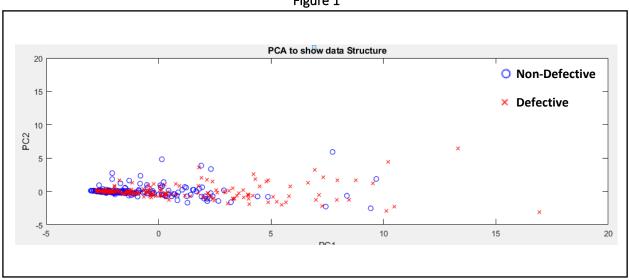
## 6COM1044-0105

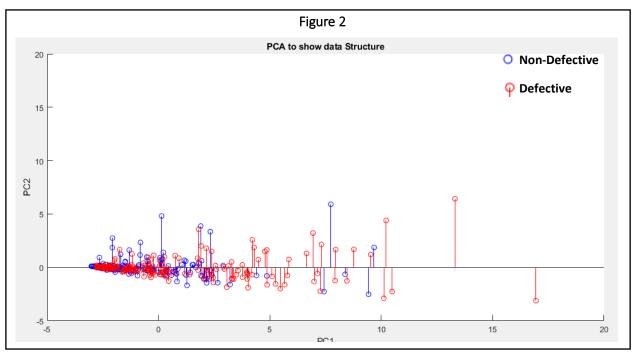
# Machine Learning and Neural Computing: Data Classification Coursework

Task 1 – Data Exploration

Code can be found in Appendix







## Critical Analysis

In Figure 1, we can observe that a big majority of the data of both classes is concentrated in a single area. Subsequently this would make it difficult for a trained model to predict and may not be always accurate. Further, it appears that the defective data is further spread out in comparison to the non-defective data and also holds both extremes for the X and Y axis which would suggest a higher variance in the defective data.

## Task 2 - Normalisation

To normalise (scale) the data, I used the function sym-scale included in LibSVM as seen below in Figure 3:

```
Figure 3

C:\Docs\libsvm-3.23\windows>svm-scale -l 0 -u 1 -s range1 trainingSet.dat > NormalisedTrainingSet

C:\Docs\libsvm-3.23\windows>svm-scale -r range1 testingSet.dat > NormalisedTestingSet
```

## Task 3 - Cross Validation

## **Given parameters**

 $C = 5 G = 0.01 \Rightarrow Cross\ Validation\ Accuracy = 62.6374\%$ 

```
C:\Docs\libsvm-3.23\windows>svm-train -c 5 -g 0.01 -v 5 normalisedtrainingDATA.data

potimization finished, #iter = 148

nu = 0.914131
obj = -1271.290936, rho = 0.221471
nSV = 268, rBSV = 266
Total nSV = 268

potimization finished, #iter = 142

nu = 0.894147
obj = -1246.397824, rho = 0.131891
nSV = 262, rBSV = 260
Total nSV = 262

potimization finished, #iter = 147

nu = 0.931607
obj = -1289.788835, rho = 0.363748
nSV = 272, rBSV = 269
Total nSV = 272

potimization finished, #iter = 139
nu = 0.902791
obj = -1243.683663, rho = 0.256575
nSV = 264, rBSV = 262
Total nSV = 264

potimization finished, #iter = 143
nu = 0.909895
obj = -1255.701502, rho = 0.306781
nSV = 264, rBSV = 262
Total nSV = 264
Cross Validation Accuracy = 62.6374%
```

#### $C = 5 G = 10 \Rightarrow Cross\ Validation\ Accuracy = 64.011\%$

```
C:\Docs\libsvm-3.23\windows>svm-train -c 5 -g 10 -v 5 normalisedtrainingDATA.data
.*.*

optimization finished, #iter = 697

nu = 0.627169

obj = -797.102927, rho = -0.502085

nSV = 238, rBSV = 154

Total nSV = 238
.*.*

optimization finished, #iter = 708

nu = 0.627562

obj = -798.419847, rho = -0.548758

nSV = 244, rBSV = 151

Total nSV = 244
.*.*

optimization finished, #iter = 664

nu = 0.638513

obj = -815.533121, rho = -0.395133

nSV = 238, rBSV = 151

Total nSV = 238
.*.*

optimization finished, #iter = 708

nu = 0.587439

obj = -748.978662, rho = -0.558161

nSV = 226
.*.*

optimization finished, #iter = 821

nu = 0.634073

obj = -810.702631, rho = -0.499388

nSV = 241, rBSV = 151

Total nSV = 241

Cross Validation Accuracy = 64.011%
```

