

Assignment No. 3

Title

Apply a-priori algorithm to find frequently occurring items from given data and generate strong association rules using support and confidence thresholds.

Problem Definition:

Market Basket Analysis

Prerequisite:

Basic Concepts of ETL

Software Requirements:

Rapid Miner/ Weka

Learning Objectives:

Model associations between products by determining sets of items frequently purchased together and building association rules to derive recommendations.

Outcomes:

Create association rules which can be used for product recommendations depending on the confidences of the rules

Theory Concepts:

Association rule for mining:

- Proposed by R Agrawal and R Srikant in 1994.
- It is an important data mining model studied extensively by the database and data mining community.
- Assume all data are categorical.
- Initially used for Market Basket Analysis to find how items purchased by customers are related.

The Apriori algorithm:

- The best known algorithm
- Two steps:
 - Find all item sets that have minimum support (frequent item sets, also called large item sets).
 - It Create Association rule with support and Confidence.
 - E.g. if we buy tooth brush : it suggest Colgate and tongue cleaner

Data Set

T-Id	Item Set
T-1000	M,O,N,K,E,Y
T-1001	D,O,N,K,E,Y
T-1002	M,A,K,E
T-1003	M,U,C,K,Y
T-1004	C,O,O,K,E

Table 3.1: Data Set

Given: Minimum **Support** = 60%
Minimum **Confidence** = 80%

Candidate Table C1: Now find support count of each item set

Item Set	Support Count
M	3
O	4
N	2
E	4
Y	3
D	1
A	1
U	1
C	2
K	5

Table 3.2: Candidate Table C1

- Now find out minimum Support
- $\text{Support} = 60/100 * 5$
=3
- Where 5 is Number of entry
- Compare Min Support with each item set

L1 Support Count

Item Set	Support Count
M	3
O	4
K	5
E	4
Y	3

Table 3.3: L1 Support Count

Candidate Table C2:

Item Set	Support Count
MO	1
MK	3
ME	2
MY	2
OK	3
OE	3
OY	2
KE	4
KY	3
EY	2

Table 3.4: Candidate Table C2

- Now again Compare C2 with Min Support 3

L2 Support Count

Item Set	Support Count
MK	3
OK	3

OE	3
KE	4
KY	3

Table 3.5: L2 Support Count

- After satisfied minimum support criteria
- Make Pair to generate C3

Candidate Table C3

Item Set	Support count
M,K,O	1
M,K,E	2
M,K,Y	2
O,K,E	3
O,K,Y	2

Table 3.6: Candidate Table C3

L3 Support Count

Now again compare the item set with min support 3

Item Set	Support Count
O,K,E	3

Table 3.7: L3 Support Count

Now create association rule with support and Confidence for {O,K,E}

- Confidence = Support/No. of time it Occurs

Association Rule	Support	Confidence	Confidence (%)
$O \wedge K \Rightarrow E$	3	$3/3 = 1$	$1*100=100$
$O \wedge E \Rightarrow K$	3	$3/3 = 1$	$1*100=100$
$K \wedge E \Rightarrow O$	3	$3/4 = 0.75$	$0.75*100=75$
$E \Rightarrow O \wedge K$	3	$3/4 = 0.75$	$0.75*100=75$
$K \Rightarrow O \wedge E$	3	$3/5 = 0.6$	$0.6*100=60$
$O \Rightarrow K \wedge E$	3	$3/4 = 0.75$	$0.75*100=75$

Table 3.8: Association Rule

- Compare this with Minimum Confidence=80%

Rule	Support	Confidence
$O \wedge K \Rightarrow E$	3	100
$O \wedge E \Rightarrow K$	3	100

Table 3.9: Support and Confidence

Hence final Association rule are

$\{O \wedge K \Rightarrow E\}$

$\{O \wedge E \Rightarrow K\}$

- From first observation we predict that if the customer buy item O and item K then defiantly he will by item E
- From Second observation we predict that the customer buy item O and item E then defiantly he will by item K

Market Basket Analysis using Rapid Miner

Rapid Miner is a data science software platform developed by the company of the same name that provides an integrated environment for data preparation, machine learning, deep learning, text mining, and predictive analytics. It is used for business and commercial applications as well as for research, education, training, rapid prototyping, and application development and supports all steps of the machine learning process including data preparation, results visualization, model validation and optimization Rapid Miner is developed on an open core model. The Rapid Miner Studio Free Edition, which is limited to 1 logical processor and 10,000 data rows, is available under the AGPL license.

