# Lab1: back-propagation

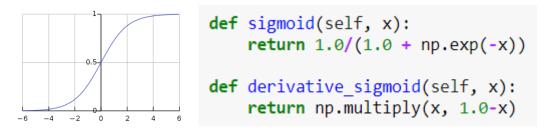
0516069 翁英傑

## 1. Introduction

In the training stage of a neural network, it relies on a technic called gradient decent. While calculating the gradient of the loss regarding to each weight, using back propagation will ease the computation, which the gradient is calculate from the output layer to the input layer.

# 2. Experiment setups

## A. Sigmoid functions



Sigmoid function is used for making each layer none linear so that it's meaningful for to create a multi-hidden layer network.

### B. Neural network

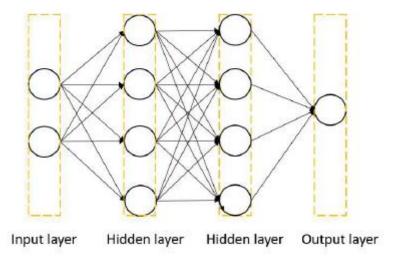
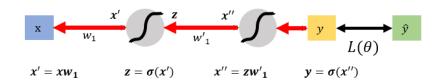


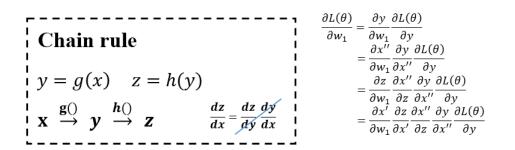
Figure 1. Two layers neural network

A picture of a network with two hidden layers, which each layer contains

four nodes.

## C. Backpropagation





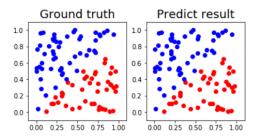
With the chain rule, if we calculate the gradient beginning from the output layer, the previous gradient will be used in the computation of the next gradient.

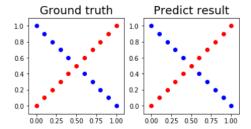
# 3. Results of your testing

#### 1. Linear distribution

#### 2. XOR distribution

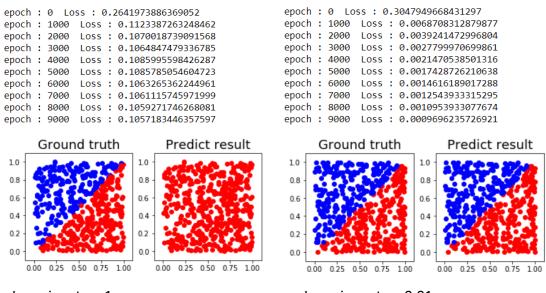
```
epoch: 0 Loss: 0.3014694984641999
                                       epoch: 0 Loss: 0.2559653327914768
epoch : 5000 Loss : 0.0000117492469418
                                       epoch: 5000 Loss: 0.0000295712687178
epoch : 10000 Loss : 0.0000049731729480
                                       epoch: 10000 Loss: 0.0000133654181554
epoch: 15000 Loss: 0.0000030559638145
                                       epoch: 15000 Loss: 0.0000085201522618
epoch: 20000 Loss: 0.0000021752148525
                                       epoch: 20000 Loss: 0.0000062179443973
epoch: 25000 Loss: 0.0000016755193740
                                       epoch: 25000 Loss: 0.0000048801176589
epoch: 30000 Loss: 0.0000013558388558
                                       epoch: 30000 Loss: 0.0000040083178597
epoch: 35000 Loss: 0.0000011347598883
                                       epoch: 35000 Loss: 0.0000033963572265
epoch: 40000 Loss: 0.0000009732759327
                                       epoch: 40000 Loss: 0.0000029437485499
epoch: 45000 Loss: 0.0000008504412825
                                       epoch: 45000 Loss: 0.0000025957645586
epoch : 50000
             Loss: 0.0000007540372346
                                       epoch: 50000 Loss: 0.0000023200935687
epoch :
       55000 Loss: 0.0000006764719539
                                       epoch: 55000 Loss: 0.0000020964482907
epoch : 60000 Loss : 0.0000006127880172
                                       epoch : 60000
                                                     Loss: 0.0000019114583089
epoch: 65000 Loss: 0.0000005596153966
                                       epoch: 65000 Loss: 0.0000017559591041
epoch : 70000 Loss : 0.0000005145857451
                                       epoch: 70000 Loss: 0.0000016234619090
epoch: 75000 Loss: 0.0000004759875116
                                       epoch : 75000 Loss : 0.0000015092454498
epoch: 80000 Loss: 0.0000004425539830
                                       epoch: 80000 Loss: 0.0000014097936613
epoch: 85000 Loss: 0.0000004133282027
                                       epoch: 85000 Loss: 0.0000013224349384
epoch: 90000 Loss: 0.0000003875741555 epoch: 90000 Loss: 0.0000012451034772
epoch : 95000 Loss : 0.0000003647167637 epoch : 95000 Loss : 0.0000011761771174
```





## 4. Discussion

A common tip in training is to add a learning rate to avoid the overshooting problem. In the basic setting of this LAB, which generates 100 data samples by default setting, there is no problem with gradient decent. But if we generate over 200 data samples, the gradient will be too large that it crosses the lowest point and the loss does not decrease. Adding a learning rate solves the problem.



Learning rate = 1 Learning rate = 0.01