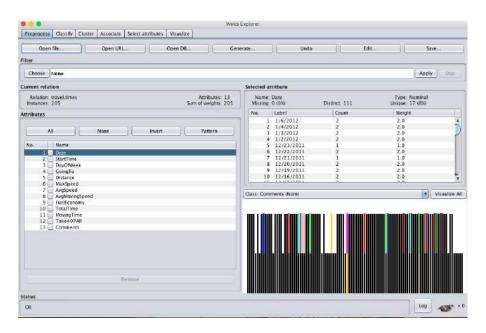
DATA PREPROCESSING AND ANALYSIS FOR DATASET USING WEKA

DESCRIPTION:

Consider a dataset of traveltimes.csv file where it contains the columns of (or) attributes as Date, StartTime, DayOfWeek, GoingTo, Distance, MaxSpeed, AvgSpeed, AvgMovingSpeed, FuelEconomy, TotalTime, MovingTime, Take407All comments.



PREPROCESS:



OBSERVATION:

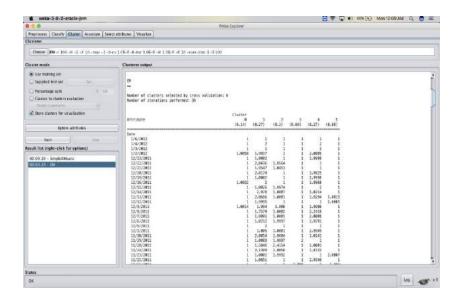
A. ATTRIBUTE TYPE:

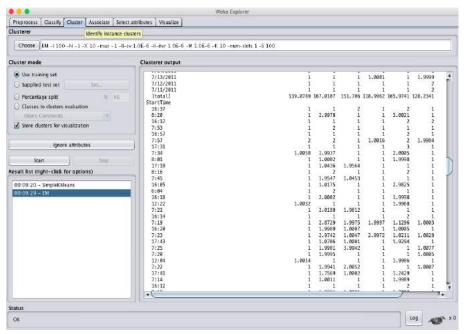
S.NO	ATTRIBUTE	ТҮРЕ
1.	Date	Nominal
2.	Start Time	Nominal
3.	Day Of Week	Nominal
4.	Going To	Nominal
5.	Distance	Numeric
6.	Max Speed	Numeric
7.	Avg Speed	Numeric
8.	Avg Moving Speed	Numeric
9.	Fuel Economy	Nominal
10.	Total Time	Numeric
11.	Moving Time	Numeric
12.	Comments	Nominal
13.	Take 407 All	Nominal

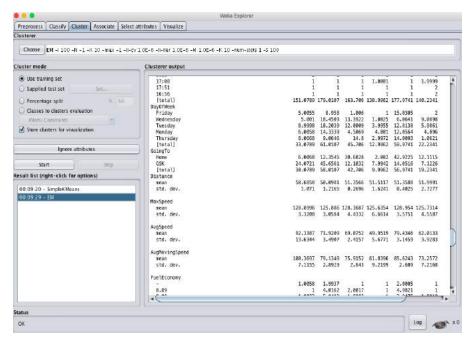
B. PERCENTAGE OF MISSING VALUES:

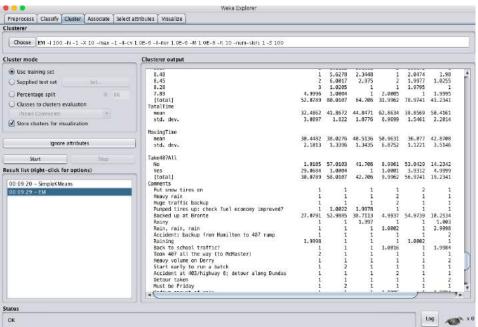
S.NO	ATTRIBUTE	Percentage Of Missing Values
1.	Date	0 %
2.	Start Time	0 %
3.	Day Of Week	0 %
4.	Going To	0 %
5.	Distance	0 %
6.	Max Speed	0 %
7.	Avg Speed	0 %
8.	Avg Moving Speed	0 %
9.	Fuel Economy	8 %
10.	Total Time	0 %
11.	Moving Time	0 %
12.	Comments	88 %
13.	Take 407 All	0 %

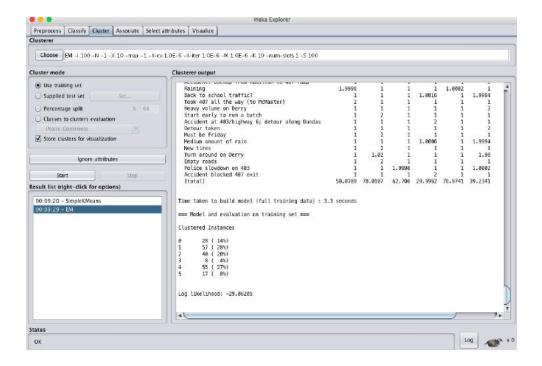
C. MIN, MAX, MEAN, STANDARD DEVIATION:











DATA SEGMENTATION BYK- MEANS CLUSTER USING WEKA AND R-TOOL

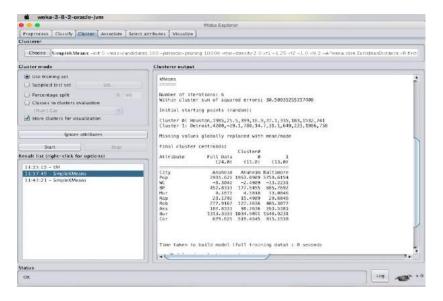
DESCRIPTION:

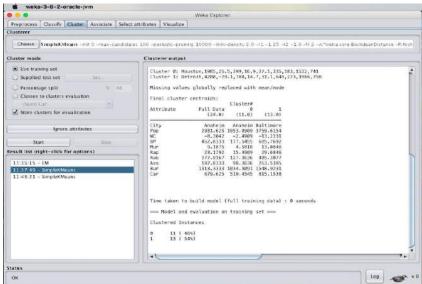
Consider a dataset of citycrimes.csv file of which it contains the attributes are City, Pop, WC, BP, Mur, Rap, Rob, Ass, Bus and car for the performance of the dataset by applying the K-means algorithm in weka and as well using R- tool.

USING WEKA TOOL:

STEPS INVOLVED:

- Choose a set of attributes for clustering and for giving a motivation.
- Choose the dataset and import the dataset into Weka tool.
- Cluster the dataset and choose simple K-means algorithm and give the motivation.

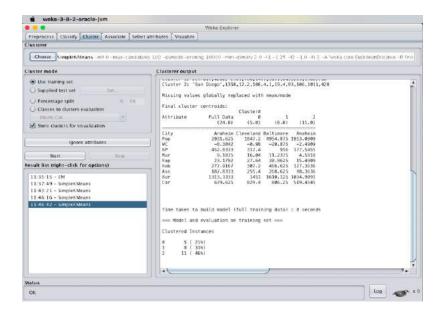


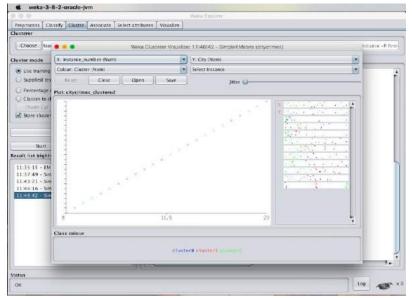


A. Experiment with atleast 2 different number of clusters but with same seed values:

STEPS INVOLVED:

- Compare the two different clusters but with the same seed values.
- Change the number of clusters value and need not to change the seed value.
- Apply the K-means algorithm and start executing the algorithm.





