### 1: The First Problem

### (a) Data1:

For the intertwined two spiral, using the standard bp algorithm can solve the problem very well. In addition, the data set is not large and the data structure is not complicated. Therefore, the main adjustment directions are  $learn_rate$  and epoch times. On the contrary, excessive adjustment of the hidden layer structure will be counterproductive.

Here is my progress report of data1:

```
Enter data filename: 1-SpiralData1
TrainNet2: IP:2 H1:5 OP:1
Params: lrn_rat:0.4 Order:0
Testing mlp:
    MinErr
               AveErr
                           MaxErr
                                      %Correct
    0.331593
               0.492433
                           0.659922
                                     46.354164
End of program.
 TrainNet3: IP:2 H1:12 H2:6 OP:1
 Params: lrn rat:0.01 Order:1
 Testing mlp:
     MinErr
                AveErr
                            MaxErr
                                       %Correct
     0.474483
                0.499983
                            0.525415
                                      50.000000
 End of program.
 TrainNet3: IP:2 H1:12 H2:6 OP:1
 Params: lrn_rat:0.01 Order:2
 Testing mlp:
    MinErr
                AveErr
                           MaxErr
                                       %Correct
     0.000000
                0.485068
                           1.000000
                                      48.437500
 End of program.
 TrainNet4: IP:2 H1:12 H2:8 H3:12 OP:1
 Params: lrn_rat:0.01 Order:3
 Testing mlp:
    MinErr
                AveErr
                           MaxErr
                                       %Correct
     0.000000
                0.488661
                           1.000000
                                      48.958336
 End of program.
```

It is obvious that using a single layer neural network and setting  $learn_{rate}$  to 0.4 can achieve the best experimental results.

#### (b) Data2:

For the abalone age problem, it has more features. So It may be a better way to combine the wrong samples with order3 and multi-layer neural network. Different models may have different performance, but multi-layer neural network should achieve good results.

Here is my progress report of data2: TrainNet4: IP:8 H1:15 H2:9 H3:6 OP:1 Params: lrn\_rat:0.01 Order:0 Testing mlp: MinErr AveErr MaxErr **%Correct** 0.016670 0.109487 0.561774 48.400002 End of program. TrainNet3: IP:8 H1:12 H2:8 OP:1 Params: lrn\_rat:0.4 Order:0 Testing mlp: MinErr AveErr MaxErr %Correct 0.000136 0.265219 0.680302 27.400000 End of program. TrainNet4: IP:44 H1:12 H2:14 H3:8 OP:1 Params: lrn\_rat:0.6 Order:3 Testing mlp: MinErr AveErr MaxErr %Correct 0.265774 0.000096 0.683462 28,200001

It can be seen that Order3 with a three-layer structure and Order0 with a two-layer structure have achieved good results. The advantages and disadvantages of the two need to be further explored.

## (c) Data3:

End of program.

For the SPECT Heart Diagnosis problem, considering that it has very input values, too many neural networks may affect the calculation, so try to adjust the rate and single layer structure. Here is my progress report of data3:

```
TrainNet3: IP:44 H1:33 H2:14 OP:1
Params: lrn rat:0.6 Order:0
Testing mlp:
   MinErr
               AveErr
                          MaxErr
                                     %Correct
   0.245555
              0.550889
                          0.754445
                                    60.000004
End of program.
Enter data filename: data3.txt
TrainNet2: IP:44 H1:33 OP:1
Params: lrn_rat:0.01 Order:1
Testing mlp:
   MinErr
              AveErr
                          MaxErr
                                     %Correct
   0.336831
              0.467771
                          0.650450
                                    36.000000
End of program.
```

It can be seen that a single-layer neural network achieves better results than a multi-layer.

### 2: The Second Problem

## (a) Data1:

To be able to learn classification correctly the experiment shows that the mlp that is most suitable for problem 1 is a structure of 5 neurons in a single layer, and the learning rate is 0.4 and a suitable number of iterations is 10000 at the same time, each setting can see in Figure 1:

### (b) Data2:

To be able to learn classification correctly, by comparison, the mlp that is most suitable for problem 2 is a two-layer neural network with nodes 12 and 8, and a suitable number of iterations is 8000, each set can see in Figure 2 and Figure 3:

# (c) Data3:

To be able to learn classification correctly, the experiment shows that the MLP that is most suitable for Problem 3 is a single-layer hidden layer but contains a large number of nodes. The appropriate number of iterations is 6000, and each setting can see in Figure 4:

NumIPs: 2 NumOPs: 1

NumTrnPats: 192
NumTstPats: 192

NumIts: 10000

NumHN: 1
NumHN1: 5

LrnRate: 0.4

Mtm1: 1.2 Mtm2: 0.4

ObjErr: 0.005

Ordering: 0

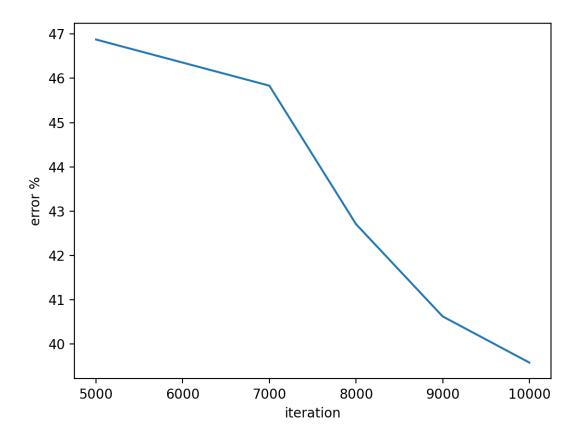


Figure 1: d1 parameters and error vs iteration

NumIPs: 8
NumOPs: 1

NumTrnPats: 3677 NumTstPats: 500

NumIts: 10000

NumHN: 3 NumHN1: 12 NumHN2: 8 NumHN3: 12

LrnRate: 0.4
Mtml: 1.2

Mtm2: 0.4

ObjErr: 0.005

Ordering: 0

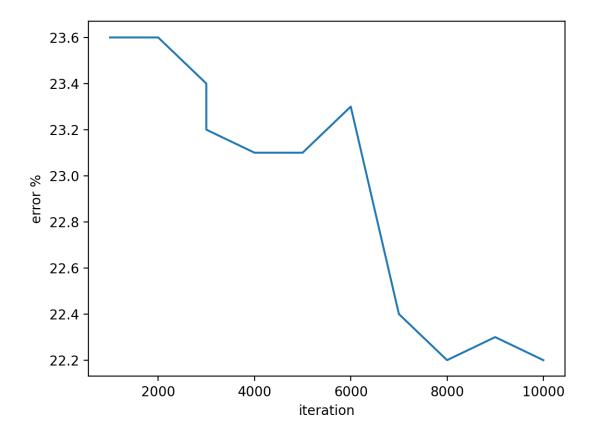


Figure 2: d2.1 parameters and error vs iteration

NumIPs: 8 NumOPs: 1

NumTrnPats: 3677 NumTstPats: 500

NumIts: 10000

NumHN: 2
NumHN1: 12
NumHN2: 8

LrnRate: 0.4

Mtm1: 1.2 Mtm2: 0.4

ObjErr: 0.005

Ordering: 0

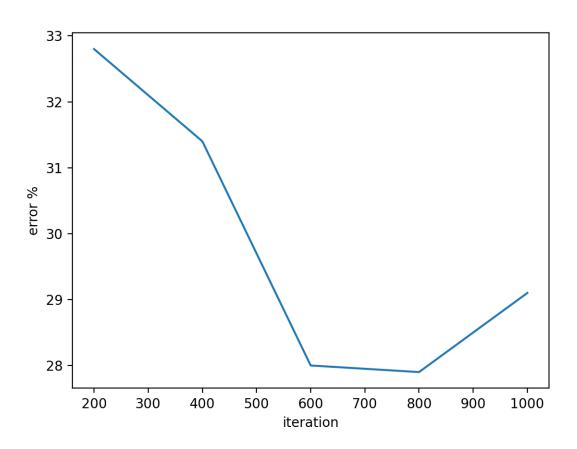


Figure 3: d2.2 parameters and error vs iteration

NumIPs: 44
NumOPs: 1
NumTrnPats: 299
NumTstPats: 50
NumIts: 6000
NumHN: 1
NumHN: 33
LrnRate: 0.01
Mtm1: 1.5
Mtm2: 0.4
ObjErr: 0.005
Ordering: 1

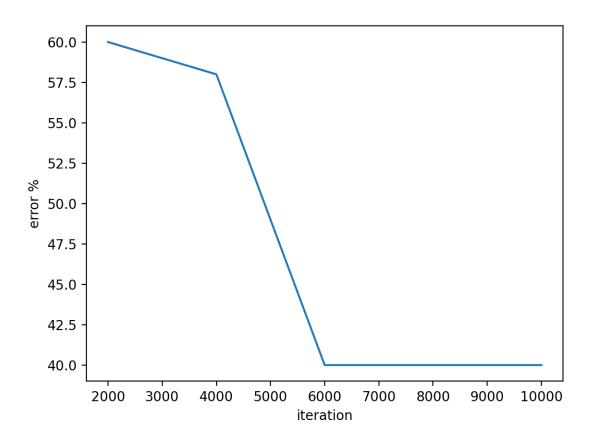


Figure 4: d3 parameters and error vs iteration