Introduction to Deep Learning

Neural Network

Simple Neural Network Model

• Logistic regression model:

$$\hat{y} = \sigma(w_0 + w_1 * x_1 + w_2 * x_2)$$

• Two steps:

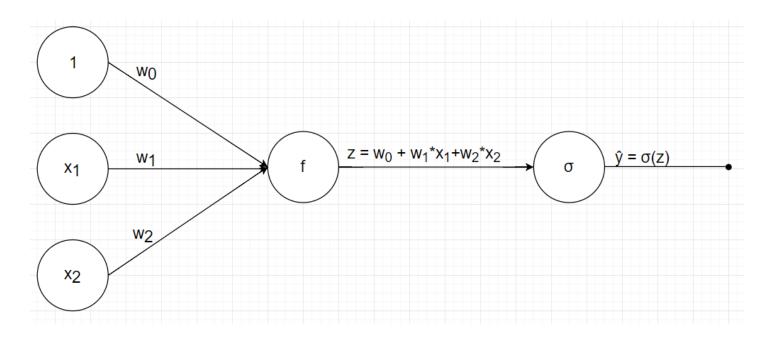
(1) Linear sum:

$$z = 1 * w_0 + x_1 * w_1 + x_2 * w_2$$

(2) Apply sigmoid function:

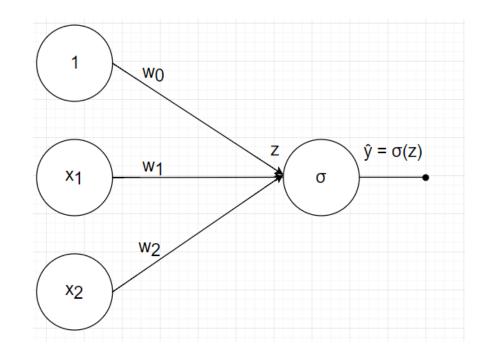
$$\hat{y} = \boldsymbol{\sigma}(z)$$

Simple Neural Network Model



Flow chart of logistic regression model

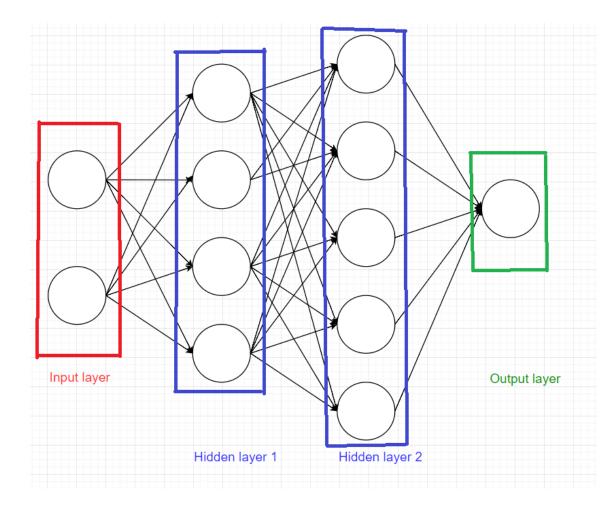
Simple Neural Network Model

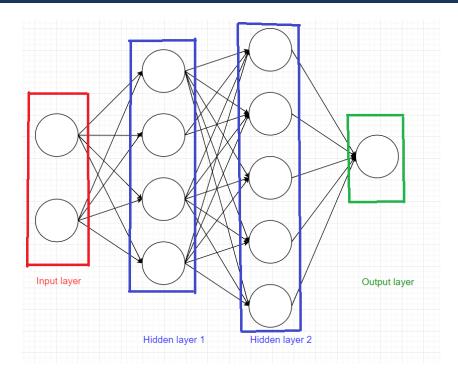


Flow chat of logistic regression model

- W₀ is called bias coefficient, or free coefficient
- Sigmoid function is called activation function