DATASEGMENTATION BY EXPECTATION MAXIMISATION ALGORITHM THROUGH WEKA

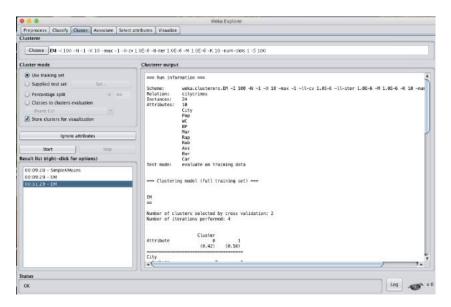
DESCRIPTION:

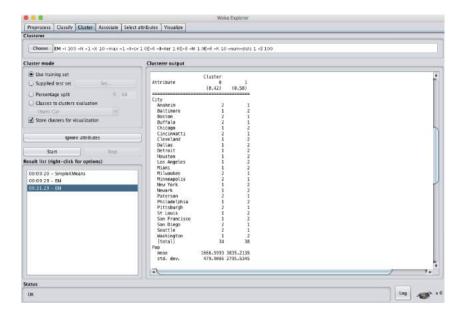
Consider a dataset of citycrimes.csv file of which it contains the attributes are City, Pop, WC, BP, Mur, Rap, Rob, Ass, Bus and car for the performance of the dataset by applying the K-means algorithm in weka and as well using R- tool.

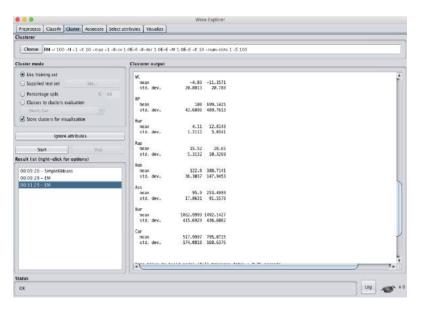
When the clustering is been made through the expectation maximization algorithm by setting minimum standard deviation values then the results will be of the following :

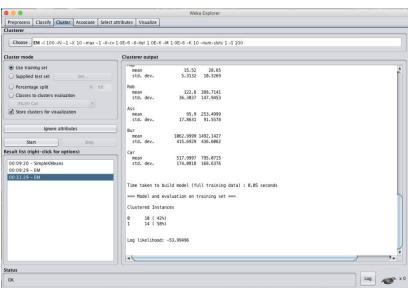
Steps Involved:

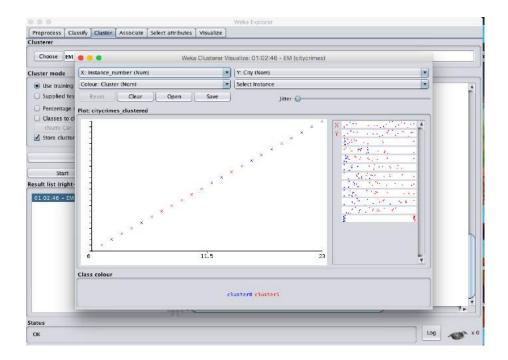
- Initially, load the dataset into the weka tool and check for all the attributes present in the dataset.
- Then move to cluster panel and apply the EM algorithm technique for the datasheet.
- Finally, Observe the results that are obtained.



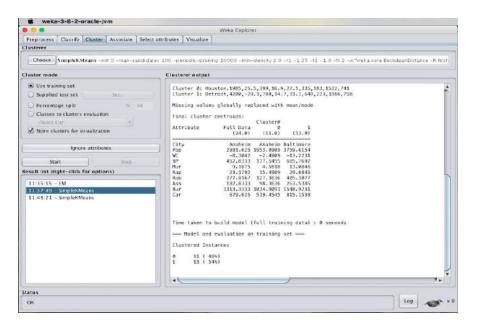








***** K- MEANS ALGORITHM:



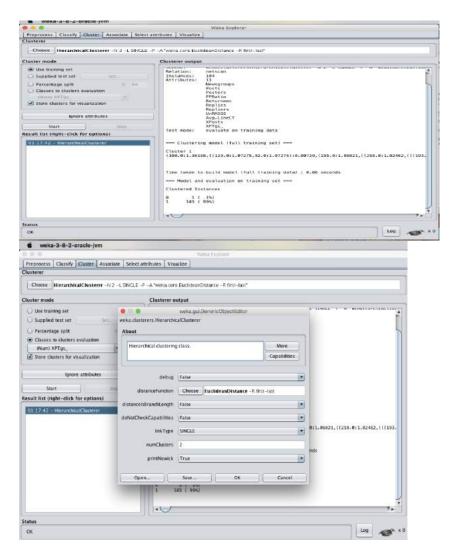
DATA SEGMENTATION BY COBWEB – HIERARCHIAL CLUSTERING ALGORITHM USING WEKA TOOL

.

DESCRIPTION:

Consider a dataset netscan.csv where it contains the attributes of Newsgroups, posts, posters, PPRatio, Returnees, Replies, Repliers, UnRMsgs, Avg.LineCT, Xports, XPTgs. Each attribute will have different types of the meanings.

HIERARCHIIAL CLUSTERING:



FREQUENT PATTERN MINING USING ASSOCIATION RULE THROUGH WEKA AND R TOOLS

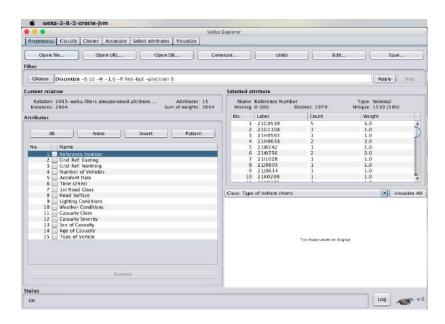
DESCRIPTION:

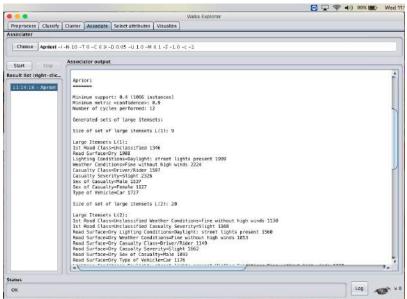
Consider a dataset of 2015.csv file of which it contains the attributes are Reference Number, Grid ref: Easting, Grid Ref: Northing, Number of vehicles, Accident date, Time(24 hr), 1st Road class, Road Surface, Lighting conditions, Weather conditions, casuality class, Sex of casuality, Age of casuality, Type of casuality for the performance of the dataset by applying the Apriori algorithm in weka and as well using R- tool.