#### C = 5 G = 2048 => Cross Validation Accuracy = 60.1648%

```
C:\Docs\libsvm-3.23\windows>svm-train -c 5 -g 2048 -v 5 normalisedtrainingDATA.data
.*

optimization finished, #iter = 567

nu = 0.205648

obj = -154.779131, rho = -0.054791

nSV = 287, rBSV = 3

Total nSV = 287
.**

optimization finished, #iter = 674

nu = 0.207774

obj = -156.201291, rho = -0.068125

nSV = 290, rBSV = 3

Total nSV = 290
.**

optimization finished, #iter = 618

nu = 0.207939

obj = -157.638962, rho = -0.055755

nSV = 288, rBSV = 3

Total nSV = 288
.*.*

optimization finished, #iter = 639

nu = 0.205324

obj = -152.914148, rho = -0.074548

nSV = 289, rBSV = 2

Total nSV = 289
.*.*

optimization finished, #iter = 603

nu = 0.211167

obj = -160.789562, rho = -0.065431

nSV = 288, rBSV = 2

Total nSV = 288

Cross Validation Accuracy = 60.1648%
```

#### C = 100 G = 0.01 = Cross Validation Accuracy = 64.2857%

```
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 0.01 -v 5 normalisedtrainingDATA.data

""" optimization finished, #iter = 218
nu = 0.803972
obj = -22895.294480, rho = 0.603546
nSV = 238, nBSV = 231
Total nSV = 238

""" optimization finished, #iter = 198
nu = 0.791186
obj = -22485.924287, rho = -0.204390
nSV = 233, nBSV = 228
Total nSV = 233

""" optimization finished, #iter = 208
nu = 0.816497
obj = -23027.205490, rho = 2.061166
nSV = 239, nBSV = 233
Total nSV = 239

""" optimization finished, #iter = 233
nu = 0.778597
obj = -22140.455243, rho = 2.245228
nSV = 231, nBSV = 225
Total nSV = 231

""" optimization finished, #iter = 251
nu = 0.792939
obj = -22351.516901, rho = 2.503308
nSV = 233, nBSV = 227
Total nSV = 233
Cross Validation Accuracy = 64.2857%
```

## C = 100 G = 10 => Cross Validation Accuracy = 67.3077% ← Highest

## C = 100 G = 2048 => Cross Validation Accuracy = 60.4396%

```
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 2048 -v 5 normalisedtrainingDATA.data
.*.*
optimization finished, #iter = 586
nu = 0.011488
obj = -167.720734, rho = -0.052361
nSV = 286, rBSV = 0
Total nSV = 286
.*.*
optimization finished, #iter = 659
nu = 0.011496
obj = -167.854265, rho = -0.067163
nSV = 287, rBSV = 0
Total nSV = 287
.*.*
optimization finished, #iter = 606
nu = 0.012119
obj = -175.730786, rho = -0.055026
nSV = 287, rBSV = 0
Total nSV = 287
.*.*
optimization finished, #iter = 659
nu = 0.010791
obj = -157.542408, rho = -0.074249
nSV = 289, rBSV = 0
Total nSV = 289
.*.*
optimization finished, #iter = 594
nu = 0.012309
obj = -178.481574, rho = -0.063241
nSV = 287, rBSV = 0
Total nSV = 287
Cross Validation Accuracy = 60.4396%
```

## **Own Parameters**

Screenshots for this will be added in appendix at end of the report.

The value of G=10 gave the most accurate reading, for that reason I started experimenting with it first, keeping C at a value of 100 as that gave the most accurate reading beforehand.

. C = 100 G = 20 => Cross Validation Accuracy = 66.7582%

Outcome was lower than when G was 10 so lowered it to G = 0

*C* = 100 *G* = 0 => Cross Validation Accuracy = 65.9341%

This was even lower so decided to keep the G value in between 10 and 20 since they were the best results

C = 100 G = 15 => Cross Validation Accuracy = 67.5824%

This was the highest accuracy so far. Decided to take G = 14

C = 100 G = 14 => Cross Validation Accuracy = 66.7582%

As the accuracy for this was lower than G=15, I tried G=16 next

C = 100 G = 16 => Cross Validation Accuracy = **67.8571%** 

As this increased even more next was G=17

*C* = 100 *G* = 17 => Cross Validation Accuracy = 66.7582%

As this now decreased the most optimal value of G found was 16

Now I will be training with different C values and G being 16.