Commercial pricing starts at \$2,500 and is available from the developer.

MARKET BASKET ANALYSIS

Model associations between products by determining sets of items frequently purchased together and building association rules to derive recommendations.

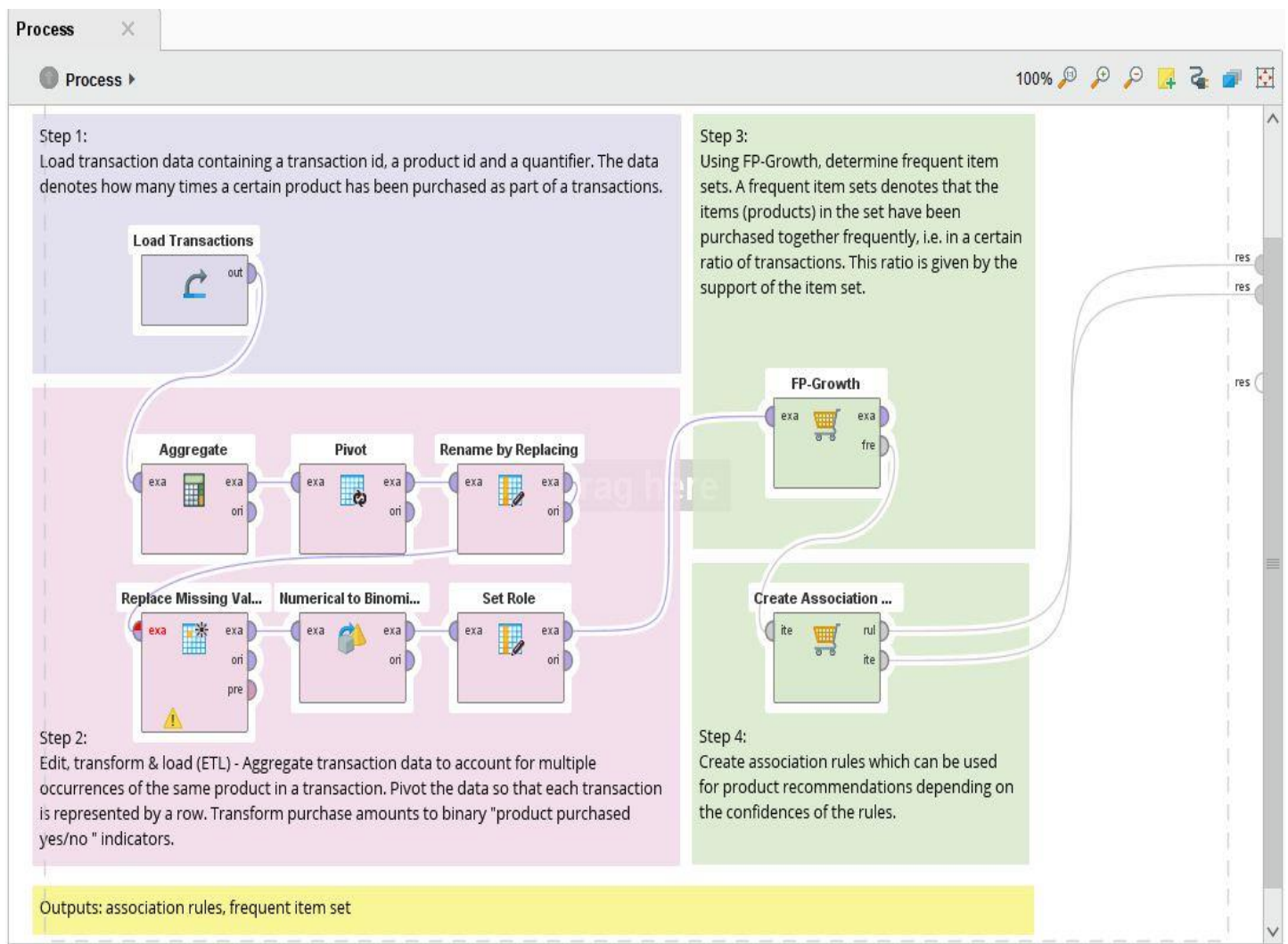


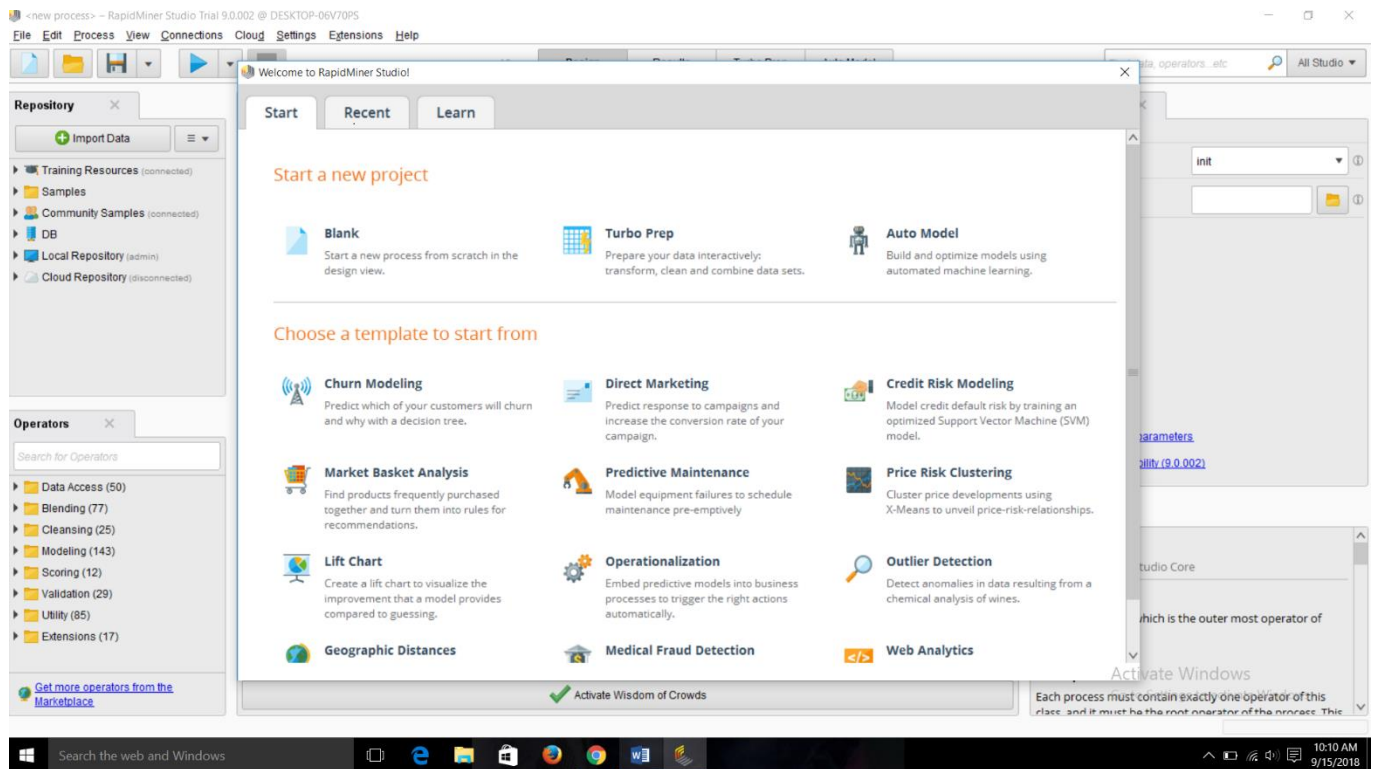
Figure 3.1: MARKET BASKET ANALYSIS

No. of Sets: 47					
Total Max. Size: 3					
Min. Size: 1					
Max. Size: 3					
Contains Item:					
Update View					
Size	Support	Item 1	Item 2	Item 3 ↓	
3	0.006	Product 12	Product 20	Product 27	
3	0.006	Product 11	Product 12	Product 20	
3	0.006	Product 11	Product 20	Product 19	
1	0.138	Product 11			
1	0.136	Product 12			
1	0.103	Product 20			
1	0.079	Product 10			
1	0.079	Product 18			
1	0.079	Product 23			
1	0.073	Product 15			
1	0.071	Product 26			
1	0.067	Product 13			
1	0.059	Product 21			

