

# MLP Exercise

COMP47590

## Question

Using the network setup in Figure 1 perform the following tasks:

1. Perform a forward propagation pass through the network
2. Calculate the loss associated with the training instance
3. Perform a backward propagation pass through the network
4. Perform a weight update within the network using a learning rate,  $\alpha$ , of 0.001

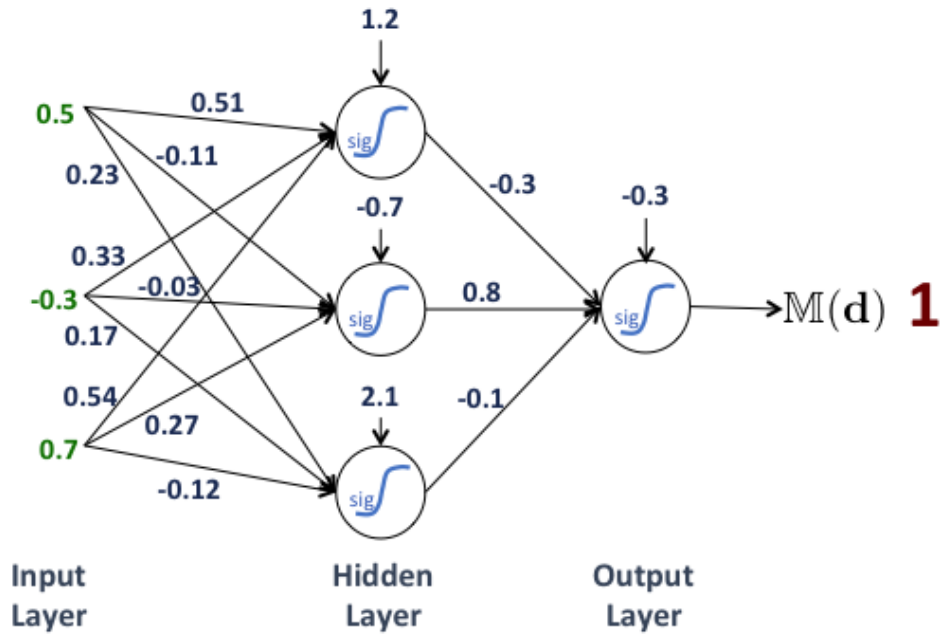


Figure 1: An MLP network and association inputs, weights, biases, and target

## Solution

### Setup Input, Weight and Bias Matrices

The network inputs:

$$\mathbf{d} = \begin{bmatrix} 0.5 \\ -0.3 \\ 0.7 \end{bmatrix} \quad (1)$$

Weights and biases for **Layer 1**:

$$\mathbf{W}^{[1]} = \begin{bmatrix} 0.51 & 0.33 & 0.54 \\ -0.11 & -0.03 & 0.27 \\ 0.23 & 0.17 & -0.12 \end{bmatrix} \quad (2)$$

$$\mathbf{b}^{[1]} = \begin{bmatrix} 1.2 \\ -0.7 \\ 2.1 \end{bmatrix} \quad (3)$$

Weights and biases for **Layer 2**:

$$\mathbf{W}^{[2]} = [-0.3 \quad 0.8 \quad -0.1] \quad (4)$$

$$\mathbf{b}^{[2]} = [-0.3] \quad (5)$$

### Forward Propagate

To perform a forward propagation for the first layer in the network, first calculate  $\mathbf{z}^{[1]}$ :

$$\mathbf{z}^{[1]} = \mathbf{W}^{[1]} \mathbf{d} + \mathbf{b}^{[1]} \quad (6)$$

$$= \begin{bmatrix} 0.51 & 0.33 & 0.54 \\ -0.11 & -0.03 & 0.27 \\ 0.23 & 0.17 & -0.12 \end{bmatrix} \begin{bmatrix} 0.5 \\ -0.3 \\ 0.7 \end{bmatrix} + \begin{bmatrix} 1.2 \\ -0.7 \\ 2.1 \end{bmatrix} \quad (7)$$

$$= \begin{bmatrix} 1.734 \\ -0.557 \\ 2.08 \end{bmatrix} \quad (8)$$

then apply the activation function, in this case a **sigmoid function**, to calculate the

activation of the nodes at **Layer 1**:

$$\mathbf{a}^{[1]} = \text{sigmoid}(\mathbf{z}^{[1]}) \quad (9)$$

$$= \text{sigmoid} \left( \begin{bmatrix} 1.734 \\ -0.557 \\ 2.08 \end{bmatrix} \right) \quad (10)$$

$$= \begin{bmatrix} 0.84992335 \\ 0.36424189 \\ 0.88894403 \end{bmatrix} \quad (11)$$

