

MBA

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```
#Load the libraries
#library(readr)
library(arules)
```

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

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

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

```
library(arulesViz)
```

```
## Loading required package: grid
```

```
## Registered S3 method overwritten by 'seriation':
##   method             from
##   reorder.hclust      gclus
```

```
library(magrittr)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:arules':
##
##      intersect, recode, setdiff, setequal, union
```

```
## The following objects are masked from 'package:stats':
##
##      filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
## intersect, setdiff, setequal, union
```

```
library(lubridate)
```

```
##  
## Attaching package: 'lubridate'
```

```
## The following object is masked from 'package:base':  
##  
## date
```

```
library(forcats)  
library(ggplot2)  
library(arulesViz)  
library(datasets)  
library(RColorBrewer)
```

```
library(readxl)  
library(tibble)  
#Importing Data set  
setwd("C:/Users/SuprasannaPradhan/Documents/My Files/Great Lakes Projects/Project-8 M  
A")  
cafe=read_excel("Cafe Great Transaction Data set.xlsx")  
cafe_df = as.data.frame(cafe)  
str(cafe_df)
```

```
## 'data.frame': 145830 obs. of 10 variables:  
## $ Date : POSIXct, format: "2011-01-25" "2011-02-20" ...  
## $ Bill Number: chr "G0522713" "N0033392" "N0032132" "N0030048" ...  
## $ Item Desc : chr "PARTY CHARGES @ 500/-" "KF DRAUGHT PITCHER (2LTR)" "SULA BRUT  
(BTL)" "RED SANGRIA (CARAFE) ÃÃÃÃÃÃÃÃÃ" ...  
## $ Time : POSIXct, format: "1899-12-31 21:21:02" "1899-12-31 21:24:40" ...  
## $ Quantity : num 23 10 3 4 9 13 2 11 8 6 ...  
## $ Rate : num 500 400 1200 850 350 225 1300 250 330 400 ...  
## $ Tax : num 2731 1250 936 884 984 ...  
## $ Discount : num 0 0 0 0 0 0 0 825 0 0 ...  
## $ Total : num 14231 5250 4536 4284 4134 ...  
## $ Category : chr "MISC" "LIQUOR" "WINES" "WINES" ...
```

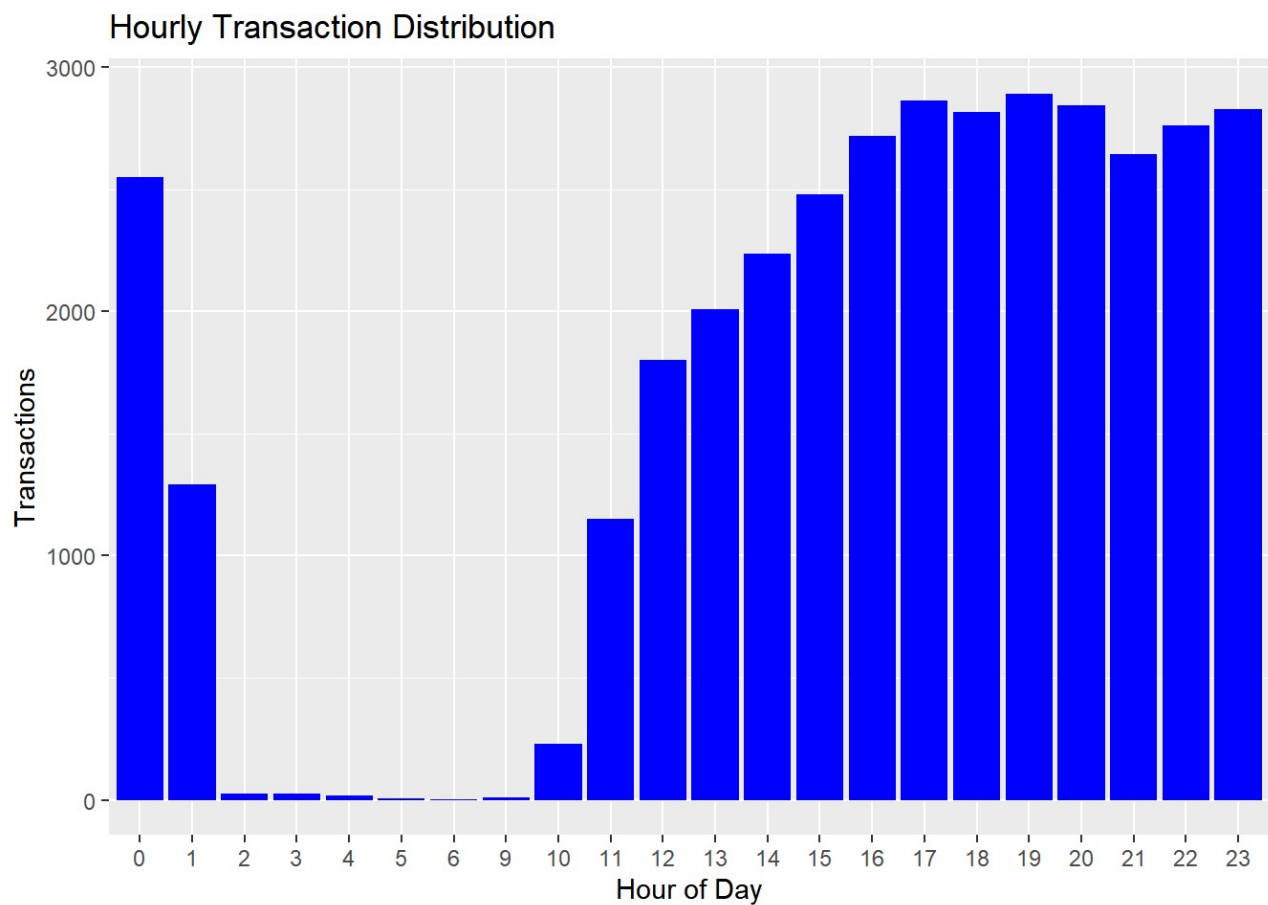
```
colnames(caffe_df)[colnames(caffe_df)=="Item Desc"] <- "Item_Desc"
colnames(caffe_df)[colnames(caffe_df)=="Bill Number"] <- "Bill_Number"
sum(is.na(caffe_df))
```