C = 150 G = 16 => Cross Validation Accuracy = 66.2088%

Lower accuracy so now attempting C =50

*C* = 50 *G* = 16 => Cross Validation Accuracy = 67.8571%

Same as C = 100 so attempted C = 25

C = 25 G = 16 => Cross Validation Accuracy = 65.6593%

Lower accuracy therefor attempting C = 75

C = 75 G = 16 => Cross Validation Accuracy = 67.5824%%

Slightly lower than C = 100 or C=50 so attempting C = 80

C = 80 G = 16 => Cross Validation Accuracy = **68.1319**%

New highest accuracy value found all other values tested were the same or lower

Optimal Values C = 80 G = 16

Task 4 - Non-linear Classification

```
C:\Docs\libsvm-3.23\windows>svm-train -c 80 -g 16 normalisedTrainingSet normalisedTrainingModel1
.....*
optimization finished, #iter = 4012
ru = 0.354629
obj = -8663.326105, rho = -0.807594
nSV = 268, rBSV = 95
Total nSV = 268
```

```
C:\Docs\libsvm-3.23\windows>svm-predict normalisedTestingSet normalisedTrainingModel1 predict
Accuracy = 57.3529% (78/136) (classification)
C:\Docs\libsvm-3.23\windows>
```

## Task 5 - Classification Analysis

#### Code:

```
falsenegativeScript.m × +
 1 -
        testingSet = csvread('testingSet.csv',1,13,[1,13,136,13]);
 2 -
       load predict;
 3
 4 -
       correctCount = sum([predict == testingSet]);
 5 -
       accuracy rate = (correctCount / size(testingSet,1))*100;
 6 -
        errorPositions = find ((predict ~= testingSet) == 1);
 7
 8 -
        falseNeg = find ((predict == -1) & (testingSet == 1))
 9
10
```

#### **Total errors: 58**

3, 7, 8, 10, 13, 14, 15, 18, 22, 24, 25, 26, 28, 31, 33, 38, 39, 40, 43, 45, 52, 53, 54, 59, 60, 61, 63, 64, 69, 72, 73, 75, 78, 83, 86, 88, 90, 91, 95, 97, 98, 103, 106, 108, 109, 114, 115, 117, 118, 121, 124, 125, 126, 128, 130, 132, 133, 135

## False negatives: 37

*3,* 10*,* 13*,* 14*,* 15*,* 18*,* 22*,* 25*,* 28*,* 31*,* 38*,* 39*,* 40*,* 43*,* 45*,* 52*,* 53*,* 60*,* 61*,* 63*,* 64*,* 69*,* 72*,* 73*,* 91*,* 95*,* 97*,* 98*,* 106*,* 108*,* 109*,* 114*,* 117*,* 124*,* 125*,* 126*,* 128

## **Critical Analysis**

Looking at the figures from Task 1 and the positions of the errors, we can see that a big majority of non-defective data were correctly classified around the large cloud of data and the defective data was correctly classified outside of it. This suggests there is a link between variance and being misclassified. Most non-defective data that was misclassified had a low variance and was mostly concentrated around the main data cloud. Defective data with high variance was mostly misclassified on the outside of the main data.

#### Task 6 - Linear Classification

```
C:\Docs\libsvm-3.23\windows>svm-train -t 0 -c 5 normalisedTrainingSet C5LinearModel

* optimization finished, #iter = 319
ru = 0.773347
obj = -1379.113699, rho = 1.471498
nSV = 286, rBSV = 280
Total nSV = 286

C:\Docs\libsvm-3.23\windows>svm-train -t 0 -c 10 normalisedTrainingSet C10LinearModel
.WARNING: using -h 0 may be faster

* optimization finished, #iter = 597
ru = 0.764125
obj = -2714.685946, rho = 1.478319
nSV = 281, rBSV = 272
Total nSV = 281

C:\Docs\libsvm-3.23\windows>svm-train -t 0 -c 100 normalisedTrainingSet C100LinearModel
......*...*
optimization finished, #iter = 5107
ru = 0.719874
obj = -25770.188339, rho = 1.380064
nSV = 266, rBSV = 255
Total nSV = 266
```

```
C:\Docs\libsvm-3.23\windows>svm-predict normalisedTestingSet C5LinearModel C5Predict
Accuracy = 61.7647% (84/136) (classification)

C:\Docs\libsvm-3.23\windows>svm-predict normalisedTestingSet C10LinearModel C10Predict
Accuracy = 60.2941% (82/136) (classification)

C:\Docs\libsvm-3.23\windows>svm-predict normalisedTestingSet C100LinearModel C100Predict
Accuracy = 61.7647% (84/136) (classification)

C:\Docs\libsvm-3.23\windows>
```

