# **SML Assignment 2 Report**

## Dhruv Jaiswal - 2023200

# **Accuracy of LDA**

	Class 0	Class 1	Class 2	Overall Accuracy
FDA then LDA (train)	0.94	0.94	0.91	0.93
FDA then LDA (test)	0.98	0.98	0.93	0.963
PCA then LDA (train)	1	0.99	0.98	0.99
PCA then LDA (test)	0.99	0.99	0.9	0.96
PCA(90%) then LDA (train)	1	0.99	0.97	0.986
PCA(90%) then LDA (test)	1	0.98	0.91	0.963
PCA(2D) then LDA (train)	0.87	0.85	0.74	0.82
PCA(2D) then LDA (test)	0.91	0.98	0.91	0.93

# **Accuracy of QDA**

	Class 0	Class 1	Class 2	Overall Accuracy
MLE (QDA, train)	1	0.99	0.99	0.993
FDA then QDA (train)	0.95	0.93	0.9	0.926
FDA then QDA (test)	0.98	0.98	0.93	0.963

PCA then QDA (train)	1	1	1	1
PCA then QDA (test)	0.99	0.74	0.99	0.906

#### **Effects of PCA**

As seen in the accuracy data above, There are 3 kinds of PCA we apply

1. PCA (95% variance, approx. 82 dimensions)

This transformation when applied on the dataset, keeps the accuracy relatively high on both the train set and test set however there is an unusual drop-off in some particular class of the trainset [Class 2 in LDA and Class 1 in QDA]. This might be because of discarding crucial dimensions needed to identify that class.

2. PCA (90% variance, approx. 50 dimensions)

This transformation performs marginally worse than the 95% PCA but is more consistent as there is no one class that lags behind. This might be because the noise that might've come inside the data is removed completely here.

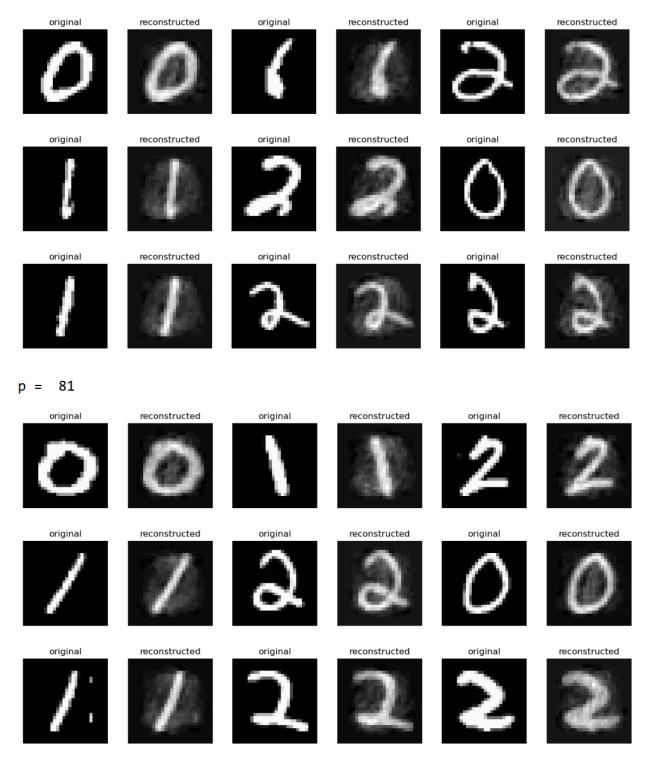
3. PCA (2 dimensions)

This transformation gives bad results on the train set but mediocre (in comparison to other).

### Visualization

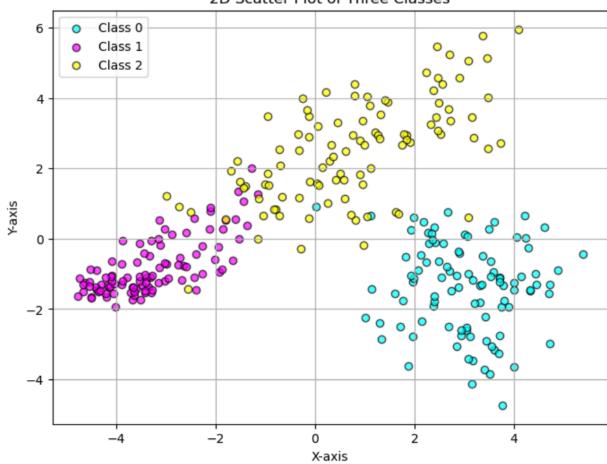
PCA [95%]

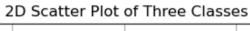
p = 82

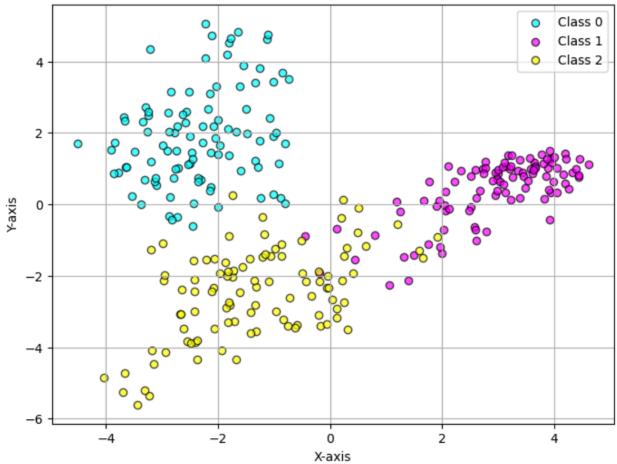


**FDA** 

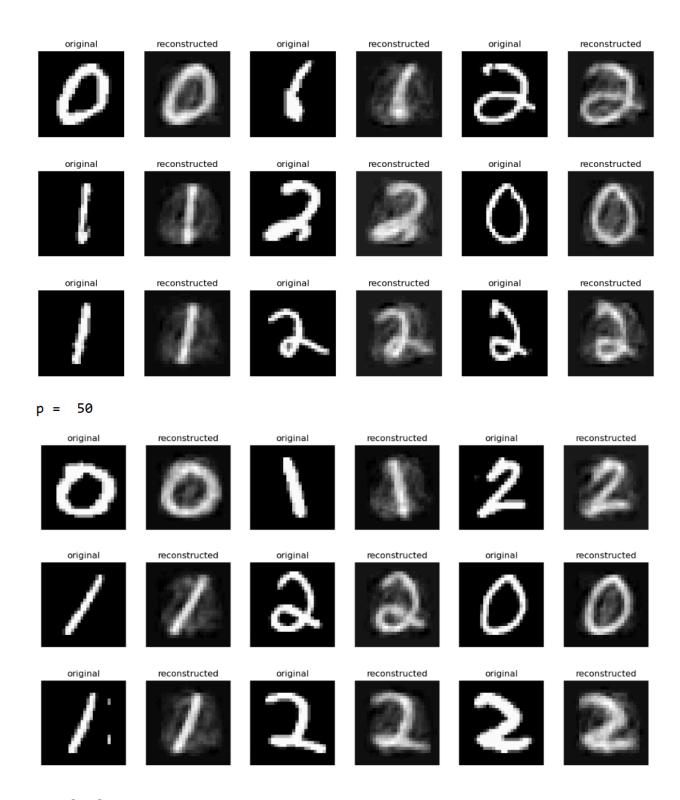
## 2D Scatter Plot of Three Classes



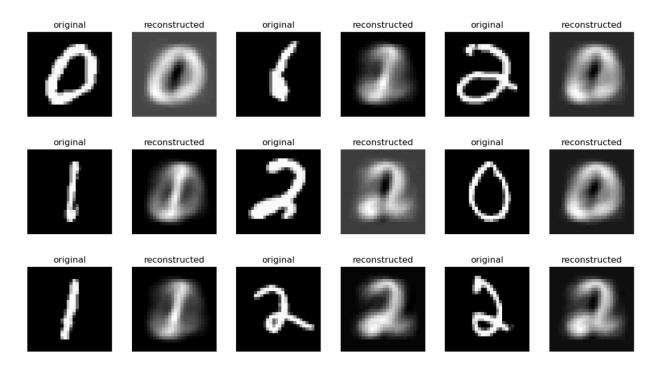




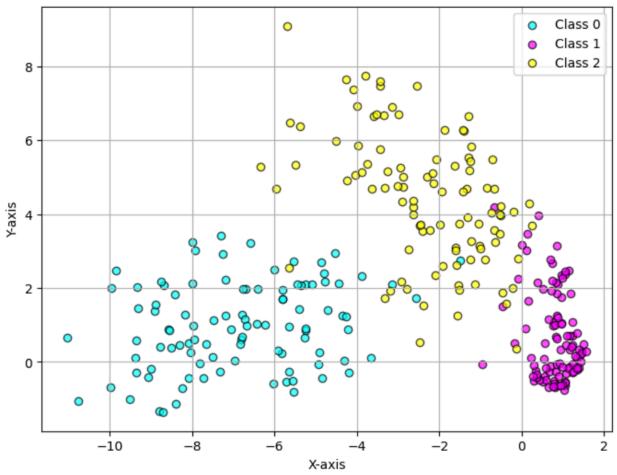
PCA [90%]

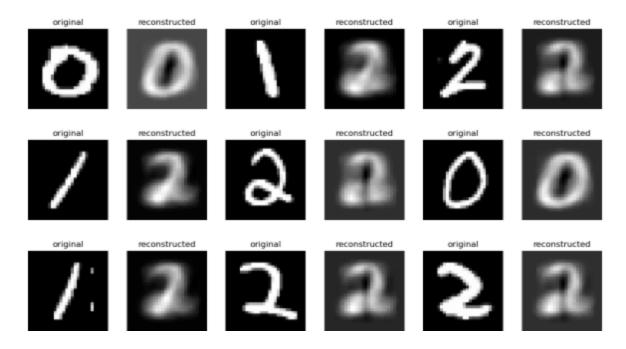


PCA [2D]



#### 2D Scatter Plot of Three Classes





#### 2D Scatter Plot of Three Classes