To perform a forward propagation for the second layer in the network, first calculate  $\mathbf{z}^{[2]}$ :

$$\mathbf{z}^{[2]} = \mathbf{W}^{[2]} \mathbf{a}^{[1]} + \mathbf{b}^{[2]} \quad (12)$$

$$= \begin{bmatrix} -0.3 & 0.8 & -0.1 \end{bmatrix} \begin{bmatrix} 0.84992335 \\ 0.36424189 \\ 0.88894403 \end{bmatrix} + \begin{bmatrix} -0.3 \end{bmatrix} \quad (13)$$

$$= \begin{bmatrix} -0.3524779 \end{bmatrix} \quad (14)$$

then apply the activation function, in this case a **sigmoid function**, to calculate the activation of the output nodes of the network:

$$\mathbf{a}^{[2]} = \text{sigmoid}(\mathbf{z}^{[2]}) \quad (15)$$

$$= \text{sigmoid}(\begin{bmatrix} -0.3524779 \end{bmatrix}) \quad (16)$$

$$= \begin{bmatrix} 0.41278167 \end{bmatrix} \quad (17)$$

## Calculate Loss

$$\text{loss} = - \left( t \times \log(\mathbf{a}^{[2]}) + (1 - t) \times \log(1 - \mathbf{a}^{[2]}) \right) \quad (18)$$

$$= - \left( 1 \times \log(\begin{bmatrix} 0.41278167 \end{bmatrix}) + (1 - 1) \times \log(1 - \begin{bmatrix} 0.41278167 \end{bmatrix}) \right) \quad (19)$$

$$= 0.8848364792770356 \quad (20)$$

## Backward Propagate

To perform a backward propagation at the second layer calculate,  $d\mathbf{z}^{[2]}$ :

$$d\mathbf{z}^{[2]} = \mathbf{a}^{[2]} - t \quad (21)$$

$$= \begin{bmatrix} 0.41278167 \end{bmatrix} - \begin{bmatrix} 1 \end{bmatrix} \quad (22)$$

$$= \begin{bmatrix} -0.58721833 \end{bmatrix} \quad (23)$$

and record  $d\mathbf{W}^{[2]}$  and  $d\mathbf{b}^{[2]}$ :

$$d\mathbf{W}^{[2]} = d\mathbf{z}^{[2]} \mathbf{a}^{[1]T} \quad (24)$$

$$= \begin{bmatrix} -0.58721833 \end{bmatrix} \begin{bmatrix} 0.84992335 & 0.36424189 & 0.88894403 \end{bmatrix} \quad (25)$$

$$= \begin{bmatrix} -0.49909057 & -0.21388951 & -0.52200423 \end{bmatrix} \quad (26)$$

$$(27)$$

$$d\mathbf{b}^{[2]} = d\mathbf{z}^{[2]} \quad (28)$$

$$= \begin{bmatrix} -0.58721833 \end{bmatrix} \quad (29)$$

To perform a backward propagation at the first layer calculate,  $d\mathbf{z}^{[1]}$ :

$$d\mathbf{z}^{[1]} = \mathbf{W}^{[2]T} d\mathbf{z}^{[2]} * g^{[1]'}(\mathbf{z}^{[1]}) \quad (30)$$

$$= \begin{bmatrix} -0.3 \\ 0.8 \\ -0.1 \end{bmatrix} \begin{bmatrix} -0.58721833 \end{bmatrix} * \text{sigmoid}' \left( \begin{bmatrix} 1.734 \\ -0.557 \\ 2.08 \end{bmatrix} \right) \quad (31)$$

$$= \begin{bmatrix} 0.1761655 \\ -0.46977467 \\ 0.05872183 \end{bmatrix} * \begin{bmatrix} 0.12755365 \\ 0.23156973 \\ 0.09872254 \end{bmatrix} \quad (32)$$

$$= \begin{bmatrix} 0.02247055 \\ -0.10878559 \\ 0.00579717 \end{bmatrix} \quad (33)$$

