## ML Assignment 1: Perceptron

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## 1 Introduction

In this assignment, we implement Fischer's Linear Discriminant, it is used for supervised learning in solving classification problems. Given the data points in M dimensional space, we project all those points to D dimensions and then try to find out a discriminant function to classify our points using a threshold found by calculating the intersection point between the normal distribution followed by the projected points. In this assignment we where given data points in 3-D and we projected it to 1-D and found out the threshold.

## 2 Implementation

After solving the optimisation problem:

$$max \frac{(W^T M_1 - W^T M_2)^2}{s_1^2 + s_2^2}$$

We get

$$W \propto S_W^{-1}(M_1 - M_2)$$
 , where  $S_W = \sum_{k=1}^2 S_k$  , and  $S_W = \sum_{k=1}^\infty (m_k - M_k)(m_k - M_k)$ 

$$S_k = \sum_{n \in c_k} (x_n - M_k)(x_n - M_k)^T$$

The threshold can be found by solving the quadratic equation  $Ax^2 + Bx + C = 0$ , where

$$\begin{split} A &= -\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \\ B &= 2(\frac{\mu_1}{\sigma_1^2} - \frac{\mu_2}{\sigma_2^2}) \\ C &= \frac{\mu_2^2}{\sigma_2^2} - \frac{\mu_1^2}{\sigma_1^2} + \log\left(\frac{\sigma_2^2}{\sigma_1^2}\right) \end{split}$$

The algorithm can be roughly outlined as:

1. Find the individual means for different classes

- 2. Find  $S_W$
- 3. Calculate unit vector W
- 4. Project all the points on to the unit vector by taking the dot product  $W^TX$
- 5. Find the Normal distribution fitting the projected points
- 6. Find the *threshold* using the intersection point of the normal distribution.

## 3 Results

Figure 1 shows the original data plot in 3 dimensions along with the decision boundary (in green). All points above the green plane are classified as belonging to class 1, while those below the plane are classified as belonging to class  $\theta$ . Figure 3 shows the data points projected to 1-dimension along with normal distributions fit on them and the threshold point.

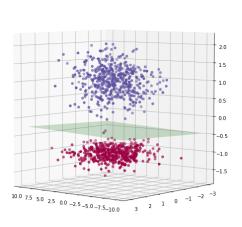


Figure 1: Scatter plot of the original data in 3-dimensional space along with the decision boundary

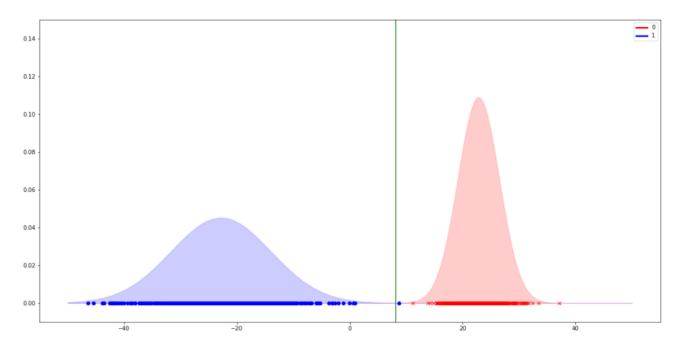


Figure 2: Normal distribution fit to each projected cluster. The green line shows the threshold point

The threshold point in **1-D** is 8.122.

The discriminating plane in original dimensions (3-D) is:  $0.148x_1 + 0.413x_2 - 22.636x_3 = 8.122$