DATA WAREHOUSING AND DATA MINING LABORATORY LIST OF EXPERIMENTS

| | NAME OF THE EXPERIMENT |
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| 1. | Listing applications for mining. |
| 2. | File format for data mining |
| 3. | Conversion of various data files |
| 4. | Training the given dataset for an application |
| 5. | Testing the given dataset for an application |
| 6. | Generating accurate models |
| 7. | Feature selection |
| 8. | Web mining |
| 9. | Text mining |
| 10. | Design of fact & dimension tables |

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| 7. | Feature selection | | |
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| 9. | Text mining | | |
| 10. | Design of fact & dimension tables | | |
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EX.NO:1

LISTING APPLICATIONS FOR MINING

AIM:

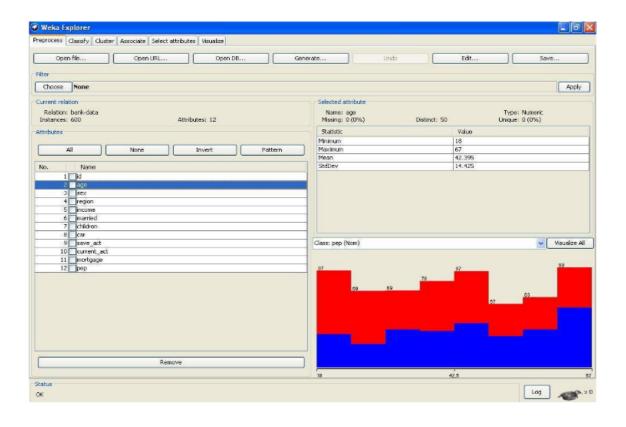
To list all the categorical (or nominal) attributes and the real-valued attributes separately.

RESOURCES: Weka mining tool1.

PROCEDURE:

- 1)Open the Weka GUI Chooser.
- 2)Select EXPLORER present in Applications.
- 3)Select Preprocess Tab.
- 4)Go to OPEN file and browse the file that is already stored in the system "bank.csv".
- 5)Clicking on any attribute in the left panel will show the basic statistics on that selected attribute.1.4

OUTPUT:



| Result: | |
|--|--|
| | |
| Thus the listing applications for the data mining was studied. | |
| | |

EX.NO:2

FILE FORMAT FOR DATA MINING

Aim: To study the file formats for the data mining.

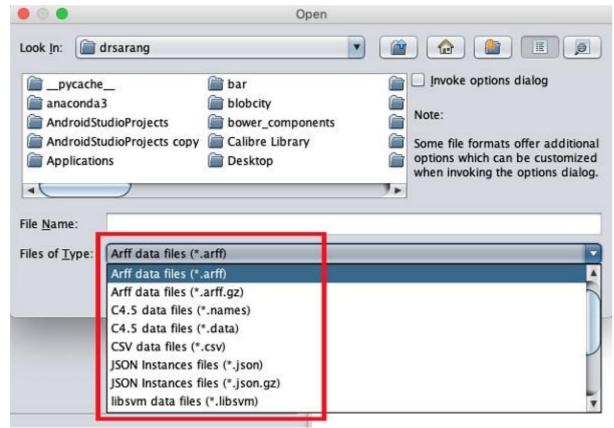
Introduction:

WEKA supports a large number of file formats for the data. The complete list of file formats are given here:

- 1. arff
- 2. arff.gz
- 3. bsi
- 4. csv
- 5. dat
- 6. data
- 7. json
- 8. json.gz
- 9. libsvm
- 10. m
- 11. names
- 12. xrff
- 13. xrff.gz

The types of files that it supports are listed in the drop-down list box at the bottom of the screen.

This is shown in the screenshot given below.



As you would notice it supports several formats including CSV and JSON.

The default file type is Arff.

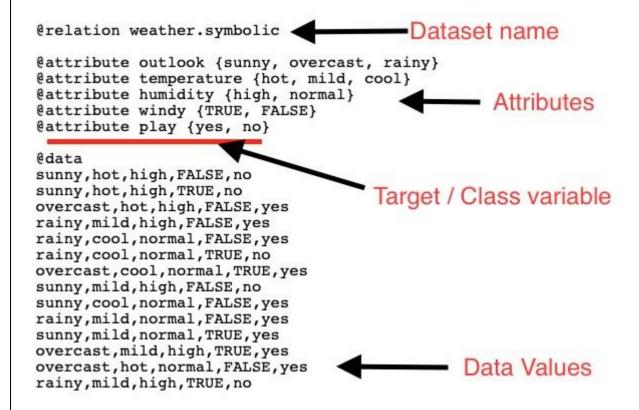
Arff Format

An Arff file contains two sections - header and data.

The header describes the attribute types.

The data section contains a comma separated list of data.

As an example for Arff format, the Weather data file loaded from the WEKA sample databases is shown below:



From the screenshot, you can infer the following points –

The @relation tag defines the name of the database.

The @attribute tag defines the attributes.

The @data tag starts the list of data rows each containing the comma separated fields.

The attributes can take nominal values as in the case of outlook shown here –

@attribute outlook (sunny, overcast, rainy)

The attributes can take real values as in this case –

@attribute temperature real

You can also set a Target or a Class variable called play as shown here –

@attribute play (yes, no)

The Target assumes two nominal values yes or no.

Result:

Thus the different file formats for the data mining was studied.

EX.NO:3a CONVERSION OF TEXT FILE INTO ARFF FILE

Aim:

To convert a text file to ARFF(Attribute-Relation File Format) using Weka3.8.2 tool.

Objectives:

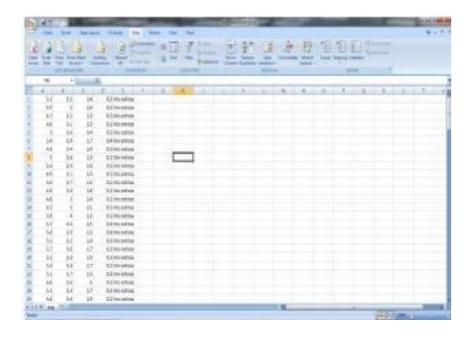
Most of the data that we have collected from public forum is in the text format that cannot be read by Weka tool. Since Weka (Data Mining tool) recognizes the data in ARFF format only we have to convert the text file into ARFF file.

Algorithm:

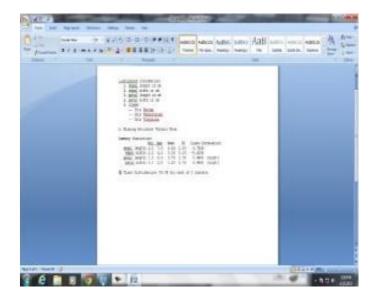
- 1. Download any data set from UCI data repository.
- 2. Open the same data file from excel. It will ask for delimiter (which produce column) in excel.
- 3. Add one row at the top of the data.
- 4. Enter header for each column.
- 5. Save file as .CSV (Comma Separated Values) format.
- 6. Open Weka tool and open the CSV file.
- 7. Save it as ARFF format.

