Assignment - 1 Big Data Analytics (CSOE17) REPORT

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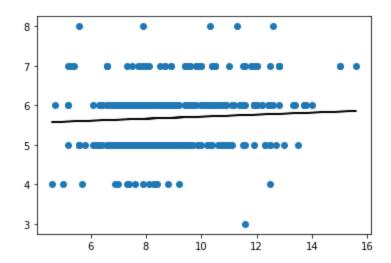
TASK 1

The Quality attributes of a wine given in a dataset, are estimated through linear regression. The training dataset are used to train the linear regression model and Tested with testing data, the output which we get is the predicted data. I have formed a Graphical representation of a line fitted over the testing data considering for a single feature (Fixed acidity).

Intercept: 5.457538106303752

Slope: [0.02557288]

fixed acidity x_test vs y_pred



Considering multi variable linear regression, the sum of squared errors are calculated for the test dataset. The output is attached below. The line is found to have a slope of 0.0255 and its Y-intercept is around 5.458.

Mean Absolute Error: 0.5224781243375761
Mean Squared Error: 0.4463021333444595

Root Mean Squared Error: 0.6680584804824047

TASK 2

In the given dataset, we treat quality to be the class attribute and by taking those tuples with quality value greater than 7 to be "good" and others "bad", we now handle it as a two-class problem.

Out[2]:												
Jul[2]:		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
	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	2.8	0.080	6.0	17.0	0.9964	3.15	0.92	11.7
	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	0.60	2.2	0.152	7.0	18.0	0.9986	3.04	1.06	9.4
In [3]:	# (CLASSIFING		bad AND .	1-good and 1	forming \	6.0 Y TRAINING DATA if q<7 else 1)	24.0 ************************************			0.66	13.4
In [3]:	pri # (y_1	nt("***** CLASSIFING rain = tr	********** i quality 0- ain_wine["q	******* bad AND uality"]	**************************************	******** forming	**************************************	*******				
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As seen here, the quality attribute of training data has been converted to a 2-class problem with 0 and 1 being the 2 classes. Similarly we classify the quality attribute for testing data as well.

We apply the 4 types of classifier algorithms and observe the outcome.

We must now create classification algorithms and train them with training data and apply them on the testing data to predict the classes of each tuple. By comparing these predicted values and actual outputs, we can evaluate the effectiveness of our algorithms.

(i)LINEAR REGRESSION CLASSIFIER:

We first apply the Linear Regression based classifier on the test data, and then classify the prediction data based on the mean.

(ii)LOGISTIC REGRESSION CLASSIFIER:

```
Y_PREDICTION: (first 10 values)
[0 0 0 0 0 0 0 0 0]
```

(iii) SVM CLASSIFIER:

```
Y_PREDICTION: (first 10 values)
[0 0 0 0 0 0 0 0 0]
```

(iv) NAÏVE BAYES CLASSIFIER:

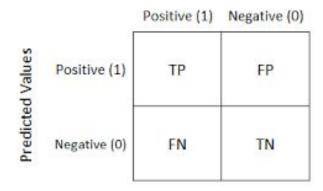
```
BernoulliNB(binarize=True)
Y_PREDICTION:(first 10 values)
[0 0 0 0 0 0 0 0 0]
```

These are the predicted outcomes of the first 10 tuples generated by the 4 different classifier algorithms.

TASK 3

We must compare these predicted outcomes to the actual outcomes and determine the performance measures of the 4 algorithms, namely the accuracy, precision, recall, f measure, sensitivity and specificity of each algorithm, derived from the confusion matrix.

Actual Values



We derive the mentioned parameters from the confusion matrix given.

PRECISION(P) = TP/(TP+FP)

RECALL(R) = TP/(TP+FN)

F-MEASURE = 2*P*R/(P+R)

ACCURACY = (TP+TN)/(TP+TN+FP+FN)

SENSITIVITY = TP/(TP+FN)

SPECIFICITY = TN/(TN+FP)

We look at these measures to determine how effective the classifier is.

PERFORMANCE MEASURES

(i) PERFORMANCE OF LINEAR REGRESSION CLASSIFIER:

```
Confusion Matrix :
  [[280 146]
  [ 3 51]]
Accuracy : 0.68958333333333333
Precision : 0.9893992932862191
Recall : 0.8459214501510574
```

(ii) PERFORMANCE OF LOGISTIC REGRESSION CLASSIFIER:

Confusion Matrix :
 [[414 12]
 [43 11]]
Accuracy : 0.885416666666666
Precision : 0.9059080962800875
Recall : 0.9741176470588235
f measure: 0.9387755102040816
Sensitivity : 0.971830985915493
Specificity : 0.2037037037037

(iii) PERFORMANCE OF SVM CLASSIFIER:

```
Confusion Matrix:
[[280 146]
[ 3 51]]
Accuracy: 0.6895833333333333
Precision: 0.9893992932862191
Recall: 0.8459214501510574
f measure: 0.9120521172638437
Sensitivity: 0.6572769953051644
Specificity: 0.944444444444444
```

(iv) PERFORMANCE OF NAÏVE BAYES CLASSIFIER:

```
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
```

These determine the performance of each classifier in predicting the outcome of the given dataset tested for quality.

TASK 4

PCA is dimensionality-reduction method used to reduce dimension of larger dataset, by converting larger dataset into smaller dataset but that still contains almost all the information.

We need to standardize the data, then find a covariance matrix to compute eigenvectors and eigenvalues of the covariance matrix to identify principal components. And we implement pca by using library sklearn which has inbuilt functions to apply PCA to the dataset.

PCA for 7 attributes

For Linear Regression Classifier

specificity is: 0.72222222222222 For Logistic Regression Classifier

Accuracy is: 0.875

Precision is: 0.8877118644067796 Recall is: 0.9976190476190476 F-measure is: 0.9394618834080718 sensitivity is: 0.9835680751173709 specificity is: 0.018518518518517

For SVM Classifier

Accuracy is: 0.88125

Precision is: 0.888421052631579
Recall is: 0.9976359338061466
F-measure is: 0.9398663697104677
sensitivity is: 0.9906103286384976
specificity is: 0.018518518518518517

For Naive Bayesian Classifier

PCA for 4 attributes

For Linear Regression Classifier

Accuracy is: 0.435416666666667
Precision is: 0.9015544041450777
Recall is: 0.8325358851674641
F-measure is: 0.8656716417910447
sensitivity is: 0.4084507042253521
specificity is: 0.6481481481481481

For Logistic Regression Classifier

Accuracy is: 0.8875 Precision is: 0.8875

Recall is: 1.0

F-measure is: 0.9403973509933775

sensitivity is: 1.0 specificity is: 0.0 For SVM Classifier Accuracy is: 0.8875 Precision is: 0.8875

Recall is: 1.0

F-measure is: 0.9403973509933775

sensitivity is: 1.0 specificity is: 0.0

For Naive Bayesian Classifier

Accuracy is: 0.8875 Precision is: 0.8875

Recall is: 1.0

F-measure is: 0.9403973509933775

sensitivity is: 1.0 specificity is: 0.0

And as we observe,

Accuracy has reduced compared with dataset passed to modal without PCA, and also PCA with 7 components has more accuracy compared with PCA of 4 components. But not very large differences are observed.