C$$

$$\begin{bmatrix} 0 & 0 \\ 1 & 1 \\ 1 & 1 \\ 2 & 2 \end{bmatrix} + \begin{bmatrix} 0 & -1 \\ 0 & -1 \\ 0 & -1 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix}$$

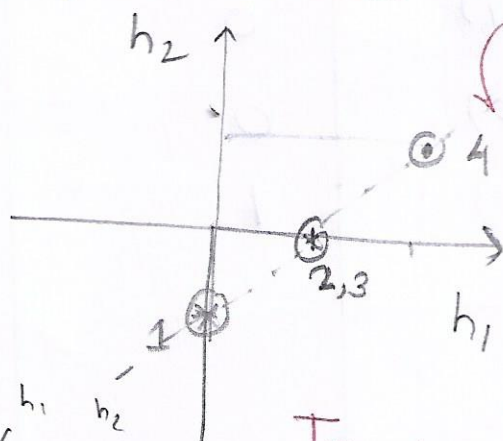
Transformed Space.

Plot Original Space and Transformed Space



- 1  $(0, 0)$
- 2  $(0, 1)$
- 3  $(1, 0)$
- 4  $(1, 1)$

ORIGINAL  
INPUT SPACE



- 1  $(0, -1)$
- 2  $(1, 0)$
- 3  $(1, 0)$
- 4  $(2, 1)$

Transformed  
SPACE

Step 3] Apply ReLU Activation

⑧

$$g(z) = \max\{0, z\} = \max(0, xw + c)$$

§ Considers 1st node of hidden layer. ( $h_1$ )

original input	$h_1$	ReLU	$g(h_1)$
(0,0)	0	$g(h_1^1)$	0
(0,1)	1	$g(h_1^2)$	1
(1,0)	1	$g(h_1^3)$	1
(1,1)	2	$g(h_1^4)$	2

§ Considers for 2nd node of hidden layer ( $h_2$ )

original input	$h_2$	ReLU	$g(h_2)$
(0,0)	-1	$g(h_2)^1$	0
(0,1)	0	$g(h_2)^2$	0
(1,0)	0	$g(h_2)^3$	0
(1,1)	1	$g(h_2)^4$	1



This post ReLU we have the input bars joined as.

$x_1 \ x_2$

$$\begin{bmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}$$

→ Post ReLU

$$\begin{bmatrix} h_1' & h_2' \\ 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix}$$

Now consider mapping from hidden to output.

we have.

$$b + 0 \times w_{1,y} + 0 \times w_{2,y} = y^1$$

$$b + 1 \times w_{1,y} + 0 \times w_{2,y} = y^2$$

$$b + 1 \times w_{1,y} + 0 \times w_{2,y} = y^3$$

$$b + 2 \times w_{2,y} + 1 \times w_{2,y} = y^4$$

Putting it all together.

XOR Table

$x_1$	$x_2$	$y$
0	0	0
0	1	1
1	0	1
1	1	0

hidden Layer to output

MATRIX FORM

$$\begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} w_{1,y} \\ w_{2,y} \end{bmatrix} + [b] = \begin{bmatrix} y^1 \\ y^2 \\ y^3 \\ y^4 \end{bmatrix}$$

Now Consider Solution parameters as proposed.

(10)

$$w = \begin{bmatrix} w_{1,y} \\ w_{2,y} \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$b = [0]$$

Thus we have.

$$\begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 1 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \end{bmatrix} + [0] = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix} \quad \text{Output}$$

Now Compare!

XOR

$x_1$	$x_2$	$XOR(x_1, x_2)$
0	0	0
0	1	1
1	0	1
1	1	0

Output From FFN

$$\begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

Food For thought?

How do we get the bias and weights?