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##Linear Regression Classifier
In [1]:
         #importing all the packages
         import numpy as np
         import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         from statsmodels.stats.outliers influence import variance inflation factor
         from sklearn.model selection import train test split
         from sklearn.linear model import LinearRegression
         from sklearn.metrics import mean squared error,r2 score
         from sklearn.metrics import confusion matrix
         from sklearn import svm
         from sklearn import metrics
         from sklearn.linear model import LogisticRegression
         from sklearn.naive_bayes import BernoulliNB
         from sklearn.naive_bayes import GaussianNB
         from sklearn.naive bayes import MultinomialNB
         import os
         #getting the directory
         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')
         #declaring features
         fields=["fixed acidity","volatile acidity","citric acid","residual sugar","ch
         fields1=["fixed acidity","volatile acidity","citric acid","residual sugar","
In [2]:
        #X TRAINING DATASET
         train wine = pd.read csv(Dataset.train, delimiter=';', header=None, skiprows=
         x train=train wine[fields1]
         print("X TRAINING DATA:")
         x_train.head(5)
        X TRAINING DATA:
Out[2]:
                                                 free
                                                        total
            fixed volatile citric residual
                                      chlorides
                                                sulfur
                                                       sulfur
                                                             density
                                                                     pH sulphates alcohol
           acidity
                 acidity
                         acid
                                sugar
                                              dioxide dioxide
        0
             10.6
                    0.28
                         0.39
                                 15.5
                                         0.069
                                                  6.0
                                                        23.0
                                                              1.0026
                                                                    3.12
                                                                             0.66
                                                                                     9.2
        1
             9.4
                    0.30
                         0.56
                                         0.080
                                                  6.0
                                                        17.0
                                                             0.9964
                                                                    3.15
                                                                             0.92
                                                                                     11.7
                                  2.8
        2
             10.6
                    0.36
                         0.59
                                  2.2
                                         0.152
                                                  6.0
                                                        18.0
                                                             0.9986 3.04
                                                                             1.05
                                                                                     9.4
        3
             10.6
                    0.36
                                                  7.0
                                                        18.0
                                                              0.9986 3.04
                                                                             1.06
                         0.60
                                  2.2
                                         0.152
                                                                                     9.4
        4
             10.6
                    0.44
                         0.68
                                  4.1
                                                  6.0
                                                        24.0
                                                                             0.66
                                         0.114
                                                              0.9970 3.06
                                                                                    13.4
         In [3]:
         # CLASSIFING quality 0-bad AND 1-good and forming Y TRAINING DATASET
         y train = train wine["quality"].apply(lambda q:0 if q<7 else 1)
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```
print("Y_TRAINING DATA:")
          y_train.head(5)
         ************************************
         *********
         Y TRAINING DATA:
         0
              0
 Out[3]:
         1
              1
         2
              0
         3
              0
         4
              0
         Name: quality, dtype: int64
         #similarly for testing data
 In [4]:
          #X TESTING DATASET
          test wine = pd.read csv(Dataset.test, delimiter=';', header=None, skiprows=1,
          x test=test wine[fields1]
          print("X TESTING DATA:")
          x_test.head(5)
         X TESTING DATA:
                                                 free
                                                       total
 Out[4]:
             fixed volatile
                         citric
                              residual
                                      chlorides
                                                sulfur
                                                       sulfur
                                                            density
                                                                    pH sulphates alcohol
            acidity
                  acidity
                         acid
                                sugar
                                              dioxide dioxide
         0
              7.4
                    0.70
                          0.00
                                  1.9
                                         0.076
                                                 11.0
                                                         34
                                                             0.9978
                                                                   3.51
                                                                            0.56
                                                                                    9.4
         1
              7.8
