ASSOCIATION RULES MARKET BASKET ANALYSIS



Introduction to Market Basket Analysis

The most widely used area of application for association rules is Market Basket
 Analysis

Market Basket Analysis (Association Analysis) is a mathematical modeling technique based upon the theory that if you buy a certain group of items, you are likely to buy another group of items

 It is used to analyze the customer purchasing behavior and helps in increasing the sales and maintain inventory by focusing on the point of sale transaction data

Definitions and Terminology

Term	Definition
Transactions	A set of items (Item set)
	Ratio of number of times two or more items occur
Support	together to the total number of transactions
	Support can be thought of as P(A and B)
	Conditional probability that a randomly selected
Confidence	transaction will include Item B given Item A
	P(B A) (written as A => B)
	Ratio of the probability of Items A and B occurring
Lift	together (Joint probability) to the product of P(A) and
	P(B)

Rule Evaluation - Support

Transaction No.	Item 1	Item 2	Item 3	•••
100	Beer	Diaper	Chocolate	
101	Milk	Chocolate	Shampoo	
102	Beer	Wine	Vodka	
103	Beer	Cheese	Diaper	
104 A	Ic ® Cream	Diaper	Beer	

Support of {Diaper, Beer}

$$Support = \frac{\text{No.of transactions containing both A and B}}{\text{Total no.of transactions}} = \frac{3}{5} = 60\%$$

Support of {Diaper, Beer} is 3/5

Rule Evaluation - Confidence

Transaction No.	Item 1	Item 2	Item 3	•••
100	Beer	Diaper	Chocolate	
101	Milk	Chocolate	Shampoo	
102	Beer	Wine	Vodka	
103	Beer	Cheese	Diaper	
104	Ice Cream No	Diaper of transactions	Beer containing both	A and

Confidence for $\{A\} \Rightarrow \{B\} = \frac{}{\text{No. of transactions containing A}}$

Confidence for $\{Diaper\} \Rightarrow \{Beer\} \text{ is } 3/3$

When Diaper is purchased, the likelihood of Beer purchase is 100%

Confidence for {Beer} ⇒ {Diaper}is 3/4

When Beer is purchased, the likelihood of Diaper purchase is 75%

{Diaper} ⇒ {Beer}is a more important rule according to Confidence

Rule Evaluation - Lift

Transaction No.	Item 1	Item 2	Item 3	Item 4
100	Beer	Diaper	Chocolate	
101	Milk	Chocolate	Shampoo	
102	Beer	Milk	Vodka	Chocolate
103	Beer	Milk	Diaper	Chocolate
104	Mi k k	₿iaper	Beer	

 $Consider \{Chocolate\} \Rightarrow \{Milk\}$

Lift =
$$\frac{P(A \cap B)}{P(A)P(B)} = \frac{\frac{3}{5}}{\left(\frac{4}{5}\right)\left(\frac{4}{5}\right)} = 0.9375$$

Lift < 1 indicates Chocolate is decreasing the chance of Milk purchase

Using MBA for Recommendations

- Support can be used for initial recommendations or to determine the layout of the catalog of an ecommerce site
- Confidence can be used to provide recommendations based on first product purchase.
- Use rules only if lift is greater than one.

Case Study

Background

• Transactions data collected from point of sales is generally in long format. Arules package requires the data to be in wide format.

Objective

 To convert available data to a format suitable for association analysis and conduct analysis via arules package in R

Available Information

- Each transaction is given a unique ID
- Items basket contains five items, items purchased during each transaction are recorded

Data Snapshot

Transactions Data for MBA

id	item
1	В
1	С
1	D
1	E
2	Α
2	В
2	С
2	D
2	E
3	Α
3	В

Columns	Description	Measurement	Possible
		medsarement	values
id	Transaction Id	-	Positive Integers
item	Items purchased	A,B,C,D,E	5

Data Conversion

#Convert the Data

```
library(arules)
trans<-read.transactions("Transactions Data for
MBA.csv",format="single",sep=",",cols=c("id","item"),header=TRUE)</pre>
```

□ **read.transactions()** in package arules reads a transactions data file and creates a

transaction object.

format= indicates the format of the dataset. "single" implies each line corresponds to a single item, containing at least ids for the transaction and the

item. **"basket"** implies each line in the transaction data file represents a transaction where the items (item labels) are separated by the characters specified

by **sep**.

