

Creating Association Rules for Supermarket Dataset

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Loading the csv file

```
# the csv file will be loaded from a local repository and will be loaded as class transactions.  
# loading the required library  
library(arules)
```

```
## Loading required package: Matrix
```

```
##
```

```
## Attaching package: 'arules'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      abbreviate, write
```

```
# loading the dataset  
trans <- read.transactions(file.choose(), sep = ",", rm.duplicates = TRUE)
```

```
## distribution of transactions with duplicates:
```

```
## 1
```

```
## 5
```

```
# inspecting the class  
class(trans)
```

```
## [1] "transactions"
```

```
## attr(,"package")
```

```
## [1] "arules"
```

```
# checking the head of the dataset  
inspect(trans[1:5])
```

```
##      items  
## [1] {almonds,  
##      antioxydant juice,  
##      avocado,  
##      cottage cheese,
```

```
##      energy drink,
##      frozen smoothie,
##      green grapes,
##      green tea,
##      honey,
##      low fat yogurt,
##      mineral water,
##      olive oil,
##      salad,
##      salmon,
##      shrimp,
##      spinach,
##      tomato juice,
##      vegetables mix,
##      whole weat flour,
##      yams}
## [2] {burgers,
##      eggs,
##      meatballs}
## [3] {chutney}
## [4] {avocado,
##      turkey}
## [5] {energy bar,
##      green tea,
##      milk,
##      mineral water,
##      whole wheat rice}
```

```
# generating a summary of the dataset
summary(trans)
```

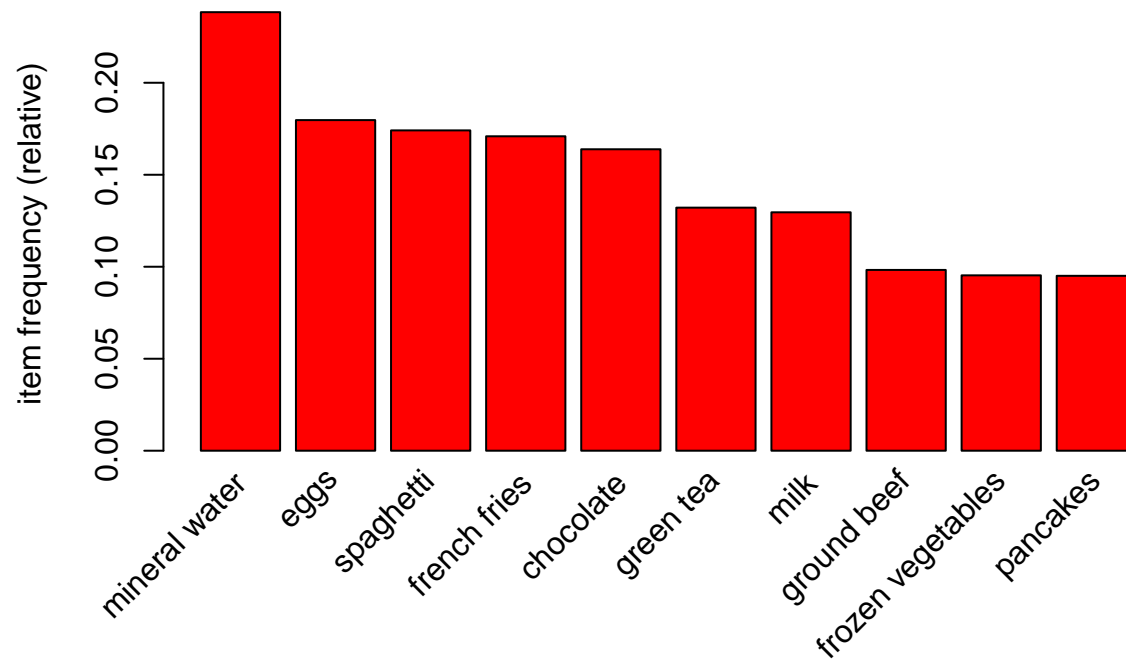
```
## transactions as itemMatrix in sparse format with
## 7501 rows (elements/itemsets/transactions) and
## 119 columns (items) and a density of 0.03288973
##
## most frequent items:
## mineral water      eggs      spaghetti  french fries      chocolate
##           1788           1348           1306           1282           1229
##      (Other)
##           22405
##
## element (itemset/transaction) length distribution:
## sizes
##      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15     16
## 1754 1358 1044  816  667  493  391  324  259  139  102   67   40   22   17    4
##      18     19     20
##      1      2      1
##
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.000  2.000   3.000   3.914   5.000  20.000
##
## includes extended item information - examples:
##           labels
## 1           almonds
```

```
## 2 antioxydant juice
## 3      asparagus
```

```
# displaying the top 10 most common items in the dataset and items whose relative importance is atleast
```