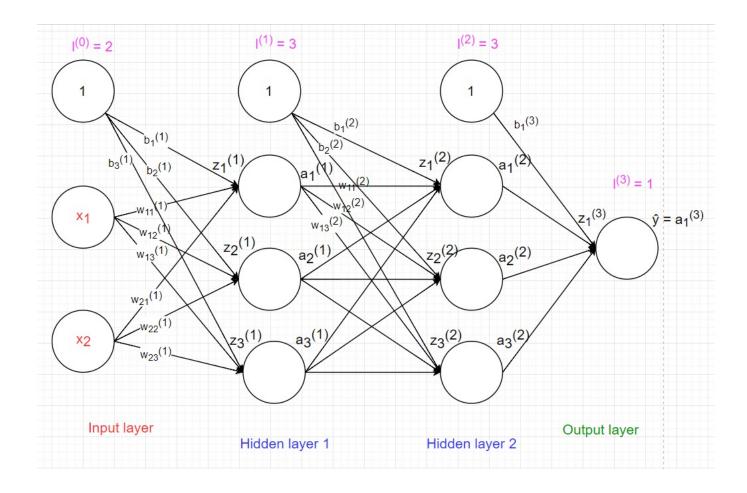
- Input layer and output layer are required
- Hidden layers are optional
- Total layers = # layers 1
- Each circle is called one node



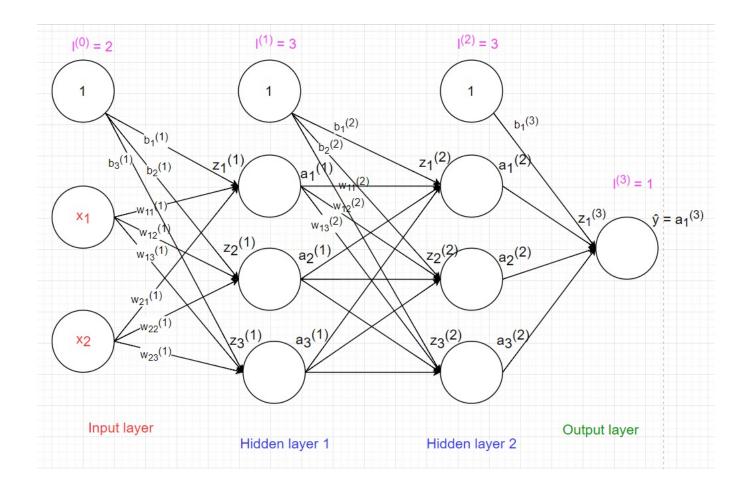


Each node in hidden layer and output layer:

- Is connected with all nodes with previous layer with the coefficents w
- Has a bias coefficient w0
- Follows two steps of linear sum and appliance of activation function (sigmoid)



The neural network has: 3 layers, 2 nodes in input layer, 3 nodes in hidden layer 1, 3 nodes in hidden layer 2, 1 node in output layer



Note: node 1 is not considered as a node since it is used to calculate bias of the node in the next layer

• Node i in layer I with bias b_i^(I) has 2 steps:

(1) Linear sum:
$$z_i^{(l)} = \sum_{j=1}^{l^{(l-1)}} a_j^{(l-1)} * w_{ji}^{(l)} + b_i^{(l)}$$

(2) Apply activation function: $a_i^{(l)} = \boldsymbol{\sigma}(z_i^{(l)})$

• At node 2 of layer 1, we have:

$$z_{2}^{(1)} = x_{1} * w_{12}^{(1)} + x_{2} * w_{22}^{(1)} + b_{2}^{(1)}$$
$$a_{2}^{(1)} = \boldsymbol{\sigma}(z_{2}^{(1)})$$

• At node 3 of layer 2, we have:

$$z_3^{(2)} = a_1^{(1)} * w_{13}^{(2)} + a_2^{(1)} * w_{23}^{(2)} + a_3^{(1)} * w_{33}^{(2)}$$
$$a_3^{(2)} = \sigma(z_3^{(2)})$$

Feedforward

• Let call input layer $x = a^{(0)}$, with size 2*1, we have:

$$\begin{split} z^{(1)} &= \begin{bmatrix} z_1^{(1)} \\ z_2^{(1)} \\ z_3^{(1)} \end{bmatrix} = \begin{bmatrix} a_1^{(0)} * w_{11}^{(1)} + a_2^{(0)} * w_{21}^{(1)} + a_3^{(0)} * w_{31}^{(1)} + b_1^{(1)} \\ a_1^{(0)} * w_{12}^{(1)} + a_2^{(0)} * w_{22}^{(1)} + a_3^{(0)} * w_{32}^{(1)} + b_2^{(1)} \\ a_1^{(0)} * w_{13}^{(1)} + a_2^{(0)} * w_{23}^{(1)} + a_3^{(0)} * w_{33}^{(1)} + b_3^{(1)} \end{bmatrix} \\ &= (W^{(1)})^T * a^{(0)} + b^{(1)} \end{split}$$