Output:

Data Text File:



Data ARFF File:



Result:

 $Thus, conversion \ of \ a \ text \ file \ to \ ARFF (Attribute-Relation \ File \ Format) \ using \ Weka 3.8.2 \ tool \ is \ implemented.$

EX.NO:3b. CONVERSION OF ARFF TO TEXT FILE

Aim:

To convert ARFF (Attribute-Relation File Format) into text file.

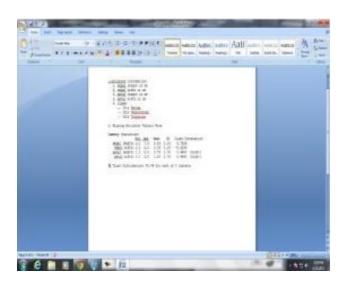
Objectives:

Since the data in the Weka tool is in ARFF file format we have to convert the ARFF file to text format for further processing.

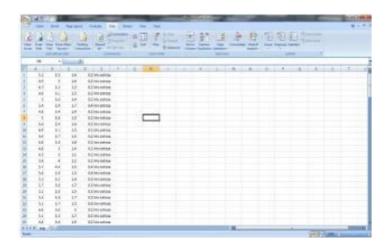
Algorithm:

- 1. Open any ARFF file in Weka tool.
- 2. Save the file as CSV format.
- 3. Open the CSV file in MS-EXCEL.
- 4. Remove some rows and add coreseponding header to the data.
- 5. Save it as text file with the desire delimiter.

Data ARFF File:



Data Text File:



Result: Thus conversion of ARFF (Attribute-Relation File Format) into text file is implemented.

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EX. No: 4 TRAINING THE GIVEN DATASET FOR AN APPLICATION

Aim:

To apply the concept of Linear Regression for training the given dataset.

Algorithm:

- 1. Open the weka tool.
- 2. Download a dataset by using UCI.
- 3. Apply replace missing values.
- 4. Apply normalize filter.
- 5. Click the Classify Tab.
- 6. Choose the Simple Linear Regression option.
- 7. Select the training set of data.
- 8. Start the validation process.
- 9. Note the output.

LINEAR REGRESSION:

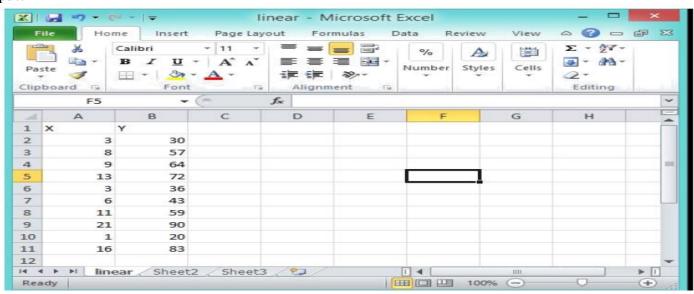
In statistics, Linear Regression is an approach for modeling a relationship between a scalar dependent variable Y and one or more explanatory variables denoted X.the case of explanatory variable is called Simple Linear Regression.

Coefficient of Linear Regression is given by: Y=ax+b

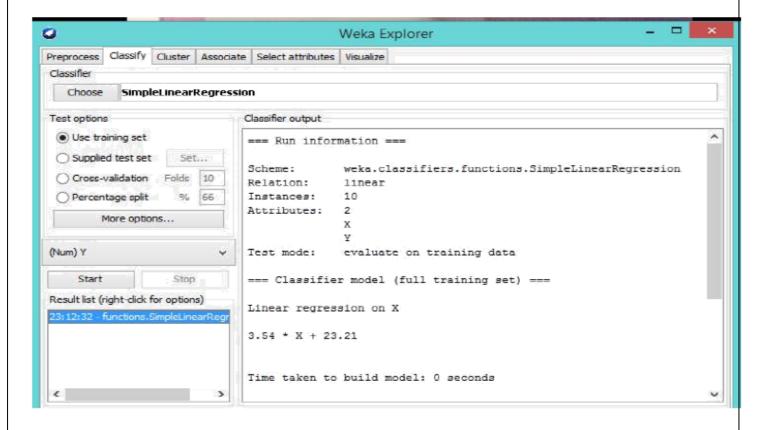
PROBLEM:

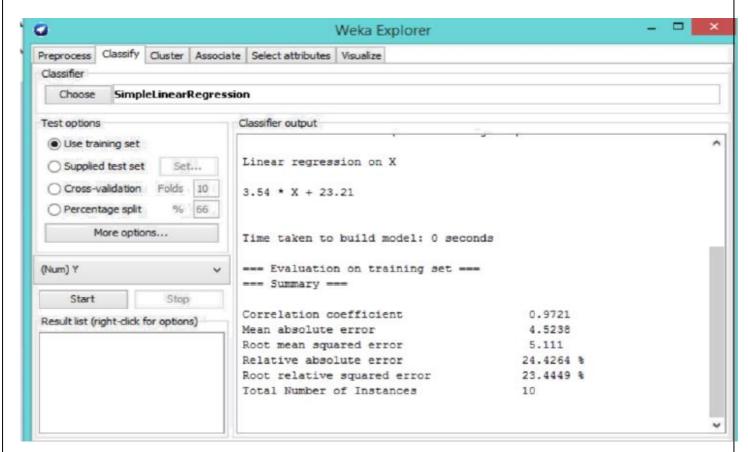
Consider the dataset below where x is the number of working experience of a college graduate and y is the corresponding salary of the graduate. Build a regression equation and predict the salary of college graduate whose experience is 10 years.

Input:



Output:





Result: Thus the concept of Linear Regression for training the given dataset is applied and implemented.

EX. No: 5 TESTING THE GIVEN DATASET FOR AN APPLICATION

Aim:

To apply the Navie Bayes Classification for testing the given dataset.

Algorithm:

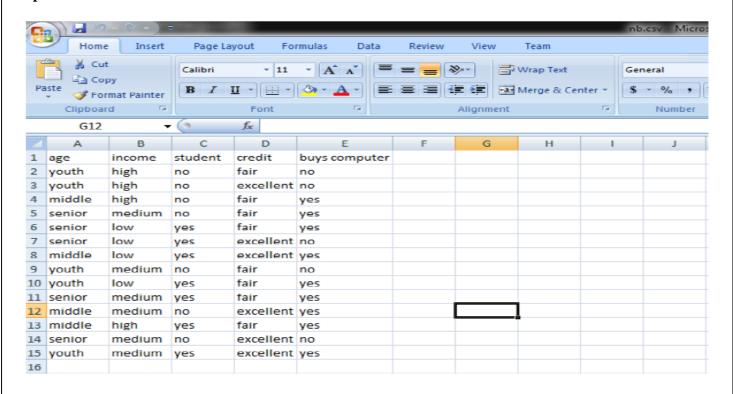
- 1. Open the weka tool.
- 2. Download a dataset by using UCI.
- 3. Apply replace missing values.
- 4. Apply normalize filter.
- 5. Click the Classification Tab.
- 6. Apply Navie Bayes Classification.
- 7. Find the Classified Value.
- 8. Note the output.