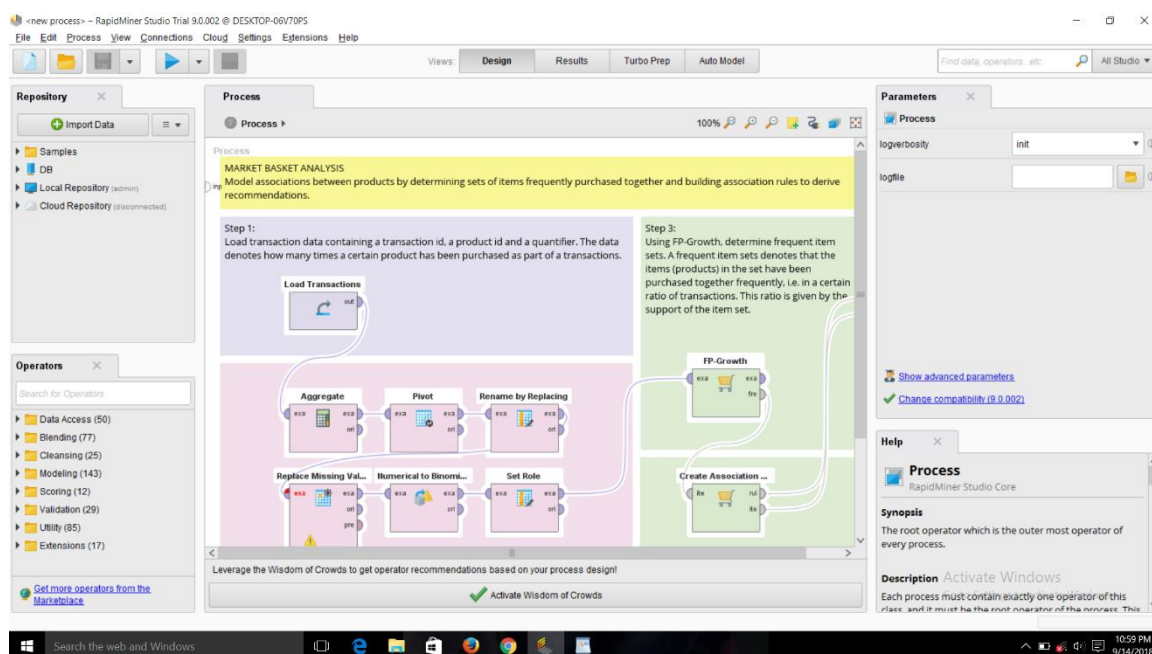
Figure 3.2: Frequent Item Sets (FP Growth)

Screenshot and Explanation of Generating Association Rules without using Readymade Template using RapidMiner

1. Select Market Basket Analysis



2. Open Market basket analysis



3. Select A-priori

The screenshot shows the RapidMiner Studio 9.0.002 interface. The main workspace displays a process design for "MARKET BASKET ANALYSIS". The process starts with a "Load Transactions" operator, followed by "Aggregate", "Pivot", and "Rename by Replacing" operators. These are connected to "FP-Growth" and "Create Association ..." operators. The "Parameters" panel on the right shows settings for "logverbosity" (init) and "logfile". The "Help" panel on the right provides information about the "Process" operator, including its synopsis and description. The "Repository" panel on the left shows the "Local Repository" with folders for "Samples", "DB", "data", "processes", and "A-priori". The "Operators" panel on the left lists various operators such as "Data Access", "Blending", "Cleansing", "Modeling", "Scoring", "Validation", "Utility", and "Extensions".

The screenshot shows the RapidMiner Studio 9.0.002 interface. The main workspace displays a process design for "A-priori". The process starts with a "Retrieve Transactions" operator, followed by an "Execute Market basket analysis" operator. The "Parameters" panel on the right shows settings for "logverbosity" (init) and "logfile". The "Help" panel on the right provides information about the "Process" operator, including its synopsis and description. The "Repository" panel on the left shows the "Local Repository" with folders for "Samples", "DB", "data", "processes", and "A-priori". The "Operators" panel on the left lists various operators such as "Data Access", "Blending", "Cleansing", "Modeling", "Scoring", "Validation", "Utility", and "Extensions".

4. Click Run

MARKET BASKET ANALYSIS
Model associations between products by determining sets of items frequently purchased together and building association rules to derive recommendations.

Step 1: Load transaction data containing a transaction id, a product id and a quantifier. The data denotes how many times a certain product has been purchased as part of a transactions.

Step 2: Using FP-Growth, determine frequent item sets. A frequent item sets denotes that the items (products) in the set have been purchased together frequently, i.e. in a certain ratio of transactions. This ratio is given by the support of the item set.

Parameters:
logverbosity: init
logfile: [empty]

Help:
Process
Synopsis: The root operator which is the outer most operator of every process.
Description: Activate Windows
Each process must contain exactly one operator of this class, and it must be the root operator of the process. This...

5. Click Result

FP-Growth

Result History
FrequentItemSets (FP-Growth)

Size	Support	Item 1	Item 2	Item 3
1	0.138	Product 11		
1	0.136	Product 12		
1	0.103	Product 20		
1	0.079	Product 10		
1	0.079	Product 18		
1	0.079	Product 23		
1	0.073	Product 15		
1	0.071	Product 26		
1	0.067	Product 13		
1	0.059	Product 21		
1	0.057	Product 24		
1	0.049	Product 19		
1	0.049	product 1		
1	0.047	Product 16		
1	0.043	Product 14		
1	0.037	Product 29		
1	0.028	Product 25		
1	0.028	Product 27		
1	0.024	Product 17		

Association Rules

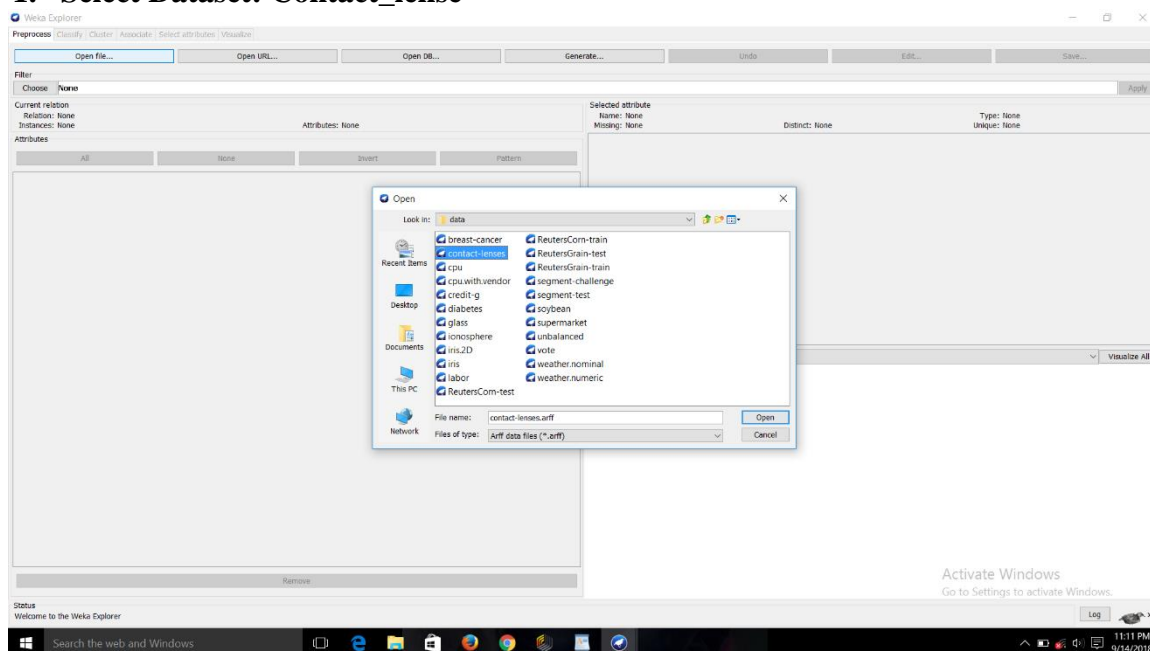
Result History

AssociationRules (Create Association Rules)

No.	Premises	Conclusion	Support	Confidence
18	Product 12	Product 20	0.026	0.194
19	Product 27	Product 12	0.006	0.214
20	Product 27	Product 12, Product 20	0.006	0.214
21	Product 29	Product 20	0.008	0.222
22	Product 12, Product 20	Product 11	0.006	0.231
23	Product 12, Product 20	Product 27	0.006	0.231
24	Product 11	Product 20	0.034	0.250
25	Product 19	Product 20	0.012	0.250
26	Product 20	Product 12	0.026	0.255
27	product 1	Product 12	0.014	0.292
28	Product 20	Product 11	0.034	0.333
29	Product 12	Product 15	0.047	0.343
30	Product 22	Product 12	0.008	0.364
31	Product 31	Product 12	0.010	0.417
32	Product 20, Product 27	Product 12	0.006	0.429
33	Product 27	Product 20	0.014	0.500
34	Product 20, Product 19	Product 11	0.006	0.500
35	Product 11, Product 12	Product 20	0.006	0.600
36	Product 15	Product 12	0.047	0.639