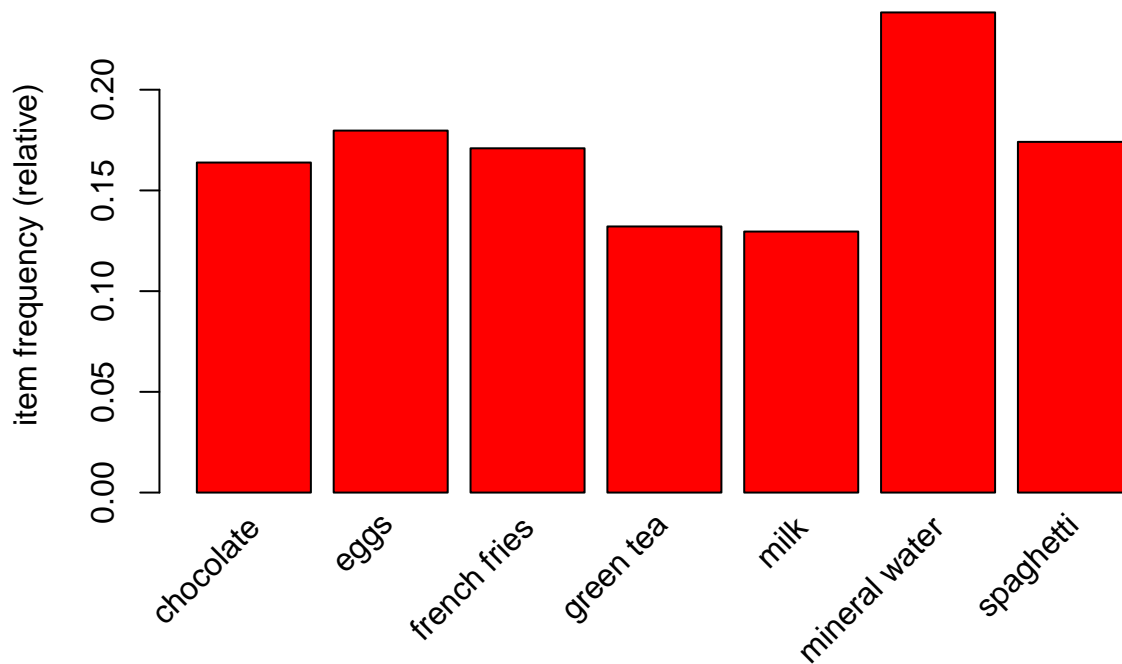
```
# showing the top 10 common items
```

```
itemFrequencyPlot(trans, topN=10, col="red", type="relative")
```



```
# showing the items whose importance is at least 10%
```

```
itemFrequencyPlot(trans, support=0.1, col="red", type="relative")
```



The top most most frequent item in the data set is mineral water followed by eggs and spaghetti while the 10th most frequent item was pancakes

The items whose popularity was at least 10% were 7 with mineral water having the highest popularity followed by eggs.