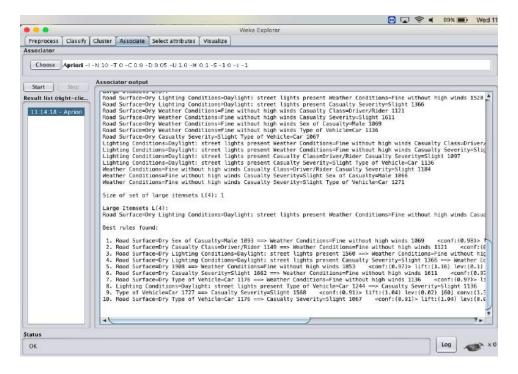
***** USING WEKA TOOL:

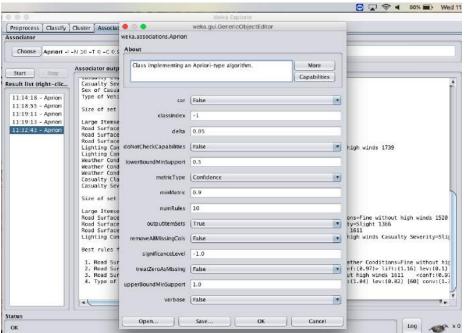
STEPS INVOLVED:

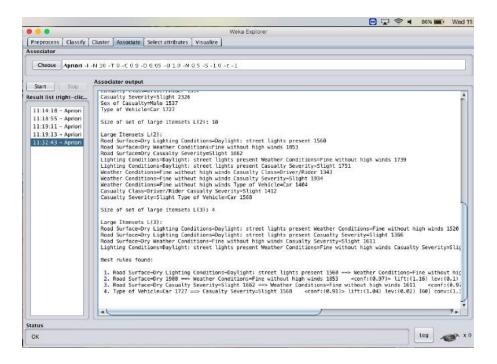
- Choose a set of attributes for clustering and for giving a motivation.
- Choose the dataset and import the dataset into Weka tool.
- Discretize the attributes from numeric to nominal to perform the algorithm.
- Cluster the dataset and choose simple Apriori algorithm.
- Set the Upper bound min sup and lower bound min sup values.

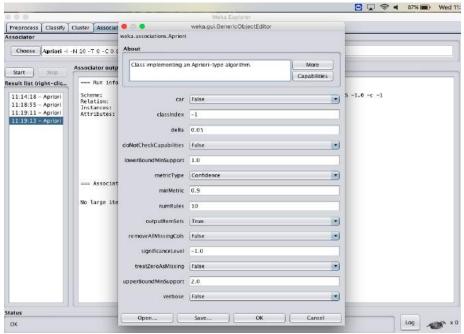


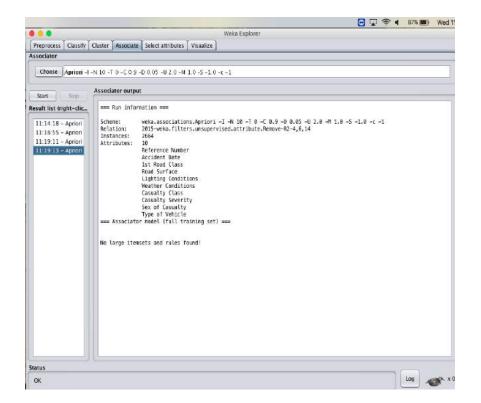












FREQUENT PATTERN MINING USING FP GROWTH THROUGH WEKA TOOL

DESCRIPTION:

Consider a dataset of 2015.csv file of which it contains the attributes are Reference Number, Grid ref: Easting, Grid Ref: Northing, Number of vehicles, Accident date, Time(24 hr), 1st Road class, Road Surface, Lighting conditions, Weather conditions, casuality class, Sex of casuality, Age of casuality, Type of casuality for the performance of the dataset by applying the FP algorithm in weka tool.

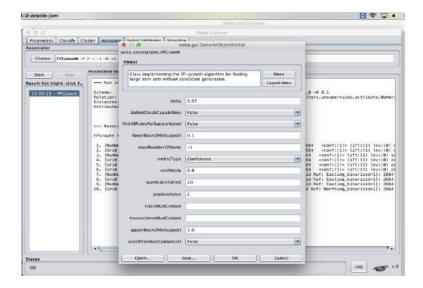
***** USING WEKA TOOL:

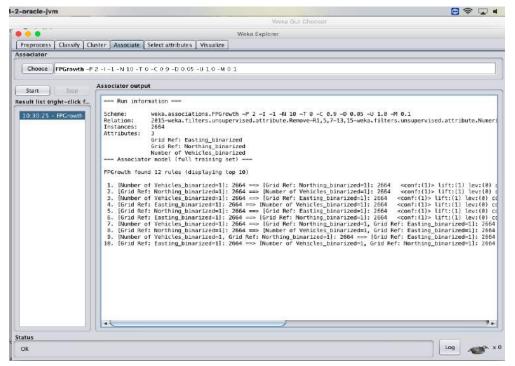
STEPS INVOLVED:

- Choose a set of attributes for clustering and for giving a motivation.
- Choose the dataset and import the dataset into Weka tool.
- Discretize the attributes from all data types to nominal to perform the algorithm.
- Associate the attributes with the FP growth algorithm.
- Set the Upper bound min_sup and lower bound min_sup values.

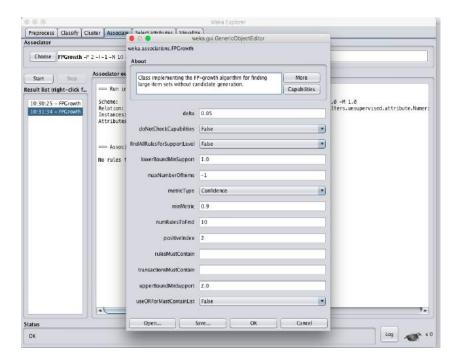
OBSERVATIONS:

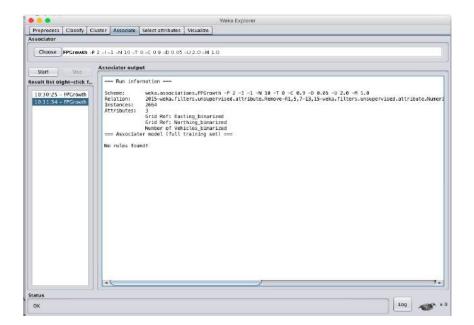
- 1) When the association rules are of values:
 - a) Upper bound min_sup = 1.0
 - b) Lower bound min_sup = 0.1
 - c) Metric type = confidence.





- 2) When the association rules are of values:
 - a) Upper bound min_sup = 2.0
 - b) Lower bound min_sup = 1.0
 - c) Metric type = confidence.

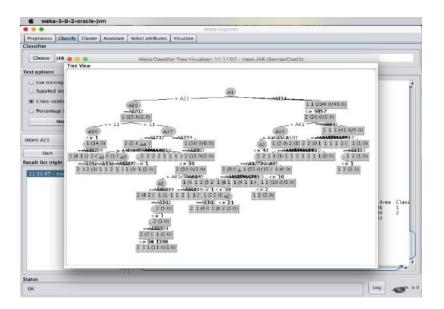


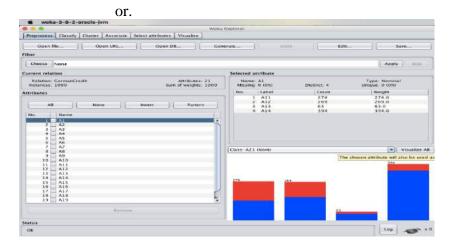


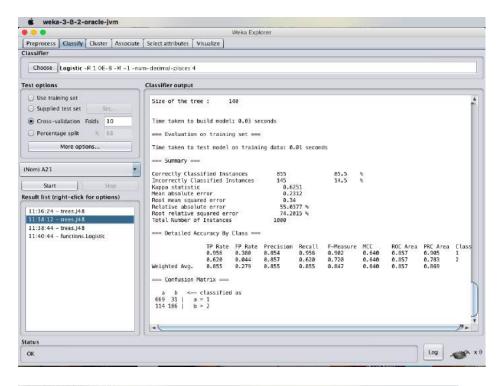
PREDICTION OF CATEGORICAL DATA USING DECISION TREE ALGORTHM THROUGH WEKA

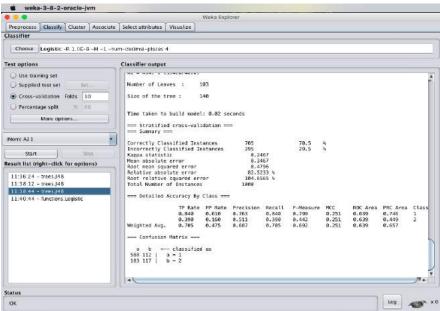
Decision Tree:

Visualize the decision tree for the given dataset.



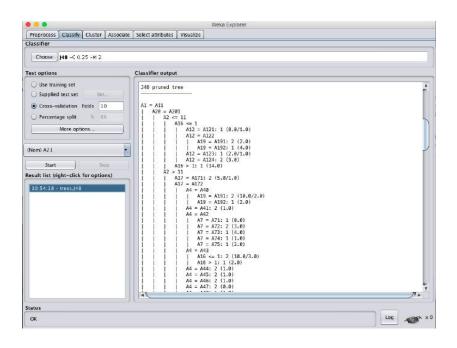


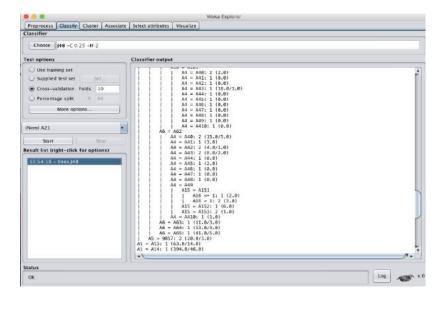


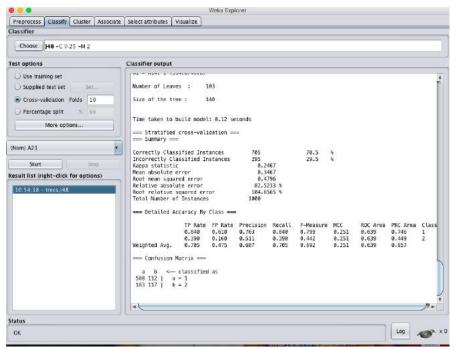


> CROSS VALIDATION ANALYSIS:

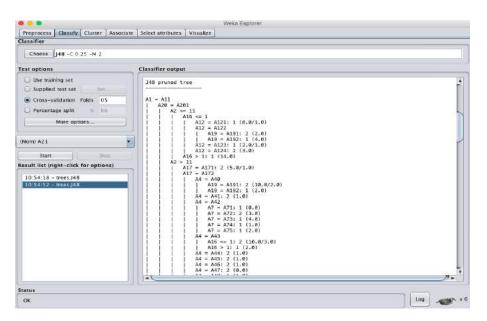
• When cross validation folds are 10:

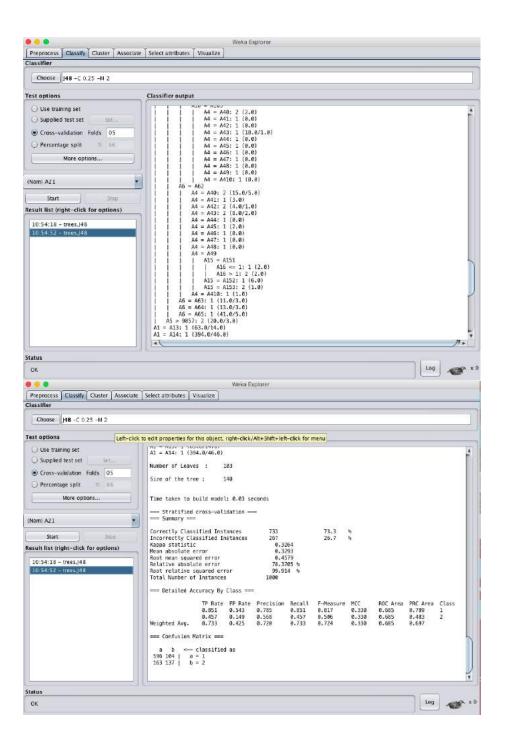






• When cross validation folds are: 05:-





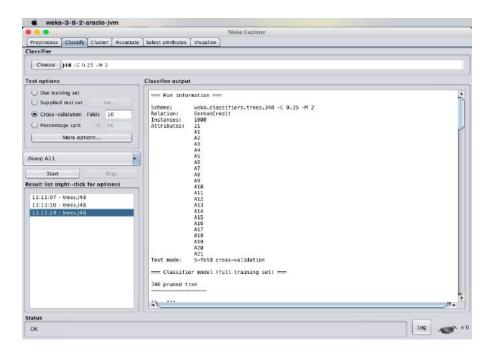
PREDICTION OF CATEGORICAL DATA USING SMO ALGORTHM THROUGH WEKA

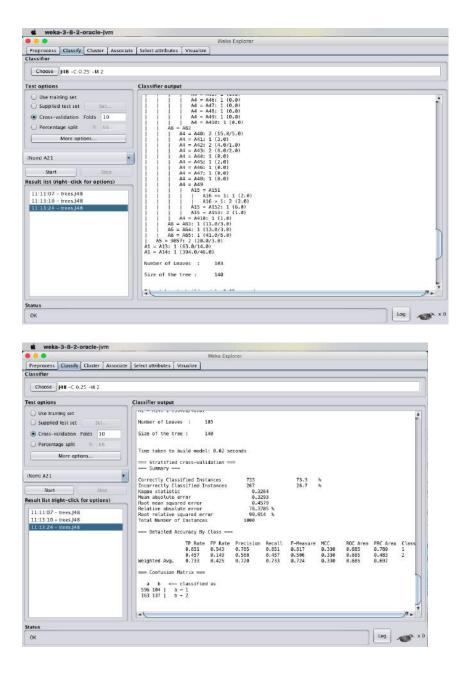
DESCRIPTION:

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

DECISION TREE:

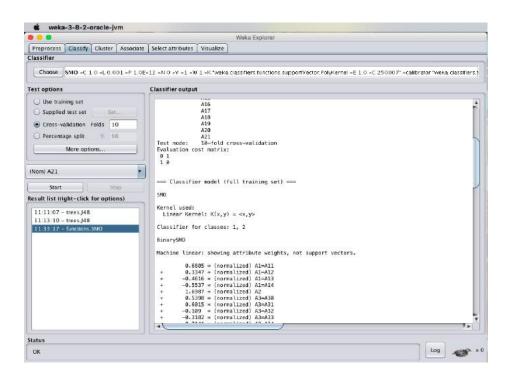
A tree has many analogies in real life, and turns out that it has influenced a wide area of machine learning, covering both classification and regression. In decision analysis, a decision tree can be used to visually and explicitly represent decisions and decision making. As the name goes, it uses a tree-like model of decisions. Though a commonly used tool in data mining for deriving a strategy to reach a particular goal, its also widely used in machine learning, which will be the main focus of this article.

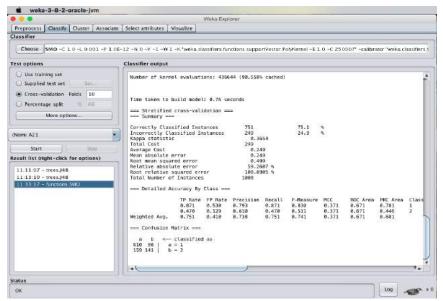




SMO ALGORITHM:

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.





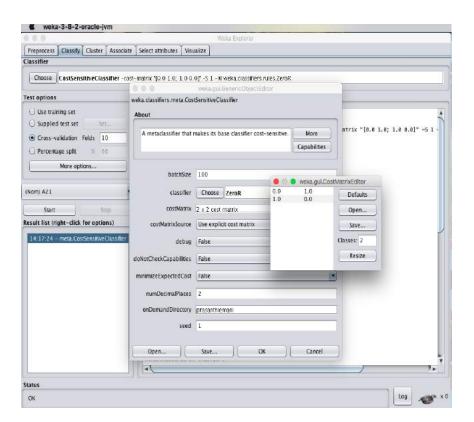
1. Set the cost sensitive evaluation and compare the obtained results.

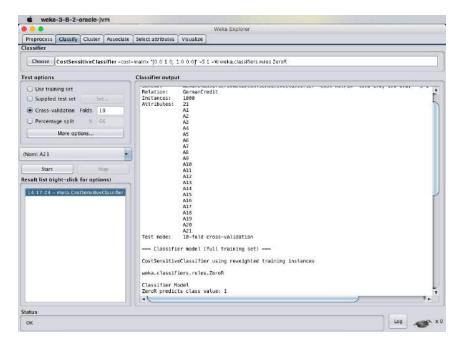
Cost-Sensitive Learning is a type of learning in data mining that takes the misclassification costs (and possibly other types of cost) into consideration. The goal of this type of learning is to minimize the total cost. The key difference between cost-sensitive learning and cost-insensitive learning is that cost-sensitive learning treats the different misclassifications differently. Costinsensitive learning does not take the misclassification costs into consideration. The goal of this type of learning is to pursue a high accuracy of classifying examples into a set of known classes.

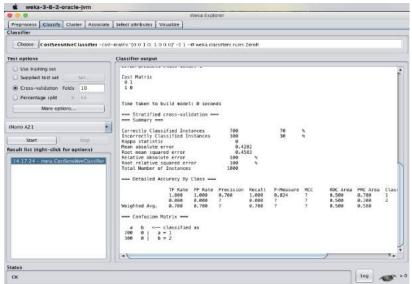
STEPS:

- Classifythe dataset with the cost sensitive classifier technique.
- ➤ Change the cost matrix to 2*2 matrix and execute.

ANALYSIS:







2. What is the significance of the following parameters :

a) Mean Absolute Error:

Mean Absolute Error (MAE) is similar to the Mean Squared Error, but it uses absolute values instead of squaring. This measure is not as popular as MSE, though its meaning is more intuitive (the "average error").

b) Total Number of Instances:

The data present consists of various instances of the class. In the case of german_credit dataset, the total number of instances present in the german credit dataset are 1000 instances.

EVALUATING ACCURACY OF THE CLASSIFIERS

DESCRIPTION:

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

ANALYSIS:

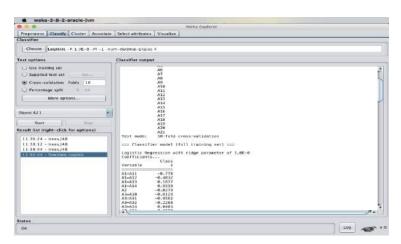
A) Logistic Regression:

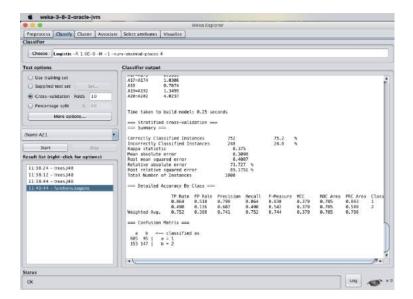
Logistic regression predicts the probability of an outcome that can only have two values (i.e. a dichotomy). The prediction is based on the use of one or several predictors (numerical and categorical).

Steps:

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

Output:





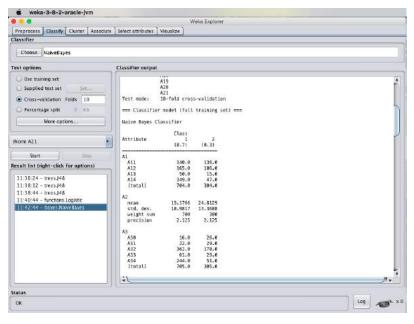
B) Naïve Bayes Algorithm:

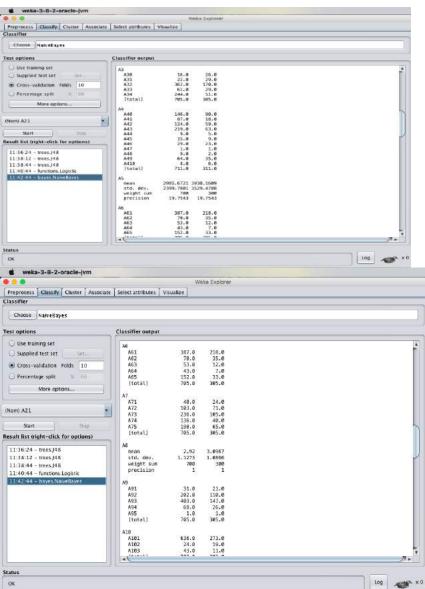
The Naive Bayesian classifier is based on Bayes' theorem with the independence assumptions between predictors. A Naive Bayesian model is easy to build, with no complicated iterative parameter estimation which makes it particularly useful for very large datasets. Despite its simplicity, the Naive Bayesian classifier often does surprisingly well and is widely used because it often outperforms more sophisticated classification methods.

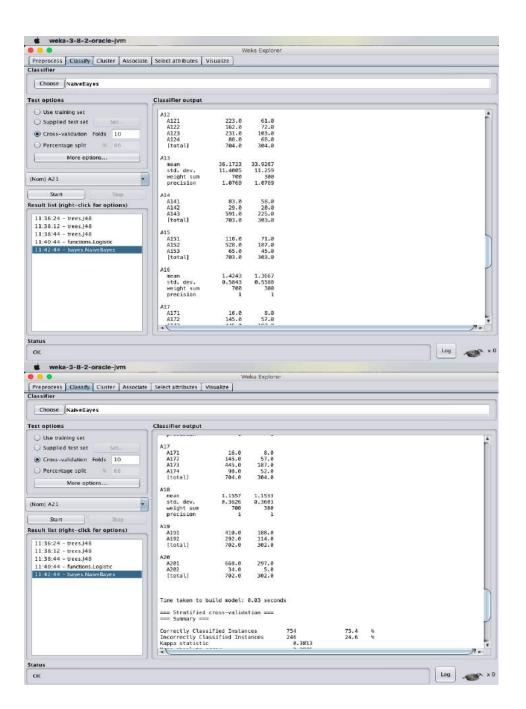
Steps:

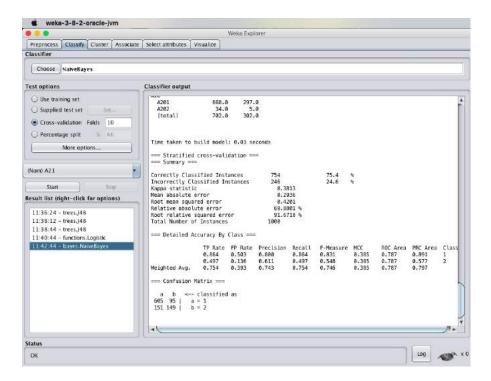
- Load the dataset into the weka tool and preprocess it.
- Apply the classification the Naïve bayes technique and execute for the result.

Output:









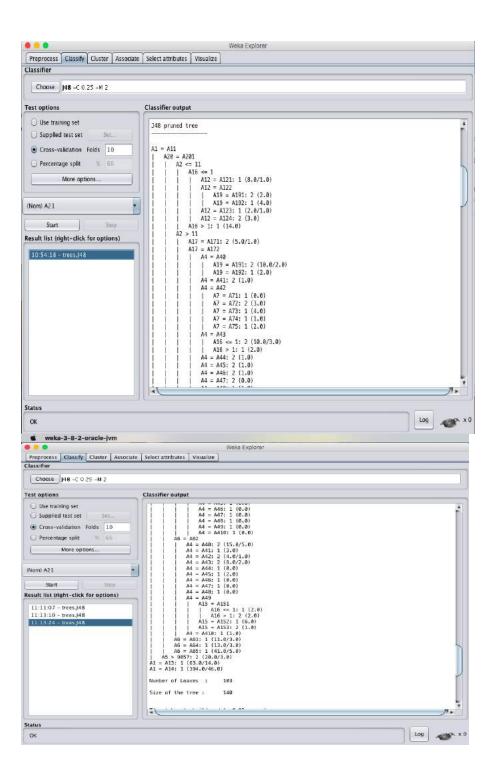
C) J48 Algorithm:

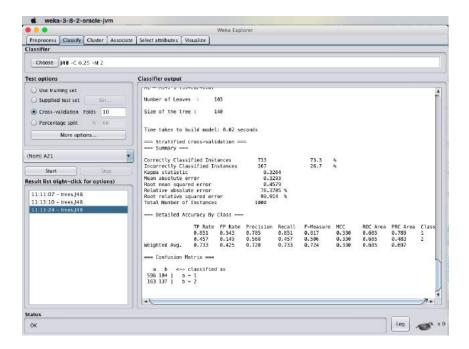
Classification is the process of building a model of classes from a set of records that contain class labels. Decision Tree Algorithm is to find out the way the attributes-vector behaves for a number of instances. Also on the bases of the training instances the classes for the newly generated instances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

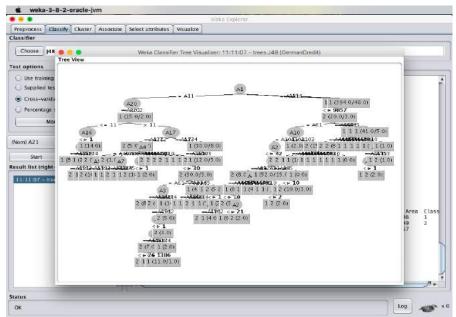
Steps:

- 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.

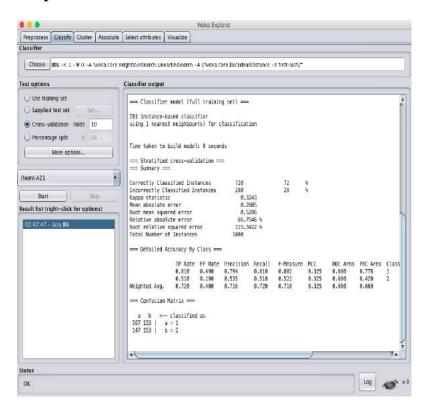
It is widely disposable in real-life scenarios since it is non-parametric, meaning, it does not make any underlying assumptions about the distribution of data (as opposed to other algorithms such as GMM, which assume a Gaussian distribution of the given data).

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:

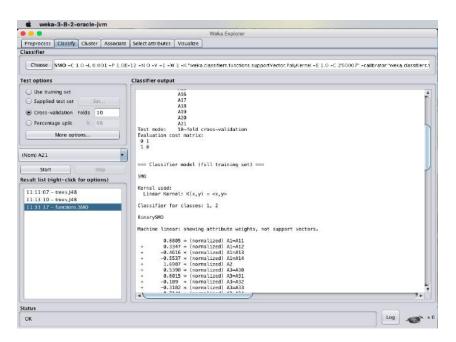


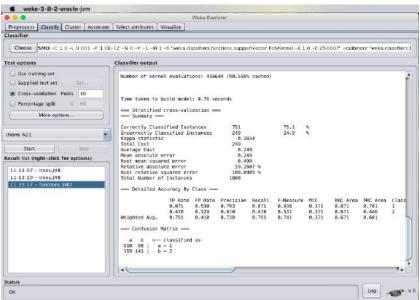
E) SMO Algorithm:

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.





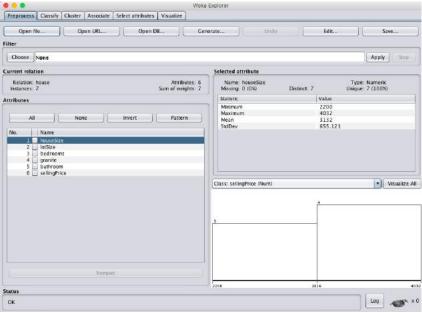
NUMERICAL PREDICTION ANALYSIS USING LINEAR REGRESSION THROUGH WEKA

DESCRIPTION:

Consider a dataset of house.arff where it contains the attributes as house size, lot size, bedrooms, granite, bathroom and the selling price.