```
## [1] 0
```

```
str(caffe_df)
```

```
## 'data.frame': 145830 obs. of 10 variables:
## $ Date : POSIXct, format: "2011-01-25" "2011-02-20" ...
## $ Bill_Number: chr "G0522713" "N0033392" "N0032132" "N0030048" ...
## $ Item_Desc : chr "PARTY CHARGES @ 500/-" "KF DRAUGHT PITCHER (2LTR)" "SULA BRUT
(BTL)" "RED SANGRIA (CARAFE) Ã;Ã;Ã;Ã;Ã;Ã;Ã;Ã;" ...
## $ Time : POSIXct, format: "1899-12-31 21:21:02" "1899-12-31 21:24:40" ...
## $ Quantity : num 23 10 3 4 9 13 2 11 8 6 ...
## $ Rate : num 500 400 1200 850 350 225 1300 250 330 400 ...
## $ Tax : num 2731 1250 936 884 984 ...
## $ Discount : num 0 0 0 0 0 0 0 825 0 0 ...
## $ Total : num 14231 5250 4536 4284 4134 ...
## $ Category : chr "MISC" "LIQUOR" "WINES" "WINES" ...
```

```
library(dplyr)
purchase_time <-
  caffe_df %>%
  group_by(Time) %>%
  slice(1) %>%
  mutate(time_of_day = hour(Time)) %>%
  pull(time_of_day) %>%
  as.factor() %>%
  fct_count()
purchase_time %>%
  ggplot() +
  geom_col(aes(x = f, y = n), fill = "blue") +
  xlab("Hour of Day") + ylab("Transactions") +
  ggtitle("Hourly Transaction Distribution")
```



```
library(dplyr)
#How many items are purchased on an average?
items <-
  cafe_df %>%
    group_by(Item_Desc) %>%
    summarize(count = n()) %>%
    pull(count)
mean(items)
```

```
## [1] 251.431
```

```
median(items)
```

```
## [1] 37
```

```
#Most Purchased Items
cafe_df %>%
  group_by(Item_Desc)%>%
  summarize(count = n()) %>%
  arrange(desc(count))
```

```
## # A tibble: 580 x 2
##   Item_Desc          count
##   <chr>             <int>
## 1 NIRVANA HOOKAH SINGLE    8553
## 2 MINT FLAVOUR SINGLE    5817
## 3 CAPPUCCINO              5495
## 4 GREAT LAKES SHAKE      4895
## 5 SAMBUCA                 4425
## 6 POUTINE WITH FRIES      3464
## 7 QUA MINERAL WATER(1000ML) 3331
## 8 CALCUTTA MINT           3318
## 9 JR.CHL AVALANCHE        3314
## 10 B.M.T. PANINI           2724
## # ... with 570 more rows
```

#Average Order Value

```
total_revenue <-
  cafe_df %>%
  group_by(Bill_Number) %>%
  summarize(order_sum = sum(Total)) %>%
  pull(order_sum) %>%
  sum()

total_transactions <-
  cafe_df%>%
  group_by(Bill_Number) %>%
  summarize(n()) %>%
  nrow()

total_revenue / total_transactions
```

```
## [1] 468.7762
```

```
summary(cafe_df)
```

```

##      Date                Bill_Number      Item_Desc
## Min.   :2010-04-01 00:00:00 Length:145830 Length:145830
## 1st Qu.:2010-07-10 00:00:00 Class :character Class :character
## Median :2010-10-07 00:00:00 Mode  :character Mode  :character
## Mean   :2010-10-04 15:52:45
## 3rd Qu.:2011-01-01 00:00:00
## Max.   :2011-03-31 00:00:00
##      Time                Quantity          Rate
## Min.   :1899-12-31 00:00:01 Min.    : 1.000 Min.    :  0.01
## 1st Qu.:1899-12-31 14:48:53 1st Qu.: 1.000 1st Qu.: 95.00
## Median :1899-12-31 18:17:24 Median : 1.000 Median : 125.00
## Mean   :1899-12-31 16:39:16 Mean   : 1.121 Mean   : 161.78
## 3rd Qu.:1899-12-31 21:02:20 3rd Qu.: 1.000 3rd Qu.: 225.00
## Max.   :1899-12-31 23:59:57 Max.    :30.000 Max.    :2100.00
##      Tax                Discount          Total
## Min.   :  0.00 Min.    : 0.0000 Min.    :  0.01
## 1st Qu.: 22.56 1st Qu.: 0.0000 1st Qu.: 117.56
## Median : 32.06 Median : 0.0000 Median : 167.06
## Mean   : 48.93 Mean   : 0.0951 Mean   : 224.96
## 3rd Qu.: 72.00 3rd Qu.: 0.0000 3rd Qu.: 315.00
## Max.   :2731.25 Max.    :825.0000 Max.    :14231.25
##      Category
## Length:145830
## Class :character
## Mode  :character
##
##
##

```

```

#cafe_df$ID = as.factor(cafe_df$ID)
#cafe_df$Date = as.factor(cafe_df$Date)
#cafe_df$Bill_Number = as.factor(cafe_df$Bill_Number)
cafe_df$Item_Desc = as.factor(cafe_df$Item_Desc)
#cafe_df$Time = as.factor(cafe_df$Time)
#cafe_df$Quantity = as.factor(cafe_df$Quantity)
#cafe_df$Rate = as.factor(cafe_df$Rate)
#cafe_df$Tax = as.factor(cafe_df$Tax)
#cafe_df$Discount = as.factor(cafe_df$Discount)
#cafe_df$Total = as.factor(cafe_df$Total)
cafe_df$Category = as.factor(cafe_df$Category)
str(cafe_df)

```

