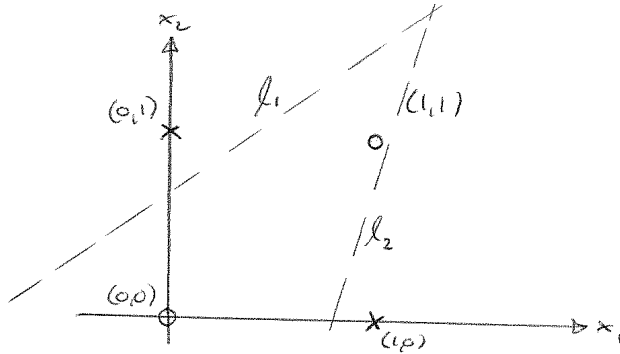


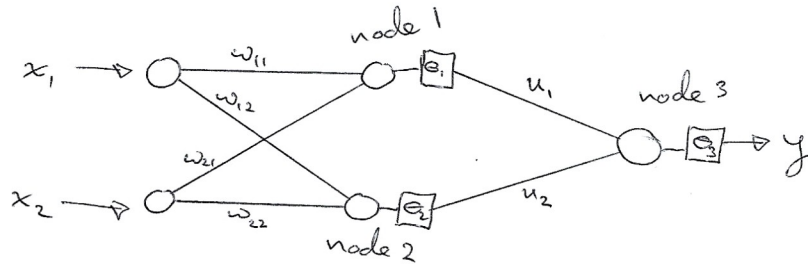
Adaptive Computation and Machine Learning

The dataset $D = \{((0,0),0), ((0,1),1), ((1,0),1), ((1,1),0)\}$ is not linearly separable, i.e., it is impossible to draw a straight line in \mathbb{R}^2 so that all the 1's are on one side of the line and all the 0's are on the other. However, it is possible to separate the dataset using **two** straight lines, as shown below, where a \circ represents a target of 0 and a \times represents a target of 1.



If a point lies above line ℓ_1 and above line ℓ_2 then it is a \times . If it lies below ℓ_1 and above ℓ_2 then it is a \circ ; and if it lies below line ℓ_1 and below line ℓ_2 then it is a \times .

We can use a 2-layer perceptron to model the separation by two straight lines as follows:



EXERCISE: Try to find suitable values for w_{11} , w_{12} , w_{21} , w_{22} , u_1 , u_2 , θ_1 , θ_2 and θ_3 so that the above 2-layer perceptron correctly outputs the data in D .

Solution:

There are many possible solutions. For example,

$$w_{11} = 1, w_{12} = 1, w_{21} = -1, w_{22} = -1, u_1 = 2, u_2 = -1, \theta_1 = \frac{1}{2}, \theta_2 = -\frac{1}{2} \text{ and } \theta_3 = -\frac{1}{2}$$

EXERCISES

- (1) Consider a network with 4 input nodes, one hidden layer with 4 nodes and 2 output nodes. The weights between layer 0 and layer 1, and the biases at layer 1 are given by:

$$W_1 = \begin{bmatrix} 4 & -5 & 0 & 1 \\ -3 & 6 & -1 & 2 \\ 0 & 1 & 1 & -2 \\ 2 & 0 & -3 & 4 \end{bmatrix}, \quad \mathbf{b}_1 = (1, -2, 0, -1).$$

The weights between layer 1 and layer 2, and the biases at layer 2 are given by:

$$W_2 = \begin{bmatrix} -2 & 1 \\ -1 & 0 \\ 1 & -3 \\ 5 & -1 \end{bmatrix}, \quad \mathbf{b}_2 = (-2, 2).$$

- (a) Using the sigmoid activation function σ at each node, calculate the output from the network for the following inputs:

- (i) $(1, 0, 2, -3)$
- (ii) $(-2, 3, 1, 0)$
- (iii) $(1, 0, 1, 1)$

- (b) Using the activation function *relu* at each node, calculate the output from the network with same inputs as above.

- (c) Using *relu* at the hidden layer and σ at the output layer, calculate the output from the network with same inputs as above.

Solutions: (a) (i)

$$(1, 0, 2, -3)W_1 = (1, 0, 2, -3) \begin{bmatrix} 4 & -5 & 0 & 1 \\ -3 & 6 & -1 & 2 \\ 0 & 1 & 1 & -2 \\ 2 & 0 & -3 & 4 \end{bmatrix} = (-2, -3, 11, -15)$$

$$(-2, -3, 11, -15) + \mathbf{b}_1 = (-2, -3, 11, -15) + (1, -2, 0, -1) = (-1, -5, 11, -16).$$

$$\sigma(-1, -5, 11, -16) = (0.269, 0.001, 1.000, 0.000)$$

$$(0.269, 0.001, 1.000, 0.000)W_2 = (0.269, 0.001, 1.000, 0.000) \begin{bmatrix} -2 & 1 \\ -1 & 0 \\ 1 & -3 \\ 5 & -1 \end{bmatrix} = (0.461, -2.731)$$

$$(0.461, -2.731) + \mathbf{b}_2 = (0.461, -2.731) + (-2, 2) = (-1.539, -0.731).$$

$$\sigma(-1.539, -0.731) = (0.177, 0.325).$$

(b) (i)

$$(1, 0, 2, -3)W_1 = (1, 0, 2, -3) \begin{bmatrix} 4 & -5 & 0 & 1 \\ -3 & 6 & -1 & 2 \\ 0 & 1 & 1 & -2 \\ 2 & 0 & -3 & 4 \end{bmatrix} = (-2, -3, 11, -15)$$