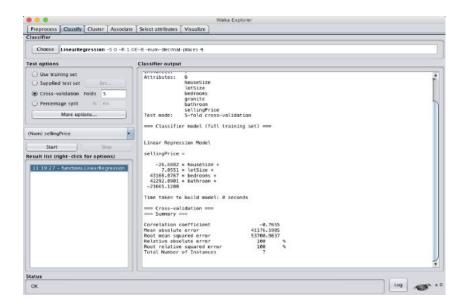
Steps:

- Load the dataset into the weka tool and check for the attributes.
- Classify the data using linear regression analysis method (or) technique.
- Check for the cross-validation folds where the value of the folds should be less than the value of the instances present in the dataset.
- Observe the cross validation summary after applying the linear regression technique for the price of the house.

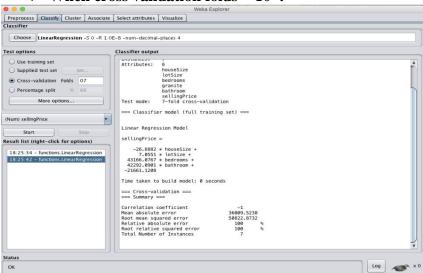


OBSERVATION:

***** When cross validation folds = 05:



\\$ When cross validation folds = :



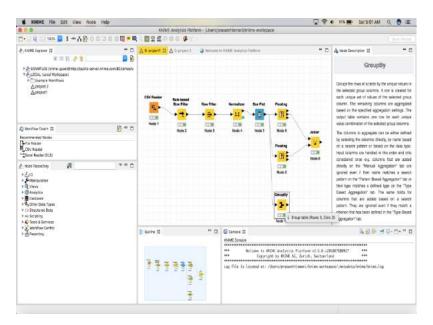
EXTRACT TRANSFORM LOAD (ETL) AND OLAP OPERATION USING KNIME TOOL

DESCRIPTION:

Consider a dataset movies.csv where it contains the attributes ad title, genre, director, year, duration, gross, budget, cast_facebook_likes, votes, reviews, rating.

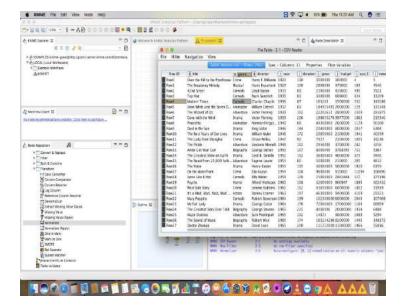
STEPS:

Import the dataset into the knime tool using csv reader.



CSV Reader:

Execute the CSV reader, look at the table of the loaded dataset.



❖ Rule-Based Row Filter:

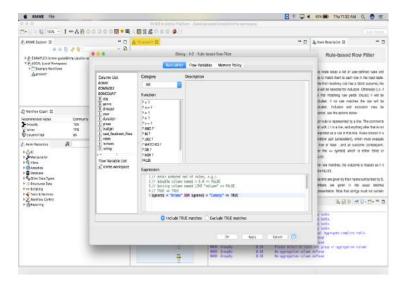
This node takes a list of user-defined rules and tries to match them to each row in the input table. If the first matching rule has a TRUE outcome, the row will be selected for inclusion. Otherwise (i.e. if the first matching rule yields FALSE) it will be excluded. If no rule matches the row will be excluded.

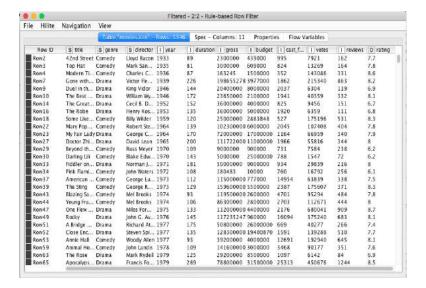
Steps:

- Make a connection between CSV reader and rule based row filter.
- Configure rule based row filter.
- Execute and Check out for the table after applying rule based row filter.

Code:

\$genre\$ = "Drama" XOR \$genre\$ = "Comedy" => TRUE





* Row Filter:

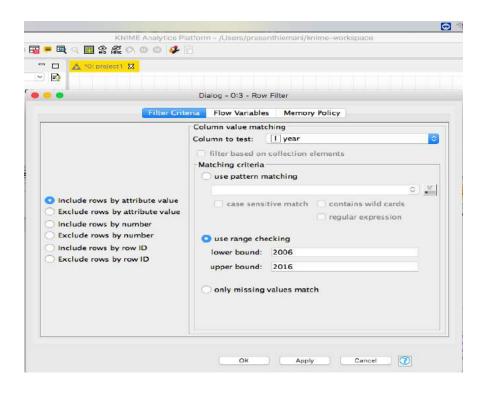
3 matching criteria on data colums: on String by full or partial pattern matching, on numbers by range, on missing values, all of them also on collection columns. 1 matching criterion on row numbers: from row number to row number. 1 matching criterion on RowID: full and partial pattern matching. Partial pattern matching is obtained through wild cards and RegEx. All matching criteria can be used in Include or Exclude mode. Include keeps the match results. Exclude excludes it.

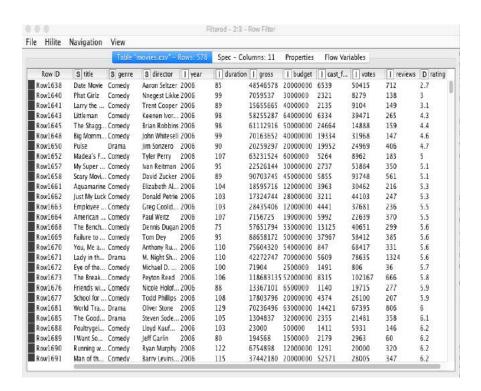
Steps:

- Make a connection between rule based row filter and row filter.
- Configure row filter.
- Execute and Check out for the table after applying row filter.

Use range checking :

Lower Bound: 2006 Upper Bound: 2016

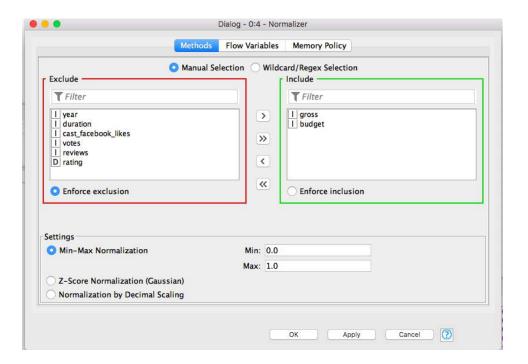




❖ Normalizer:

Steps:

- Connect normalizer with the row filter.
- Configure normalizer as methods which are to be included for normalization technique and set min and max values.
- Include:
 - a) Gross
 - b) Budget
 - c) Min: 0.0
 - d) Max: 0.1
- Execute the normalizer and check for the values in the table where you will find the normalized values of the table.