```
## 'data.frame':   145830 obs. of  10 variables:
## $ Date          : POSIXct, format: "2011-01-25" "2011-02-20" ...
## $ Bill_Number: chr  "G0522713" "N0033392" "N0032132" "N0030048" ...
## $ Item_Desc    : Factor w/ 580 levels "1 AXE TWIST",...: 411 322 504 448 494 321 490
492 69 322 ...
## $ Time         : POSIXct, format: "1899-12-31 21:21:02" "1899-12-31 21:24:40" ...
## $ Quantity     : num  23 10 3 4 9 13 2 11 8 6 ...
## $ Rate         : num  500 400 1200 850 350 225 1300 250 330 400 ...
## $ Tax          : num  2731 1250 936 884 984 ...
## $ Discount     : num  0 0 0 0 0 0 0 825 0 0 ...
## $ Total        : num  14231 5250 4536 4284 4134 ...
## $ Category     : Factor w/ 8 levels "BEVERAGE","FOOD",...: 6 3 8 8 3 3 3 3 2 3 ...
```

convert to transaction

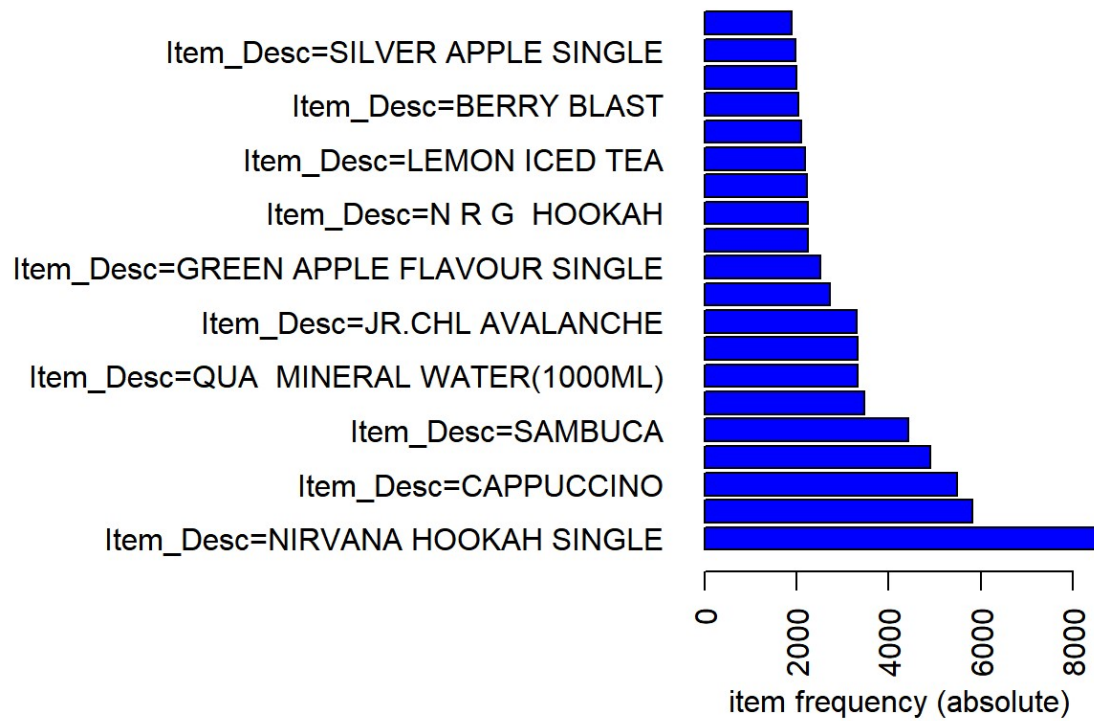
```
cafe_trasn<- subset(cafe_df ,select= c(3))
trans1 <- as(cafe_trasn, "transactions")
summary(trans1)
```

```
## transactions as itemMatrix in sparse format with
## 145830 rows (elements/itemsets/transactions) and
## 580 columns (items) and a density of 0.001724138
##
## most frequent items:
## Item_Desc=NIRVANA HOOKAH SINGLE      Item_Desc=MINT FLAVOUR SINGLE
##                                8553                                5817
##      Item_Desc=CAPPUCCINO      Item_Desc=GREAT LAKES SHAKE
##                                5495                                4895
##      Item_Desc=SAMBUCA                                (Other)
##                                4425                                116645
##
## element (itemset/transaction) length distribution:
## sizes
##      1
## 145830
##
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1         1         1         1         1         1
##
## includes extended item information - examples:
##                                labels variables          levels
## 1      Item_Desc=1 AXE TWIST Item_Desc      1 AXE TWIST
## 2 Item_Desc=1+1 BTL4 SEASON WHITE Item_Desc 1+1 BTL4 SEASON WHITE
## 3  Item_Desc=1+1 GLS 4SEASON RED Item_Desc   1+1 GLS 4SEASON RED
##
## includes extended transaction information - examples:
## transactionID
## 1             1
## 2             2
## 3             3
```

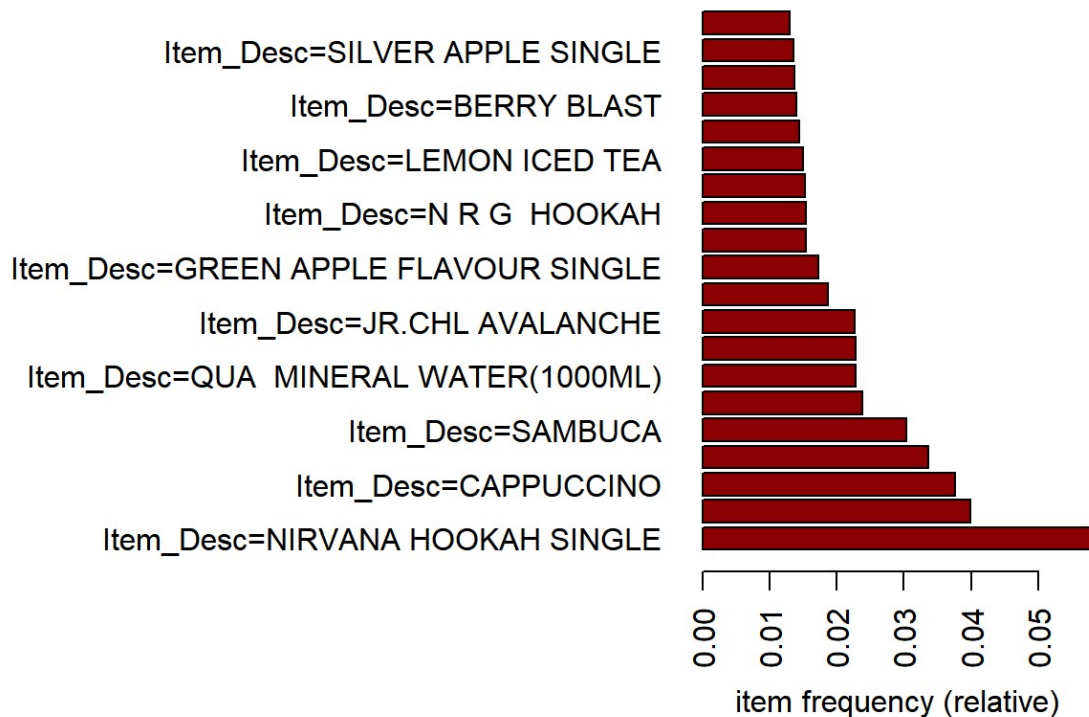
```
inspect(trans1[20])
```