Feedforward

• Similarly, we have:

$$z^{(2)} = (W^{(2)})^T * a^{(1)} + b^{(2)}$$

$$a^{(2)} = \boldsymbol{\sigma}(z^{(2)})$$

$$z^{(3)} = (W^{(3)})^T * a^{(2)} + b^{(3)}$$

$$\hat{y} = a^{(3)} = \boldsymbol{\sigma}(z^{(3)})$$

Feedforward



feedforward neural network

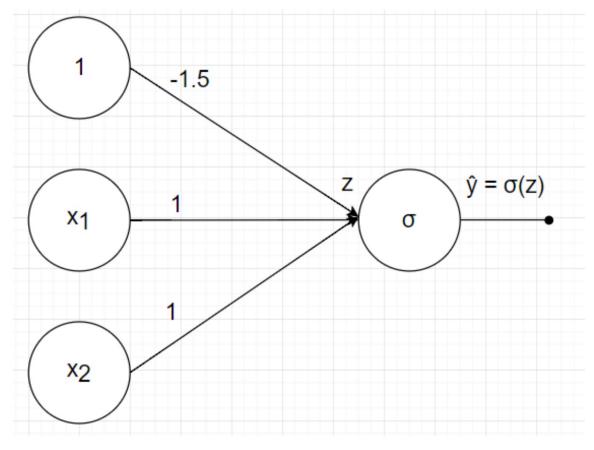
Loss function

- Gradient descent algorithm
- Step of derivative calculation of coefficients of loss function is done with the backpropagation algorithm Will be taught in the next lecture

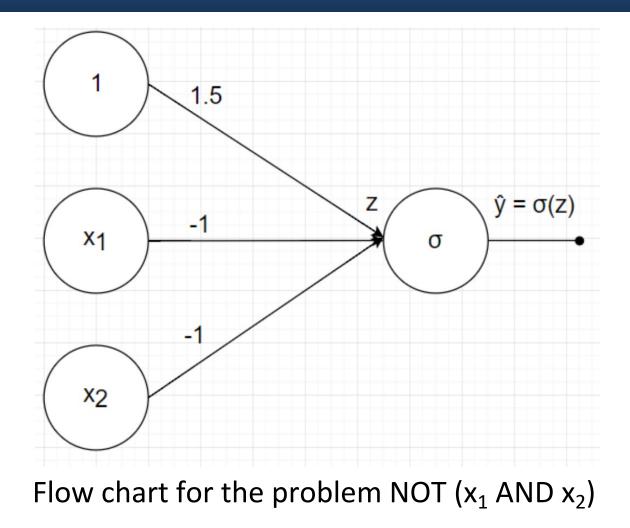
Problem: AND

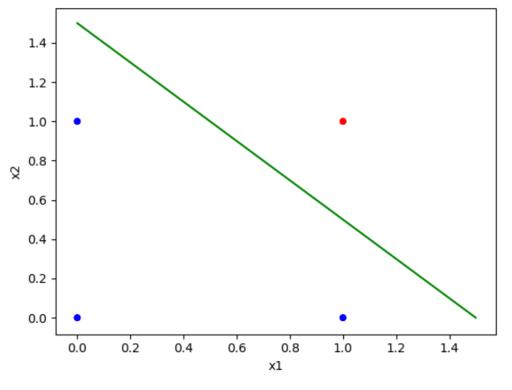
x_1	<i>x</i> ₂	У
0	0	0
0	1	0
1	0	0
1	1	1