Bayes' Theorem In the Classification Context:

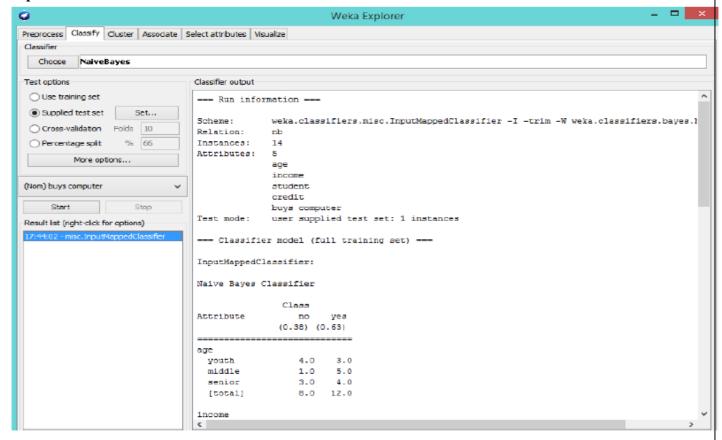
X is a data tuple. In Bayesian term it is considered "evidence". H is some hypothesis that X belongs to a specified class C. P(H|X) is the posterior probability of H conditioned on X.

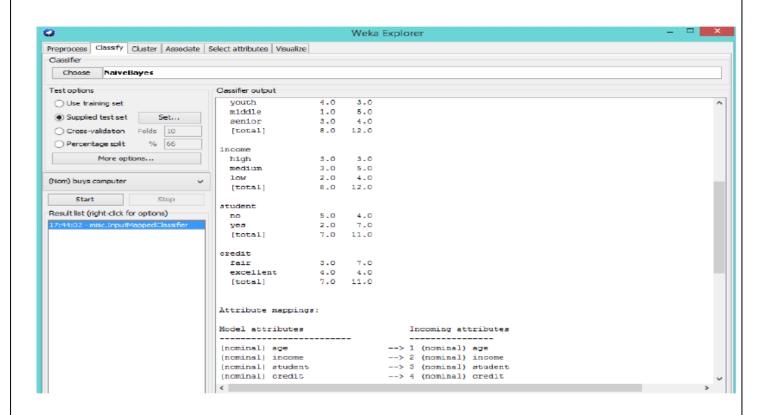
Example: predict whether a costumer will buy a computer or not "Costumers are described by two attributes: age and income "X is a 35 years-old costumer with an income of 40k" X is the hypothesis that the costumer will buy a computer "X is a 35 years-old costumer X will buy a computer given that we know the costumers age and income.

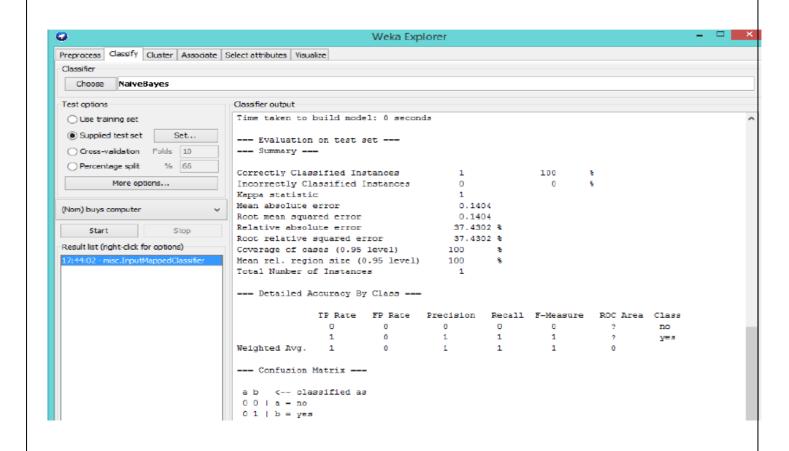
Input Data:



Output data:







Result:

Thus the Navie Bayes Classification for testing the given dataset is implemented.

EX. No: 6 GENERATE ACCURATE MODEL

Aim:

To find the good result (by improving the performance) using the training set and testing data set for numerical values.

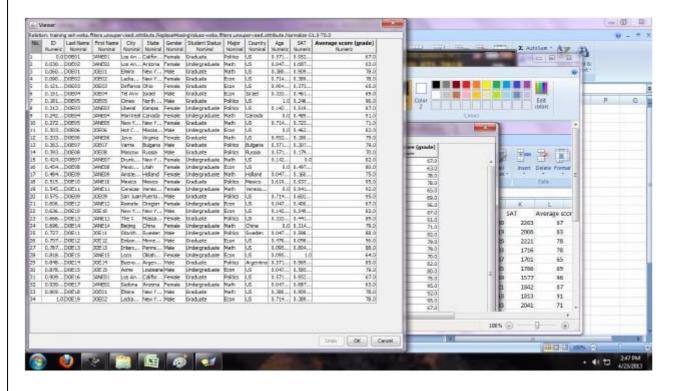
Objectives:

To develop training and testing data using numerical data set in order to get accurate model for classification.

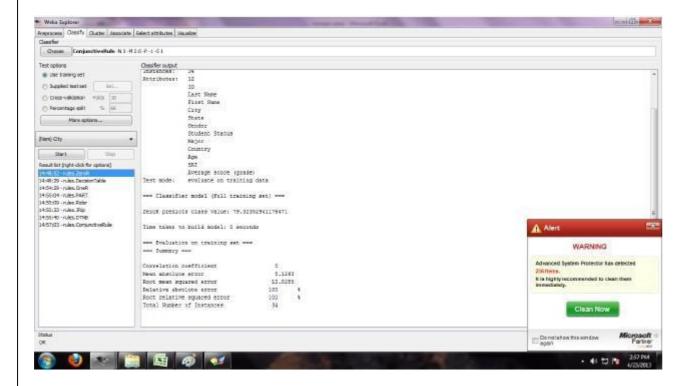
ALGORITHM:

- 1. Download any data set.
- 2. Save the file with .ARFF format.
- 3. Apply 'Replace Missing Values' filter.
- 4. Normalize the values by applying normalize flter.
- 5. Go to unsupervised instance remove percentage
- 6. Right click on that (show properties) option then select 70% true and save it as training.arff
- 7. Select the original data set then right click on show properties then give 70% false and save it as testing.arff
- 8. Select classification and apply various algorithms.