```
##      items                                transactionID
## [1] {Item_Desc=HOEGAARDEN MUG (1 LITRE)} 20
```

```
itemFrequencyPlot(trans1, topN = 20, horiz = TRUE, type = 'absolute', col = "blue")
```

```
itemFrequencyPlot(trans1, topN = 20, type = 'relative',horiz = TRUE, col = "dark red")
```



```
cafe_df$bill_number <- gsub("(\\D+)", "",cafe_df$Bill_Number)
cafe_df$bill_number = as.factor(cafe_df$bill_number)
cafe_sorted <- cafe_df[order(cafe_df$bill_number),]
cafe_mba <- subset(cafe_sorted,select= c(3,11))
str(cafe_mba)
```

```
## 'data.frame':  145830 obs. of  2 variables:
## $ Item_Desc : Factor w/ 580 levels "1 AXE TWIST",...: 132 132 539 539 3 132 539 53
9 132 132 ...
## $ bill_number: Factor w/ 69982 levels "0027835","0027836",...: 1 2 3 4 5 6 7 8 9 10
...
```

```
cafe_mba$bill_number <- as.numeric(cafe_mba$bill_number)
head(cafe_mba)
```

```
##           Item_Desc bill_number
## 80310      CARLSBERG          1
## 80311      CARLSBERG          2
## 48201      TUBORG            3
## 99550      TUBORG            4
## 53538 1+1 GLS 4SEASON RED    5
## 80312      CARLSBERG          6
```

```
cafe_mba$bill_number <- as.numeric(cafe_mba$bill_number)
tail(cafe_mba)
```

```
##           Item_Desc bill_number
## 21159  NIRVANA HOOKAH SINGLE    69982
## 66534  RED BULL ENERGY DRINK   69982
## 78061   KIT KAT SHAKE           69982
## 78062   GREAT LAKES SHAKE       69982
## 78063   SUNSHINE SHAKE          69982
## 139211 MOROCCAN MINT TEA        69982
```

```
library(plyr)
```

```
## -----
```

```
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
```

```
## -----
```

```
##
## Attaching package: 'plyr'
```

```
## The following object is masked from 'package:lubridate':
##
##     here
```

```
## The following objects are masked from 'package:dplyr':
##
##     arrange, count, desc, failwith, id, mutate, rename, summarise,
##     summarize
```

```
df_itemList <- ddply(caffe_mba,'bill_number', function(df1)paste(df1$Item_Desc,collapse
= ","))
head(df_itemList)
```

```
##   bill_number      V1
## 1          1  CARLSBERG
## 2          2  CARLSBERG
## 3          3    TUBORG
## 4          4    TUBORG
## 5          5 1+1 GLS 4SEASON RED
## 6          6  CARLSBERG
```

Now, we remove the column 'bill- number'.

```
df_itemList$bill_number <- NULL
head(df_itemList)
```

```
##           V1
## 1  CARLSBERG
## 2  CARLSBERG
## 3    TUBORG
## 4    TUBORG
## 5 1+1 GLS 4SEASON RED
## 6  CARLSBERG
```

```
colnames(df_itemList) <- c("Item List")
head(df_itemList)
```

```
##           Item List
## 1  CARLSBERG
## 2  CARLSBERG
## 3    TUBORG
## 4    TUBORG
## 5 1+1 GLS 4SEASON RED
## 6  CARLSBERG
```

```
write.csv(df_itemList,"ItemList.csv", quote = FALSE, row.names = TRUE)
```

```
library(arules)
trans2 = read.transactions(file="ItemList.csv", rm.duplicates= FALSE, format="basket",
sep="," ,cols= 1);
```

```
## Warning in asMethod(object): removing duplicated items in transactions
```

```
inspect(head(trans2))
```

```
##      items      transactionID
## [1] {Item List}
## [2] {CARLSBERG}      1
## [3] {CARLSBERG}      2
## [4] {TUBORG}        3
## [5] {TUBORG}        4
## [6] {1+1 GLS 4SEASON RED} 5
```

```
#trans$items <- gsub("\\d+", "", df),trans$items)
#trans@itemInfo$Labels <- gsub("", "",trans@itemInfo$Labels)
```

```
print(head(itemInfo(trans2)))
```

```
##      labels
## 1      1 AXE TWIST
## 2 1+1 BTL4 SEASON WHITE
## 3 1+1 GLS 4SEASON RED
## 4 1+1 GLS 4SEASONS WHITE
## 5      1+1 KF 1 LITER
## 6      1+1 KF 1/2 LITER
```

```
print(levels(itemInfo(trans2)[["level1"]]))
```

```
## NULL
```

```
print(levels(itemInfo(trans2)[["level6"]]))
```

```
## NULL
```

```
print(dim(trans2)[1])
```

```
## [1] 69983
```

```
print(dim(trans2)[2])
```

```
## [1] 581
```

Generate Rules

```
first.rules <- apriori(trans2,  
                      parameter = list(support = 0.001, confidence = 0.01))
```