#### Cost = 5

Accuracy rate = 61.7646

**Total Errors: 52** 

1,3,10,13,14,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,61,63,64,65,69,72,73,78,80,84,86,88,91,95,97,98,99,108,109,112,115,117,119,120,121,123,125,128

Total False Negatives: 44

1,3,10,13,14,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,61,63,64,65,69,72,73,80,84,91,95,97,98,99,108,109,117,120,125,128

## **Cost = 10**

Accuracy rate = 60.2941

**Total Errors: 54** 

1,3,10,13,14,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,60,61,63,64,65,69,72,73,78,80,84,86,88,91,95,97,98,99,108,109,112,115,117,119,120,121,123,125,127,128

Total False Negatives: 46

1,3,10,13,14,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,60,61,63,64,65,69,72,73,80,84,91,95,97,98,99,108,109,117,120,125,127,128

## Cost = 100

Accuracy rate = 61.7647

Total Errors: 52

1,3,10,13,14,15,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,60,61,63,64,65,69,72,73,78,80,84,86,88,91,97,98,99,108,109,112,115,117,121,123,125,127,128

Total False Negatives: 45

1,3,10,13,14,15,16,18,21,22,23,25,28,31,32,35,37,39,43,44,45,47,49,52,53,60,61,63,64,65,69,72,73,80,84,91,97,98,99,108,109,117,125,127,128

## Critical Analysis

From the 3 models above, we can see that the parameters C = 5 and C = 100 resulted in the best accuracy % and both resulted in the same amount of errors. As C = 5 produced one less False Negative and thus I would use it over the others as false negatives make up the majority of the errors and having less would make it more balanced.

Surprisingly, all 3 linear models resulted in higher accuracy rates in comparison to the Gaussian model in task 4. All linear models also had fewer overall errors 52 and 54 in comparison to the Gaussian model with 58. The Gaussian model had more of a balance between false negatives and false positives as the linear models all had a high concentration of the errors being false negatives.

As Gaussian models usually result in higher accuracy %, these results suggest that the C and  $\gamma$  values used in task 4 were not optimal.

## Appendix:

## Task 1 Code:

```
COURSEWORK.m × falsenegativeScript.m × +
         training = csvread('trainingSet.csv',1,0,[1,0,364,12]);
dataLabels = csvread('trainingset.csv',1,13,[1,13,364,13]);
label_name = {'LOC_BLANK','BRANCH_COUNT','LOC_CODE_AND_COMMENT','LOC_COMMENTS','CYCLOMATIC_COMPLEXITY','DESIGN_COMPLEXITY',
testingLabels = csvread('testingLabels.csv')
2 -
 5 -
          normData = zscore(training);
testing = xlsread('testingSet.csv');
          [pcvectors, pcScores, pcEigenValues] = pca(normData);
8
9 –
10 -
          hold on
          hl = plot(pcScores(dataLabels == -1,1), pcScores(dataLabels == -1,2), 'bo');
11 -
          h2 = plot(pcScores(dataLabels == 1,1), pcScores(dataLabels == 1,2), 'rx');
13 -
14
15
16 -
          figure(2)
17 -
          h1 = stem(pcScores(dataLabels == -1,1), pcScores(dataLabels == -1,2), 'bo');
h2 = stem(pcScores(dataLabels == 1,1), pcScores(dataLabels == 1,2), 'ro');
18 -
19 -
20
          normData2 = zscore(testing);
22
23 -
         [pcvectors2, pcScores2, pcEigenValues2] = pca(normData2);
errors = find((predict~=testingLabels)==1);
24 -
```

