

EX.NO:29

Date:

EVALUATING ACCURACY OF THE CLASSIFIERS

AIM:

To create evaluating accuracy of the classifiers using weka tool.

DESCRIPTION:

Consider the german credit dataset which can be downloaded from the UCI repository.

PROCEDURE:

- 1.Download WEKA And Install
- 2.Start WEKA
- 3.Open The Data/iris.arff Dataset
- 4.Select And Run An Algorithm
- 5.Review The Results

ANALYSIS :

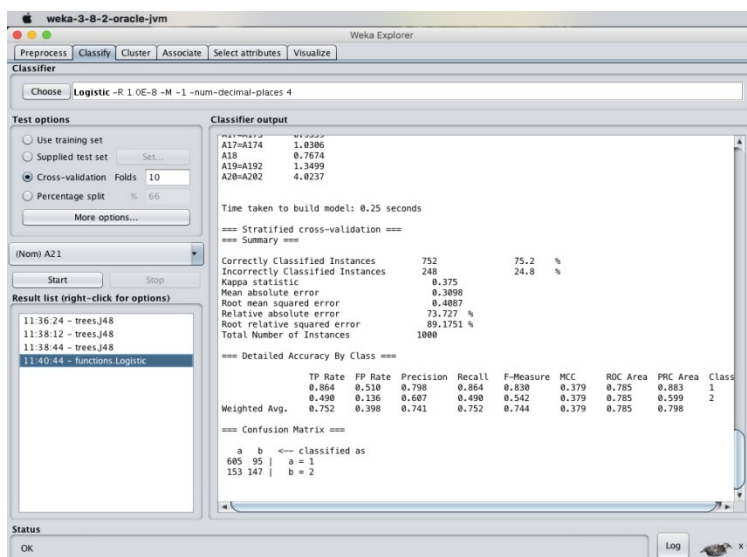
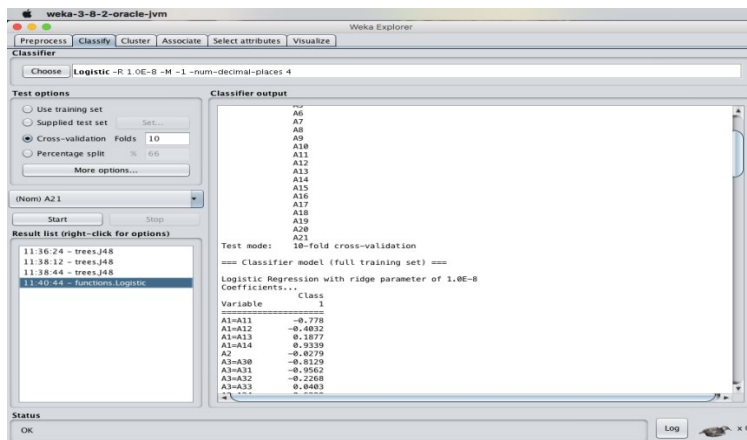
A) Logistic Regression :

Logistic regression predicts the probability of an outcome that can only have two values (i.e. a dichotomy). The prediction is based on the use of one or several predictors (numerical and categorical).

Steps :

- Load the dataset into the weka tool and preprocess it.
- Apply the classification the logistic regression technique and execute for the result.

Output :



B) Naïve Bayes Algorithm :

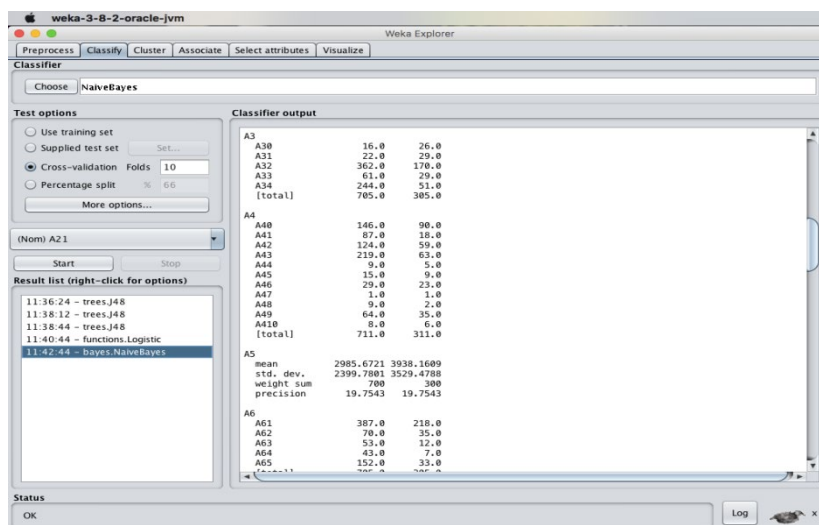
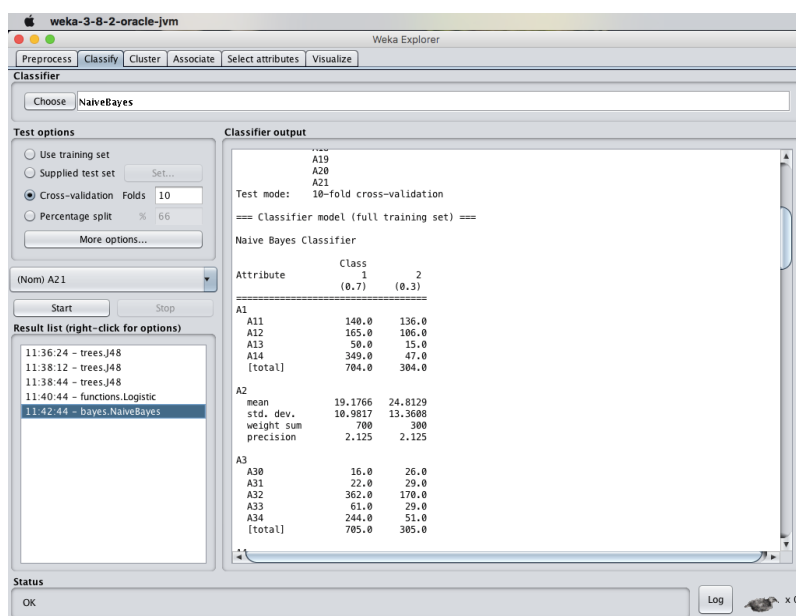
The Naive Bayesian classifier is based on Bayes' theorem with the independence assumptions between predictors. A Naive Bayesian model is easy to build, with no complicated iterative parameter estimation which makes it particularly useful for very large datasets. Despite its simplicity, the

Naive Bayesian classifier often does surprisingly well and is widely used because it often outperforms more sophisticated classification methods.

Steps :

- Load the dataset into the weka tool and preprocess it.
- Apply the classification the Naïve bayes technique and execute for the result.

Output :



weka-3-8-2-oracle-jvm Weka Explorer

Preprocess Classify Cluster Associate Select attributes Visualize

Classifier Choose NaiveBayes

Test options

☐ Use training set
☐ Supplied test set Set...
☒ Cross-validation Folds 10
☐ Percentage split % 66
 More options...

(Nom) A21

Start Stop

Result list (right-click for options)

- 11:36:24 - trees.J48
- 11:38:12 - trees.J48
- 11:38:44 - trees.J48
- 11:40:44 - functions.Logistic
- 11:42:44 - bayes.NaiveBayes

Classifier output

A6

A61	387.0	218.0
A62	70.0	35.0
A63	53.0	12.0
A64	43.0	7.0
A65	152.0	33.0
[total]	705.0	305.0

A7

A71	40.0	24.0
A72	103.0	71.0
A73	236.0	105.0
A74	136.0	40.0
A75	190.0	65.0
[total]	705.0	305.0

A8

mean	2.92	3.0967
std. dev.	1.1273	1.0866
weight sum	700	300
precision	1	1

A9

A91	31.0	21.0
A92	202.0	110.0
A93	403.0	147.0
A94	68.0	26.0
A95	1.0	1.0
[total]	705.0	305.0

A10

A101	636.0	273.0
A102	24.0	19.0
A103	45.0	11.0
[total]	705.0	305.0

Status OK Log x 0

weka-3-8-2-oracle-jvm Weka Explorer

Preprocess Classify Cluster Associate Select attributes Visualize

Classifier Choose NaiveBayes

Test options

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Classifier output

A12

A121	223.0	61.0
A122	162.0	72.0
A123	231.0	103.0
A124	88.0	68.0
[total]	704.0	304.0

A13

mean	36.1723	33.9267
std. dev.	11.4005	11.259
weight sum	700	300
precision	1.0769	1.0769