```
## Apriori  
##  
## Parameter specification:  
## confidence minval  smax  arem  aval originalSupport  maxtime support minlen  
##      0.01    0.1    1 none FALSE               TRUE      5   0.001    1  
## maxlen target   ext  
##     10  rules FALSE  
##  
## Algorithmic control:  
## filter tree heap memopt load sort verbose  
##    0.1 TRUE TRUE  FALSE TRUE    2    TRUE  
##  
## Absolute minimum support count: 69  
##  
## set item appearances ...[0 item(s)] done [0.00s].  
## set transactions ...[581 item(s), 69983 transaction(s)] done [0.04s].  
## sorting and recoding items ... [210 item(s)] done [0.00s].  
## creating transaction tree ... done [0.05s].  
## checking subsets of size 1 2 3 done [0.01s].  
## writing ... [677 rule(s)] done [0.00s].  
## creating S4 object ... done [0.01s].
```

Rules Summary

```
summary(first.rules)
```

```

## set of 677 rules
##
## rule length distribution (lhs + rhs):sizes
##   1   2
## 50 627
##
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  1.000   2.000   2.000   1.926   2.000   2.000
##
## summary of quality measures:
##      support      confidence      lift      count
## Min.   :0.001000   Min.   :0.01007   Min.   : 0.1464   Min.   : 70.0
## 1st Qu.:0.001200   1st Qu.:0.02939   1st Qu.: 0.9843   1st Qu.: 84.0
## Median :0.001529   Median :0.04755   Median : 1.2857   Median : 107.0
## Mean   :0.003817   Mean   :0.05607   Mean   : 1.8382   Mean   : 267.1
## 3rd Qu.:0.002343   3rd Qu.:0.07025   3rd Qu.: 1.7865   3rd Qu.: 164.0
## Max.   :0.119872   Max.   :0.54589   Max.   :79.3655   Max.   :8389.0
##
## mining info:
##   data ntransactions support confidence
## trans2          69983   0.001         0.01

```

```

second.rules <- apriori(trans2,
                        parameter = list(support = 0.025, confidence = 0.04))

```

```

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##      0.04    0.1    1 none FALSE          TRUE        5   0.025    1
## maxlen target  ext
##      10 rules FALSE
##
## Algorithmic control:
## filter tree heap memopt load sort verbose
##    0.1 TRUE TRUE  FALSE TRUE    2    TRUE
##
## Absolute minimum support count: 1749
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[581 item(s), 69983 transaction(s)] done [0.07s].
## sorting and recoding items ... [22 item(s)] done [0.00s].
## creating transaction tree ... done [0.04s].
## checking subsets of size 1 2 done [0.00s].
## writing ... [9 rule(s)] done [0.00s].
## creating S4 object ... done [0.01s].

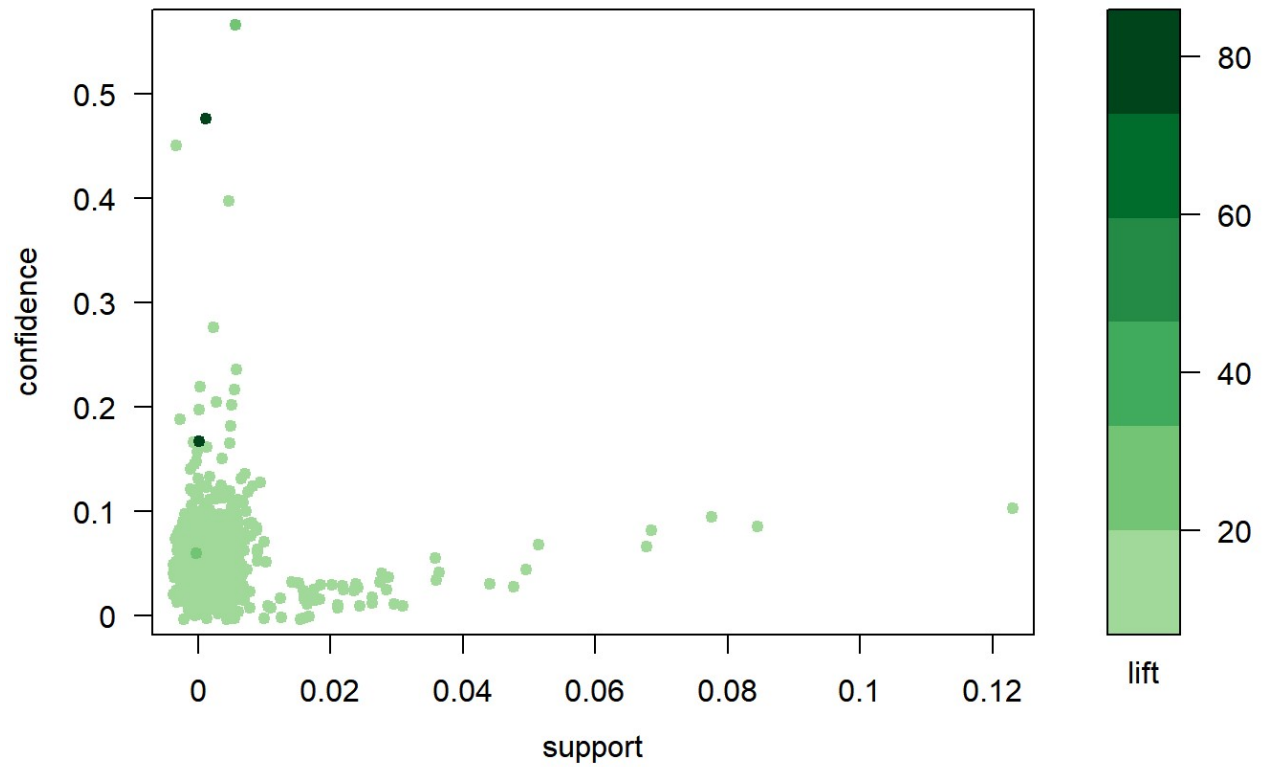
```

```
print(summary(second.rules))
```

```
## set of 9 rules
##
## rule length distribution (lhs + rhs):sizes
## 1
## 9
##
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1      1      1      1      1      1
##
## summary of quality measures:
##      support      confidence      lift      count
## Min.   :0.04725  Min.   :0.04725  Min.   :1  Min.   :3307
## 1st Qu.:0.04755  1st Qu.:0.04755  1st Qu.:1  1st Qu.:3328
## Median :0.06320  Median :0.06320  Median :1  Median :4423
## Mean   :0.06719  Mean   :0.06719  Mean   :1  Mean   :4702
## 3rd Qu.:0.07813  3rd Qu.:0.07813  3rd Qu.:1  3rd Qu.:5468
## Max.   :0.11987  Max.   :0.11987  Max.   :1  Max.   :8389
##
## mining info:
##      data ntransactions support confidence
## trans2      69983    0.025      0.04
```