## Task 3 Screenshots Below

```
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 20 -v 5 normalisedtrainingDATA.data
 C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 0 -v 5 normalisedtrainingDATA.data
 .
WARNING: using -h 0 may be faster
 optimization finished, #iter = 466
 optimization frished, #fter = 466
nu = 0.767532
obj = -21507.344196, rho = 0.449159
nSV = 230, nBSV = 218
Total nSV = 230
 WARNING: using -h 0 may be faster
optimization finished, #iter = 543

nu = 0.743246

nu = 0.20827.607953, rho = -3.957476

nSV = 223, rBSV = 211

Total nSV = 223

.*.*

optimization finished, #iter = 650

nu = 0.774409

obj = -21661.799901, rho = -2.878544

nSV = 230, rBSV = 217

Total nSV = 230

.*.*

optimization finished, #iter = 603

nu = 0.738413

obj = -20790.693680, rho = -2.922256

nSV = 225, rBSV = 209

Total nSV = 225

.*

optimization finished, #iter = 397
 ."
optimization finished, #iter = 397
nu = 0.750643
obj = -21137.126399, rho = -3.153576
nSV = 224, nBSV = 215
Total nSV = 224
Cross Validation Accuracy = 65.9341%
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 1 -v 5 normalisedtrainingDATA.data
```

```
:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 1 -v 5 normalisedtrainingDATA.data
 optimization finished, #iter = 2521
nu = 0.632479
obj = -17376.812381, rho = -1.483210
nSV = 213, nBSV = 167
Total nSV = 213
 optimization finished, #iter = 1708

nu = 0.613096

obj = -16813.402377, rho = -1.021613

nSV = 200, nBSV = 165

Total nSV = 200
 lotal nSV = 200

...*..*

optimization finished, #iter = 2310

ru = 0.644669

obj = -17648.576638, rho = -3.573698

nSV = 211, rBSV = 170

Total nSV = 211

....*.*

optimization finished, #iter = 1962

ru = 0.608987

obj = -16580.813826, rho = -3.278438

nSV = 206, rBSV = 161

Total nSV = 206
 optimization finished, #iter = 1824

nu = 0.648452

obj = -17635.628038, rho = -4.447456

nSV = 212, nBSV = 173

Total nSV = 212

Cross Validation Accuracy = 64.011%
   C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 15 -v 5 normalisedtrainingDATA.data
 optimization finished, #iter = 4921

nu = 0.345004

obj = -8333.801211, rho = -0.802866

nSV = 208, rBSV = 68

Total nSV = 208
nSV = 208, nBSV = 68
Total nSV = 208
.....*

optimization finished, #iter = 3455
nu = 0.329954
obj = -7919.753065, rho = -1.106604
nSV = 186, nBSV = 71
Total nSV = 186
.....*

optimization finished, #iter = 3901
nu = 0.359930
obj = -8660.397382, rho = -0.556403
nSV = 222, rBSV = 72
Total nSV = 222
.....*

optimization finished, #iter = 3596
nu = 0.322225
obj = -7838.596155, rho = -0.785334
nSV = 207, rBSV = 63
Total nSV = 207
.....*

optimization finished, #iter = 3162
nu = 0.343376
obj = -8196.329724, rho = -0.938261
nSV = 210, rBSV = 71
Total nSV = 210
Cross Validation Accuracy = 67.5824%
C:\Docs\libsym-3.23\windows>sym-train
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 14 -v 5 normalisedtrainingDATA.data
```

```
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 16 -v 5 normalisedtrainingDATA.data
optimization finished, #iter = 3714
nu = 0.336303
obj = -8088.950673, rho = -0.824981
nSV = 211, nBSV = 68
Total nSV = 211
nSV = 211, r8SV = 68
Total nSV = 211
.....*

optimization finished, #iter = 3584

nu = 0.322387
obj = -7685.337585, rho = -1.037867
nSV = 192, r8SV = 68
Total nSV = 192
optimization finished, #iter = 3958
nu = 0.350180
obj = -8397.662497, rho = -0.551701
nSV = 223, r8SV = 72
Total nSV = 223
.....*
optimization finished, #iter = 3579
