```
import pandas as pd
In [11]:
          import numpy as np
          import statsmodels.api as sm
          import matplotlib.pyplot as plt
          from sklearn.preprocessing import StandardScaler
          from sklearn.decomposition import PCA
          from sklearn.linear model import LogisticRegression
          from sklearn.linear model import LinearRegression
          from sklearn.naive_bayes import BernoulliNB
          from sklearn import svm
          from sklearn import metrics
          from sklearn.metrics import confusion matrix
          import os
          THIS FOLDER = os.path.abspath('')
          #setting the training and testing dataset path
          class Dataset:
            train=os.path.join(THIS_FOLDER, 'wineQualityRed_train.csv')
            test=os.path.join(THIS_FOLDER, 'wineQualityRed test.csv')
          pd.options.display.max rows = None
          pd.options.display.max columns = None
          fields=["fixed acidity", "volatile acidity", "citric acid", "residual sugar", "ch
                  "free sulfur dioxide", "total sulfur dioxide", "density", "pH", "sulphate
          train_wine = pd.read_csv(Dataset.train, delimiter=';', header=None, skiprows=
          test_wine = pd.read_csv(Dataset.test, delimiter=';', header=None, skiprows=1,
          # CLASSIFING quality 0-bad AND 1-good and forming Y_TRAINING DATASET
          temp = train wine["quality"].replace([3,4,5,6] , 0)
          temp1 = temp.replace([7,8], 1)
          train wine["quality"] = temp1
          tempo = test wine["quality"].replace([3,4,5,6] , 0)
          tempo1 = tempo.replace([7,8] , 1)
          test_wine["quality"] = tempo1
          ##Training Dataset
          x train = train wine.iloc[:,:11]
          Y train = train wine.iloc[:,-1:]
          y train = Y train.values.reshape(-1,1)
          ## Testing Dataset
          x test = test wine.iloc[:,:11]
          Y test = test wine.iloc[:,-1:]
          y_test = Y_test.values.reshape(-1,1)
          ## Feature Scaling
          SC = StandardScaler()
          x train scaled = SC.fit transform(x train)
          x test scaled = SC.fit transform(x test)
          ## Principal Components Analysis to get "7" attributes
          Principal_7_comp = PCA(n_components = 7)
          x_redwine_7_training = Principal_7_comp.fit_transform(x_train_scaled)
          x_redwine_7_testing = Principal_7_comp.fit_transform(x_test_scaled)
          ## Principal Components Analysis to get "4" attributes
```

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In [12]:
        ## Linear Regression Classifier
         ## "7" Attributes.
         print("-----")
         reg1 = LinearRegression()
         regl.fit(x redwine 7 training , y train.ravel())
         x1 test pred = reg1.predict(x redwine 7 testing).round(3)
         ## HERE WE FIND OUT THE THRESHOLD ABOVE WHICH WE CAN CONSIDER THE QUALITY VAL
         ## "mid" IS THE VARIABLE CONTAINING THAT THRESHOLD VALUE....
         ## CONVERTING OUR PREDICTED RESULT AS PER THE THRESHOLD.....
         #MEAN OF THE Y PREDICTION
         mid = np.mean(x1 test pred)
         print("mean: ", mid)
         pred= []
         #Classifing y prediction based on the mean value
         for p in x1 test pred :
             if p < mid:</pre>
                pred.append(0)
            elif p > mid:
                pred_append(1)
         print("Y PREDICTION DATA after classificaion:(first 10 values)")
         print(pred[:10])
         #Getting the confusion matrix
         cm1 = confusion matrix(y test,pred)
         print('Confusion Matrix : \n', cm1)
         #prints [[TP,FN],[FP,TN]]
         total1=sum(sum(cm1))
         accuracy1=(cm1[0,0]+cm1[1,1])/total1
         print ('Accuracy : ', accuracy1)
         precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
         print ('Precision : ', precision1)
         recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
         print ('Recall : ', recall1)
         f_measure=(2*recall1*precision1)/(precision1+recall1)
         print('f measure: ', f_measure)
         sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
         print('Sensitivity : ', sensitivity1 )
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specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
        **********
            mean: 0.145670833333333333
        Y PREDICTION DATA after classificaion:(first 10 values)
        [1, 0, 0, 0, 1, 1, 0, 1, 1, 0]
        Confusion Matrix :
         [[179 247]
         [ 15 39]]