```
plot(first.rules,
      control=list(jitter=2, col = rev(brewer.pal(9, "Greens")[4:9])),
      shading = "lift")
```


Scatter plot for 677 rules



```
plot(first.rules, method="grouped",  
      control=list(col = rev(brewer.pal(9, "Greens")[4:9])))
```

Items in LHS Group

1 rules: {ADD HERB ROAST CHICKEN, 1 AXE TWIST}
 1 rules: {LEMON INFUSED CHAR GRILLED VEG, 1 AXE TWIST}
 2 rules: {ADD HAZELNUT FLAVOUR, 1 AXE TWIST}
 1 rules: {VANILLA ICECREAM, 1 AXE TWIST}
 6 rules: {ADD FRIES, 1 AXE TWIST}
 3 rules: {COOL CALIFORNICA, 1 AXE TWIST}
 8 rules: {CAFFE LATTE, 1 AXE TWIST}
 7 rules: {MAGGI NDL ARRABIATA, FRENCH FRIES}
 9 rules: {RED BULL 2+1, 1 AXE TWIST}
 18 rules: {COUNTRY ROAST CHICKEN PANINI, SATAY CHICKEN PANINI}
 31 rules: {B.M.T. PANINI, 1 AXE TWIST}
 37 rules: {BERRY BLAST, PINK LEMONADE, +1 items}
 91 rules: {POUTINE WITH FRIES, OREO COOKIE SHAKE, +2 items}
 85 rules: {QUA MINERAL WATER(1000ML), PHILLYCREAM CHEESE & CHILLY F
 20 rules: {LEMON ICED TEA, 1 AXE TWIST}
 12 rules: {MASALA CHAI CUTTING, 1 AXE TWIST}
 64 rules: {SAMBUCA, CALCUTTA MINT, +2 items}
 210 rules: {NIRVANA HOOKAH SINGLE, JR.CHL AVALANCHE, +23 items}
 37 rules: {GREAT LAKES SHAKE, 1 AXE TWIST}
 34 rules: {CAPPUCCINO, 1 AXE TWIST}
 1 rules: {51 suppressed (1LTR)}
 1 rules: {ADD HERB ROAST CHICKEN, 1 AXE TWIST}
 1 rules: {LEMON INFUSED CHAR GRILLED VEG, 1 AXE TWIST}
 2 rules: {ADD HAZELNUT FLAVOUR, 1 AXE TWIST}
 1 rules: {VANILLA ICECREAM, 1 AXE TWIST}
 6 rules: {ADD FRIES, 1 AXE TWIST}
 3 rules: {COOL CALIFORNICA, 1 AXE TWIST}
 8 rules: {CAFFE LATTE, 1 AXE TWIST}
 7 rules: {MAGGI NDL ARRABIATA, FRENCH FRIES}
 9 rules: {RED BULL 2+1, 1 AXE TWIST}
 18 rules: {COUNTRY ROAST CHICKEN PANINI, SATAY CHICKEN PANINI}
 31 rules: {B.M.T. PANINI, 1 AXE TWIST}
 37 rules: {BERRY BLAST, PINK LEMONADE, +1 items}
 91 rules: {POUTINE WITH FRIES, OREO COOKIE SHAKE, +2 items}
 85 rules: {QUA MINERAL WATER(1000ML), PHILLYCREAM CHEESE & CHILLY F
 20 rules: {LEMON ICED TEA, 1 AXE TWIST}
 12 rules: {MASALA CHAI CUTTING, 1 AXE TWIST}
 64 rules: {SAMBUCA, CALCUTTA MINT, +2 items}
 210 rules: {NIRVANA HOOKAH SINGLE, JR.CHL AVALANCHE, +23 items}
 37 rules: {GREAT LAKES SHAKE, 1 AXE TWIST}
 34 rules: {CAPPUCCINO, 1 AXE TWIST}

Size: support
Color: lift

```
bev.rules <- subset(first.rules, subset = rhs %pin% "CAFFE LATTE")
inspect(bev.rules)
```

##	lhs	rhs	support	confidence
## [1]	{}	=> {CAFFE LATTE}	0.029935842	0.02993584
## [2]	{ADD HAZELNUT FLAVOUR}	=> {CAFFE LATTE}	0.001614678	0.54589372
## [3]	{ULTIMATE HOT CHOCOLATE}	=> {CAFFE LATTE}	0.001014532	0.06641721
## [4]	{MASALA CHAI CUTTING}	=> {CAFFE LATTE}	0.001400340	0.04386750
## [5]	{B.M.T. PANINI}	=> {CAFFE LATTE}	0.001057400	0.02840691
## [6]	{QUA MINERAL WATER(1000ML)}	=> {CAFFE LATTE}	0.002286270	0.04807692
## [7]	{CAPPUCCINO}	=> {CAFFE LATTE}	0.004243888	0.05431602
## [8]	{GREAT LAKES SHAKE}	=> {CAFFE LATTE}	0.002529186	0.03633751
##	lift	count		
## [1]	1.0000000	2095		
## [2]	18.2354559	113		
## [3]	2.2186519	71		
## [4]	1.4653840	98		
## [5]	0.9489264	74		
## [6]	1.6059987	160		
## [7]	1.8144143	297		
## [8]	1.2138462	177		

```
fv.rules <- subset(first.rules, subset = rhs %pin% "ADD FRIES")
inspect(fv.rules)
```

```
##      lhs                                rhs      support    confidence
## [1] {}                                => {ADD FRIES} 0.010745467 0.01074547
## [2] {SATAY CHICKEN PANINI}           => {ADD FRIES} 0.001128846 0.08036623
## [3] {COUNTRY ROAST CHICKEN PANINI}   => {ADD FRIES} 0.001243159 0.06580938
## [4] {COTTAGE CHEESE PANINI}          => {ADD FRIES} 0.001043110 0.04528536
## [5] {PHILLYCREAM CHEESE &CHILLY PAN} => {ADD FRIES} 0.001171713 0.04408602
## [6] {B.M.T. PANINI}                  => {ADD FRIES} 0.002157667 0.05796545
## [7] {QUA MINERAL WATER(1000ML)}     => {ADD FRIES} 0.001286027 0.02704327
##      lift      count
## [1] 1.000000 752
## [2] 7.479082 79
## [3] 6.124385 87
## [4] 4.214369 73
## [5] 4.102755 82
## [6] 5.394410 151
## [7] 2.516714 90
```

```
top.bev.rules <- head(sort(bev.rules, decreasing = TRUE, by = "lift"), 10)
inspect(top.bev.rules)
```

```
##      lhs                                rhs      support    confidence
## [1] {ADD HAZELNUT FLAVOUR}           => {CAFFE LATTE} 0.001614678 0.54589372
## [2] {ULTIMATE HOT CHOCOLATE}         => {CAFFE LATTE} 0.001014532 0.06641721
## [3] {CAPPUCCINO}                     => {CAFFE LATTE} 0.004243888 0.05431602
## [4] {QUA MINERAL WATER(1000ML)}     => {CAFFE LATTE} 0.002286270 0.04807692
## [5] {MASALA CHAI CUTTING}           => {CAFFE LATTE} 0.001400340 0.04386750
## [6] {GREAT LAKES SHAKE}              => {CAFFE LATTE} 0.002529186 0.03633751
## [7] {}                                => {CAFFE LATTE} 0.029935842 0.02993584
## [8] {B.M.T. PANINI}                  => {CAFFE LATTE} 0.001057400 0.02840691
##      lift      count
## [1] 18.2354559 113
## [2] 2.2186519 71
## [3] 1.8144143 297
## [4] 1.6059987 160
## [5] 1.4653840 98
## [6] 1.2138462 177
## [7] 1.0000000 2095
## [8] 0.9489264 74
```

```
plot(top.bev.rules, method="graph",
      control=list(type="items"),
      shading = "lift")
```