nu = 0.314765
obj = -7594.041536, rho = -0.787299
nu = 0.314765
obj = -7594.041536, rho = -0.787299
nu = 0.314765
obj = -7947.264906, rho = -0.941163
nSV = 207
.....*
...*
optimization finished, #iter = 2990
nu = 0.335115
obj = -7947.264906, rho = -0.941163
nSV = 209, rBSV = 69
Total nSV = 209
Cross Validation Accuracy = 67.8571%
C:\Docs\libsym-3.23\windows>sym-train
C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 17 -v 5 normalisedtrainingDATA.data
 optimization Thissed, #Tter = 2823

nu = 0.327566

obj = -7710.324323, rho = -0.936519

nSV = 209, nBSV = 65

Total nSV = 209

Cross Validation Accuracy = 66.7582%
C:\Docs\libsvm-3.23\windows>svm-train -c 150 -g 16 -v 5 normalisedtrainingDATA.data
```

```
C:\Docs\libsvm-3.23\windows>svm-train -c 50 -g 16 -v 5 normalisedtrainingDATA.data
.....*...*
optimization finished, #iter = 2409
nu = 0.382574
obj = -4662.260872, rho = -0.638957
nSV = 219, rBSV = 81
Total nSV = 219
  ....*..*
optimization finished, #iter = 1932
  optimization finished, #iter = 1932

nu = 0.366134

obj = -4450.343732, rho = -0.894758

nSV = 215, nBSV = 83

Total nSV = 215

.....*

optimization finished, #iter = 2810
  obj = -4863.030954, rho = -0.506410

nSV = 229, rBSV = 86

Total nSV = 229
 Total nSV = 229
....*..*
.*
optimization finished, #iter = 2256
nu = 0.359125
obj = -4372.441807, rho = -0.678322
nSV = 219, rBSV = 75
Total nSV = 219
.....*
optimization finished, #iter = 2555
ru = 0.383322
obj = -4605.062824, rho = -0.813200
nSV = 222, rBSV = 79
Total nSV = 222
Cross Validation Accuracy = 67.8571%
C:\Docs\libsvm-3.23\windows>svm-train -c 75 -g 16 -v 5 normalisedtrainingDATA.data
  C:\Docs\libsvm-3.23\windows>svm-train -c 25 -g 16 -v 5 normalisedtrainingDATA.data
 optimization finished, #iter = 1488

nu = 0.438237

obj = -2672.434168, rho = -0.509453

nSV = 231, nBSV = 96

Total nSV = 231

* *
optimization finished, #iter = 1463
nu = 0.431044
obj = -2575.741731, rho = -0.740546
nSV = 226, nBSV = 96
Total nSV = 226
...*.*
optimization finished, #iter = 1208
nu = 0.456013
obj = -2797.552839, rho = -0.429486
nSV = 228, nBSV = 102
Total nSV = 228
...*.*
optimization finished, #iter = 2185
nu = 0.409110
obj = -2507.616075, rho = -0.600032
nSV = 227, nBSV = 86
Total nSV = 227
...*.*
optimization finished, #iter = 1809
nu = 0.446982
obj = -2663.042601, rho = -0.690251
nSV = 227, nBSV = 90
Total nSV = 227
Cross Validation Accuracy = 65.6593%
  ...*..*
optimization finished, #iter = 1463
```

```
optimization finished, #iter = 2895
nu = 0.366433
obj = -7037.852634, rho = -0.542012
nSV = 222, rBSV = 76
Total nSV = 222
.....*..*
optimization finished, #iter = 2990
nu = 0.327530
obj = -6359.708625, rho = -0.753798
nSV = 211, rBSV = 64
Total nSV = 211
.....*
optimization finished, #iter = 2563
nu = 0.350133
obj = -6676.570157, rho = -0.885319
nSV = 215, rBSV = 74
Total nSV = 215
Cross Validation Accuracy = 68.1319%
  C:\Docs\libsym-3.23\windows>sym-train -c 84 -g 16 -v 5 normalisedtrainingDATA.data
  optimization finished, #iter = 3805

nu = 0.347842

obj = -7050.663647, rho = -0.768806

nSV = 212, rBSV = 69

Total nSV = 212
nSV = 212, r8SV = 69
Total nSV = 212
......*..*

optimization finished, #iter = 2778
ru = 0.333402
obj = -6710.518100, rho = -1.008197
nSV = 191, r8SV = 71
Total nSV = 191
......*

optimization finished, #iter = 3566
ru = 0.362868
obj = -7315.032319, rho = -0.544153
nSV = 223, r8SV = 74
Total nSV = 223
......*

