

Week – 5 Demonstrate performing classification on data sets Load each dataset into Weka and run 1d3, J48 classification algorithm. Study the classifier output. Compute entropy values, Kappa statistic.

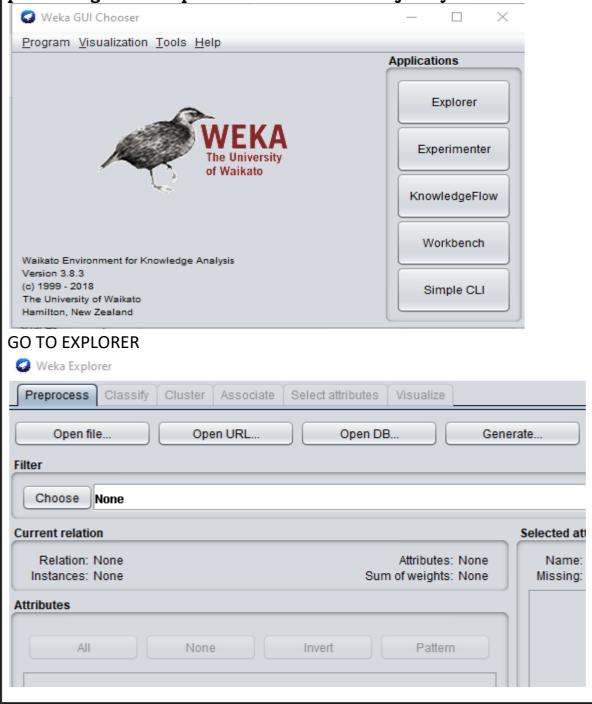
Extract if-then rules from the decision tree generated by the classifier, Observe the confusion matrix.

Load each dataset into Weka and perform Naïve-bayes classification and k-Nearest Neighbour classification.

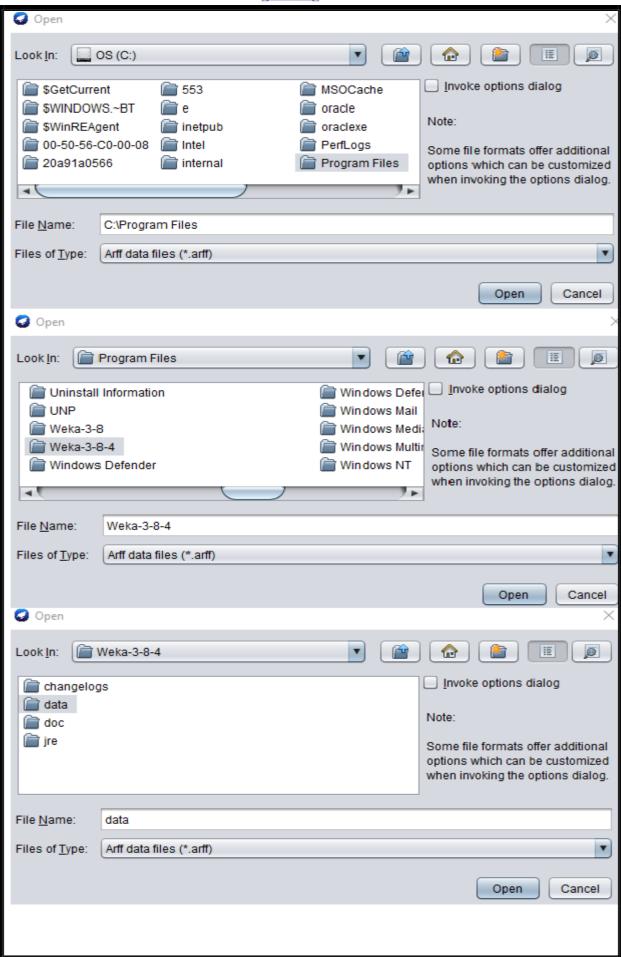
Interpret the results

obtained. Plot RoC Curves

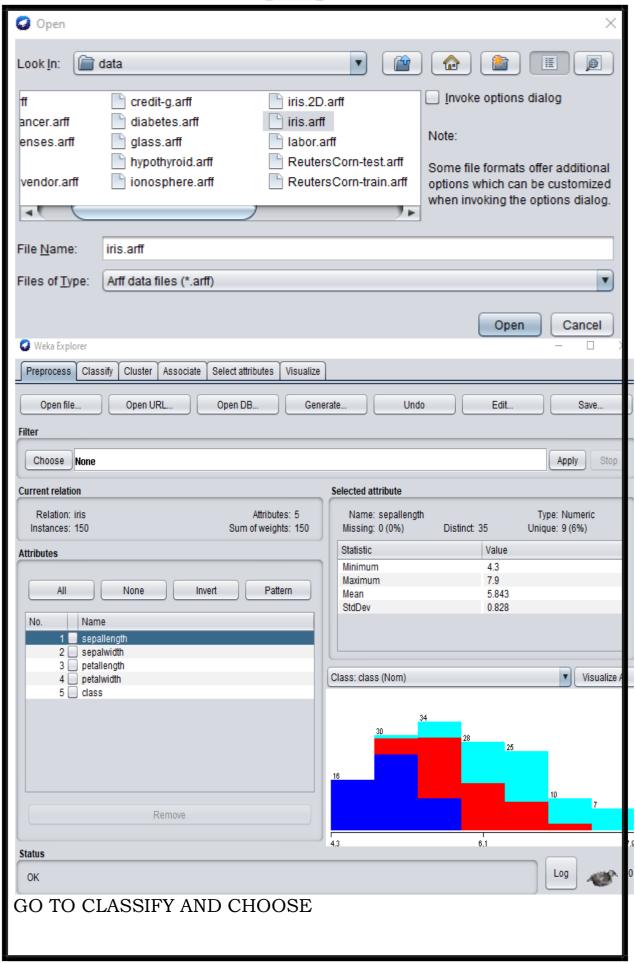
Compare classification results of ID3, J48, Naïve-Bayes and k-NN classifiers for each dataset, and deduce which classifier is performing best and poor for each dataset and justify.