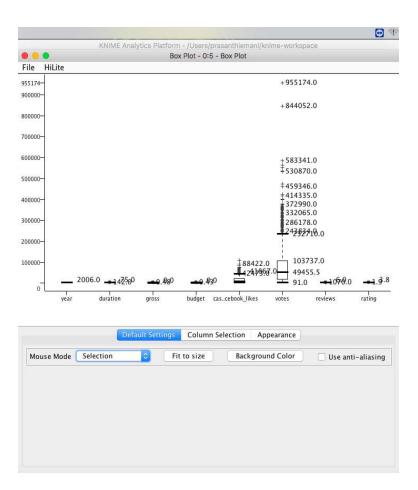
Boxplot :

A box plot displays robust statistical parameters: minimum, lower quartile, median, upper quartile, and maximum. A box plot for one numerical attribute is constructed in the following way: The box itself goes from lower quartile (Q1) to upper quartile (Q3). Median is drawn as horizontal bar inside box. Distance between Q1 and Q3 is called interquartile range (IQR). Above and below box are so-called whiskers. They are drawn at minimum and maximum value as horizontal bars and are connected with the box by a dotted line.

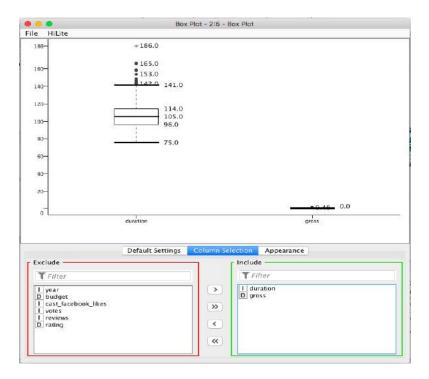
Steps:

- Make connection between normalizer and boxplot.
- View for the boxplot directly.
- We can select the specific columns for the individual boxplot through column selection.

> Boxplot for the whole of dataset:



> Boxplot for the two of the attributes : gross and duration :

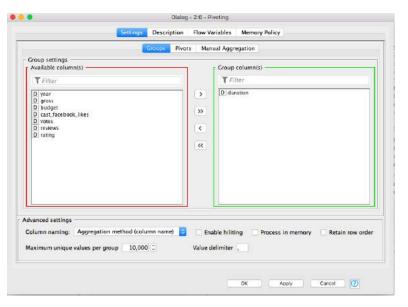


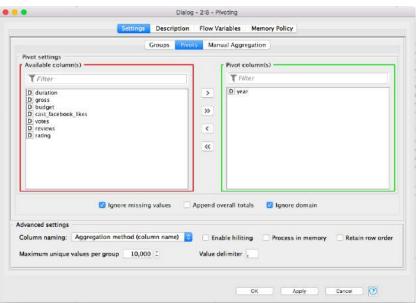
Pivoting:

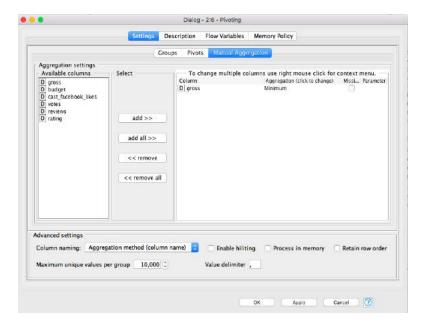
Performs a pivoting on the given input table using a selected number of columns for grouping and pivoting. The group columns will result into unique rows, whereby the pivot values turned into columns for each set of column combinations together with each aggregation method. In addition, the node returns the total aggregation (a) based on only the group columns and (b) based on only the pivoted columns resulting in a single row; optionally, with the total aggregation without pivoting.

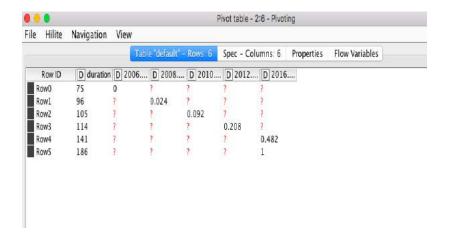
Steps:

- Connect pivot with boxplot and have the connection between them.
- Configure pivoting with 3 different columns for same data type for : Groups duration
 - Pivots year
 - Manual Aggregation gross
- Execute pivoting and checkout for the changes in the table.









❖ Joiner:

A Joiner node joins two tables together on one or more common key values. Possible join modes: inner join, left outer join, right outer join, full outer join. Two tabs: "Joiner Settings" and "Column Selection". "Joiner Settings" defines the parameters for the join operation: join mode and column keys. "Column Selection" sets which columns to keep and/or drop and strategies to deal with duplicate columns.

Steps:

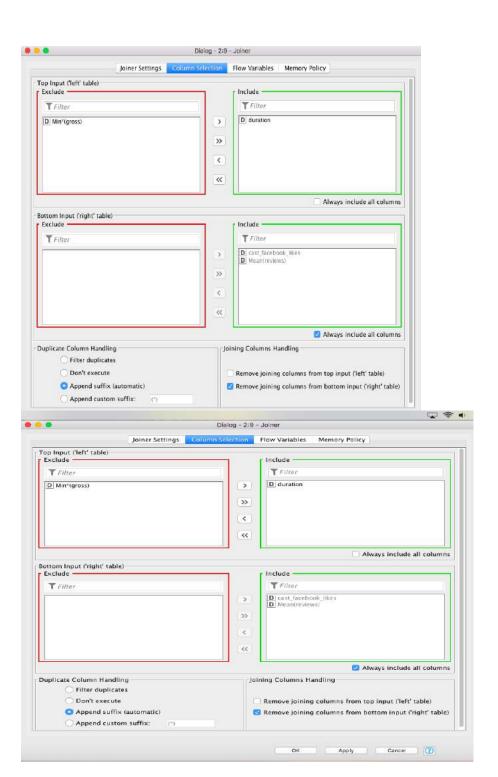
- Have the connection between joiner and pivot so that it is easy to analyse.
- Configure joiner with columns if necessary as:
- ➤ Top Input (left table):

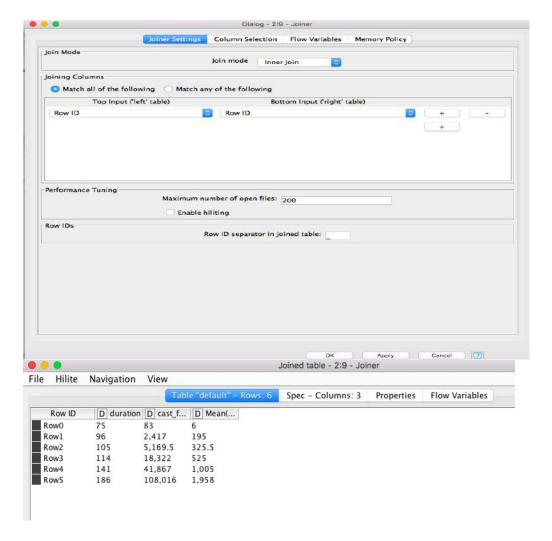
Include: duration

➤ Bottom Input(right table):

Include:

- →cast_facebook_likes
- → Mean
 - Execute the joiner and check for the final table.





Group By:

Groups the rows of a table by the unique values in the selected group columns. A row is created for each unique set of values of the selected group column. The remaining columns are aggregated based on the specified aggregation settings. The output table contains one row for each unique value combination of the selected group columns.

Steps:

- Make the connection to group by with the boxplot directly.
- Configure group by as the following:
- > Groups : year
- ➤ Manual Aggregation : gross
 - Execute the group by and check for the analysed table.