optimization finished, #iter = 3326
ru = 0.324385
obj = -6611.830902, rho = -0.761531
nSV = 211, r8SV = 64
Total nSV = 211
......*

optimization finished, #iter = 2822
ru = 0.346654
obj = -6936.546989, rho = -0.893068
nSV = 215, r8SV = 73
Total nSV = 215
Cross Validation Accuracy = 68.1319%
C:\Docs\libsvm-3.23\windows>svm-train
  C:\Docs\libsvm-3.23\windows>svm-train -c 85 -g 16 -v 5 normalisedtrainingDATA.data
 C:\Docs\Tibsym-3.23\Windows>sym-trai
......*

optimization finished, #iter = 3480

ru = 0.347054

obj = -7116.870819, rho = -0.771956

rsV = 212, rmsV = 69

Total rsV = 212
.....*

optimization finished, #iter = 2934

ru = 0.332665

obj = -6772.839154, rho = -1.010102

rsV = 191, rmsV = 71

Total rsV = 191
......*
```

```
C:\Docs\libsvm-3.23\windows>svm-train -c 95 -g 16 -v 5 normalisedtrainingDATA.data
  lotal nSV = 223

......*

optimization finished, #iter = 3728

nu = 0.317130

obj = -7291.818956, rho = -0.778901

nSV = 209, nBSV = 63

Total nSV = 209
 C:\Docs\libsvm-3.23\windows>svm-train -c 100 -g 16 -v 5 normalisedtrainingDATA.data
  optimization finished, #iter = 3714
nu = 0.336303
obj = -8088.950673, rho = -0.824981
nSV = 211, nBSV = 68
Total nSV = 211
nSV = 211, nBSV = 68
Total nSV = 211
.....*

optimization finished, #iter = 3584
nu = 0.322387
obj = -7685.337585, rho = -1.037867
nSV = 192, nBSV = 68
Total nSV = 192
optimization finished, #iter = 3958
nu = 0.350180
obj = -8397.662497, rho = -0.551701
nSV = 223, nBSV = 72
Total nSV = 223
.....*
optimization finished, #iter = 3579
nu = 0.314765
obj = -7594.041536, rho = -0.787299
nSV = 207, nBSV = 64
Total nSV = 207
.....*
optimization finished, #iter = 2990
nu = 0.335115
obj = -7947.264906, rho = -0.941163
nSV = 209, nBSV = 69
Total nSV = 209
Cross Validation Accuracy = 67.8571%
```

```
C:\Docs\libsvm-3.23\windows>svm-train -c 99 -g 16 -v 5 normalisedtrainingDATA.data
 C:\docs(1189\m 31.3\m)
optimization finished, #iter = 3946
ru = 0.336909
obj = -8025.288396, rho = -0.820658
nSV = 211, nBSV = 68
Total nSV = 211
 notal nsv = 211

.....*...*

optimization finished, #iter = 3417

nu = 0.323029

obj = -7625.689894, rho = -1.035878

nSV = 191, nBSV = 68

Total nSV = 191
nsv = 191, nBsv = 68
Total nsv = 191
.....*

optimization finished, #iter = 4068
nu = 0.350847
obj = -8331.180341, rho = -0.551658
nsv = 223, rBsv = 71
Total nsv = 223
.....*

optimization finished, #iter = 3367
nu = 0.315209
obj = -7533.976579, rho = -0.785596
nsv = 207, rBsv = 64
Total nsv = 207
.....*

optimization finished, #iter = 2834
nu = 0.335794
obj = -7885.427116, rho = -0.938119
nsv = 209, rBsv = 70
Total nsv = 209
Cross Validation Accuracy = 67.8571%
  C:\Docs\libsvm-3.23\windows>svm-train -c 79 -g 16 -v 5 normalisedtrainingDATA.data
 optimization finished, #iter = 3451
nu = 0.351976
obj = -6716.784991, rho = -0.754147
nSV = 213, nBSV = 72
Total nSV = 213
 Total nSV = 213
.....*..*
optimization finished, #iter = 2295
nu = 0.337055
obj = -6395.793558, rho = -0.999544
nSV = 197, nBSV = 73
Total nSV = 197
 Total nSV = 197
.....*...*
optimization finished, #iter = 3293
nu = 0.367337
obj = -6968.070741, rho = -0.541096
nSV = 223, nBSV = 76
Total nSV = 223
......*
optimization finished, #iter = 3211
nu = 0.328393
obj = -6296.277452, rho = -0.751358
nSV = 211, nBSV = 64
Total nSV = 211
.....*
 optimization finished, #iter = 2587

nu = 0.350907

obj = -6611.081675, rho = -0.883860

nSV = 215, rBSV = 74

Total nSV = 215

Cross Validation Accuracy = 68.1319%
```