# CS6690: Pattern Recognition Assignment #2

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## 1 Bayesian Classifiers

According to Bayes Theorem, for a dataset x with classes  $\omega_i$ , Probability of a datapoint belonging to class  $\omega_i$  is defined as:

$$P(\omega_i|x) = \frac{(P(x|\omega_i)P((\omega_i)))}{P(x)}$$
(1)

- Here,  $P(x|\omega_i)$  is known as the class likelihood. To estimate this value, we require the distribution of  $\omega_i$ . Based on the central limit theorem, we can assume that this would be Gaussian distribution for large datasets.
- The value  $P(\omega_i)$  is the class prior and is calculated using:

$$P(\omega_i) = N_i/N \tag{2}$$

This term becomes irrelevant if the classes have equal probabilities.

• P(x) is termed as 'evidence' and can be calculated as:

$$P(x) = \sum_{i} P(x|\omega_i)P(\omega_i)$$
 (3)

#### 2 Gaussian Likelihood Distribution

For multi-dimensional data, the Gaussian Distribution is:

$$P(x; \mu, \Sigma) = \frac{1}{2\pi^{k/2} |\Sigma|^{1/2}} e^{-(x-m)^T \Sigma^{-1}(x-m)}$$
(4)

where

- $\mu$  is the mean
- $\Sigma$  is the covariance matrix

The above parameters are calculated for the following cases:

- 2.1 Bayes Classifier with Covariance same for all classes
- 2.2 Bayes Classifier with Covariance different for all classes
- 2.3 Naive Bayes Classifier with  $C = \Sigma^2 * I$
- 2.4 Naive Bayes Classifier with C same for all classes
- 2.5 Naive Bayes Classifier with C different for all classes

## 3 Bayes Classification

If 
$$P(\omega_1|x) > P(\omega_2|x)$$
 then x belongs to class  $\omega_1$   
If  $P(\omega_1|x) < P(\omega_2|x)$  then x belongs to class  $\omega_2$ 

Using equation (1), this can be written as:

$$P(x|\omega_1)P(\omega_1) \geqslant P(x|\omega_2)P(\omega_2) \tag{5}$$

This classification rules minimizes number of misclassifications.

## 4 Experiments

- 4.1 Data
- 4.2 Parameters
- 4.3 DET Curves
- 4.4 Decision Boundaries
- 4.5 Confusion Matrices
- 5 Cases

#### 5.1 Subsection Heading Here

Write your subsection text here.

#### 6 Conclusion

Write your conclusion here.

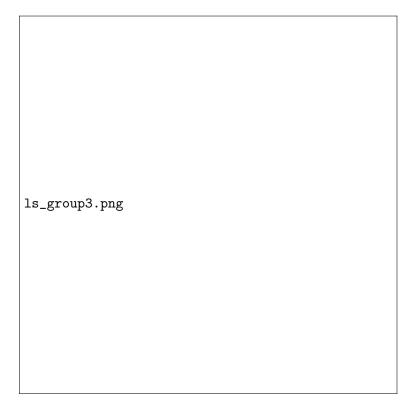


Figure 1: Linearly Separable Data

Figure 2: Simulation Results