                    0.88
                          0.00
                                  2.6
                                         0.098
                                                 25.0
                                                         67
                                                             0.9968
                                                                   3.20
                                                                            0.68
                                                                                    9.8
         2
              7.8
                    0.76
                         0.04
                                  2.3
                                         0.092
                                                 15.0
                                                         54
                                                             0.9970 3.26
                                                                            0.65
                                                                                    9.8
         3
             11.2
                    0.28
                                         0.075
                                                             0.9980 3.16
                                                                            0.58
                         0.56
                                  1.9
                                                 17.0
                                                         60
                                                                                    9.8
         4
              7.4
                    0.70
                         0.00
                                  1.9
                                         0.076
                                                 11.0
                                                         34
                                                             0.9978 3.51
                                                                            0.56
                                                                                    9.4
                                                                                    •
 In [5]:
          #Y TESTING DATASET
          y_test = test_wine["quality"].apply(lambda q:0 if q<7 else 1)</pre>
          print("Y TESTING DATA:")
          y_test.head(10)
         **********
         Y_TESTING DATA:
         0
              0
 Out[5]:
         1
              0
         2
              0
         3
              0
         4
              0
         5
         6
              0
         7
              1
         8
              1
         Name: quality, dtype: int64
         #declaring linear regression model
In [17]:
          l_r= LinearRegression()
          # training the model with X TRAINING AND Y TRAINING
          l_r.fit(x_train, y_train)
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#Getting Y_PREDICTION
          pred1=l_r.predict(x_test)
          print("Y PREDICTION DATA:(first 10 values)")
          pred1[:10]
         **********
         Y PREDICTION DATA: (first 10 values)
\begin{array}{c} \text{Out} \texttt{[17]: array}(\texttt{[-0.12480267, -0.04486957, -0.02591986, 0.17223449, -0.12480267, -0.12215259, -0.07652599, 0.01531073, -0.01059442, 0.22524849])} \end{array}
         #MEAN OF THE Y PREDICTION
In [18]:
          mid = np.mean(pred1)
          print("mean: ", mid)
         mean: 0.10580178551233062
In [19]:
          pred= []
          #Classifing y prediction based on the mean value
          for p in pred1 :
              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])
         Y PREDICTION DATA after classificaion:(first 10 values)
         [0, 0, 0, 1, 0, 0, 0, 0, 0, 1]
         #Getting the confusion matrix
In [20]:
          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 )
          specificity1 = cm1[1,1]/(cm1[1,0]+cm1[1,1])
          print('Specificity : ', specificity1)
         Confusion Matrix:
           [[280 146]
           [ 3 51]]
```

Accuracy: 0.68958333333333333 Precision: 0.9893992932862191 Recall: 0.8459214501510574 f measure: 0.9120521172638437 Sensitivity: 0.6572769953051644 ##Logistic Regression Classifier In [9]: #declaring logistic regression model logisticRegr = LogisticRegression() #training the model with X TRAINING AND Y TRAINING logisticRegr.fit(x train,y train) **#Y PREDICTION** pred=logisticRegr.predict(x test) print("Y PREDICTION:(first 10 values)") print(pred[:10]) ******** Y PREDICTION: (first 10 values) $[0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$ /home/thrishik/.local/lib/python3.6/site-packages/sklearn/linear model/ logis tic.py:764: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT. Increase the number of iterations (max iter) or scale the data as shown in: https://scikit-learn.org/stable/modules/preprocessing.html Please also refer to the documentation for alternative solver options: https://scikit-learn.org/stable/modules/linear model.html#logistic-regres extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG) In [10]: cm1 = confusion_matrix(y_test,pred) print('Confusion Matrix : \n', cm1) total1=sum(sum(cm1)) accuracy1=(cm1[0,0]+cm1[1,1])/total1print ('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) Confusion Matrix: [[414 12] [43 11]] Accuracy: 0.885416666666666 Precision: 0.9059080962800875 Recall: 0.9741176470588235

f measure: 0.9387755102040816

Sensitivity: 0.971830985915493 Specificity: 0.2037037037037037

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In [11]:
        ##SVM Classifier
        #declaring svm model
        cls=svm.SVC(kernel="linear")
        #training the model with X TRAINING AND Y TRAINING
        cls.fit(x train,y train)
        #Y PREDICTION
        pred=cls.predict(x test)
        print("Y PREDICTION:(first 10 values)")
        print(pred[:10])
       **********
       Y PREDICTION: (first 10 values)
       [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
In [21]:
        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)
       Confusion Matrix :
        [[280 146]
        [ 3 51]]
       Accuracy: 0.6895833333333333
       Precision: 0.9893992932862191
       Recall: 0.8459214501510574
       f measure: 0.9120521172638437
       Sensitivity: 0.6572769953051644
       In [22]: | ##Naive Bayes Classifier
        #declaring Naive Bayes model
        BernNB= BernoulliNB(binarize=True)
        #training the model with X TRAINING AND Y TRAINING
        BernNB.fit(x train,y train)
        print(BernNB)
```

```
#Y PREDICTION
         y_pred=BernNB.predict(x_test)
         print("Y PREDICTION:(first 10 values)")
         print(y_pred[:10])
        **********
        BernoulliNB(binarize=True)
        Y PREDICTION: (first 10 values)
        [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]
In [23]:
        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)
        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 [ ]:
```