        Accuracy: 0.4541666666666666
        Precision: 0.9226804123711341
        Recall: 0.8211009174311926
        f measure: 0.8689320388349516
        Sensitivity: 0.42018779342723006
        Specificity: 0.72222222222222
        ## "4" Attributes.
In [13]:
        print("-----")
        rege1 = LinearRegression()
        regel.fit(x redwine 4 training , y train.ravel())
        x1 test pred = rege1.predict(x redwine 4 testing).round(3)
        ## HERE WE FIND OUT THE THRESHOLD ABOVE WHICH WE CAN CONSIDER THE QUALITY VAL
         ## "mid" IS THE VARIABLE CONTAINING THAT THRESHOLD VALUE....
        mid = np.mean(x1 test pred)
        print("mean: ", mid)
        pred= []
         #Classifing y prediction based on the mean value
        for p in x1 test pred:
            if p < mid:</pre>
                pred.append(0)
            elif p > mid:
                pred.append(1)
        print("Y PREDICTION DATA after classification:(first 10 values)")
        print(pred[:10])
        #Getting the confusion matrix
        cm1 = confusion_matrix(y_test,pred)
        print('Confusion Matrix : \n', cm1)
        #prints [[TP,FN],[FP,TN]]
        total1=sum(sum(cm1))
        accuracy1=(cm1[0,0]+cm1[1,1])/total1
        print ('Accuracy : ', accuracy1)
        precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
        print ('Precision : ', precision1)
         recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
         print ('Recall : ', recall1)
        f measure=(2*recall1*precision1)/(precision1+recall1)
```

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print('f measure: ', f_measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
       -----4 Attributes-----
       mean: 0.145683333333333333
       Y PREDICTION DATA after classificaion:(first 10 values)
       [1, 0, 0, 1, 1, 1, 1, 1, 1, 0]
       Confusion Matrix:
        [[174 252]
        [ 19 35]]
       Accuracy: 0.435416666666667
       Precision: 0.9015544041450777
       Recall: 0.8325358851674641
       f measure: 0.8656716417910447
       Sensitivity: 0.4084507042253521
       Specificity: 0.6481481481481481
       In [14]:
        ## Logistic Regression As Classifier
        ## "7" Attributes.
        print("-----")
        reg = LogisticRegression()
        reg.fit(x redwine 7 training , y train.ravel())
        pred = reg.predict(x redwine 7 testing)
        print("Y PREDICTION:(first 10 values)")
        print(pred[:10])
        cm1 = confusion matrix(y test,pred)
        print('Confusion Matrix : \n', cml)
        total1=sum(sum(cm1))
        accuracy1=(cm1[0,0]+cm1[1,1])/total1
        print ('Accuracy : ', accuracy1)
        precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
        print ('Precision : ', precision1)
        recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
        print ('Recall : ', recall1)
        f measure=(2*recall1*precision1)/(precision1+recall1)
        print('f measure: ', f_measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
       ********
         ------7 Attributes-----
       Y PREDICTION: (first 10 values)
       [0 0 0 0 0 0 0 0 0]
       Confusion Matrix :
```

[[419

71

```
[ 53
             1]]
       Accuracy: 0.875
       Precision: 0.8877118644067796
       Recall: 0.9976190476190476
       f measure: 0.9394618834080718
       Sensitivity: 0.9835680751173709
       Specificity: 0.018518518518517
       ## "4" Attributes.
In [15]:
        print("-----")
        rege = LogisticRegression()
        rege.fit(x redwine 4 training , y train.ravel())
        pred = rege.predict(x redwine 4 testing)
        print("Y PREDICTION:(first 10 values)")
        print(pred[:10])
        cm1 = confusion_matrix(y_test,pred)
        print('Confusion Matrix : \n', cm1)
        total1=sum(sum(cm1))
        accuracy1=(cm1[0,0]+cm1[1,1])/total1
        print ('Accuracy : ', accuracy1)
        precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
        print ('Precision : ', precision1)
        recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
        print ('Recall : ', recall1)
        f measure=(2*recall1*precision1)/(precision1+recall1)
        print('f measure: ', f measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
         Y PREDICTION: (first 10 values)
       [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
       Confusion Matrix :
        [[426 0]
        [ 54
             0]]
       Accuracy : 0.8875
       Precision: 0.8875
       Recall: 1.0
       f measure: 0.9403973509933775
       Sensitivity: 1.0
       Specificity: 0.0
In [16]:
       ## 3. SVM Classifier....
        ## "7" Attributes.
        print("-----")
        classi svm = svm.SVC(kernel="linear")
        classi_svm.fit(x_redwine_7_training , y_train.ravel())
        pred = classi svm.predict(x redwine 7 testing)
```

```
print("Y PREDICTION:(first 10 values)")
         print(pred[:10])
         cm1 = confusion matrix(y test,pred)
         print('Confusion Matrix : \n', cm1)
         total1=sum(sum(cm1))
         accuracy1=(cm1[0,0]+cm1[1,1])/total1
         print ('Accuracy : ', accuracy1)
         precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
         print ('Precision : ', precision1)
         recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
         print ('Recall : ', recall1)
         f measure=(2*recall1*precision1)/(precision1+recall1)
         print('f measure: ', f measure)
         sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
         print('Sensitivity : ', sensitivity1 )
         specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
         print('Specificity : ', specificity1)
        ********
        -----7 Attributes-----
        Y PREDICTION: (first 10 values)
        [0 0 0 0 0 0 0 0 0]
        Confusion Matrix :
         [[422
               41
         [ 53
              1]]
        Accuracy: 0.88125
        Precision: 0.888421052631579
        Recall: 0.9976359338061466
        f measure: 0.9398663697104677
        Sensitivity: 0.9906103286384976
        Specificity: 0.018518518518517
In [17]:
        ## "4" Attributes.
         print("-----")
         classi_svm = svm.SVC(kernel="linear")
         classi_svm.fit(x_redwine_4_training , y_train.ravel())
         pred = classi svm.predict(x redwine 4 testing)
         print("Y PREDICTION:(first 10 values)")
         print(pred[:10])
         cm1 = confusion_matrix(y_test,pred)
         print('Confusion Matrix : \n', cm1)
         total1=sum(sum(cm1))
         accuracy1=(cm1[0,0]+cm1[1,1])/total1
         print ('Accuracy : ', accuracy1)
         precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
         print ('Precision : ', precision1)
         recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
```

```
print ('Recall : ', recall1)
        f measure=(2*recall1*precision1)/(precision1+recall1)
        print('f measure: ', f_measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity: ', specificity1)
        Y PREDICTION: (first 10 values)
       [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
       Confusion Matrix :
        [[426
             01
        [ 54
             0]]
       Accuracy : 0.8875
       Precision: 0.8875
       Recall: 1.0
       f measure: 0.9403973509933775
       Sensitivity: 1.0
       Specificity: 0.0
       In [18]:
        ## 4. Naïve Bayesian....
        ## "7" Attributes...
        print("-----")
        Berno= BernoulliNB(binarize=True)
        Berno.fit(x redwine 7 training, y train.ravel())
        y pred = Berno.predict(x redwine 7 testing)
        ## Measures List....
        print("Y PREDICTION:(first 10 values)")
        print(y pred[:10])
        cm1 = confusion_matrix(y_test,y_pred)
        print('Confusion Matrix : \n', cm1)
        total1=sum(sum(cm1))
        accuracy1=(cm1[0,0]+cm1[1,1])/total1
        print ('Accuracy : ', accuracy1)
        precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
        print ('Precision : ', precision1)
        recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
        print ('Recall : ', recall1)
        f measure=(2*recall1*precision1)/(precision1+recall1)
        print('f measure: ', f measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
```

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**********
          Y PREDICTION: (first 10 values)
       [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
       Confusion Matrix :
        [[424 2]
        [ 53
             1]]
       Accuracy: 0.885416666666666
       Recall: 0.9976470588235294
       f measure: 0.9401330376940134
       Sensitivity: 0.9953051643192489
       Specificity: 0.018518518518517
       ## "4" Attributes
In [19]:
        print("-----")
        Berno= BernoulliNB(binarize=True)
        Berno.fit(x redwine 4 training, y train.ravel())
        y pred = Berno.predict(x redwine 4 testing)
        print("Y PREDICTION:(first 10 values)")
        print(y pred[:10])
        cm1 = confusion_matrix(y_test,y_pred)
        print('Confusion Matrix : \n', cm1)
        total1=sum(sum(cm1))
        accuracy1=(cm1[0,0]+cm1[1,1])/total1
        print ('Accuracy : ', accuracy1)
        precision1=(cm1[0,0])/(cm1[0,0]+cm1[1,0])
        print ('Precision : ', precision1)
        recall1=(cm1[0,0])/(cm1[0,0]+cm1[1,1])
        print ('Recall : ', recall1)
        f_measure=(2*recall1*precision1)/(precision1+recall1)
        print('f measure: ', f_measure)
        sensitivity1 = cm1[0,0]/(cm1[0,0]+cm1[0,1])
        print('Sensitivity : ', sensitivity1 )
        specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
        print('Specificity : ', specificity1)
        ------4 Attributes-----
       Y PREDICTION: (first 10 values)
       [\overline{0} \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
       Confusion Matrix:
        [[426
             01
        [ 54
             0]]
       Accuracy: 0.8875
       Precision: 0.8875
       Recall: 1.0
       f measure: 0.9403973509933775
       Sensitivity: 1.0
       Specificity: 0.0
In [ ]:
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In []: