

# Report

## **Project Part 1: Density Estimation and Classification using Fashion-MNIST:**

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# Naïve Bayes Classifier :

## Formula used to estimate the parameters:

The features used for the classification are the **Average** and **Standard Deviation**.

The average and standard deviation are calculated for each image and stored in a list as shown below:

```
average = []  
standard_deviation = []
```

The formula used to calculate both Average and Standard deviation is shown below:

```
average.append(sum(training_data_array[i])/len(training_data_array[0]))  
standard_deviation.append(statistics.stdev(training_data_array[i]))
```

## Estimated values for the parameters:

The mean and standard deviation is calculated for X1 and X2 for each class and the values are as below:

```
Mean of X1 for class 0: 0.3256077664399094  
Standard Deviation of X1 for Class 0: 0.11338436369948433  
Mean of X1 for class 1: 0.22290531462584984  
Standard Deviation of X1 for Class 1: 0.056955755259147804  
Mean of X2 for class 0: 0.32024038721169107  
Standard Deviation of X2 for Class 0: 0.0880463127795082  
Mean of X2 for class 1: 0.3341548890116032  
Standard Deviation of X2 for Class 1: 0.05707345032584217
```

## Expression for the estimated normal distribution:

The 2-D normal (gaussian) distribution is estimated by using the scipy library. It is found for both the parameters of each class.

Calculated as shown below:

*# Gaussian Distribution for x1 and x2 for both the classes 0 and 1:*

```
gaussian_dist_x1_class_0 = scipy.stats.norm(mean_x1_class_0, std_x1_class_0)
gaussian_dist_x1_class_1 = scipy.stats.norm(mean_x1_class_1, std_x1_class_1)
gaussian_dist_x2_class_0 = scipy.stats.norm(mean_x2_class_0, std_x2_class_0)
gaussian_dist_x2_class_1 = scipy.stats.norm(mean_x2_class_1, std_x2_class_1)
```

## Classification explanation by using the distribution:

Mean and standard deviation is calculated for both x1 and x2 for each class.

The gaussian distribution is calculated for each input parameters for both the classes where the inputs are the mean and standard deviation obtained from the above step.

Then, the PDF is calculated for each input variable by using **.pdf()** function of the gaussian distribution.

```
prob_class_0_x1_x2                                     =
gaussian_dist_x1_class_0.pdf(testing_list_x1_class_0[i]) *
gaussian_dist_x2_class_0.pdf(testing_list_x2_class_0[i]) * prob_class_0
```

Then, the class is predicted based on the PDF value for each input variable.

## Classification Accuracy:

The accuracy for the naïve bayes classifier using normal distribution is **83.15%**

| Accuracy for the naive bayes algorithm: 83.15 %

# Logistic Regression Classifier:

## Formula used to estimate the parameters:

The features used for the classification are the **Average** and **Standard Deviation**.

The average and standard deviation are calculated for each image and stored in a list as shown below:

```
average = []  
standard_deviation = []
```

The formula used to calculate both Average and Standard deviation is shown below:

```
average.append(sum(training_data_array[i])/len(training_data_array[0]))  
standard_deviation.append(statistics.stdev(training_data_array[i]))
```

## Expression for the Sigmoid Function:

The sigmoid function is calculated for each class and the label is predicted as shown below:

```
def predict(X):  
    linear_model = np.dot(X, weights) + bias  
    y_predicted = sigmoid(linear_model)
```

```
def sigmoid(x):  
    return 1 / (1 + np.exp(-x))
```

## Classification explanation by using the distribution:

Linear model is calculated by using  $w_0 + w_1.x_1 + w_2.x_2$

This is passed to the sigmoid function and the weights are calculated for 1000 iterations as shown below:

```
for _ in range(epoch_iters):  
    linear_model = np.dot(X, weights) + bias  
    y_predicted = sigmoid(linear_model)  
  
    dw = (1 / n_samples) * np.dot(X.T, (y_predicted - y))  
    db = (1 / n_samples) * np.sum(y_predicted - y)  
  
    weights -= learning_rate * dw  
    bias -= learning_rate * db
```

Once the weights are found, predict the y values.

## Classification Accuracy:

The accuracy for the naïve bayes classifier using normal distribution is **90.45%** and it is calculated as shown below:

```
accuracy = np.sum(y_true == y_pred) / len(y_true)  
return accuracy
```