# CS 5011: Assignment 1 report

Due on Friday, September 26, 2014

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#### CS12B028

### Question 1.

Mean for class 1 = [0]\*10

Mean for class 2 = [3]\*10

Covariance matrix is same for both classes and the matrix is generated by taking each element as a random number between 0 and 3 and it is verified whether it is spherical or not.

Test data is chosen randomly from each class (400 data points from each class) and the remaining data is treated as the training data.

### Question 2.

Based on the data set created in question 1(DS1),

 $\begin{aligned} & \text{Coefficients learnt} = 0.19050926 \;,\, 0.00771853 \;,\, 0.07373751 \;,\, 0.0613499 \;,\, 0.00560171 \;, -0.00628388, \, 0.00203516 \;,\, 0.00403266 \;,\, 0.02744673 \;,\, 0.02573123 \;,\, 0.00418536 \end{aligned}$ 

Accuracy - 0.90375

Precision - 0.906801007557

Recall - 0.9

F-measure - 0.90338770389

# Question 3.

For the DS1, for k=5 the Accuracy is slightly higher compared to other values of k and the it's accuracy is slightly lower than that of regression on indicator variable.i.e; knn is doing worse than regression in this case.

for k=1

Accuracy - 0.84375

Precision - 0.846347607053

Recall - 0.84

F-measure -0.843161856964

for k=2

Accuracy - 0.82875

Precision - 0.894894894895

Recall - 0.745

F-measure -0.81309686221

for k=3

Accuracy - 0.87125

Precision - 0.872180451128

Recall - 0.87

F-measure -0.871088861076

for k=4

Accuracy - 0.8725

Precision - 0.909340659341

Recall - 0.8275

F-measure -0.866492146597

for k=5

Accuracy - 0.88625

Precision - 0.881481481481

Recall - 0.8925

F-measure -0.886956521739

for k=6

Accuracy - 0.88

Precision - 0.906417112299

Recall - 0.8475

F-measure -0.875968992248

for k=7

Accuracy - 0.885

Precision - 0.881188118812

Recall - 0.89

F-measure -0.885572139303

for k=8

Accuracy - 0.88375

Precision - 0.902887139108

Recall - 0.86

F-measure -0.880921895006

for k=9

Accuracy - 0.87875

Precision - 0.870415647922

Recall - 0.89

F-measure -0.880098887515

for k=10

Accuracy - 0.87625

Precision - 0.884910485934

Recall - 0.865

F-measure -0.874841972187

Best Fit for k=5:

Accuracy - 0.88625

Precision - 0.881481481481

Recall - 0.8925

F-measure -0.886956521739

### Question 4.

In this problem, I have taken 2 mean vectors for two classes and 3 covaiance matrices (generated randomly) The points are chosen from each covariance matrix for each class based on given probabilities.

Mean1 = [0]\*10

Mean2 = [3]\*10

We can not say that which performs better because it largely depends on what the actual data set is rather than how you generate it, but for mixture of guassians knn works better than linear classifier if we take different means for different covariance matrices, beacause there is a high chance that we can not find a line to seperate those classes. But in the case I have chosen, linear classifier works slightly better because there are only two means but different covariance matrices.

#### performance Results:

#### Using linear classifier:

 $\begin{array}{l} \text{Coefficients learnt: } 0.18627366\ 0.02848674\ 0.02265773\ 0.02526162\ 0.02790363\ 0.00238386\ 0.03325447\ 0.02182644\ 0.01192047\ 0.01345345\ 0.01415136 \end{array}$ 

Accuracy - 0.885

Precision - 0.892857142857

Recall - 0.875

F-measure - 0.883838383838

#### Using knn:

Accuracy - 0.86625

Precision - 0.865336658354

Recall - 0.8675

F-Measure - 0.866416978777

# Question 5.

In this question, we are asked to fill the missing values with mean.But mean is not a really good choice ,if the data is completely skewed(such as one noisy point which has a really large value can increase the mean of the entire data set considerably.Instead , we can use median to fill the data set so that those noisy points will not have much effect. But the given data set is not so noisy, so I have used the mean to fill the missing attributes.

# Question 6.

5 80-20 different splits are done by choosing random 20 percent from the dataset and set that as testdata and the remaining as training data and the process is repeated 5 times.

RSS1 - 0.018474853191

RSS2 - 0.0204983378639

RSS3 - 0.0209688850415

RSS4 - 0.0155163266286

RSS5 - 0.0169499521748

Mean RSS - 0.01848167098

### Question 7.

The below values are for lambda = 2 (which is observed to have less mean rss compared to others) rss1 - 0.0183201292463

rss2 - 0.01876789375

rss3 - 0.0192973920872

rss4 - 0.0148353674319

rss5 - 0.0170740764304

mean rss - 0.0176589717892

### Question 8.

Precision for class 0 - 0.611940298507

Precision for class 1 - 0.613065326633

Recall - 0.61

F-measure - 0.6115

Figure 1 is the dataset projection in 3d.

Figure 2 has both points in the derived feature space and also the classifier.

The classifier is actually a point (in Figure 2), but the point was not visible when plotted. So ,I have used a line that passes through the classifier point for the representation.

### Question 9.

Precision for class0- 1.0

Precision for class1- 1.0

Recall - 1

F-measure - 1

Figure 1 is the dataset projection in 3d.

Figure 3 represents the extracted feature set along with the classifier (red point)

Based on the above two feature extraction methods, it can be clearly seen that Ida is performing much better compared to pca. This is mainly because of the given data set, From the plot of the given data set, the points are roughly present on two parallel lines with some guassian noise, also they are present within the same range, so when we perform pca and apply linear regression, both the classes merge and classifying will not be accurate. Whereas in Ida, the data sets are linearly separable so we get 100 percent accurate measurements.

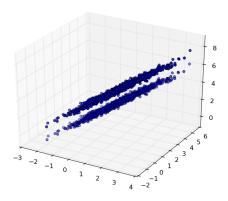


Figure 1: Data Set

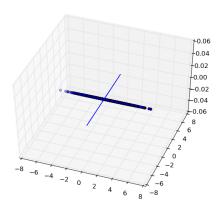


Figure 2: Data Set in the derived feature space(pca)

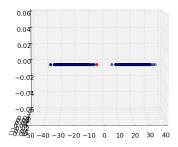


Figure 3: Data Set in the derived feature  $\operatorname{space}(\operatorname{lda})$