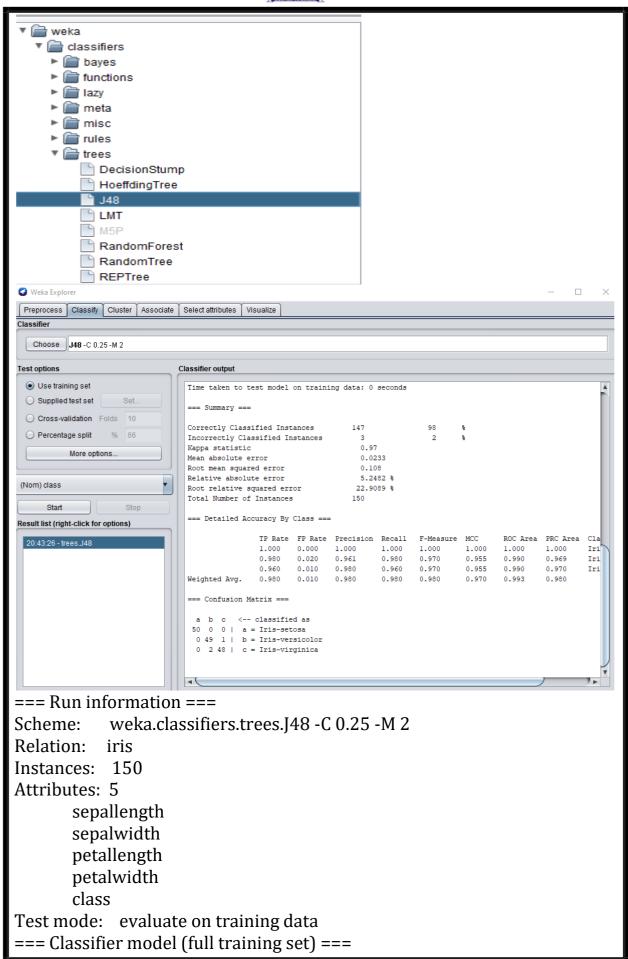




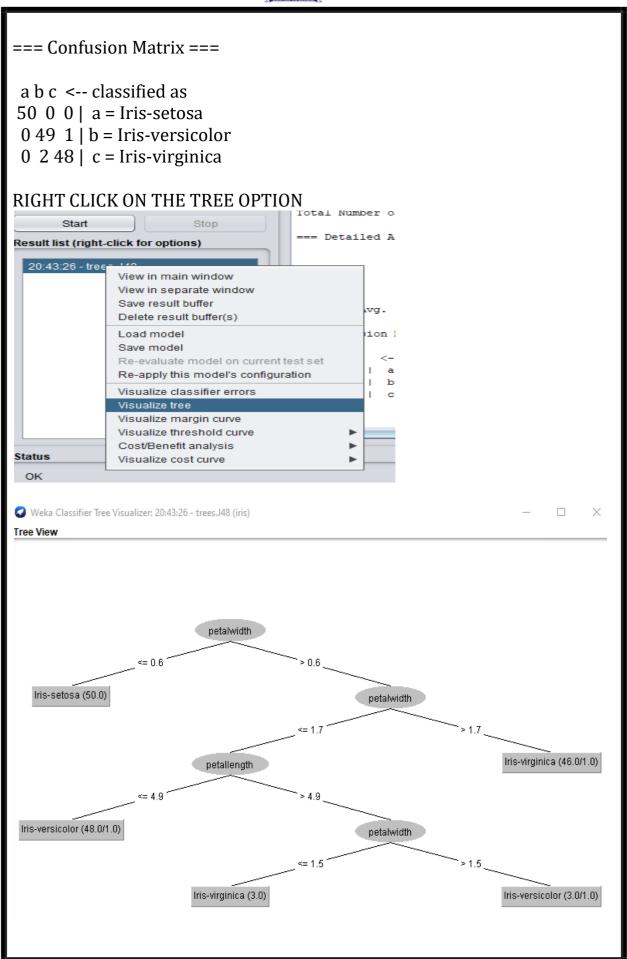








```
J48 pruned tree
petalwidth <= 0.6: Iris-setosa (50.0)
petalwidth > 0.6
| petalwidth <= 1.7
\mid petallength <= 4.9: Iris-versicolor (48.0/1.0)
| | petallength > 4.9
| | petalwidth <= 1.5: Iris-virginica (3.0)
| \ | \ | petalwidth > 1.5: Iris-versicolor (3.0/1.0)
petalwidth > 1.7: Iris-virginica (46.0/1.0)
Number of Leaves:
                       5
Size of the tree: 9
Time taken to build model: 0.01 seconds
=== Evaluation on training set ===
Time taken to test model on training data: 0 seconds
=== Summary ===
Correctly Classified Instances
                                            98
                                                 %
                                 147
                                                %
Incorrectly Classified Instances
                                  3
                                            2
Kappa statistic
                           0.97
Mean absolute error
                              0.0233
Root mean squared error
                                 0.108
Relative absolute error
                               5.2482 %
Root relative squared error
                                 22.9089 %
Total Number of Instances
                                150
=== Detailed Accuracy By Class ===
                                                             ROC Area
        TP Rate FP Rate Precision Recall F-Measure MCC
PRC Area Class
        1.000 0.000 1.000
                                1.000
                                                       1.000
                                                               1.000
                                      1.000
                                                1.000
Iris-setosa
        0.980
               0.020 0.961
                               0.980 0.970
                                               0.955
                                                      0.990
                                                               0.969
Iris-versicolor
        0.960 0.010 0.980
                                0.960 0.970
                                                0.955
                                                      0.990
                                                               0.970
Iris-virginica
Weighted Avg.
                0.980 0.010 0.980
                                       0.980
                                              0.980
                                                       0.970
                                                               0.993
0.980
```





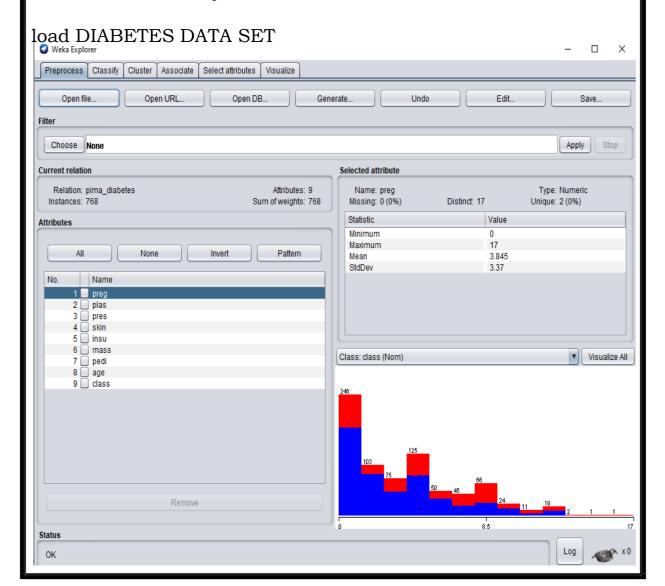
The kappa statistic, which takes into account chance agreement, is defined as **(observed agreement-expected agreement)/(1-expected agreement)**.

