1. The model is built as below:

2. Using the randomly drawn X1 & X2 with value as blew:

```
array([0.97436534, 0.31171513])
```

The loss value is:

0.6471283

3. Get the grads of the model:

```
[tensor([-0.0016, -0.0005, -0.0013, -0.0004, 0.0217, 0.0069, -0.0148, -0.0047, 0.0033, 0.0011, 0.0012, 0.0004]), tensor([-0.0017, -0.0014, 0.0074, -0.0131,
                                                                                                                                                                                                     0.0023, -0.0128, -0.0041, 0.0024, -0.0180, -0.0058,
                                                                                                                                                                      0.0073,
                                                                                                                                                                      0.0075,
    tensor([-0.0017, -0.0014, 0.0034, 0.0013]),
                                                                                                                                                                      0.0223, -0.0152,
                                                                                                                                                                                                                                      0.0077, -0.0185,
                                    0.0025, 0.0053, 0.0053,
0.0046, 0.0028, 0.0028,
     tensor([ 0.0025,
                                                                                                                                   0.0044,
                                                                                                                                                                  0.0034.
                                                                                                                                                                                                     0.0037.
                                                                                                                                                                                                                                      0.0050.
                                                                                                                                                                                                                                                                       0.0029.
                                                                                                                                    0.0057,
                                                                                                                                                                                                     0.0048,
                                                                                                                                                                   0.0058,
                                                                                                                                                                                                                                       0.0037,
                                                                                                                                                                                                                                                                       0.0040.
                                 0.0046, 0.0028, 0.0028, 0.0057, 0.0058, 0.0048, 0.0037, 0.0040, 0.0054, 0.0031, 0.0059, 0.0031, 0.0065, 0.0135, 0.0136, 0.0113, 0.0087, 0.0095, 0.0128, 0.0073, 0.0119, 0.0073, -0.0174, -0.0363, -0.0365, -0.0303, -0.0234, -0.0254, -0.0342, -0.0197, -0.0318, -0.0195, 0.0389, 0.0812, 0.0817, 0.0678, 0.0523, 0.0568, 0.0767, 0.0441, 0.0712, 0.0436, -0.0284, -0.0592, -0.0597, -0.0495, -0.0382, -0.0415, -0.0560, -0.0322, -0.0520, -0.0319, 0.0077, 0.0160, 0.0161, 0.0133, 0.0103, 0.0112, 0.0151, 0.0087, 0.0140, 0.0086, -0.0427, -0.0889, -0.0895, -0.0743, -0.0573, -0.0622, -0.0840, -0.0483, -0.0779, -0.0478, 0.0405, 0.0843, 0.0850, 0.0705, 0.0544, 0.0590, 0.0797, 0.0458
    -0.0895, -0.0743, -0.0573, -0.0622, -0.0840, -0.0483, -0.0779, -0.0478, 0.0405, 0.0843, 0.0850, 0.0705, 0.0544, 0.0590, 0.0797, 0.0458, 0.0739, 0.0454, -0.0064, -0.0133, -0.0134, -0.0111, -0.0086, -0.0093, -0.0126, -0.0072, -0.0117, -0.0071]), tensor([ 0.0073,  0.0079,  0.0187, -0.0500,  0.1119, -0.0817,  0.0220, -0.1226,  0.1163, -0.0183]), tensor([-0.7703, -0.6855, -0.7232, -0.9646, -0.8631, -0.7418, -1.0050, -0.7975, -0.8975, -0.9515]),
    -0.8975, -0
tensor([-1.6089])]
```

4. Build the forward propagation in scratch:

```
z1 = np.dot(W1, X) + b1
z1_act = my_sigmoid(z1)
z2 = np.dot(W2, z1_act) + b2
z2_act = my_sigmoid(z2)
z3 = np.dot(W3, z2_act) + b3
z3
```

The predicted value here is:

```
array([-0.28116576])
```

While the predicted value by original model is:

```
array([-0.28116578], dtype=float32)
```

5. Build the back propagation in scratch:

```
[123] Þ ►≣ M↓
          def my_loss(y_true,y_pred):
               return np.power(y_true-y_pred,2)
         def loss_gradient(y_true,y_pred):
    return 2*y_pred-2*y_true
         def sigmoid_gradient(Z):
                  urn my_sigmoid(Z) * (1-my_sigmoid(Z))
          delta3=loss_gradient(Y,prediction.data.numpy())
         b3 gradient=delta3
         w3_gradient=np.dot(delta3, z2_act.transpose().reshape((1,10)))
         w3 gradient
     array([-0.77028648, -0.68547447, -0.72318788, -0.96456754, -0.86306459, -0.74180506, -1.00498997, -0.79745549, -0.89748667, -0.95148059])
          delta2=np.dot(W3.transpose(), delta3) * sigmoid_gradient(z2)
         b2_gradient=delta2
          w2_gradient=np.dot(delta2.reshape((10,1)), z1_act.transpose().reshape(1, 10))
          delta1=np.dot(W2.transpose(),delta2)* sigmoid_gradient(z1)
         b1_gradient=delta1
         w1_gradient=np.dot(delta1.reshape((10, 1)), X.transpose().reshape((1, 2)))
```

The calculated gradients: