## <u>Data Mining – Programming Assignment 2</u> Magesh Rajasekaran - 101676478

#### **First Question:**

SVM Classifier with Class Labels 1 and 2:

```
Code: (PA2_DM)
%# Initializing labels
Labels (1:5000) = 0;
Labels (5001:10000) =1;
%# Reading Input
dataInput=load('data.txt');
%# Kfold validation
countForKFold=10;
crossValidationValue = crossvalind('Kfold', Labels, k);
classValue = classperf(Labels);
%# SVM Training with Polynomial Kernel Function of degree 3
for i = 1:countForKFold
      testIndiciesValue = (crossValidationValue == i);
      trainIndiciesValue = ~testIndiciesValue;
      svmModel = fitcsvm( dataInput(trainIndiciesValue,:),
      Labels (trainIndiciesValue), 'BoxConstraint', 2e-1,
      'KernelFunction', 'polynomial', 'PolynomialOrder', 3);
      pred = predict(svmModel, dataInput(testIndiciesValue,:));
      classValue = classperf(classValue, pred, testIndiciesValue);
end
%# accuracy
classValue.CorrectRate
%# confusion matrix
classValue.CountingMatrix
```

Parameter: Kernel Function used was Polynomial with the degree 3

#### Result:

```
>> PA2_DM
```

Accuracy =

0.9379

Confusion Matrix =

4482 103 518 4897 0 0

## Neural Network with Class 1 and 2:

## Code:

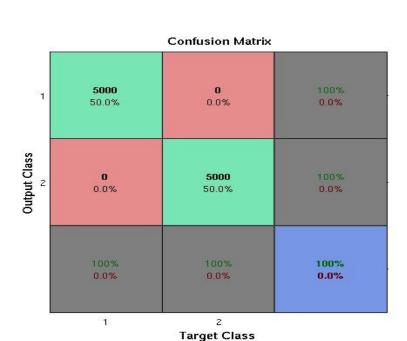
```
dataInput= load('data.txt');
%labelling the the rows as per the given problem statement.
Label (1:5000,1)=0;
Label(5001:10000,1)=1;
x = dataInput';
t = Label';
%cross validation by cvpartition
crossValidationPartition=cvpartition(10000,'kfold',10);
    indexValue(:,i) = test(crossValidationPartition,i);
end
indexValue=indexValue';
% K folding of Train and Test indicies
for Kvalue=1:10
    count1=1;
    count2=1;
    rows=1;
    while (rows <=10000)</pre>
        if (indexValue (Kvalue, rows) == 0)
            trainIndicies (Kvalue, count1) = rows;
            count1=count1+1;
        else
            testIndicies (Kvalue, count2) = rows;
            count2=count2+1;
        end
        rows=rows+1;
    end
 count1=1;
 count2=1;
end
% Create a Pattern Recognition Network
hiddenLayerSize = 10;
net = patternnet(hiddenLayerSize);
% Choose Input and Output Pre/Post-Processing Functions
% For a list of all processing functions type: help nnprocess
net.input.processFcns = {'removeconstantrows', 'mapminmax'};
net.output.processFcns = {'removeconstantrows', 'mapminmax'};
% Setup Division of Data for Training, Validation, Testing
for l=1:10
    net.divideFcn = 'divideind'; % Divide data randomly
    net.divideMode = 'sample'; % Divide up every sample
    net.divideParam.trainInd =trainIndicies(1,1:9000);
```

```
net.divideParam.testInd =testIndicies(1,:);
    %net.divideParam.valRatio = 5/100;
    % Choose a Performance Function
    net.performFcn = 'crossentropy'; % Cross-Entropy
    % Choose Plot Functions
    net.plotFcns = {'plotperform','plottrainstate','ploterrhist', ...
        'plotconfusion', 'plotroc'};
    % Train the Network
    [net, tr] = train(net, x, t);
    % Test the Network
    y = net(x);
    e = gsubtract(t, y);
    performance = perform(net,t,y)
    tind = vec2ind(t);
    yind = vec2ind(y);
    percentErrors = sum(tind ~= yind)/numel(tind);
    performance = perform(net, t,y);
    % Recalculate Training, Validation and Test Performance
    trainTargets = t .* tr.trainMask{1};
    %valTargets = t .* tr.valMask{1};
    testTargets = t .* tr.testMask{1};
    trainPerformance = perform(net,trainTargets,y)
    %valPerformance = perform(net,valTargets,y)
    testPerformance = perform(net,testTargets,y)
    [c,cm,ind,per] = confusion(t,y);
    accuracyFinal(1,1) = ((1-c)*100);
    % View the Network
    %view(net)
    figure, plotconfusion(t,y)
end
Parameter: Hidden Neurons – 10
```

Result:

Accuracy = 1.000

**Confusion Matrix Plot:** 



## **Second Question:**

#### **SVM Classifier:**

```
Code: (PA2 DM2 SVM)
%# Initializing labels
Labels=ones(10000,1);
Labels([1:500,1001:1500,2001:2500,3001:3500,4001:4500,5001:5500,6001:6500,7001:7500
,8001:8500,9001:9500])=2;
%# Reading Input
dataInput=load('data.txt');
%# Kfold validation
countForKFold=10;
crossValidationValue = crossvalind('Kfold', Labels, k);
classValue = classperf(Labels);
%# SVM Training with Polynomial Kernel Function of degree 3
for i = 1:countForKFold
      testIndiciesValue = (crossValidationValue == i);
      trainIndiciesValue = ~testIndiciesValue;
      svmModel = fitcsvm( dataInput(trainIndiciesValue,:),
      Labels (trainIndiciesValue), 'BoxConstraint', 2e-1,
      'KernelFunction', 'polynomial', 'PolynomialOrder', 3);
      pred = predict(svmModel, dataInput(testIndiciesValue,:));
      classValue = classperf(classValue, pred, testIndiciesValue);
end
%# accuracy
classValue.CorrectRate
%# confusion matrix
classValue.CountingMatrix
Result:
>> PA2 DM2 SVM
Accuracy =
  0.5018
Confusion Matrix =
    2582
            2564
    2418
            2436
     0
            0
Parameter: Kernel Function used was Polynomial with the degree 3
```

Neural Network:

#### Code:

```
%labelling the the rows as per the given problem statement
Label=zeros(10000,1);
Label([1:500,1001:1500,2001:2500,3001:3500,4001:4500,5001:5500,6001:6500,7001:7500,
8001:8500,9001:95001)=1;
Label=ones (10000, 2);
Label([1:500,1001:1500,2001:2500,3001:3500,4001:4500,5001:5500,6001:6500,7001:7500,
8001:8500,9001:9500])=0;
dataInput= load('data.txt');
x = dataInput';
t = Label';
%cross validation by cvpartition
crossValidationPartition=cvpartition(10000,'kfold',10);
for i=1:10
    indexValue(:,i) = test(crossValidationPartition,i);
end
indexValue=indexValue';
% K folding of Train and Test indicies
for Kvalue=1:10
    count1=1;
    count2=1;
    rows=1;
    while (rows <=10000)</pre>
        if (indexValue (Kvalue, rows) == 0)
            trainIndicies (Kvalue, count1) = rows;
            count1=count1+1;
        else
            testIndicies (Kvalue, count2) = rows;
            count2=count2+1;
        end
        rows=rows+1;
    end
 count1=1;
 count2=1;
end
% Create a Pattern Recognition Network
hiddenLayerSize = 10;
net = patternnet(hiddenLayerSize);
% Choose Input and Output Pre/Post-Processing Functions
% For a list of all processing functions type: help nnprocess
net.input.processFcns = {'removeconstantrows','mapminmax'};
net.output.processFcns = {'removeconstantrows', 'mapminmax'};
% Setup Division of Data for Training, Validation, Testing
for l=1:10
    net.divideFcn = 'divideind'; % Divide data randomly
    net.divideMode = 'sample'; % Divide up every sample
    net.divideParam.trainInd =trainIndicies(1,1:9000);
    net.divideParam.testInd =testIndicies(1,:);
    %net.divideParam.valRatio = 5/100;
```

```
% Choose a Performance Function
    net.performFcn = 'crossentropy'; % Cross-Entropy
    % Choose Plot Functions
    net.plotFcns = {'plotperform','plottrainstate','ploterrhist', ...
        'plotconfusion', 'plotroc'};
    % Train the Network
    [net, tr] = train(net, x, t);
    % Test the Network
    y = net(x);
    e = gsubtract(t, y);
    performance = perform(net,t,y)
    tind = vec2ind(t);
    yind = vec2ind(y);
    percentErrors = sum(tind ~= yind)/numel(tind);
    performance = perform(net, t,y);
    % Recalculate Training, Validation and Test Performance
    trainTargets = t .* tr.trainMask{1};
    %valTargets = t .* tr.valMask{1};
    testTargets = t .* tr.testMask{1};
    trainPerformance = perform(net,trainTargets,y)
    %valPerformance = perform(net, valTargets, y)
    testPerformance = perform(net,testTargets,y)
    [c,cm,ind,per] = confusion(t,y);
    accuracyFinal(1,1) = ((1-c)*100);
    % View the Network
    %view(net)
    figure, plotconfusion(t,y)
end
```

Parameter: Hidden Neurons – 10

#### Result:

Accuracy: 53.4%

**Confusion Matrix:** 

# 

2

**Target Class** 

1

Confusion Matrix

## Reference:

- [1] Matlab Neural Network Pattern Recognition tool for Neural Network Problem <a href="http://www.mathworks.com/products/neural-network/features.html#data-fitting%2C-clustering%2C-and-pattern-recognition">http://www.mathworks.com/products/neural-network/features.html#data-fitting%2C-clustering%2C-and-pattern-recognition</a>.
- [2] Fitcsvm method <a href="http://www.mathworks.com/help/stats/fitcsvm.html">http://www.mathworks.com/help/stats/fitcsvm.html</a>
- [3] <a href="http://www.robots.ox.ac.uk/~az/lectures/ml/lect2.pdf">http://www.robots.ox.ac.uk/~az/lectures/ml/lect2.pdf</a>
- [4] <a href="http://www.svms.org/parameters/">http://www.svms.org/parameters/</a>