A14

A141	83.0	58.0
A142	29.0	20.0
A143	591.0	225.0
[total]	703.0	303.0

A15

A151	110.0	71.0
A152	528.0	187.0
A153	65.0	45.0
[total]	703.0	303.0

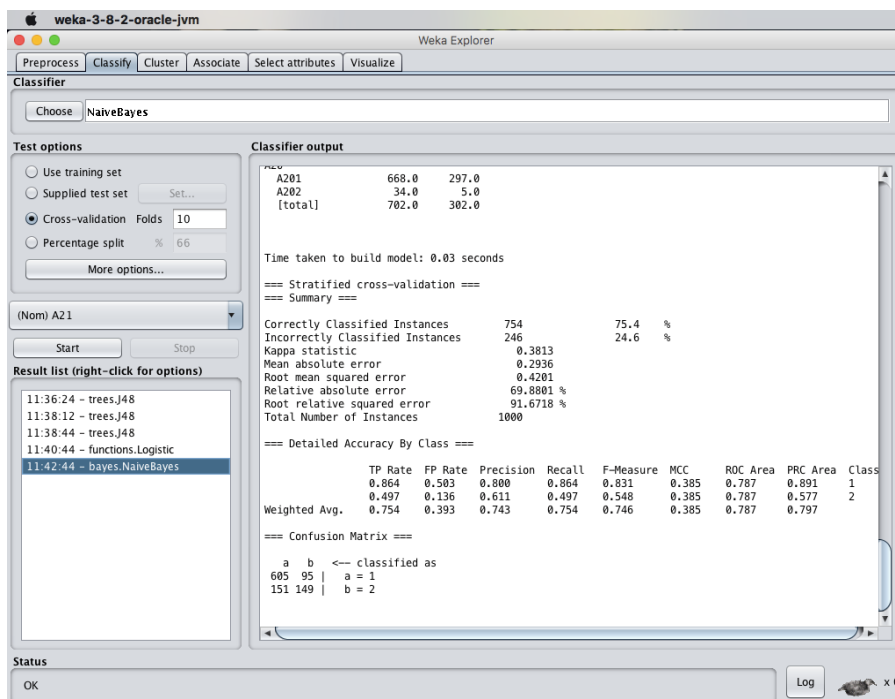
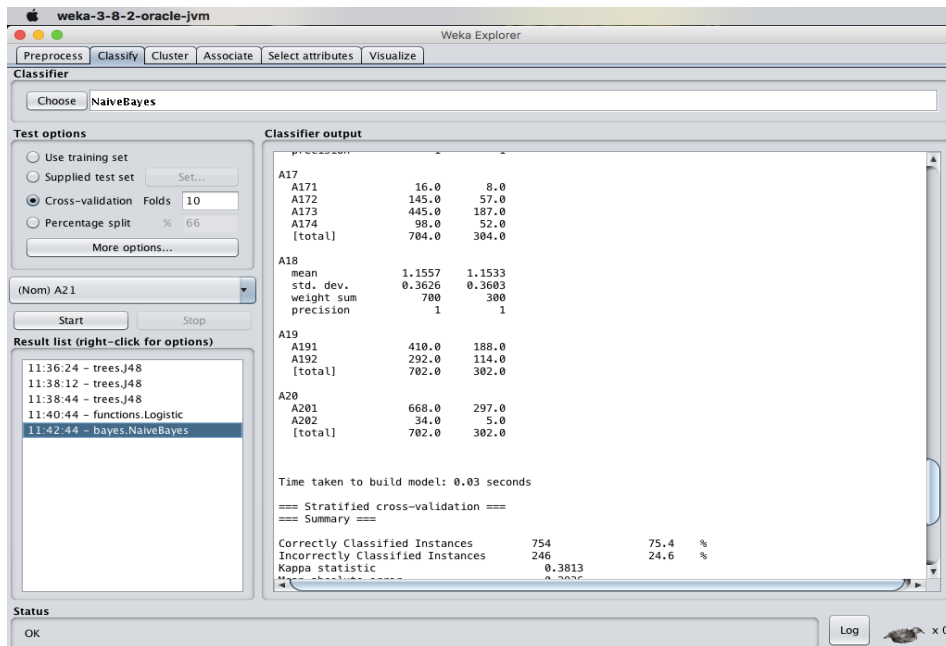
A16

mean	1.4243	1.3667
std. dev.	0.5843	0.5588
weight sum	700	300
precision	1	1

A17

A171	16.0	8.0
A172	145.0	57.0
[total]	161.0	65.0

Status OK Log x 0



C) J48 Algorithm :

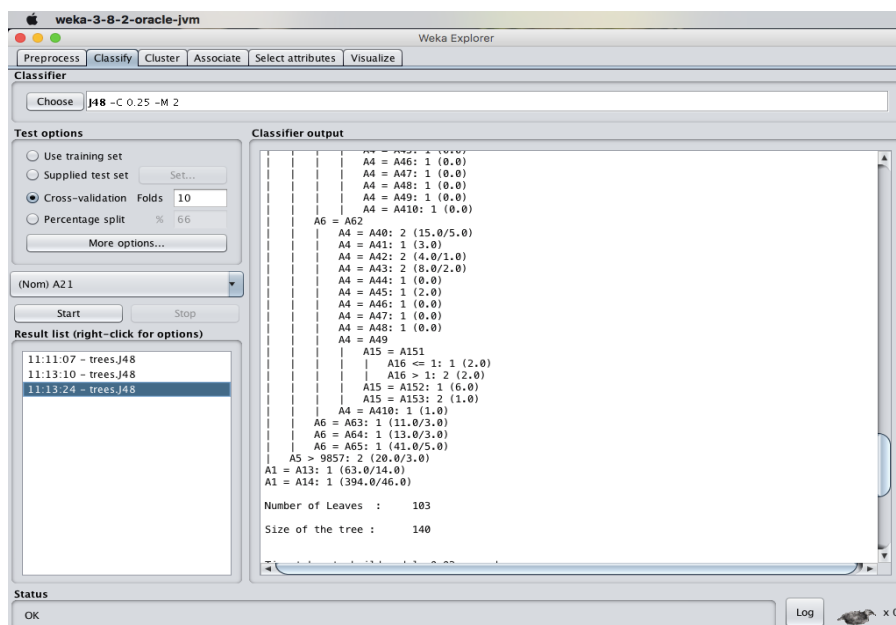
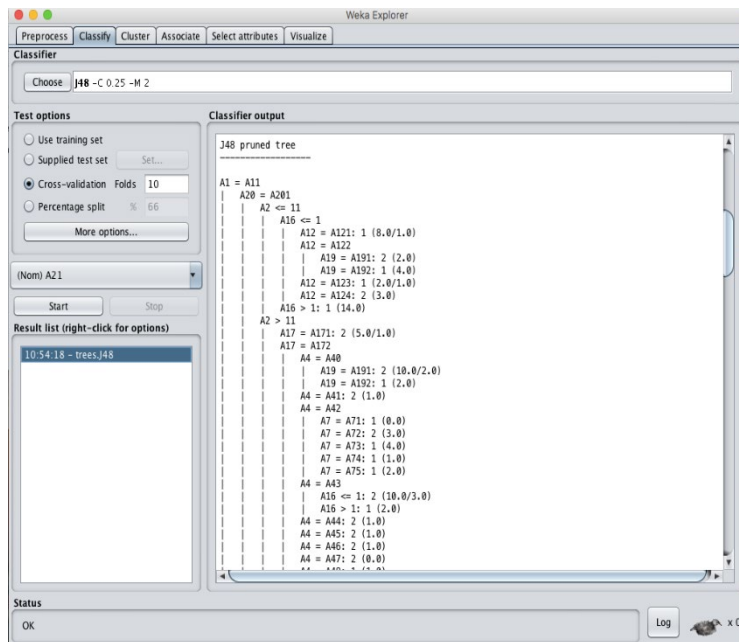
Classification is the process of building a model of classes from a set of records that contain class labels. Decision Tree Algorithm is to find out the way the attributes-vector behaves for a number of instances. Also on the bases of the training instances the classes for the newly generated instances are being found. This algorithm generates the rules for the prediction of the

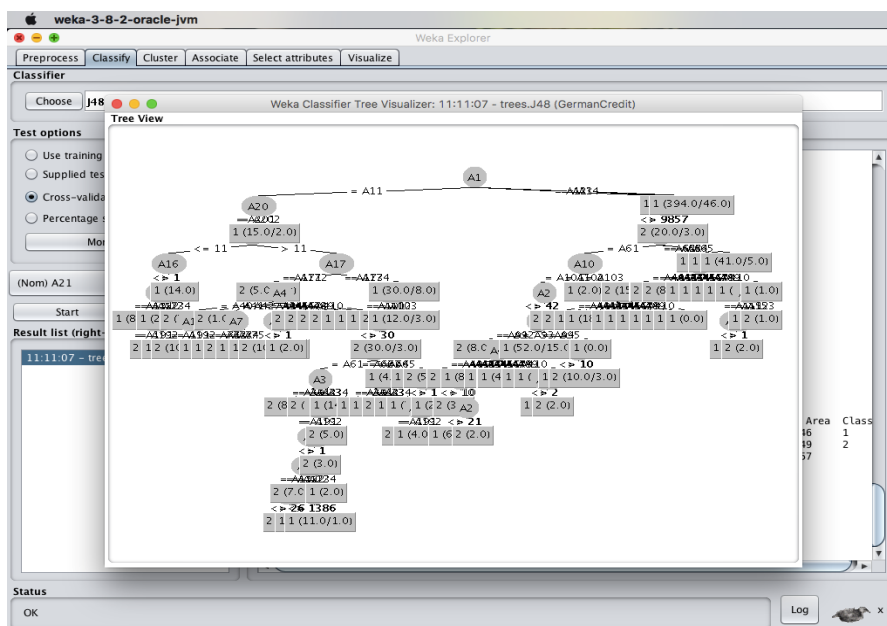
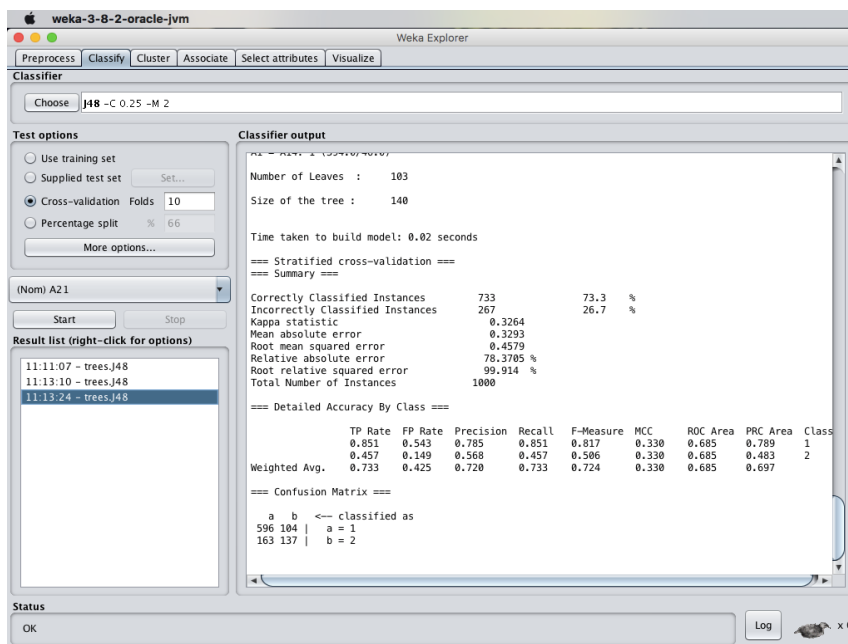
target variable. With the help of tree classification algorithm the critical distribution of the data is easily understandable.

Steps :

- Load the dataset into the weka tool and preprocess it.
- Apply the classification the J48 technique and execute for the result.

Output :





D) K-Nearest Neighbor :

K-Nearest Neighbors is one of the most basic yet essential classification algorithms in Machine Learning. It belongs to the supervised learning domain and finds intense application in pattern recognition, data mining and intrusion detection.

It is widely disposable in real-life scenarios since it is non-parametric, meaning, it does not make any underlying assumptions about the distribution of data (as opposed to other algorithms such as GMM, which assume a Gaussian distribution of the given data).

We are given some prior data (also called training data), which classifies coordinates into groups identified by an attribute.

Steps :

- Load the dataset into the weka tool and preprocess it.
- Apply the classification the K- Nearest Neighbor technique and execute for the result.

Output :

The screenshot shows the Weka Explorer interface. The 'Classifier' tab is selected, and the 'IBk' classifier is chosen. The 'Test options' section shows 'Cross-validation' with 'Folds' set to 10. The 'Classifier output' pane displays the following results:

```
=== Classifier model (full training set) ===
IB1 instance-based classifier
using 1 nearest neighbour(s) for classification

Time taken to build model: 0 seconds

=== Stratified cross-validation ===
=== Summary ===
Correctly Classified Instances      720      72 %
Incorrectly Classified Instances    280      28 %
Kappa statistic                    0.3243
Mean absolute error                 0.2885
Root mean squared error             0.5286
Relative absolute error             66.7546 %
Root relative squared error        115.3422 %
Total Number of Instances         1000

=== Detailed Accuracy By Class ===
               TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
               0.810    0.490    0.794     0.810    0.802     0.325    0.660    0.776     1
               0.510    0.190    0.535     0.510    0.522     0.325    0.660    0.420     2
Weighted Avg.   0.720    0.400    0.716     0.720    0.718     0.325    0.660    0.669

=== Confusion Matrix ===
  a    b  <-- classified as
567 133 | a = 1
147 153 | b = 2
```