Mean Absolute Error calculates the average difference between the calculated values and actual values. It is also known as scale- dependent accuracy as it calculates error in observations taken on the same scale. It is used as evaluation metrics for regression models in machine learning. It calculates errors between actual values and valuespredicted by the model. It is used to predict the accuracy of the machine learning model.

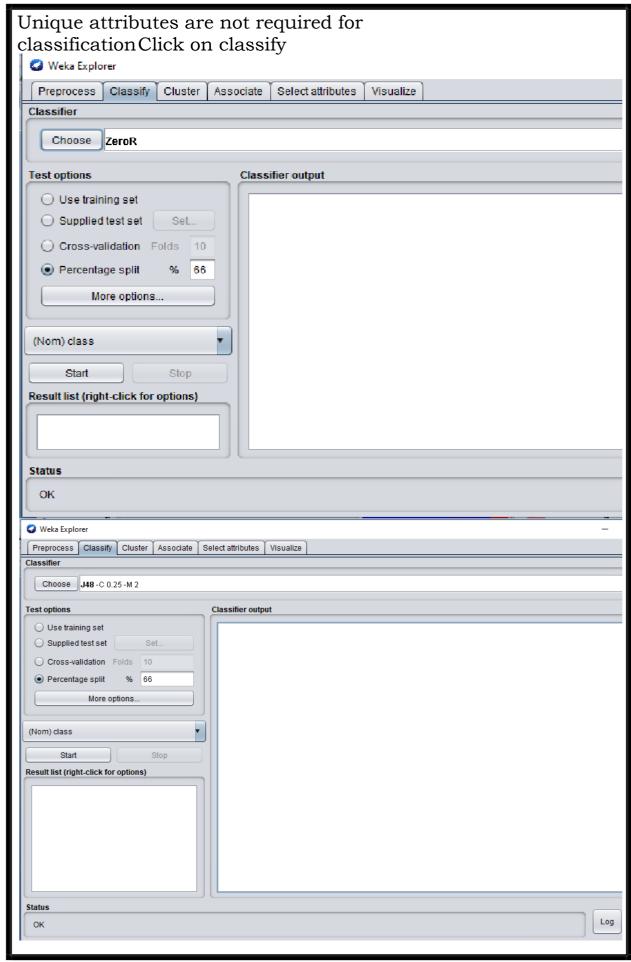
Formula:

Mean Absolute Error = $(1/n) * \sum |y_i - x_i|$ *where*,

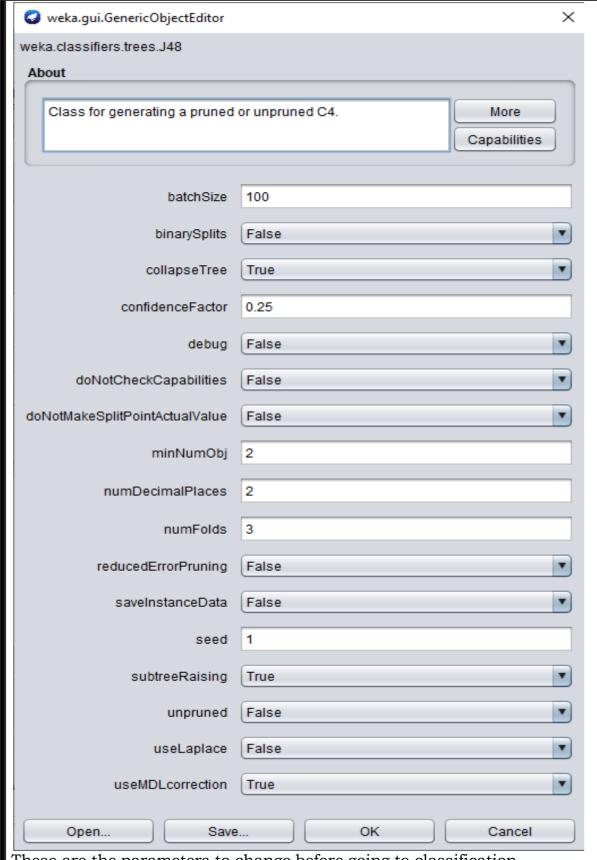
- Σ: Greek symbol for summation
- y_i : Actual value for the ith observation
- x_i : Calculated value for the ith observation
- n: Total number of observations







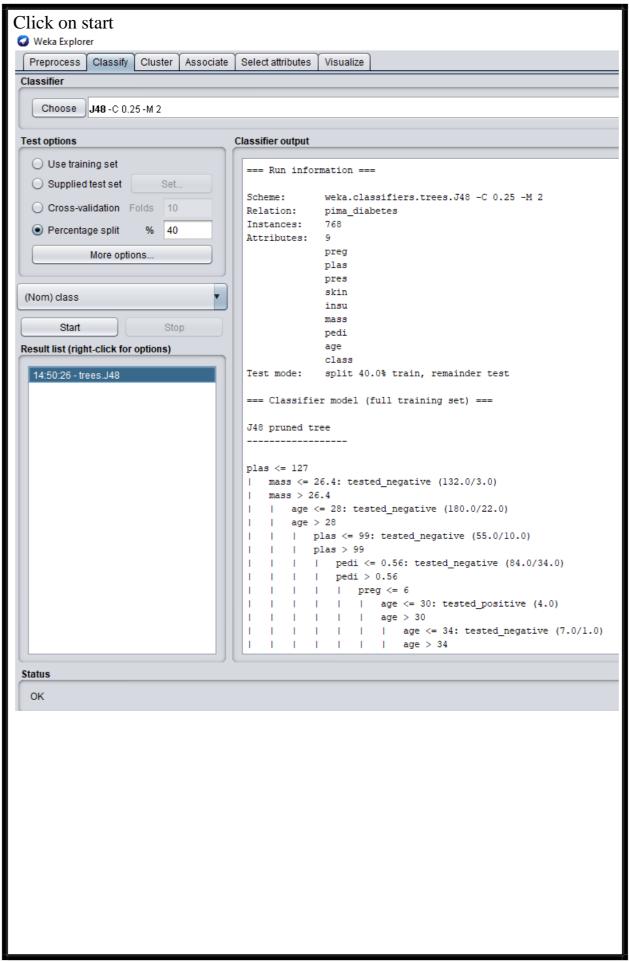




These are the parameters to change before going to classification Percentage split: 40%