 $x_1 AND x_2$



Flow chart for the problem x_1 AND x_2



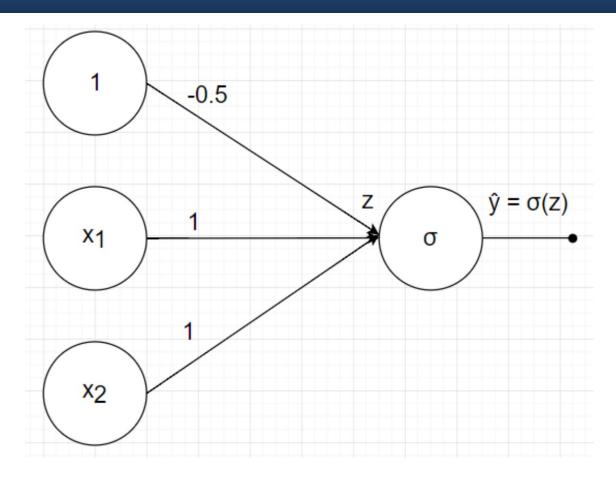


Separation line y = $1.5 - 1 * x_1 - 1 * w_2$

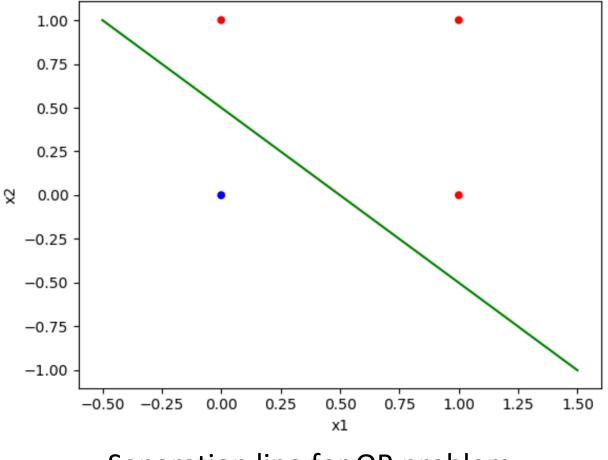
Problem: OR

A	В	A OR B
0	0	0
0	1	1
1	1 0 1	
1	1	1

 $x_1 \text{ OR } x_2$



Flow chart for the problem $x_1 OR x_2$

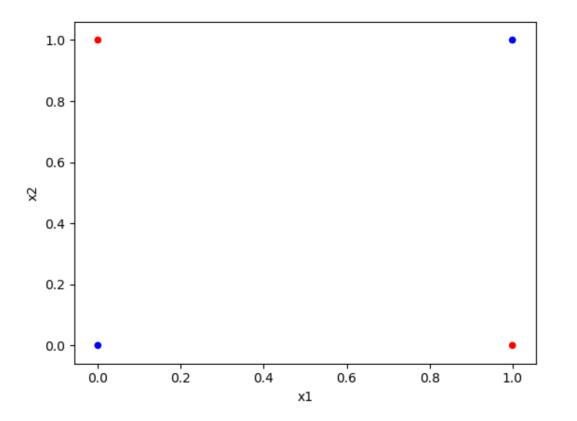


Separation line for OR problem

Problem: XOR

Α	В	A XOR B	
0	0	0	
0	1	1	
1	0	1	
1	1	0	

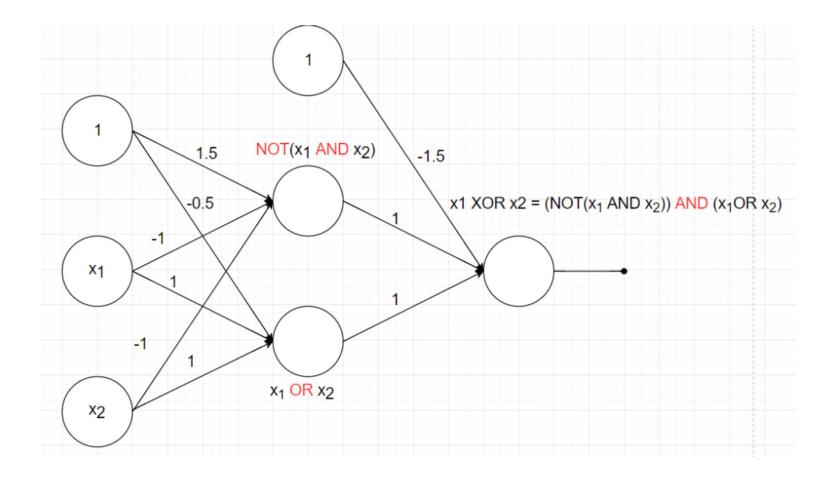
 $x_1 XOR x_2$



Cannot draw a separation line for XOR problem → Cannot solve the XOR problem using logistic regression → Neural network ???

Rewrite XOR problem:

A	B	A XOR B	A AND B	NOT (A AND B)	A OR B	(NOT(A AND B) AND (A OR B))
0	0	0	0	1	0	0
0	1	1	0	1	1	1
1	0	1	0	1	1	1
1	1	0	1	0	1	0



Solve XOR problem with several logistic regression models