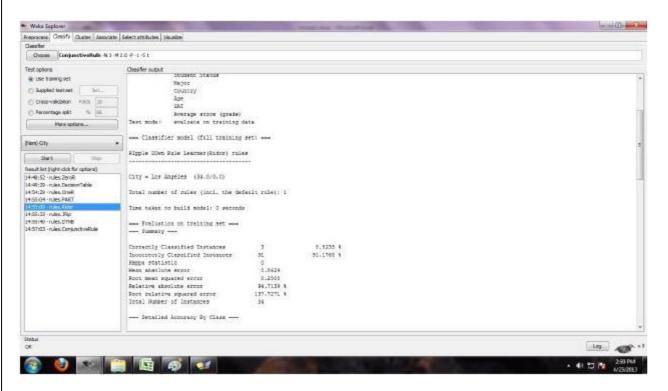
TRAINING DATA:



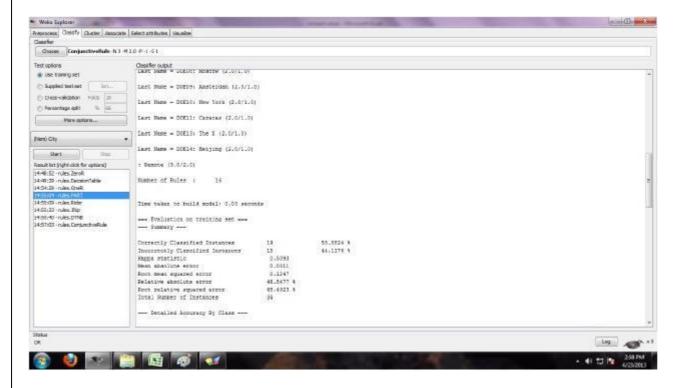
ZeroR:



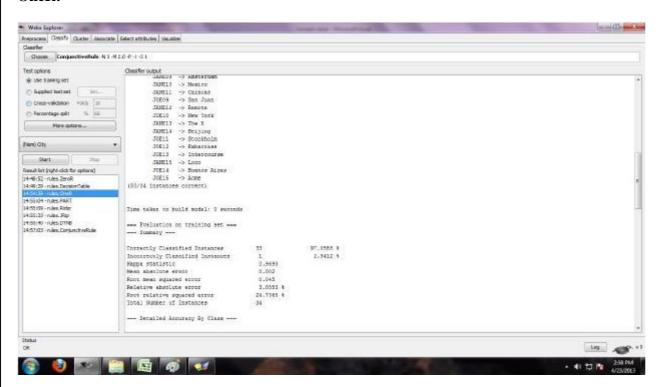
Ridor:



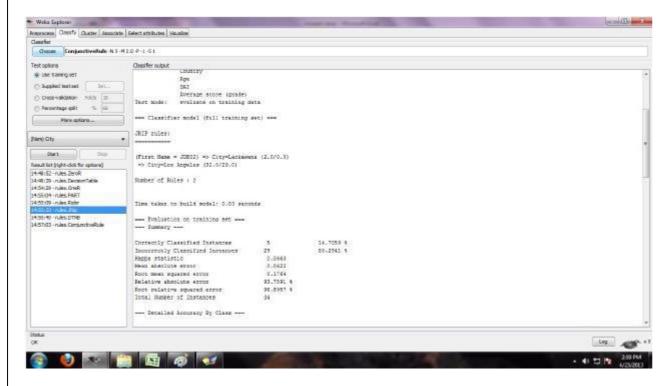
PART:



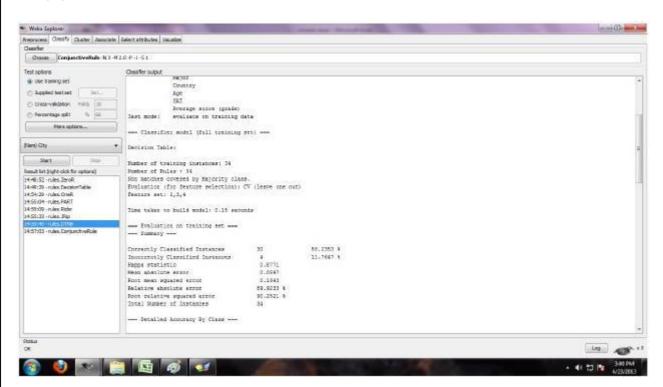
OneR:



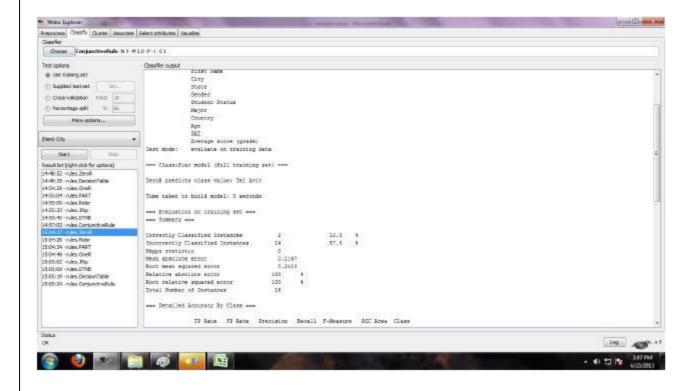
JRip:



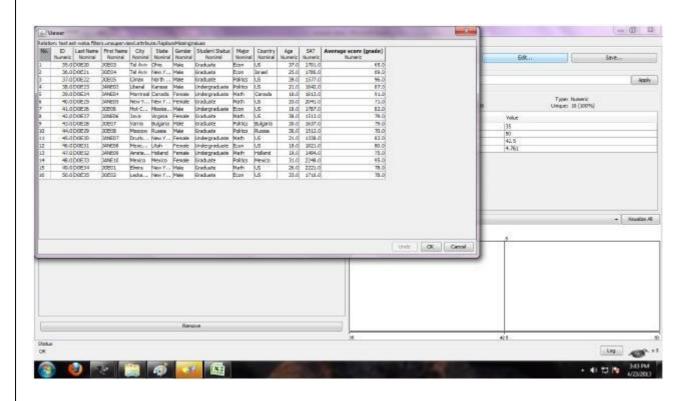
DTNB:



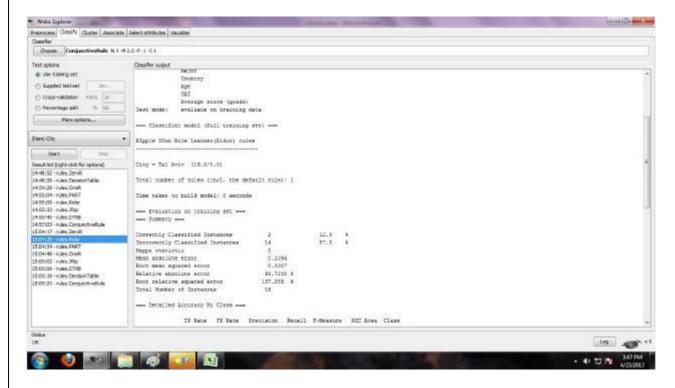
TEST DATA:



ZeroR:



Ridor:



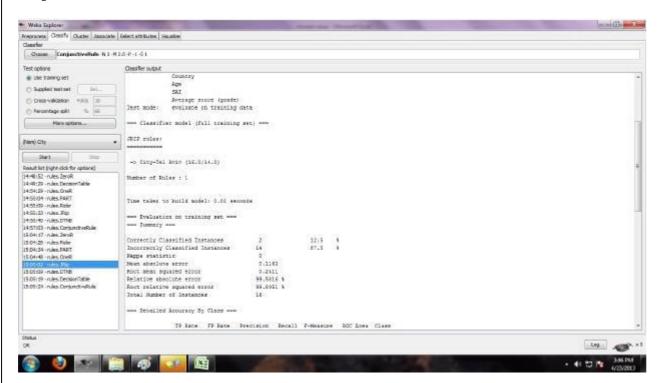
PART:



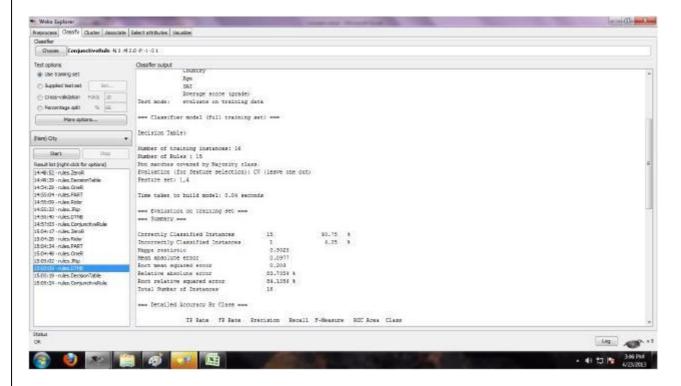
OneR:



JRip:



DTNB:



Result:

Thus, the good result (by improving the performance) using the training set and testing data set for numerical values is found out.

EX. No: 7 FEATURE SELECTION

AIM:

To find the good results by feature selection.

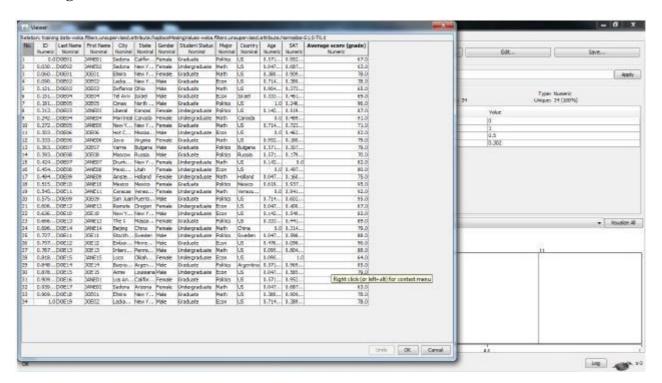
OBJECTIVES:

Any classifier/model has internal feature, those feature gives more accurate and optimal result.

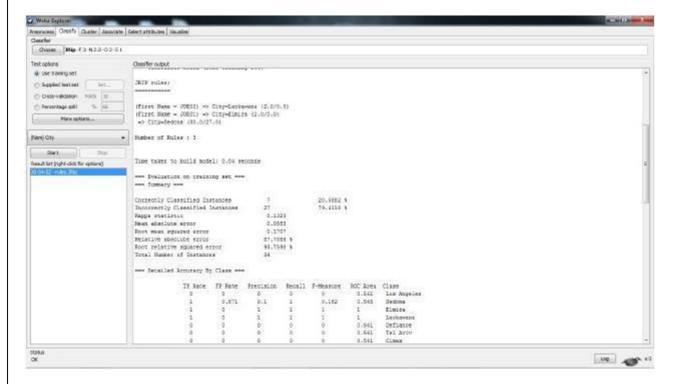
ALGORITHM:

- 1. Download any dataset with nominal values.
- 2. Save it as text.arff.
- 3. Split it into training and testing data set.
- 4. Go to unsupervised instance remove percentage.
- 5. Right click on that show properties then select 70% true and save it as training.arff
- 6. Right click on that show properties then select 70% false and save it as testing arff using original data set.
- 7. Open the parameter for classifying.
- 8. Fix the set of changing values.
- 9. Look at the performance.
- 10. Go to step 3 until the expected values of maximum value is reached.

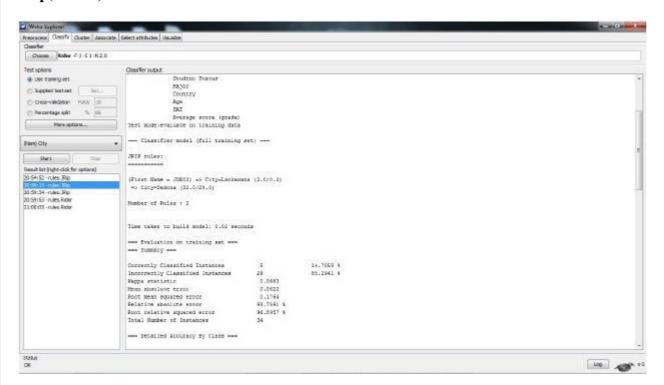
Training Data:



JRip(seed=1):



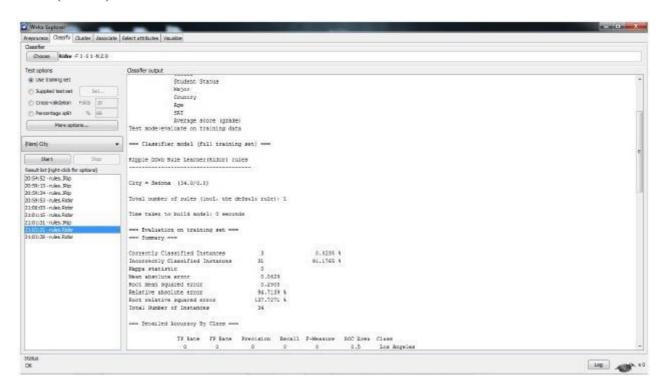
JRip(seed=2):



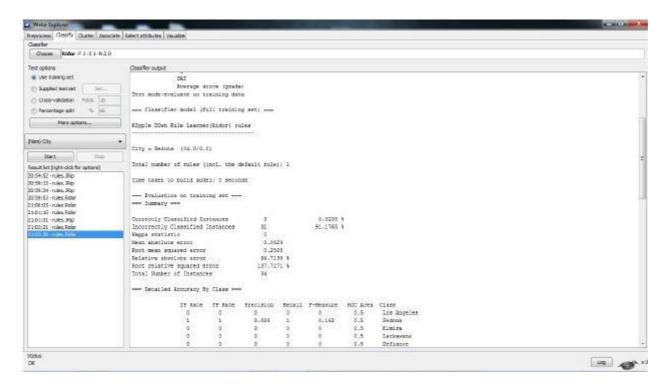
JRip(seed=3):



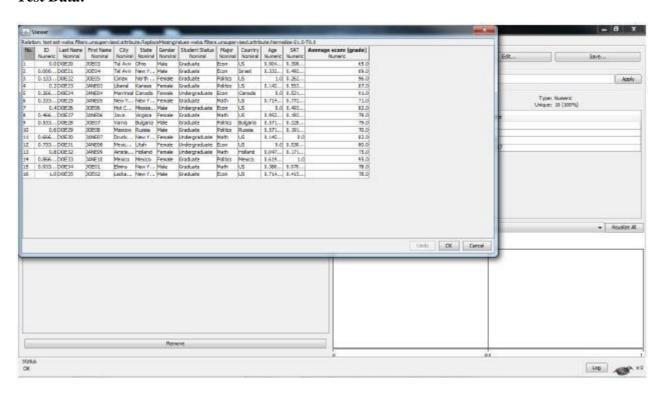
Ridor(seed=1):



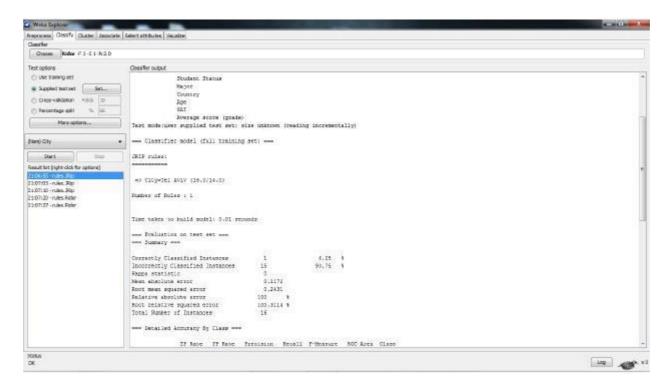
Ridor(seed=2):



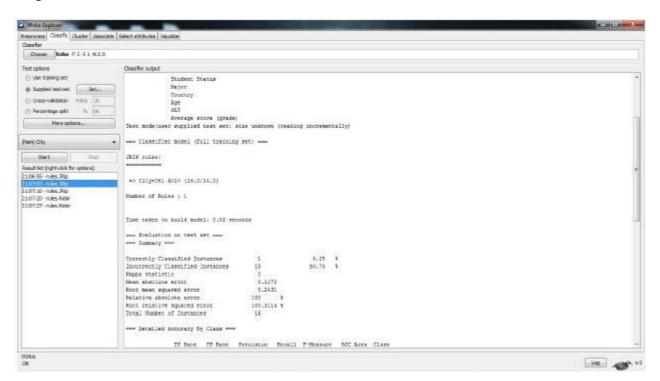
Test Data:



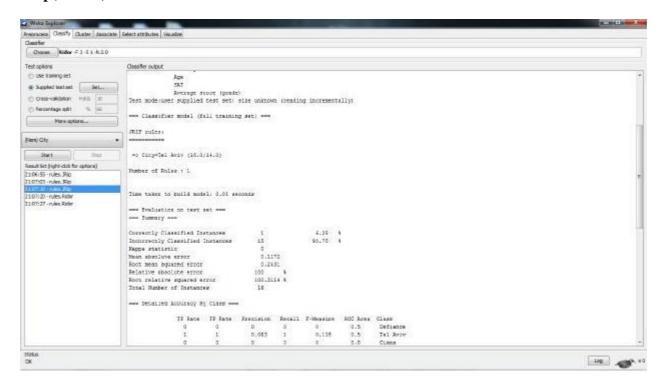
JRip(seed=1):



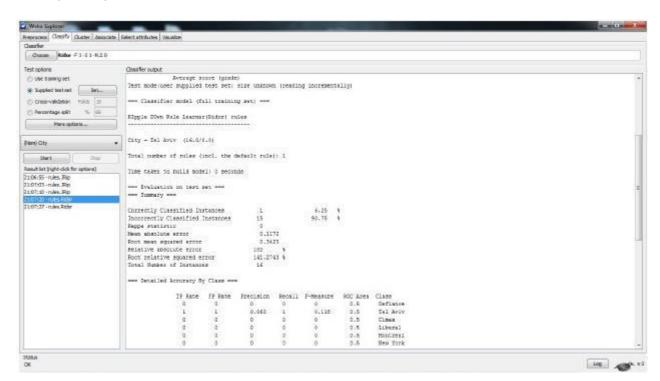
JRip(seed=2):



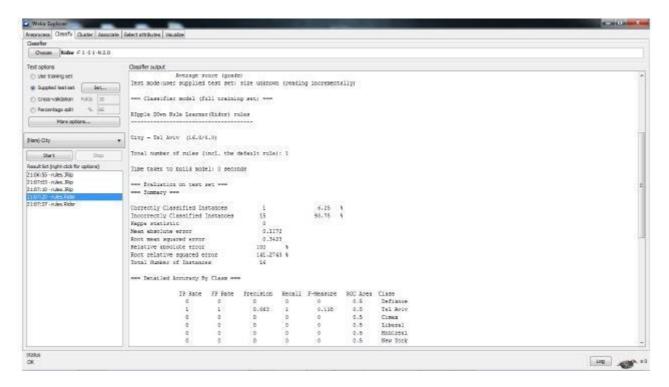
JRip(seed=3):



Ridor(seed=1):



Ridor(seed=2):



Training Data Set Performance:

| TRAINING SET | | | |
|--------------|----------------------|--|--|
| CLASSIFIER | PARAMETER SETTING | PERFORMANCE | |
| JRip | Seed=1 | Root Mean Squared Error=0.1707 Mean Absolute Error=0.0583 | |
| JRip | Seed =2 | Root Mean Squared Error=0.1764 Mean Absolute Error=0.0622 | |
| JRip | Seed =3 | Root Mean Squared Error=0.1764 Mean Absolute Error=0.0622 | |
| Ridor | Seed =1 | Root Mean Squared Error=0.2508 Mean Absolute Error=0.0629 | |
| Ridor | Seed=2 | Root Mean Squared Error=0.2508 Mean Absolute Error=0.0629 | |

Testing Data set Performance:

| TEST SET | | | |
|------------|----------------------|--|--|
| CLASSIFIER | PARAMETER SETTING | PERFORMANCE | |
| JRip | Seed=1 | Root Mean Squared Error=0.2431 Mean Absolute Error=0.1172 | |
| JRip | Seed =2 | Root Mean Squared Error=0.2431 Mean Absolute Error=0.1172 | |
| JRip | Seed =3 | Root Mean Squared Error=0.2431 Mean Absolute Error=0.1172 | |
| Ridor | Seed =1 | Root Mean Squared Error=0.3423 Mean Absolute Error=0.1172 | |
| Ridor | Seed=2 | Root Mean Squared Error=0.3423 Mean Absolute Error=0.1172 | |

Comparison between training and testing data set:

| TRAINING | | | |
|----------|---------|--------------------------------|--|
| JRip | Seed=1 | Root Mean Squared Error=0.1707 | |
| | | Mean Absolute Error=0.0583 | |
| Ridor | Seed =1 | Root Mean Squared Error=0.2508 | |
| | | Mean Absolute Error=0.0629 | |

| TEST | | | |
|-------|---------|--------------------------------|--|
| JRip | Seed=1 | Root Mean Squared Error=0.2431 | |
| | | Mean Absolute Error=0.1172 | |
| Rider | Seed =1 | Root Mean Squared Error=0.3423 | |
| | | Mean Absolute Error=0.1172 | |

Result:

Thus the good results by feature selection were found.

Aim:

To apply the web mining technique clustering algorithm for the given dataset.

Introduction to Web Mining:

Web mining is an application of data mining techniques to find information patterns from the web data. Web mining helps to improve the power of web search engine by identifying the web pages and classifying the web documents. Web mining is very useful to e-commerce websites and e-services.

Web Content Mining:

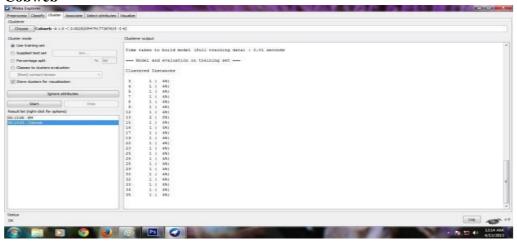
Web content mining can be used for mining of useful data, information and knowledge from web page content. Web structure mining helps to find useful knowledge or information pattern from the structure of hyperlinks. Due to heterogeneity and absence of structure in web data, automated discovery of new knowledge pattern can be challenging to some extent. Web content mining performs scanning and mining of the text, images and groups of web pages according to the content of the input (query), by displaying the list in search engines. For example: If an user wants to search for a particular book, then search engine provides the list of suggestions.

ALGORITHM:

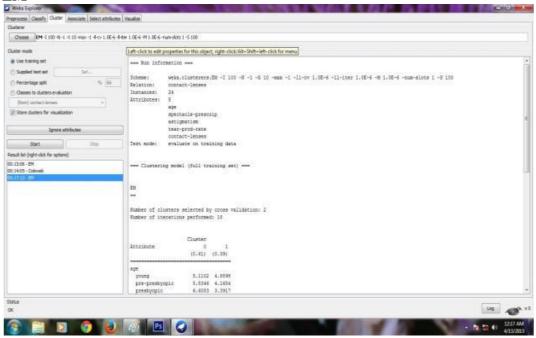
- 1. Open the weka tool.
- 2. Download a dataset by using UCI.
- 3. Apply replace missing values.
- 4. Apply normalize filter.
- 5. Click the cluster tab.
- 6. Apply all algorithms one by one.
- 7. Find the no of clusters that are formed
- 8. Note the output.

Output:

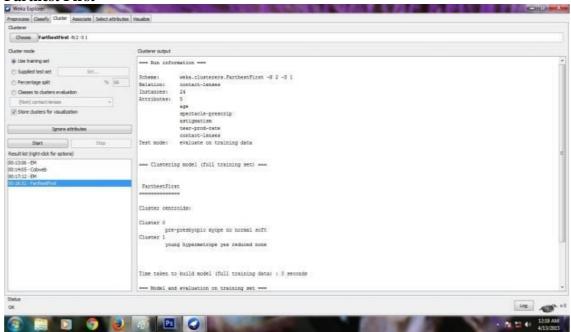
Cobweb



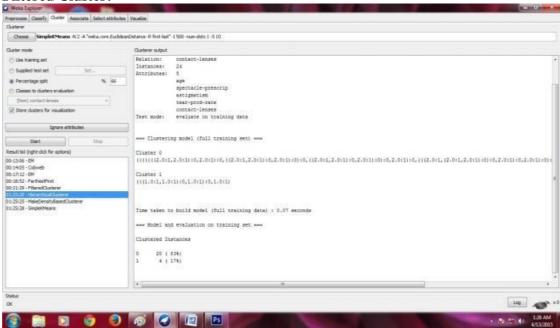
$\mathbf{E}\mathbf{M}$



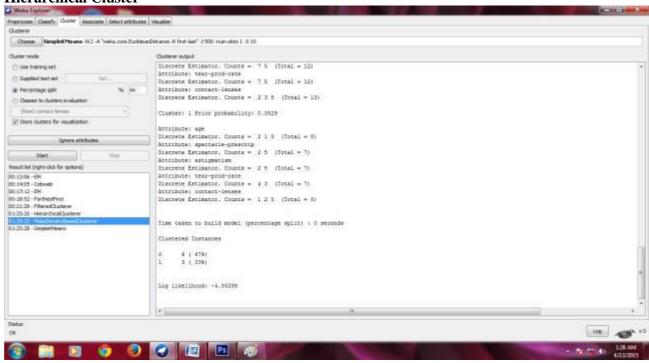
Farthest First



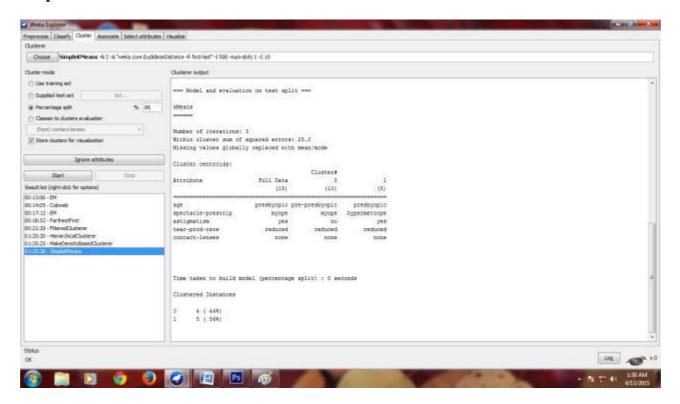
Filtered Cluster:



Hierarchical Cluster



Simple KMeans:



Result:

Thus the web mining technique clustering algorithm for the given dataset is implemented.

TEXT MINING

Aim:

To find association between data and to find the frequent item set for text mining.

Text Data Mining

Text data mining can be described as the process of extracting essential data from standard language text. All the data that we generate via text messages, documents, emails, files are written in common language text. Text mining is primarily used to draw useful insights or patterns from such data. The purchasing of one product when another product is purchased represents an association rule. Association rules are frequently used by retail store to assist in marketing, advertising, floor placement, and inventory control. Association rules are used to show the relationship between data items.

Keyword-based Association Analysis in text mining:

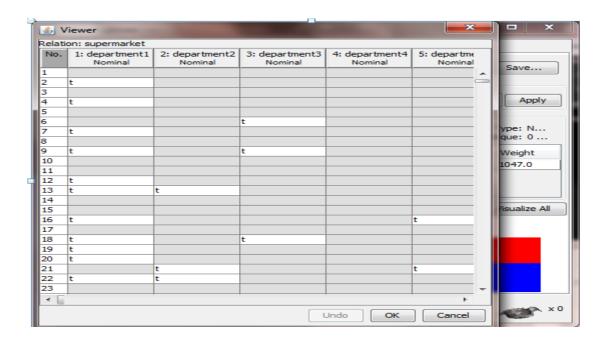
It collects sets of keywords or terms that often happen together and afterward discover the association relationship among them. First, it preprocesses the text data by parsing, stemming, removing stop words, etc. Once it pre-processed the data, then it induces association mining algorithms. Here, human effort is not required, so the number of unwanted results and the execution time is reduced.

ALGORITHM:

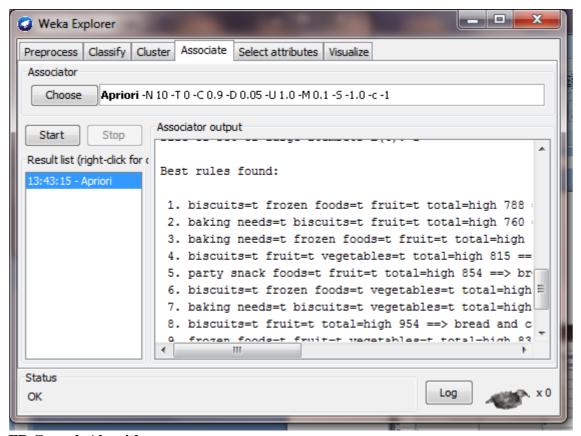
- 1. Open dataset
- 2. Select associate
- 3. Choose different algorithm for association
- 4. Observe the performance
- 5. Select the association rule with the maximum confidence rule.

INPUT:

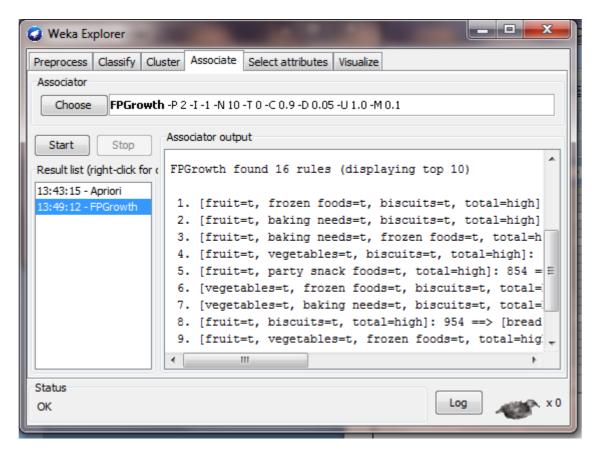
SuperMarket data set

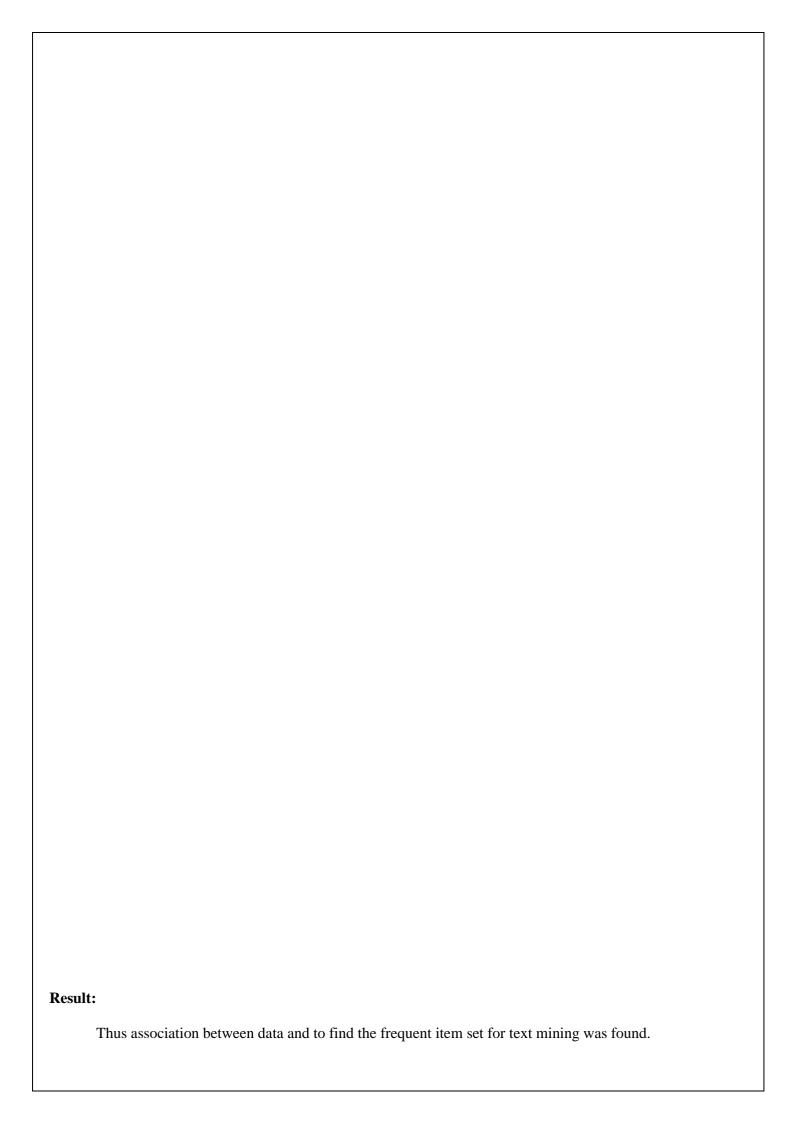


OUTPUT: Apriori Algorithm



FP-Growth Algorithm:





EX. No: 110 DESIGN OF FACT AND DIMENSION TABLES

Aim:

To design fact and dimension tables.

Fact Table:

A fact table is used in the dimensional model in data warehouse design. A fact table is found at the center of a star schema or snowflake schema surrounded by dimension tables. A fact table consists of facts of a particular business process e.g., sales revenue by month by product. Facts are also known as measurements or metrics. A fact table record captures a measurement or a metric.

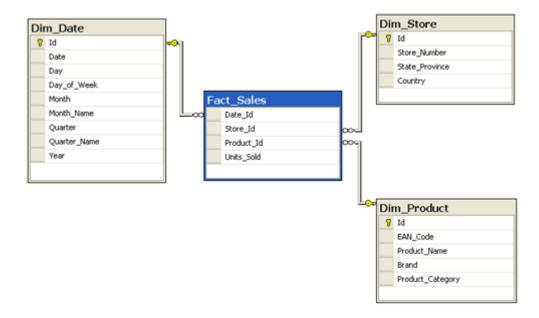
Designing fact table steps

Here is overview of four steps to designing a fact table:

- 1. **Choosing business process to model** The first step is to decide what business process to model by gathering and understanding business needs and available data
- 2. **Declare the grain** by declaring a grain means describing exactly what a fact table record represents
- 3. **Choose the dimensions** once grain of fact table is stated clearly, it is time to determine dimensions for the fact table.
- 4. **Identify facts** identify carefully which facts will appear in the fact table.

Fact table FACT_SALES that has a grain which gives us a number of units sold by date, by store and by product.

All other tables such as DIM_DATE, DIM_STORE and DIM_PRODUCT are dimensions tables. This schema is known as the star schema.



Result: Thus design fact and dimension tables are created.