

# M8-L1-P1

November 4, 2023

## 1 M8-L1 Problem 1

In this problem you will solve for  $\frac{\partial L}{\partial W_2}$  and  $\frac{\partial L}{\partial W_1}$  for a neural network with two input features, a hidden layer with 3 nodes, and a single output. You will use the sigmoid activation function on the hidden layer. You are provided an input sample  $x_0$ , the current weights  $W_1$  and  $W_2$ , and the ground truth value for the sample,  $t = -2$

$$L = \frac{1}{2}e^T e$$

```
[ ]: import numpy as np

x0 = np.array([[ -2], [ -6]])

W1 = np.array([[ -2,  1], [ 3,  8], [-12, 7]])
W2 = np.array([[ -11,  2,  5]])

t = np.array([[ -2]])
```

### 1.1 Define activation function and its derivative

First define functions for the sigmoid activation functions, as well as its derivative:

```
[ ]: def sigmoid(x):
      return 1 / (1 + np.exp(-x))

def del_sigmoid(x):
    s = sigmoid(x)
    return s*(1-s)
```

## 2 Forward propagation

Using your activation function, compute the output of the network  $y$  using the sample  $x_0$  and the provided weights  $W_1$  and  $W_2$

```
[ ]: a1 = W1 @ x0
     x1 = sigmoid(a1)
     a2 = W2 @ x1
     y = a2
```

```
print(y)
```

```
[[-1.31123207]]
```

## 2.1 Backpropagation

Using your calculated value of  $y$ , the provided value of  $t$ , your  $\sigma$  and  $\sigma'$  function, and the provided weights  $W_1$  and  $W_2$ , compute the gradients  $\frac{\partial L}{\partial W_2}$  and  $\frac{\partial L}{\partial W_1}$ .

```
[ ]: e = t - y
      L = 0.5*(e.T @ e)

      delta_2 = -e
      dLdw2 = delta_2 * x1
      delta_1 = delta_2 @ W2 @ del_sigmoid(a1)
      dLdw1 = delta_1 * x0

      print(dLdw2)
      print(dLdw1)
```

```
[[8.21031503e-02]
 [2.43316128e-24]
 [1.04899215e-08]]
[[1.59095662]
 [4.77286987]]
```