$$(-2, -3, 11, -15) + \mathbf{b}_1 = (-2, -3, 11, -15) + (1, -2, 0, -1) = (-1, -5, 11, -16).$$

$$\text{relu}(-1, -5, 11, -16) = (0, 0, 11, 0)$$

$$(0, 0, 11, 0)W_2 = (0, 0, 11, 0) \begin{bmatrix} -2 & 1 \\ -1 & 0 \\ 1 & -3 \\ 5 & -1 \end{bmatrix} = (11, -33)$$

$$(11, -33) + \mathbf{b}_2 = (11, -33) + (-2, 2) = (9, -31).$$

$$\text{relu}(9, -31) = (9, 0).$$

(c) (i)

$$(1, 0, 2, -3)W_1 = (1, 0, 2, -3) \begin{bmatrix} 4 & -5 & 0 & 1 \\ -3 & 6 & -1 & 2 \\ 0 & 1 & 1 & -2 \\ 2 & 0 & -3 & 4 \end{bmatrix} = (-2, -3, 11, -15)$$

$$(-2, -3, 11, -15) + \mathbf{b}_1 = (-2, -3, 11, -15) + (1, -2, 0, -1) = (-1, -5, 11, -16).$$

$$\text{relu}(-1, -5, 11, -16) = (0, 0, 11, 0)$$

$$\begin{aligned}
(0, 0, 11, 0)W_2 &= (0, 0, 11, 0) \begin{bmatrix} -2 & 1 \\ -1 & 0 \\ 1 & -3 \\ 5 & -1 \end{bmatrix} = (11, -33) \\
(11, -33) + \mathbf{b}_2 &= (11, -33) + (-2, 2) = (9, -31). \\
\sigma(9, -31) &= (1, 0).
\end{aligned}$$

- (2) Consider a network with 2 input nodes, 3 nodes in the first hidden layer, 3 nodes in the second hidden layer and 2 output nodes. The weights between the respective layers are given by the following matrices:

$$W_1 = \begin{bmatrix} 1 & -1 & 0 \\ -2 & 4 & -1 \end{bmatrix} \quad W_2 = \begin{bmatrix} 2 & 0 & -1 \\ 1 & -2 & -3 \\ 5 & 0 & 0 \end{bmatrix} \quad W_3 = \begin{bmatrix} 0 & 2 \\ -1 & -1 \\ 4 & -3 \end{bmatrix}$$

The bias values at the two hidden layers and output layer are given by:

$$\mathbf{b}_1 = (0, -3, 2), \quad \mathbf{b}_2 = (3, 3, -1), \quad \mathbf{b}_3 = (0, 1).$$

- (a) Using the *relu* activation function at each node, calculate the output from the network for the following inputs:

(i) $(1, -1)$

(ii) $(3, -2)$

- (b) Using the sigmoid activation function σ at each node, calculate the output from the network with same inputs as above.

Solutions:

- (a) (i)

$$(1, -1)W_1 = (1, -1) \begin{bmatrix} 1 & -1 & 0 \\ -2 & 4 & -1 \end{bmatrix} = (3, -5, 1)$$

$$(3, -5, 1) + \mathbf{b}_1 = (3, -5, 1) + (0, -3, 2) = (3, -8, 3)$$

$$\text{relu}(3, -8, 3) = (3, 0, 3)$$

$$(3, 0, 3)W_2 = (3, 0, 3) \begin{bmatrix} 2 & 0 & -1 \\ 1 & -2 & -3 \\ 5 & 0 & 0 \end{bmatrix} = (21, 0, -3)$$

$$(21, 0, -3) + \mathbf{b}_2 = (21, 0, -3) + (3, 3, -1) = (24, 3, 2)$$

$$\text{relu}(24, 3, 2) = (24, 3, 2)$$

$$(24, 3, 2)W_3 = (24, 3, 2) \begin{bmatrix} 0 & 2 \\ -1 & -1 \\ 4 & -3 \end{bmatrix} = (5, -9)$$

$$(5, -9) + \mathbf{b}_3 = (5, -9) + (0, 1) = (5, -8)$$

$$\text{relu}(5, -8) = (5, 0).$$

(b) (i)

$$(1, -1)W_1 = (1, -1) \begin{bmatrix} 1 & -1 & 0 \\ -2 & 4 & -1 \end{bmatrix} = (3, -5, 1)$$

$$(3, -5, 1) + \mathbf{b}_1 = (3, -5, 1) + (0, -3, 2) = (3, -8, 3)$$

$$\sigma(3, -8, 3) = (0.953, 0.000, 0.953)$$

$$(0.953, 0.000, 0.953)W_2 = (0.953, 0.000, 0.953) \begin{bmatrix} 2 & 0 & -1 \\ 1 & -2 & -3 \\ 5 & 0 & 0 \end{bmatrix} = (6.671, 0.000, -0.953)$$

$$(6.671, 0.000, -0.953) + \mathbf{b}_2 = (6.671, 0.000, -0.953) + (3, 3, -1) = (9.671, 3.000, -1.953)$$

$$\sigma(9.671, 3.000, -1.953) = (1.000, 0.953, 0.124)$$

$$(1.000, 0.953, 0.124)W_3 = (1.000, 0.953, 0.124) \begin{bmatrix} 0 & 2 \\ -1 & -1 \\ 4 & -3 \end{bmatrix} = (-0.457, 0.675)$$

$$(-0.457, 0.675) + \mathbf{b}_3 = (-0.457, 0.675) + (0, 1) = (-0.457, 1.675)$$

$$\sigma(-0.457, 1.675) = (0.388, 0.842)$$