For the "single" format, cols= is a numeric or character vector of length two giving the numbers or names of the columns (fields) with the transaction and

item ids, respectively.

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Data Conversion

#The converted data looks as follows:

inspect(trans)

```
transactionID
      items
[1]
      {B, C, D, E}
[2]
      \{A, C, D\}
                        10
[3]
      \{A, B, E\}
                       100
[4]
      {B, D, E}
                        11
[5]
      {B, D, E}
                        12
[6]
      {A, B, C, D, E} 13
    {A, C, D, E}
[7]
[8] {A, C}
                        15
[9] \{A, B, D\}
                        16
[10] \{B, C, E\}
                       17
[11] \{A, B, C, D\}
                       18
[12] {B, C}
                        19
[13]
      {A, B, C, D, E}
      \{A, B, C, D\}
[14]
                        20
[15]
      {A, C, D}
                        21
[16]
      \{A, E\}
                        22
[17]
      \{A, C, D\}
                        23
[18]
      \{A, E\}
                        24
      \{A, C, E\}
[19]
                        25
```

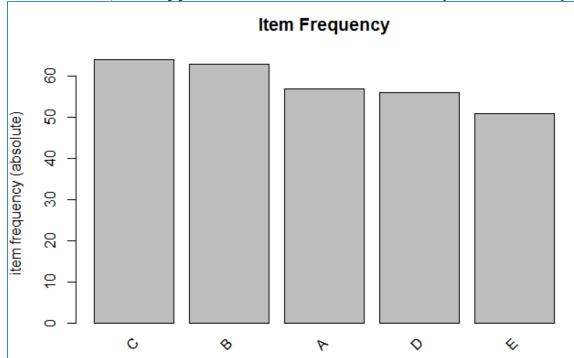
#Visualise Frequency

itemFrequencyPlot(trans,topN=5,type="absolute")

- □ **itemFrequencyPlot()** calculates item frequency and returns a barplot.
- topN= instructs R to plot only top N highest item frequency or lift

Output

type= is a character string indicating whether item



Interpretation:

The plot shows items by frequency in a descending order.

#Get the Rules

```
rules<-apriori(trans,parameter=list(supp=0.001,conf=0.8))
inspect(rules[1:5])</pre>
```

- **apriori()** is used to mine frequent itemsets, association rules or
- association hyperedges using this algorithm with specified support
 - and confidence
- inspect() in package arules displays association and additional
 - information formatted for online inspection

Output

```
Parameter specification:
 confidence minval smax arem aval original Support maxtime support
        0.8
               0.1
                      1 none FALSE
                                              TRUE
maxlen target ext
    10 rules FALSE
Algorithmic control:
filter tree heap memopt load sort verbose
    0.1 TRUE TRUE FALSE TRUE
                                      TRUE
Absolute minimum support count: 0
set item appearances ...[0 item(s)] done [0.00s].
set transactions ... [5 item(s), 100 transaction(s)] done [0.00s].
sorting and recoding items ... [5 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 done [0.00s].
writing ... [5 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
```

Interpretation:

- The output displays parameter specification, algorithmic control and absolute minimum
 - support count.
- It also lists down tasks performed and time taken to complete them.
- We are interested in knowing how many rules are created; Here 5 rules are

Output

```
rhs support confidence lift count
    1hs
[1] \{A,D\} => \{C\} \ 0.25
                             0.81
                                             25
[2] \{A,D,E\} => \{C\} 0.11
                            0.85
                                             11
[3] \{A,B,E\} => \{C\} 0.10
                         0.83
                                      1.3 10
[4] \{A,B,D\} => \{C\} 0.16
                          0.84
                                             16
                                        1.3
[5] {A.B.D.E} => {C} 0.06
                             1.00
                                        1.6
```

Interpretation:

- inspect() returns list of lhs and rhs items, their support, confidence and lift values
- Questions: 1) Find rules any with confidence=12) Find top 3 rules based on highest

Lift

THANK YOU!!