# **Experiment 1.3**

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Subject Name: Data Mining Lab Subject Code: 20CSP-376

#### 1. Aim/Overview of the practical:

Demonstration of associate rule mining using Apriori algorithm on supermarket data

## 2. Objective:

- a. To learn how to import.
- b. To learn how to perform associate rule mining using Apriori algorithm.
- c. To learn the utilization of arules, arulesviz, RcolorBrewer.

#### 3. Code:

library(arules)
library(arulesViz)
library(RColorBrewer)

data("Groceries")

rules <- apriori(Groceries, parameter = list(supp = 0.01, conf = 0.2))

#### 4. Output:

```
Console Terminal × Background Jobs ×
R 4.2.2 · D:/DataMining/
> rules <- apriori(Groceries, parameter = list(supp = 0.01, conf = 0.2))
Apriori
Parameter specification:
 confidence minval smax arem aval original Support maxtime support minlen maxlen target ext
               0.1
                      1 none FALSE
                                               TRUE
                                                          5
                                                                0.01
                                                                         1
                                                                                10 rules TRUE
Algorithmic control:
 filter tree heap memopt load sort verbose
    0.1 TRUE TRUE FALSE TRUE
Absolute minimum support count: 98
set item appearances ...[0 item(s)] done [0.00s].
set transactions ...[169 item(s), 9835 transaction(s)] done [0.00s].
sorting and recoding items ... [88 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 done [0.00s].
writing ... [232 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
> inspect(rules[1:10])
     1hs
                        rhs
                                                       confidence coverage
                                                                              lift
                                            support
                                                                                       count
[1]
                     => {whole milk}
                                            0.25551601 0.2555160 1.00000000 1.000000 2513
     {hard cheese}
                                            0.01006609 0.4107884 0.02450432 1.607682
[2]
                    => {whole milk}
[3]
     {butter milk}
                    => {other vegetables} 0.01037112 0.3709091 0.02796136 1.916916
     {butter milk}
                                            0.01159126 0.4145455
                                                                   0.02796136 1.622385
[4]
                    => {whole milk}
                                                                                        114
[5]
                                            0.01148958 0.4414062
                                                                   0.02602949 1.727509
     {ham}
                     => {whole milk}
                                                                                        113
                                            0.01077783 0.4398340
[6]
     {sliced cheese} => {whole milk}
                                                                   0.02450432 1.721356
                                                                                        106
[7]
     {oil}
                    => {whole milk}
                                            0.01128622 0.4021739
                                                                   0.02806304 1.573968
                                                                                        111
[8]
                    => {other vegetables} 0.01423488 0.4590164
                                                                  0.03101169 2.372268
     {onions}
                                                                                        140
[9]
     {onions}
                    => {whole milk}
                                            0.01209964 0.3901639
                                                                   0.03101169 1.526965
                                                                                        119
[10] {berries}
                    => {yogurt}
                                            0.01057448 0.3180428 0.03324860 2.279848
                                                                                        104
> arules::itemFrequencyPlot(Groceries, topN = 20,
+ col = brewer.pal(8, 'Pastel2'),
                            main = 'Relative Item Frequency Plot',
                             type = "relative"
                            ylab = "Item Frequency (Relative)")
```



### **Graph output:**

