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Subject :- Data Mining & Warehousing

Experiment 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.

Objectives :

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

Hardware Requirement :

Pentium or higher processor, 2GB RAM and 500 GB HDD.

Software Requirement :

Rapid Miner

Outcomes :

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

Theory :

- 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 : 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: 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: 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: 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: 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: 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: 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: 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: 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

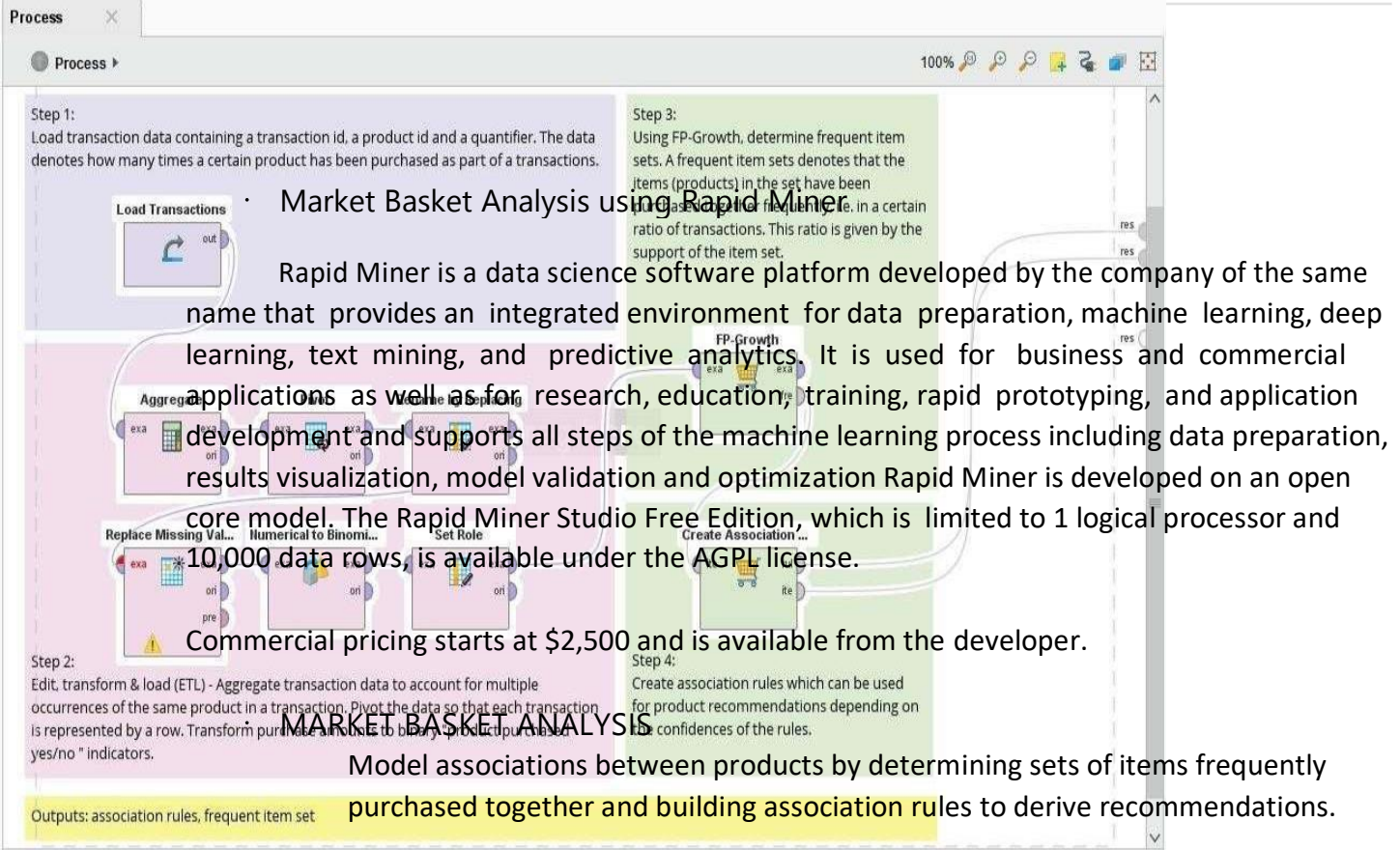


Figure: MARKET BASKET ANALYSIS

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: Frequent Item Sets (FP Growth)

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.