and record  $d\mathbf{W}^{[1]}$  and  $d\mathbf{b}^{[1]}$ :

$$d\mathbf{W}^{[1]} = d\mathbf{z}^{[1]} \mathbf{d}^T \quad (34)$$

$$= \begin{bmatrix} 0.02247055 \\ -0.10878559 \\ 0.00579717 \end{bmatrix} \begin{bmatrix} 0.5 & -0.3 & 0.7 \end{bmatrix} \quad (35)$$

$$= \begin{bmatrix} 0.01123528 & -0.00674117 & 0.01572939 \\ -0.0543928 & 0.03263568 & -0.07614992 \\ 0.00289858 & -0.00173915 & 0.00405802 \end{bmatrix} \quad (36)$$

$$(37)$$

$$d\mathbf{b}^{[1]} = d\mathbf{z}^{[1]} \quad (38)$$

$$= \begin{bmatrix} 0.02247055 \\ -0.10878559 \\ 0.00579717 \end{bmatrix} \quad (39)$$

## Update Weights

Update the weight and bias terms in the first layer:

$$\mathbf{W}^{[1]} = \mathbf{W}^{[1]} - \alpha d\mathbf{W}^{[1]} \quad (40)$$

$$= \begin{bmatrix} 0.51 & 0.33 & 0.54 \\ -0.11 & -0.03 & 0.27 \\ 0.23 & 0.17 & -0.12 \end{bmatrix} - 0.001 \times \begin{bmatrix} 0.01123528 & -0.00674117 & 0.01572939 \\ -0.0543928 & 0.03263568 & -0.07614992 \\ 0.00289858 & -0.00173915 & 0.00405802 \end{bmatrix} \quad (41)$$

$$= \begin{bmatrix} 0.50987641 & 0.33007415 & 0.53982698 \\ -0.10940168 & -0.03035899 & 0.27083765 \\ 0.22996812 & 0.17001913 & -0.12004464 \end{bmatrix} \quad (42)$$

$$\mathbf{b}^{[1]} = \mathbf{b}^{[1]} - \alpha d\mathbf{b}^{[1]} \quad (43)$$

$$= \begin{bmatrix} 1.2 \\ -0.7 \\ 2.1 \end{bmatrix} - 0.001 \times \begin{bmatrix} 0.02247055 \\ -0.10878559 \\ 0.00579717 \end{bmatrix} \quad (44)$$

$$= \begin{bmatrix} 1.19975282 \\ -0.69880336 \\ 2.09993623 \end{bmatrix} \quad (45)$$

Update the weight and bias terms in the second layer:

$$\mathbf{W}^{[2]} = \mathbf{W}^{[2]} - \alpha d\mathbf{W}^{[2]} \quad (46)$$

$$= \begin{bmatrix} -0.3 & 0.8 & -0.1 \end{bmatrix} - 0.001 \times \begin{bmatrix} -0.49909057 & -0.21388951 & -0.52200423 \end{bmatrix} \quad (47)$$

$$= \begin{bmatrix} -0.29451 & 0.80235278 & -0.09425795 \end{bmatrix} \quad (48)$$

$$\mathbf{b}^{[2]} = \mathbf{b}^{[2]} - \alpha d\mathbf{b}^{[2]} \quad (49)$$

$$= \begin{bmatrix} -0.3 \end{bmatrix} - 0.001 \times \begin{bmatrix} -0.58721833 \end{bmatrix} \quad (50)$$

$$= \begin{bmatrix} -0.2935406 \end{bmatrix} \quad (51)$$

## Check Impact of Updating Weights

Performing a forward pass with these new weight values gives:

$$\mathbf{a}^{[2]} = \begin{bmatrix} 0.4166458 \end{bmatrix} \quad (52)$$

from which we can calculate a new loss:

$$\text{loss} = - \left( t \times \log \left( \mathbf{a}^{[2]} \right) + (1 - t) \times \log \left( 1 - \mathbf{a}^{[2]} \right) \right) \quad (53)$$

$$= - \left( 1 \times \log \left( [0.4166458] \right) + (1 - 1) \times \log \left( 1 - [0.4166458] \right) \right) \quad (54)$$

$$= 0.8755188083022127 \quad (55)$$

which is ever so slightly less than the previous value of 0.8848364792770356 so our back-propagation and weight update step worked!