Screenshot and Explanation of Generating Association Rules using Apriori algorithm using Weka

1. Select Dataset: Contact_lense



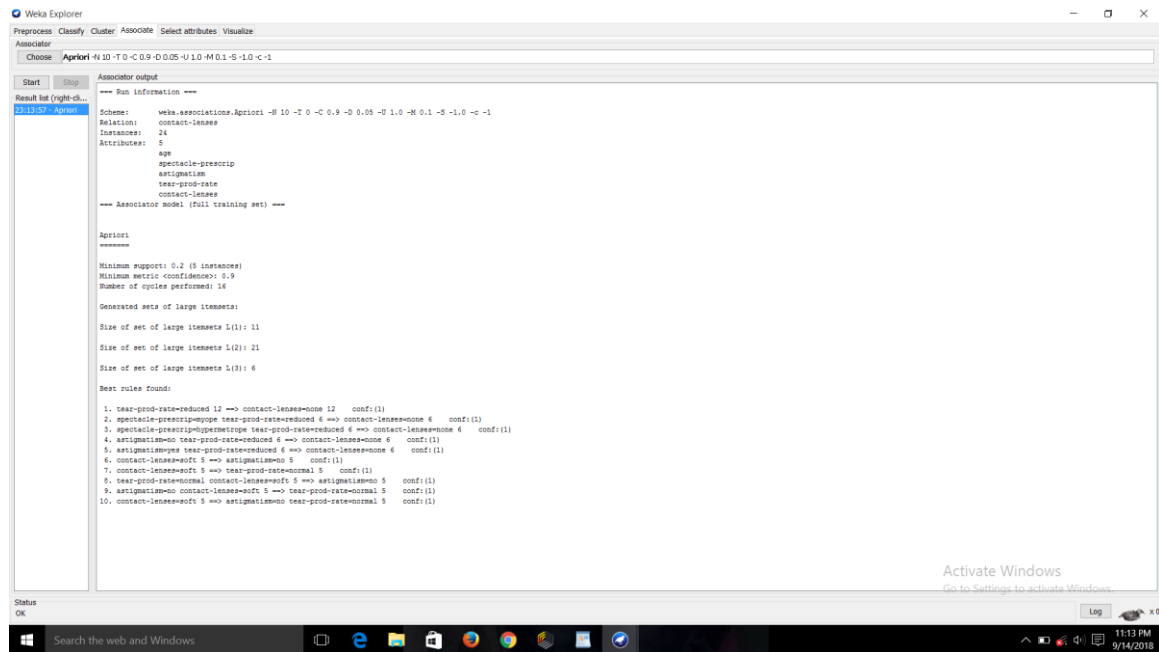
2.Dataset open

The Weka Explorer window displays the 'contact-lenses' dataset. The 'Attributes' list on the left includes 'age', 'sex', 'astigmatism', 'tear-prod-rate', and 'contact-lenses'. The 'Selected attribute' pane on the right shows the 'age' attribute with 3 distinct values: 'young' (8 instances), 'pre-presbyopic' (8 instances), and 'presbyopic' (8 instances). The 'Visualize All' button is visible, and the resulting stacked bar chart shows three bars, each representing an age group, with segments for 'young' (cyan), 'pre-presbyopic' (red), and 'presbyopic' (blue).

3.Click on A-priori

The Weka Explorer window shows the 'Classify' tab selected. The 'weka.gui.GenericObjectEditor' dialog box is open, displaying the configuration for the 'A-priori' algorithm. The 'Class' field is set to 'weka.associations.Apriori'. The 'Metric' is set to 'Confidence'. The 'Lower bound min support' is set to 0.1. The 'Upper bound min support' is set to 1.0. The 'Significance level' is set to -1.0. The 'Verbose' option is checked. The 'OK' button is highlighted.

4. Select Start Result



Expected Oral Questions: Answers

1. Explain Association Rule
2. What is the Application of A-Priori algorithm?
3. What is Market Basket Analysis? Explain with suitable example?

Conclusion

Thus we learn that to find frequently occurring items from given data and generate strong association rules using support and confidence thresholds using a-priori algorithm