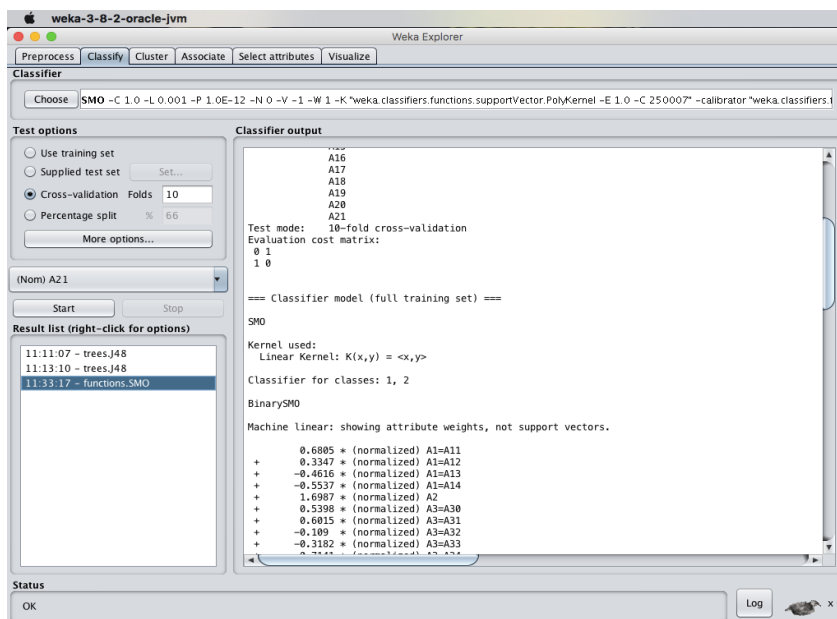
The 'Result list' on the left shows a single entry: '02:47:47 - lazy.IBk'.

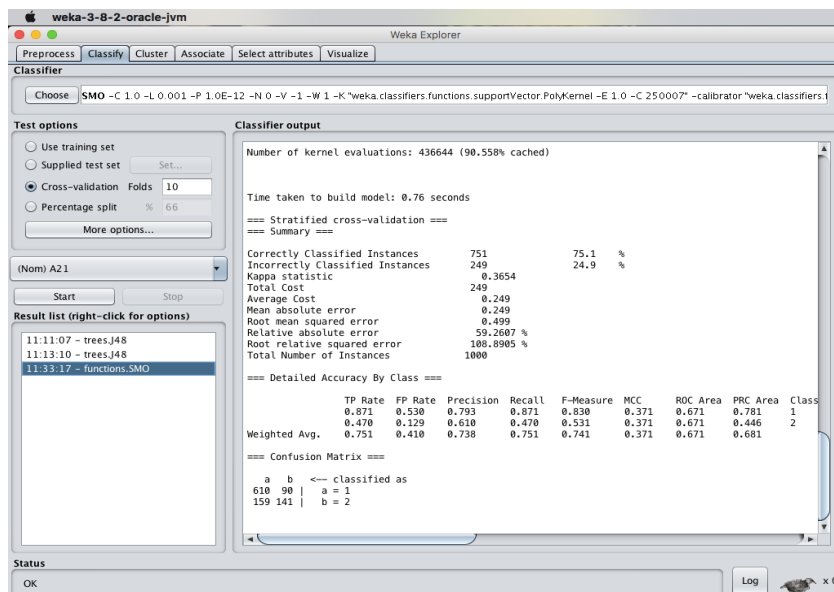
E) SMO Algorithm :

The iterative algorithm Sequential Minimal Optimization (SMO) is used for solving quadratic programming (QP) problems. One example where QP problems are relevant is during the training process of support vector machines (SVM). The SMO algorithm is used to solve in this example a constraint optimization problem. John Platt proposed this algorithm in 1998 and it was successfully used since then. We describe here the basics of the algorithm in the light of big data.

Steps :

- Load the dataset into the weka tool and preprocess it.
- Apply the classification the Sequential Minimal Optimization (SMO) technique and execute for the result.





RESULT :

Thus, the comparison of the confusion matrix for all the methods and techniques. Out of the comparing matrix with all the techniques there is a change in instances. Naïve bayes has more number of correct instances than other but when compared to time K-nearest neighbor is best. The above graphs will show the variations of values in the parameters.