## BITS F464 - Machine Learning

Assignment-1A: Fisher's Linear Discriminant

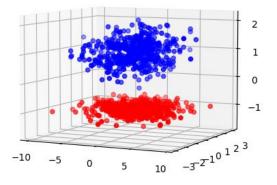
#### Made by -

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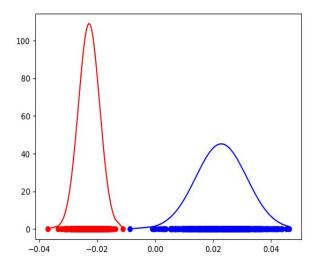
# **Model Description**

- The main idea is to maximize the difference between the means of the classes and minimize sum of variances.
- The expression boils down to maximize w<sup>T</sup>S<sub>b</sub>w/ w<sup>T</sup>S<sub>w</sub>w where S<sub>b</sub> is the between class covariance matrix and S<sub>w</sub> is the within class covariance matrix.
- W comes proportional to the expression S<sub>w</sub>(M<sub>1</sub>- M<sub>2</sub>)
- This algorithm works only if points are linearly separable else there will be a large number of misclassifications.
- When the points are projected onto w, we get a 1D plot of the points. Now we need a threshold to discriminate between the points in 1D.
- This is done by fitting normal distributions in the 1D classes.
- The intersection of these distributions gives us the threshold.
- If threshold is b and let w be [w1, w2,w3] then => w1\*x+w2\*y+w3\*z >=b is negative point and w1\*x+w2\*y+w3\*z < b is a positive point or vice versa.
- Thus  $w^Tx = b$  gives us the discriminating boundary in original space.

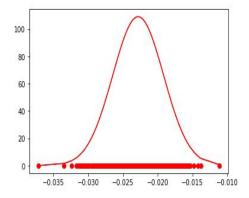
### **Plots**



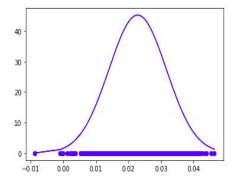
Plot visualizing the points in 3D space. Red represents 0 and blue represents 1.



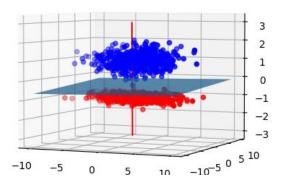
Plot visualizing the clusters in 1D and their respective normal curves.



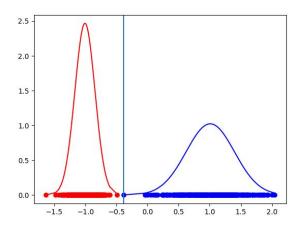
Scatter plot and normal curve for class 0 in 1D



Scatter plot and normal curve for class 1 in 1D



3D plane separating the classes in original space



Line separating the classes in 1D x=-0.3893028

# Results

- The unit vector w [-0.00655686 -0.01823739 0.99981218]
- The intersection point of normal distributions in 1D -0.3893028 , 0
- The line separating the classes in 1D is x=-0.3893028.
- The discriminant function -0.00655686x -0.01823739y + 0.99981218z= -0.3893028.
- Accuracy is 100%