

# Homework 5 Report

Brian King

## Question 1

Using 5 folds with the Naïve Bayes algorithm obtains the following accuracies for the validation sets:

```
Scores: [93.33333333333333, 96.66666666666667, 100.0, 93.33333333333333, 93.33333333333333]
Mean Accuracy: 95.333%
```

## Question 2

The following functions were modified in order to implement a Multivariate Bayes classifier in place of the Naive Bayes classifier:

### Calculate\_class\_probabilities function

```
145 # Calculate the probabilities of predicting each class for a given row
146 def calculate_class_probabilities(summaries, row):
147     del(row[-1])
148     total_rows = sum([summaries[label][2] for label in summaries])
149     probabilities = dict()
150     for class_value, class_summaries in summaries.items():
151         probabilities[class_value] = summaries[class_value][2]/float(total_rows)
152         mean, cov, _ = class_summaries
153         probabilities[class_value] *= calculate_probability(row, mean, cov) # discriminant function
154     return probabilities
```

### Calculate\_probability function

```
136 # Calculate the Gaussian probability distribution function for x
137 def calculate_probability(x, mean, cov):
138     inv_cov = np.linalg.inv(cov)
139     multiply1 = np.dot(np.transpose(x-mean), inv_cov)
140     multiply2 = np.dot(multiply1, (x-mean))
141     exponent = exp(-.5*multiply2)
142     return (1 / ((2 * pi)**(4/2) * (np.linalg.det(cov))**.5)) * exponent
```

### Summarize\_dataset function

```
114 # Calculate the mean, cov and count for each dataset
115 def summarize_dataset(dataset):
116     columns = [column for column in zip(*dataset)]
117     del(columns[-1]) # remove classification from array
118     mean = means(columns)
119     length = len(columns[1])
120     cov = covariance_creator(columns, mean)
121     summaries = [mean, cov, length]
122     return summaries
```

### Covariance function

```
103     # Calculate covariance arrays
104     def covariance_creator(datalist, mean):
105         covariances = np.zeros((4, 4))
106         for i in range(len(datalist)):
107             for j in range(len(datalist)):
108                 for k in range(len(datalist[1])):
109                     r = ((datalist[i][k] - mean[i]) * (datalist[j][k] - mean[j]))
110                     covariances[i][j] = covariances[i][j] + r/len(datalist[1])
111         return covariances
```

### Mean function

```
95     # Calculate mean arrays
96     def means(datalist):
97         mean = np.array([])
98         for i in range(len(datalist)):
99             this_mean = sum(datalist[i])/len(datalist[i])
100             mean = np.append(mean, this_mean)
101         return mean
```

## Question 3

Utilizing the aforementioned Multivariate Bayes classifier, the following accuracies were obtained.

```
Scores: [96.66666666666667, 96.66666666666667, 100.0, 100.0, 93.33333333333333]
Mean Accuracy: 97.333%
```

This is an improvement of 2% over the Naïve Bayes classifier. However, with large amounts of data (greater than the previous 150 data points used in the report), it is more efficient to use the Naïve Bayes classifier as it computationally less intensive.