If you have 10 records ,out of that 6 is used for training data set and remainingfor test data set







```
Classifier output
               pedi > 0.56
               | preg <= 6
                  | age <= 30: tested_positive (4.0)
                      age > 30
                 | | age <= 34: tested_negative (7.0/1.0)
                 | | age > 34
                 | | mass <= 33.1: tested positive (6.0)
           | | | mass > 33.1: tested_negative (4.0/1.0)
            | | preg > 6: tested_positive (13.0)
 plas > 127
    mass <= 29.9
   | plas <= 145: tested negative (41.0/6.0)
 1
    | plas > 145
       | age <= 25: tested_negative (4.0)
           age > 25
        | age <= 61
          | | mass <= 27.1: tested positive (12.0/1.0)
          - 1
              | mass > 27.1
              | | pres <= 82
  ı
           | | | | pedi <= 0.396: tested_positive (8.0/1.0)
       | | | | pedi > 0.396: tested negative (3.0)
          - 1
              | | pres > 82: tested_negative (4.0)
       | | age > 61: tested_negative (4.0)
     mass > 29.9
    | plas <= 157
 Т
        | pres <= 61: tested positive (15.0/1.0)
     | | pres > 61
    | | age <= 30: tested negative (40.0/13.0)
       | | age > 30: tested_positive (60.0/17.0)
    | plas > 157: tested positive (92.0/12.0)
 Number of Leaves :
 Size of the tree :
                      39
```



```
Number of Leaves :
Size of the tree: 39
Time taken to build model: 0.03 seconds
=== Evaluation on test split ===
Time taken to test model on test split: 0.01 seconds
=== Summary ===
Correctly Classified Instances 331
Incorrectly Classified Instances 130
                                               71.8004 %
                                                28.1996 %
                                0.3559
0.3243
Kappa statistic
Mean absolute error
                                 0.4609
Root mean squared error
Relative absolute error
                                70.9526 %
Root relative squared error
                                97.3291 %
Total Number of Instances
                               461
=== Detailed Accuracy By Class ===
              TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class
              0.807 0.458 0.777 0.807 0.792 0.357 0.733 0.816 tested negative
             0.542 0.193 0.587 0.542 0.564 0.357 0.733 0.531 tested_positive
Weighted Avg. 0.718 0.369 0.713 0.718 0.715 0.357 0.733 0.720
=== Confusion Matrix ===
  a b <-- classified as
247 59 | a = tested_negative
71 84 | b = tested_positive
```

Kappa statistic:

Cohen's kappa statistic measures interrater reliability (sometimes called interobserver agreement). Interrater reliability, or precision, happens when yourdata raters (or collectors) give the same score to the same data item. This statistic should only be calculated when:

Two raters each rate one trial on each sample, *or*. One rater rates two trials on each sample.

$$k = \frac{p0 - pe}{1 - pe} = 1 - p0$$
 $1 - pe$
 $1 - pe$

Mean Absolute Error

The Mean Absolute Error(MAE) is the average of all absolute errors. The formula is:

$$\mathrm{MAE} = \frac{1}{n} \sum_{i=1}^n |\, x_i^{} - x\,|$$

Where:

n = the number of errors,

 Σ = summation symbol (which means "add them all up"),

 $|x_i - x|$ = the absolute errors.

Root mean squared error:

Root mean square error or root mean square deviation is one of the most commonly used measures for evaluating the quality of predictions. It shows how far predictions fall from measured true values using Euclidean distance.

Root mean square error can be expressed as

$$RMSE = \sqrt{\frac{\sum_{i=1}^{N} ||y(i) - \hat{y}(i)||^2}{N}},$$

where N is the number of data points, y(i) is the i-th measurement, and y (i) isits corresponding prediction.

Relative absolute error:

It is a way to measure the performance of a predictive model. It's primarily used in machine learning, data mining, and operations management. RAE is notto be confused with <u>relative error</u>, which is a general measure of precision or accuracy for instruments like clocks, rulers, or scales.

$$\frac{|p_1 - a_1| + \dots + |p_n - a_n|}{|\overline{a} - a_1| + \dots + |\overline{a} - a_n|}$$

Root relative squared error:

The Root Relative Squared Error (RRSE) is defined as the square root of the sum of squared errors of a predictive model normalized by the sum of squarederrors of a simple model.

the root relative squared error E_i of an individual model i is evaluated by the equation:

$$E_i = \sqrt{\frac{\displaystyle\sum_{j=1}^n \bigl(P_{(ij)} - T_j\bigr)^2}{\displaystyle\sum_{j=1}^n \bigl(T_j - \overline{T}\bigr)^2}}$$

where $P_{(ij)}$ is the value predicted by the individual model i for record j (out of n records); T_i is the target value for record j; an $\overline{\mathbf{d}}$ is given by the formula:

$$\overline{T} = \frac{1}{n} \sum_{j=1}^{n} T_{j}$$

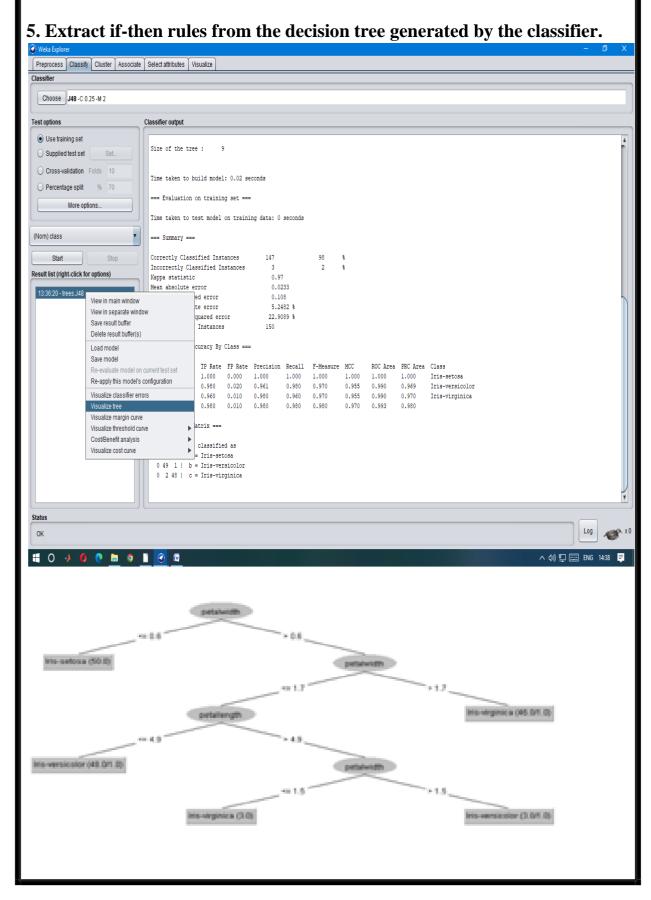
For a perfect fit, the numerator is equal to 0 and $E_i = 0$. So, the E_i index ranges from 0 to infinity, with 0 corresponding to the ideal.

, where:

n: represents the number of observations



- y_i : represents the realized value
- \hat{y}_i : represents the predicted value
- ÿ: represents the average of the realized values





6. Observe the confusion matrix.

=== Confusion Matrix ===

```
a b c <-- classified as
50 0 0 | a = Iris-setosa
0 49 1 | b = Iris-versicolor
0 2 48 | c = Iris-virginica
```

Precision: Appropriate when minimizing false positives is the

focus.Recall: Appropriate when minimizing false negatives is the

focus.

TP Rate: rate of true positives (instances correctly classified as a given class)FP Rate: rate of false positives (instances falsely classified as a given class) F measure is:

F-Measure = (2 * Precision * Recall) / (Precision + Recall)

MCC: it is used in machine learning as a measure of the quality of binary (two-class) classifications. It takes into account true and false positives and negatives and is generally regarded as a balanced measure which can be used even if the classes are of very different sizes

ROC(Receiver Operating Characteristics) area measurement: One of the most important values output by Weka. They give you an idea of how the classifiers are performing in general

PRC(Precision Recall) area:

Precision-recall curve. A plot of precision (= PPV) vs. recall (= sensitivity) for all potential cut-offs for a test.

Load each dataset into Weka and perform Naïve-bayes classification and k-Nearest Neighbour classification. Interpret the results obtained. Plot RoC Curves Compare classification results of ID3, J48, Naïve-Bayes and k-NN classifiers for each dataset, and deduce which classifier is performing best and poor for each dataset and justify.

K-Nearest Neighbors (KNN) is a standard machine-learning method that has been extended to large-scale data mining efforts. The idea is that one uses a large amount of training data, where each data point is characterized by a set of variables. KNN captures the idea of similarity (sometimes called distance, proximity, or closeness) with some mathematics we might have learned in our childhood— calculating the distance between points on a graph. There are otherways of calculating distance, and one way might be preferable depending on the problem we are solving.

However, the straight-line distance (also called the Euclidean distance) is a popular and familiar choice. 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 givendata).

Advantages & Disadvantages of KNN Algorithm Advantages

It is very easy to understand and implement

It is an instance-based learning(lazy learning) algorithm.

KNN does not learn during the training phase hence new data points can beadded with affecting the performance of the algorithm.

It is well suited for small

datasets.Disadvantages

It fails when variables have different scales. It is difficult to choose K-value.

It leads to ambiguous interpretations.

It is sensitive to outliers and missing values. Does not work well with large datasets.

It does not work well with high dimensions.K nearest neighbour:

it is also called instance based

learningit's very similar to a desktop

different names of KNN

---Memory

baseexample

instance based

lazy learning

KNN helps us to assign label to unknown data.

APPLY KNN ON DIABETES DATA SET

2) KNN Alogrithm

Different names of KNN

k-Nearest Neighbouring

Memory-Based

Reasoning

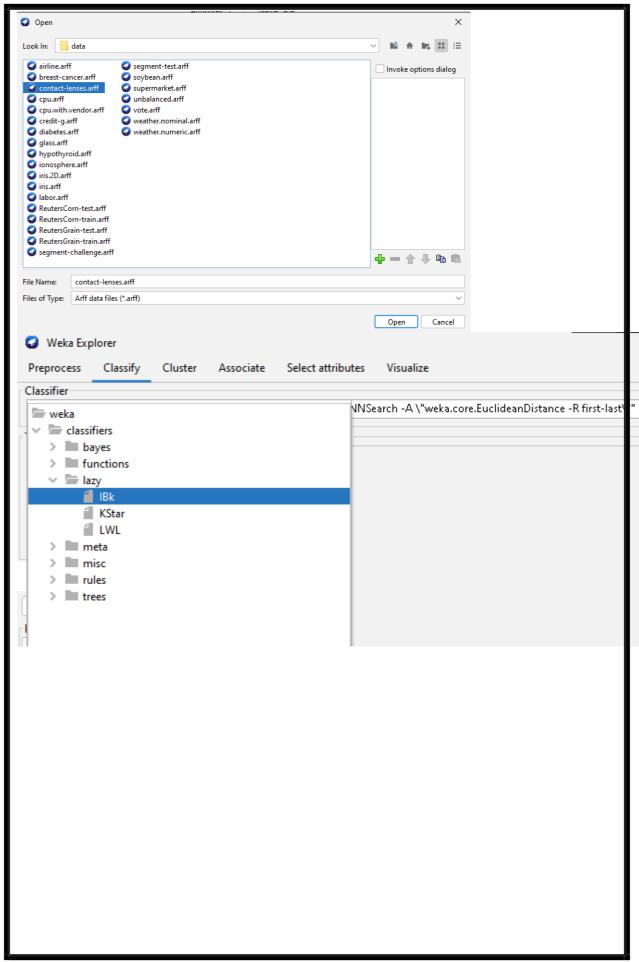
Example-Based

ReasoningInstance-

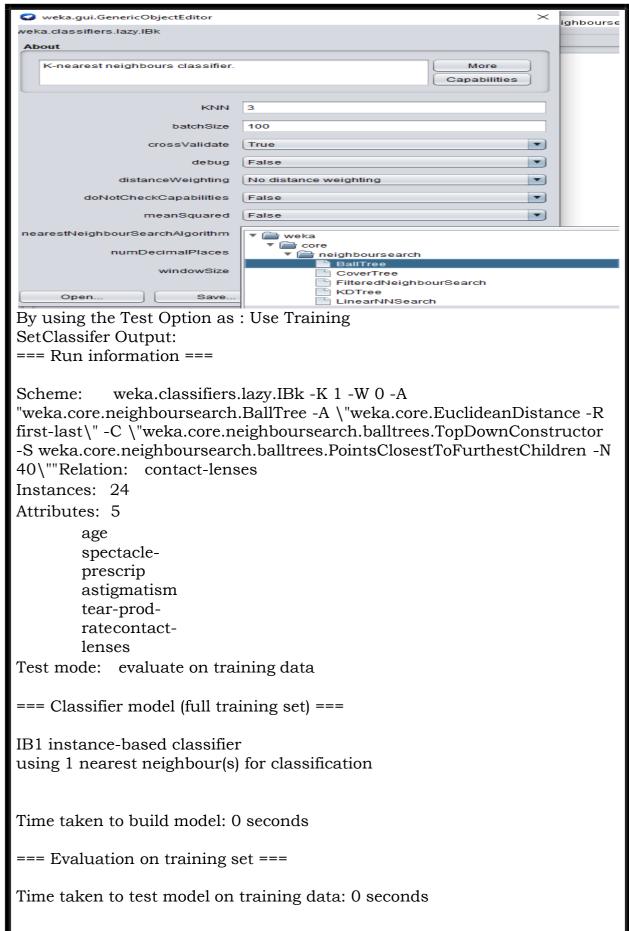
Based Learning Lazy

Learning











=== Summary ===

Correctly Classified Instances 24 100 % Incorrectly Classified Instances 0 0 %

Kappa statistic 1

Mean absolute error 0.0494
Root mean squared error 0.0524
Relative absolute error 13.4078 %
Root relative squared error 12.3482 %

Total Number of Instances 24

=== Detailed Accuracy By Class ===

ROC Area TP Rate FP Rate Precision Recall F-Measure MCC PRC Area Class 1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000 soft 1.000 0.0001.000 1.000 1.000 1.000 1.000 1.000 hard 1.000 1.000 1.000 0.0001.000 1.000 1.000 1.000 none Weighted Avg. 1.000 0.0001.000 1.000 1.000 1.000 1.000 1.000

=== Confusion Matrix ===

a b c <-- classified as

 $500 \mid a = soft$

 $0 \ 4 \ 0 \ | \ b = hard$

0.15 | c = none

By using the Test Option as : Percentage Split -60%

Classifer Output:

=== Run information ===

Scheme: weka.classifiers.lazy.IBk -K 1 -W 0 -A

"weka.core.neighboursearch.BallTree -A \"weka.core.EuclideanDistance -R first-last\" -C \"weka.core.neighboursearch.balltrees.TopDownConstructor -S weka.core.neighboursearch.balltrees.PointsClosestToFurthestChildren -N 40\""

Relation: contact-lenses

Instances: 24 Attributes: 5



age

spectacle-prescrip

astigmatism

tear-prod-rate

contact-lenses

split 60.0% train, remainder test Test mode:

=== Classifier model (full training set) ===

IB1 instance-based classifier using 1 nearest neighbour(s) for classification

Time taken to build model: 0 seconds

=== Evaluation on test split ===

Time taken to test model on test split: 0 seconds

=== Summary ===

Correctly Classified Instances 3 30 %

Incorrectly Classified Instances 70 %

Kappa statistic -0.0145

Mean absolute error 0.4301 0.564 Root mean squared error

97.0527 % Relative absolute error Root relative squared error 103.7551 %

Total Number of Instances 10

=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area

PRC Area Class

0.000? 0.920 0.886 0.0000.000soft 0.333 0.0000.000 $0.000 \quad 0.000$ -0.218 0.667 0.250

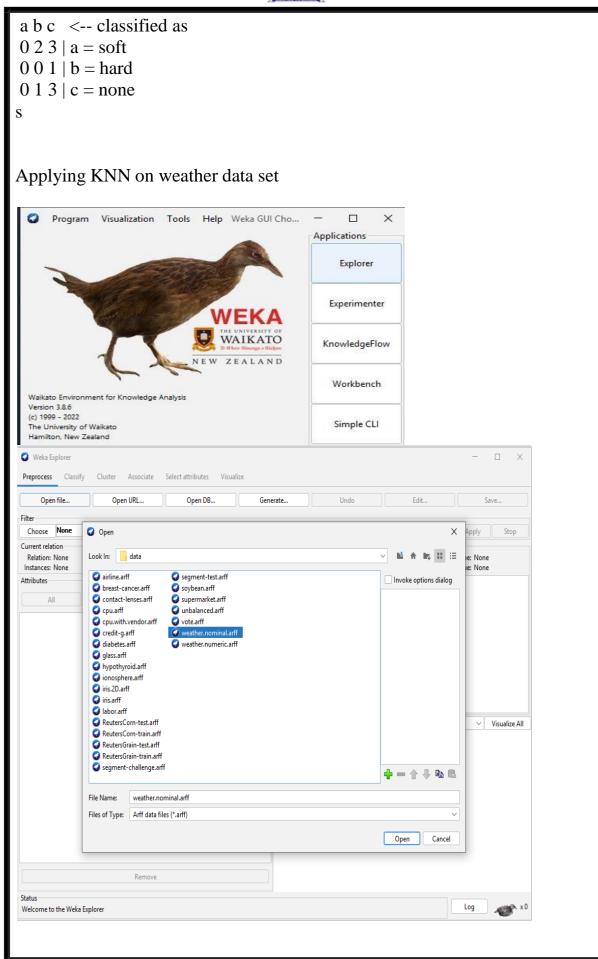
hard

0.750 0.667 0.429 0.750 0.545 0.089 0.646 0.667

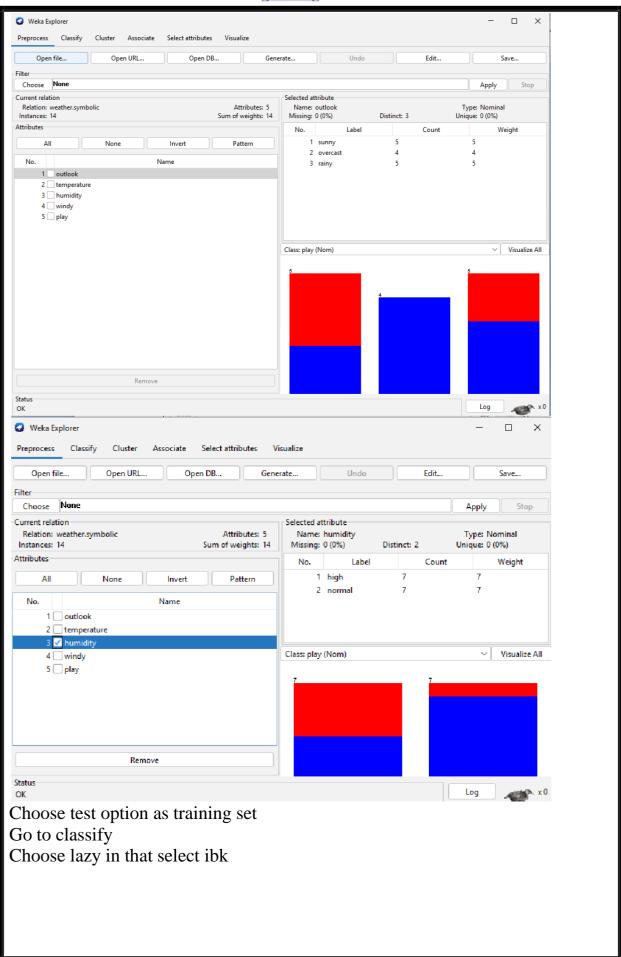
none

0.300 0.300 ? 0.300 ? ? Weighted Avg. 0.785 0.735

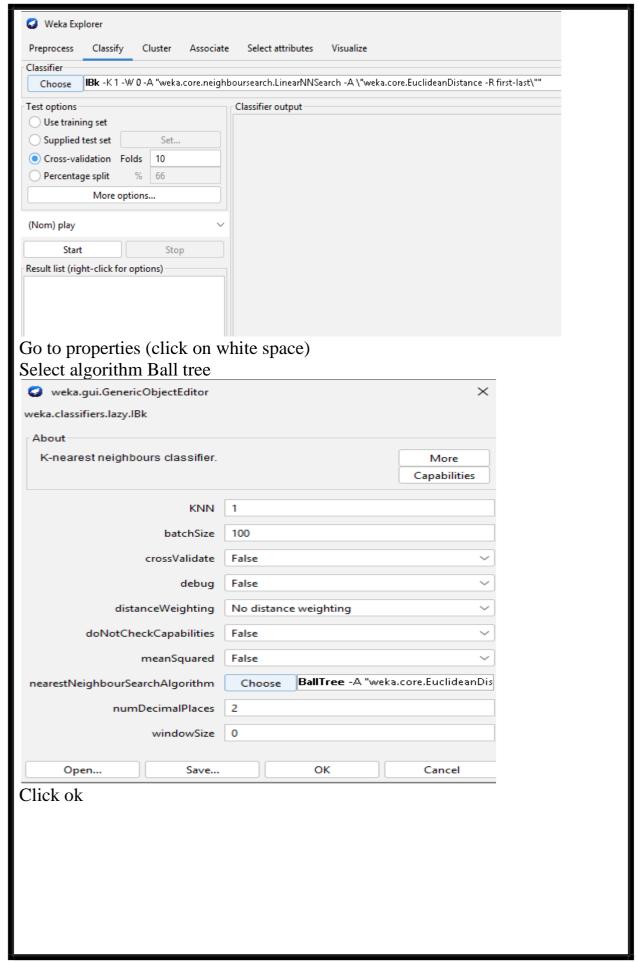
=== Confusion Matrix ===



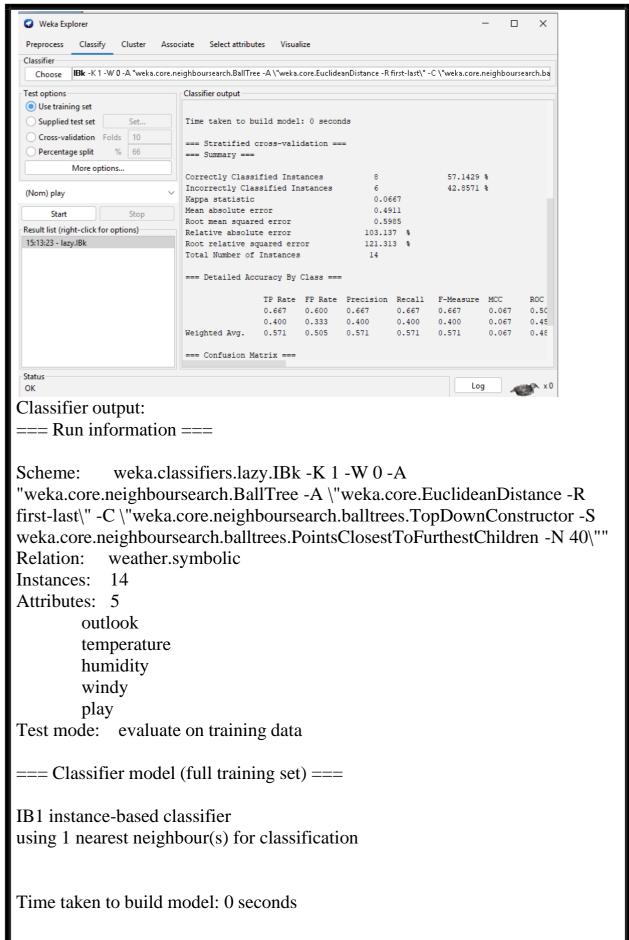














=== Evaluation on training set ===

Time taken to test model on training data: 0 seconds

=== Summary ===

Correctly Classified Instances 14 100 % Incorrectly Classified Instances 0 0 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.0625

13.4615 %

13.0347 %

=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area
PRC Area Class
1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000

yes

1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000 no

Weighted Avg. 1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000

=== Confusion Matrix ===

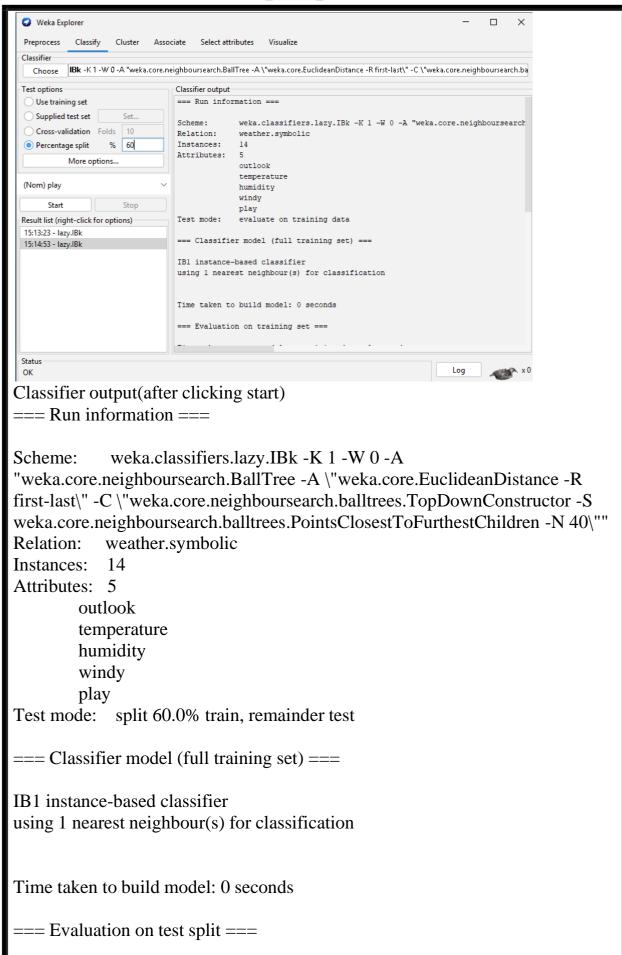
a b <-- classified as

 $9 \ 0 \ | \ a = yes$

0.5 | b = no

Choose test option as Percentage split:







Time taken to test model on test split: 0 seconds

=== Summary ===

Correctly Classified Instances 2 33.3333 % Incorrectly Classified Instances 4 66.6667 %

Kappa statistic -0.5

Mean absolute error0.5941Root mean squared error0.6782Relative absolute error127.3109 %Root relative squared error142.4592 %

Total Number of Instances 6

=== Detailed Accuracy By Class ===

TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Area Class

0.500 1.000 0.500 0.500 -0.500 0.313 0.625

yes

0.000 0.500 0.000 0.000 0.000 -0.500 0.313 0.333

no

Weighted Avg. 0.333 0.833 0.333 0.333 -0.500 0.313 0.528

=== Confusion Matrix ===

a b <-- classified as

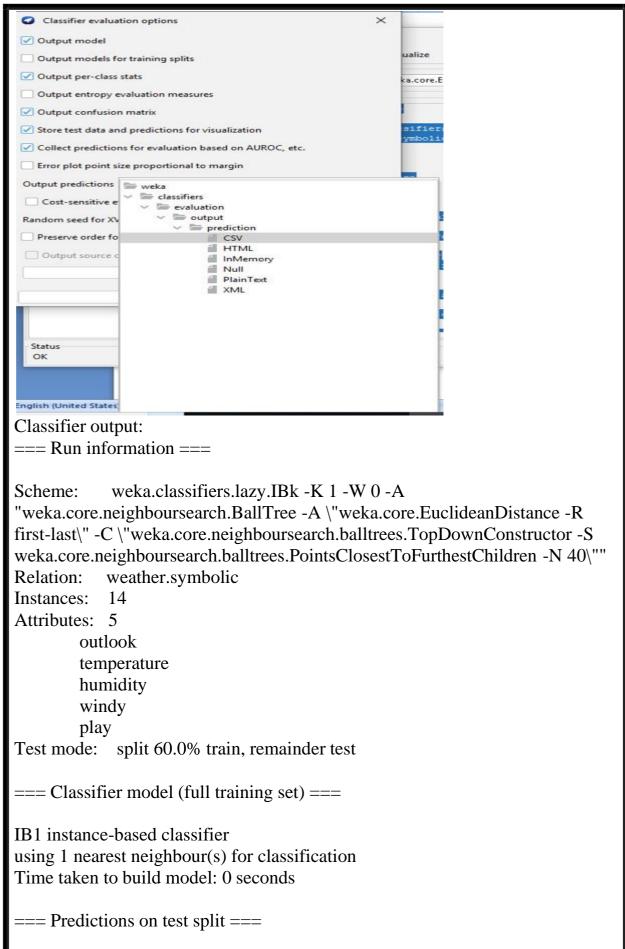
 $2 \ 2 \ | \ a = yes$

20 | b = no

Click on more options

Select csv







```
inst#,actual,predicted,error,prediction
1,1:yes,1:yes,,0.9
2,1:yes,2:no,+,0.9
3,1:yes,1:yes,,0.735
4,2:no,1:yes,+,0.9
5,2:no,1:yes,+,0.5
6,1:yes,2:no,+,0.9
=== Evaluation on test split ===
Time taken to test model on test split: 0 seconds
=== Summary ===
Correctly Classified Instances
                                  2
                                            33.3333 %
Incorrectly Classified Instances
                                             66.6667 %
Kappa statistic
                            -0.5
                               0.5941
Mean absolute error
Root mean squared error
                                 0.6782
Relative absolute error
                               127.3109 %
Root relative squared error
                               142.4592 %
Total Number of Instances
                                  6
=== Detailed Accuracy By Class ===
                                                                 ROC Area
          TP Rate FP Rate Precision Recall F-Measure MCC
PRC Area Class
          0.500
                 1.000
                        0.500
                                  0.500 0.500
                                                  -0.500 0.313
                                                                  0.625
yes
                                  0.000 \quad 0.000
                                                  -0.500 0.313
                                                                  0.333
          0.000 0.500 0.000
Weighted Avg. 0.333 0.833 0.333
                                         0.333 0.333
                                                         -0.500 0.313
0.528
=== Confusion Matrix ===
a b <-- classified as
2 \ 2 \ | \ a = yes
20 | b = no
Choose HTML in mor options
Classifier output:
=== Run information ===
```

```
weka.classifiers.lazy.IBk -K 1 -W 0 -A
Scheme:
"weka.core.neighboursearch.BallTree -A \"weka.core.EuclideanDistance -R
first-last\" -C \"weka.core.neighboursearch.balltrees.TopDownConstructor -S
weka.core.neighboursearch.balltrees.PointsClosestToFurthestChildren -N 40\""
        weather.symbolic
Relation:
Instances:
         14
Attributes: 5
      outlook
      temperature
      humidity
      windy
      play
Test mode: split 60.0% train, remainder test
=== Classifier model (full training set) ===
IB1 instance-based classifier
using 1 nearest neighbour(s) for classification
Time taken to build model: 0 seconds
=== Predictions on test split ===
<html>
<head>
<title>Predictions for dataset weather.symbolic</title>
</head>
<body>
<div align="center">
<h3>Predictions for dataset weather.symbolic</h3>
inst#actualpredictederrorpredictio
n  
align="right">0.9
21:yes2:no++
align="right">0.9
31:yes1:yes <td
align="right">0.735
41:yes+td>+
align="right">0.9
```



```
51:yes+<td
align="right">0.5
61:yes2:no+<td
align="right">0.9
</div>
</body>
</html>
=== Evaluation on test split ===
Time taken to test model on test split: 0 seconds
=== Summary ===
Correctly Classified Instances
                                      33.3333 %
                              2
Incorrectly Classified Instances
                                       66.6667 %
Kappa statistic
                        -0.5
Mean absolute error
                           0.5941
Root mean squared error
                             0.6782
Relative absolute error
                          127.3109 %
Root relative squared error
                            142.4592 %
Total Number of Instances
                              6
=== Detailed Accuracy By Class ===
        TP Rate FP Rate Precision Recall F-Measure MCC
                                                        ROC Area
PRC Area Class
        0.500
               1.000
                     0.500
                             0.500
                                   0.500
                                            -0.500 0.313
                                                          0.625
yes
        0.000
              0.500
                     0.000
                             0.000 \quad 0.000
                                            -0.500 0.313
                                                          0.333
Weighted Avg. 0.333 0.833 0.333
                                   0.333 0.333
                                                 -0.500 0.313
0.528
=== Confusion Matrix ===
a b <-- classified as
2 \ 2 \ | \ a = yes
20 | b = no
```