creating the model

```
# building the model based on apriori rules of association
rules <- apriori(trans, parameter = list(supp = 0.001, conf = 0.8))
```

```
## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##          0.8   0.1   1 none FALSE                TRUE     5   0.001     1
## maxlen target  ext
##          10 rules TRUE
##
## Algorithmic control:
## filter tree heap memopt load sort verbose
##       0.1 TRUE TRUE  FALSE TRUE     2    TRUE
##
## Absolute minimum support count: 7
```

```
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[119 item(s), 7501 transaction(s)] done [0.00s].
## sorting and recoding items ... [116 item(s)] done [0.00s].
## creating transaction tree ... done [0.00s].
## checking subsets of size 1 2 3 4 5 6 done [0.01s].
## writing ... [74 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].
```

```
rules
```

```
## set of 74 rules
```

```
# obtaining the summary of the model
summary(rules)
```

```
## set of 74 rules
##
## rule length distribution (lhs + rhs):sizes
## 3 4 5 6
## 15 42 16 1
##
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 3.000  4.000  4.000  4.041  4.000  6.000
##
## summary of quality measures:
##      support      confidence      coverage      lift
## Min.   :0.001067  Min.   :0.8000  Min.   :0.001067  Min.   : 3.356
## 1st Qu.:0.001067  1st Qu.:0.8000  1st Qu.:0.001333  1st Qu.: 3.432
## Median :0.001133  Median :0.8333  Median :0.001333  Median : 3.795
## Mean   :0.001256  Mean   :0.8504  Mean   :0.001479  Mean   : 4.823
## 3rd Qu.:0.001333  3rd Qu.:0.8889  3rd Qu.:0.001600  3rd Qu.: 4.877
## Max.   :0.002533  Max.   :1.0000  Max.   :0.002666  Max.   :12.722
##      count
## Min.   : 8.000
## 1st Qu.: 8.000
## Median : 8.500
## Mean   : 9.419
## 3rd Qu.:10.000
## Max.   :19.000
##
## mining info:
## data ntransactions support confidence
## trans          7501    0.001         0.8
```

```
# inspecting the first five rules and sorting them with the level of confidence
rules <- sort(rules, by="confidence", decreasing = TRUE)
inspect(rules[1:5])
```

```
##      lhs                                rhs      support
## [1] {french fries,mushroom cream sauce,pasta} => {escalope}      0.001066524
## [2] {ground beef,light cream,olive oil}      => {mineral water} 0.001199840
```

```
## [3] {cake,meatballs,mineral water}      => {milk}      0.001066524
## [4] {cake,olive oil,shrimp}              => {mineral water} 0.001199840
## [5] {mushroom cream sauce,pasta}        => {escalope}    0.002532996
##      confidence coverage      lift      count
## [1] 1.00          0.001066524 12.606723  8
## [2] 1.00          0.001199840  4.195190  9
## [3] 1.00          0.001066524  7.717078  8
## [4] 1.00          0.001199840  4.195190  9
## [5] 0.95          0.002666311 11.976387 19
```

The first 4 rules had 100% confidence. For the first rule, this means that customers that bought french fries, mushroom cream sauce and pasta had a 100% chance to buy escalope.

A milk promotional case

Suppose Maziwa, milk company wanted to promote milk in Carrefour. The company would like to make a decision on product placement and would like to know which other products that customers bought before buying milk

```
# generating the top five rules for milk
milk <- subset(rules, subset=rhs%pin% "milk")

# ordering the rules by confidence in descending order
milk <- sort(milk, by= "confidence", descending=TRUE)
inspect(milk[1:5])
```

```
##      lhs                                rhs      support      confidence
## [1] {cake,meatballs,mineral water}      => {milk} 0.001066524 1.0000000
## [2] {escalope,hot dogs,mineral water}    => {milk} 0.001066524 0.8888889
## [3] {meatballs,whole wheat pasta}        => {milk} 0.001333156 0.8333333
## [4] {black tea,frozen smoothie}          => {milk} 0.001199840 0.8181818
## [5] {burgers,ground beef,olive oil}     => {milk} 0.001066524 0.8000000
##      coverage      lift      count
## [1] 0.001066524 7.717078  8
## [2] 0.001199840 6.859625  8
## [3] 0.001599787 6.430898 10
## [4] 0.001466471 6.313973  9
## [5] 0.001333156 6.173663  8
```

Customers who bought cake meatballs and mineral water were 100% likely to buy milk. Therefore the milk company should place their product immediately after these products to guarantee a 100% likelihood of purchase.