```
## Warning: Unknown control parameters: type
```

```

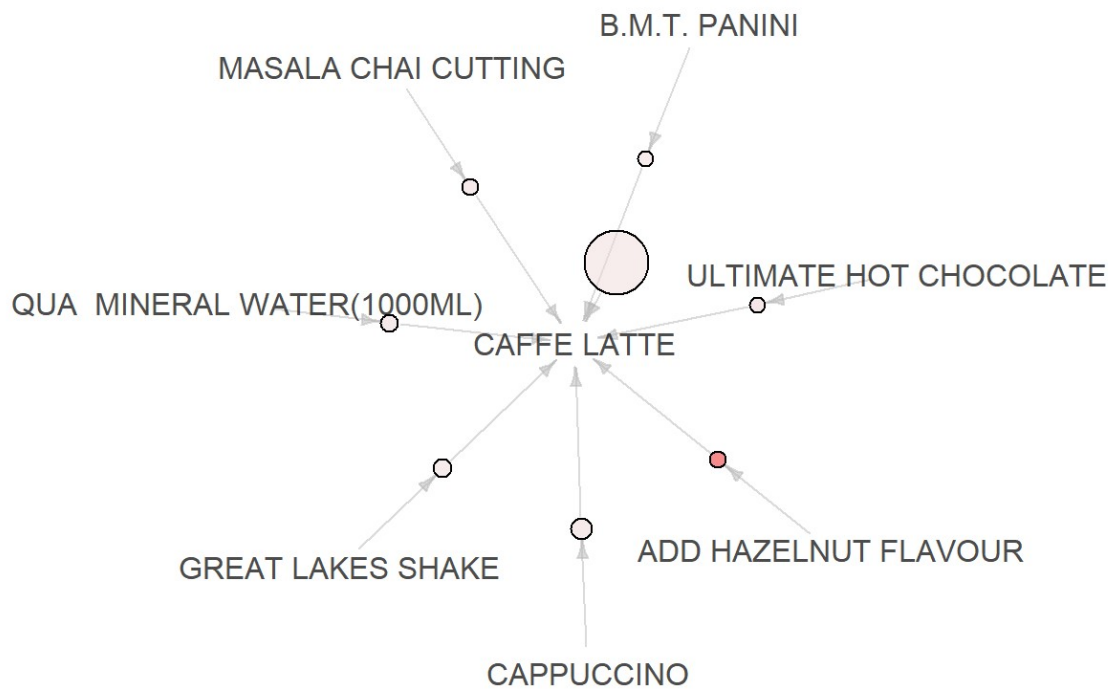
## Available control parameters (with default values):
## main = Graph for 8 rules
## nodeColors = c("#66CC6680", "#9999CC80")
## nodeCol = c("#EE0000FF", "#EE0303FF", "#EE0606FF", "#EE0909FF", "#EE0C0CFF", "#E
E0F0FFF", "#EE1212FF", "#EE1515FF", "#EE1818FF", "#EE1B1BFF", "#EE1E1EFF", "#EE2222F
F", "#EE2525FF", "#EE2828FF", "#EE2B2BFF", "#EE2E2EFF", "#EE3131FF", "#EE3434FF", "#EE
3737FF", "#EE3A3AFF", "#EE3D3DFF", "#EE4040FF", "#EE4444FF", "#EE4747FF", "#EE4A4AFF",
"#EE4D4DFF", "#EE5050FF", "#EE5353FF", "#EE5656FF", "#EE5959FF", "#EE5C5CFF", "#EE5F5F
FF", "#EE6262FF", "#EE6666FF", "#EE6969FF", "#EE6C6CFF", "#EE6F6FFF", "#EE7272FF", "#E
E7575FF", "#EE7878FF", "#EE7B7BFF", "#EE7E7EFF", "#EE8181FF", "#EE8484FF", "#EE8888F
F", "#EE8B8BFF", "#EE8E8EFF", "#EE9191FF", "#EE9494FF", "#EE9797FF", "#EE9999FF", "#EE
9B9BFF", "#EE9D9DFF", "#EE9F9FFF", "#EEA0A0FF", "#EEA2A2FF", "#EEA4A4FF", "#EEA5A5FF",
"#EEA7A7FF", "#EEA9A9FF", "#EEABABFF", "#EEACACFF", "#EEAEA EFF", "#EEB0B0FF", "#EEB1B1
FF", "#EEB3B3FF", "#EEB5B5FF", "#EEB7B7FF", "#EEB8B8FF", "#EEBABAFF", "#EEBCBCFF", "#E
EBDBDFF", "#EEBFBFFF", "#EEC1C1FF", "#EEC3C3FF", "#EEC4C4FF", "#EEC6C6FF", "#EEC8C8F
F", "#EEC9C9FF", "#EECBCBFF", "#EECD C DFF", "#EECF C FFF", "#EED0D0FF", "#EED2D2FF", "#E
ED4D4FF", "#EED5D5FF", "#EED7D7FF", "#EED9D9FF", "#EEDBDBFF", "#EEDC D CFF", "#EED E D E F
F", "#EEE0E0FF", "#EEE1E1FF", "#EEE3E3FF", "#EEE5E5FF", "#EEE7E7FF", "#EEE8E8FF", "#EE
EAEAFF", "#EECECEFF", "#EEEEEEFF")
## edgeCol = c("#474747FF", "#494949FF", "#4B4B4BFF", "#4D4D4DFF", "#4F4F4FFF", "#5
15151FF", "#535353FF", "#555555FF", "#575757FF", "#595959FF", "#5B5B5BFF", "#5E5E5E F
F", "#606060FF", "#626262FF", "#646464FF", "#666666FF", "#686868FF", "#6A6A6AFF", "#6C
6C6CFF", "#6E6E6EFF", "#707070FF", "#727272FF", "#747474FF", "#767676FF", "#787878FF",
"#7A7A7AFF", "#7C7C7CFF", "#7E7E7EFF", "#808080FF", "#828282FF", "#848484FF", "#868686
FF", "#888888FF", "#8A8A8AFF", "#8C8C8CFF", "#8D8D8DFF", "#8F8F8FFF", "#919191FF", "#9
39393FF", "#959595FF", "#979797FF", "#999999FF", "#9A9A9AFF", "#9C9C9CFF", "#9E9E9E F
F", "#A0A0A0FF", "#A2A2A2FF", "#A3A3A3FF", "#A5A5A5FF", "#A7A7A7FF", "#A9A9A9FF", "#AA
AAAAFF", "#ACACACFF", "#AEAEAEFF", "#AFAFAFFF", "#B1B1B1FF", "#B3B3B3FF", "#B4B4B4FF",
"#B6B6B6FF", "#B7B7B7FF", "#B9B9B9FF", "#BBBBBBFF", "#BCBCBCFF", "#BEBEBEFF", "#BFBFBF
FF", "#C1C1C1FF", "#C2C2C2FF", "#C3C3C4FF", "#C5C5C5FF", "#C6C6C6FF", "#C8C8C8FF", "#C
9C9C9FF", "#CACACAFF", "#CCCCC CFF", "#CDCDC DFF", "#CECECEFF", "#CFCFCFFF", "#D1D1D1F
F", "#D2D2D2FF", "#D3D3D3FF", "#D4D4D4FF", "#D5D5D5FF", "#D6D6D6FF", "#D7D7D7FF", "#D
8D8D8FF", "#D9D9D9FF", "#DADADAFF", "#DBDBDBFF", "#DCDC D CFF", "#DDDD D DFF", "#DEDEDE F
F", "#DEDEDEFF", "#DFDF D FFF", "#E0E0E0FF", "#E0E0E0FF", "#E1E1E1FF", "#E1E1E1FF", "#E2
E2E2FF", "#E2E2E2FF", "#E2E2E2FF")
## alpha = 0.5
## cex = 1
## itemLabels = TRUE
## labelCol = #000000B3
## measureLabels = FALSE
## precision = 3
## layout = NULL
## layoutParams = list()
## arrowSize = 0.5
## engine = igraph
## plot = TRUE
## plot_options = list()

```

```
## max    = 100  
## verbose = FALSE
```

Graph for 8 rules

size: support (0.001 - 0.03)
color: lift (0.949 - 18.235)



```
top.fv.rules <- head(sort(fv.rules, decreasing = TRUE, by = "lift"), 10)  
inspect(top.fv.rules)
```

```
##      lhs                                rhs      support    confidence
## [1] {SATAY CHICKEN PANINI}              => {ADD FRIES} 0.001128846 0.08036623
## [2] {COUNTRY ROAST CHICKEN PANINI}      => {ADD FRIES} 0.001243159 0.06580938
## [3] {B.M.T. PANINI}                     => {ADD FRIES} 0.002157667 0.05796545
## [4] {COTTAGE CHEESE PANINI}             => {ADD FRIES} 0.001043110 0.04528536
## [5] {PHILLYCREAM CHEESE &CHILLY PAN}   => {ADD FRIES} 0.001171713 0.04408602
## [6] {QUA MINERAL WATER(1000ML)}        => {ADD FRIES} 0.001286027 0.02704327
## [7] {}                                  => {ADD FRIES} 0.010745467 0.01074547
##      lift      count
## [1] 7.479082    79
## [2] 6.124385    87
## [3] 5.394410   151
## [4] 4.214369    73
## [5] 4.102755    82
## [6] 2.516714    90
## [7] 1.000000   752
```

```
plot(top.fv.rules, method="graph",
      control=list(type="items"),
      shading = "lift")
```

```
## Warning: Unknown control parameters: type
```

```

## Available control parameters (with default values):
## main = Graph for 7 rules
## nodeColors = c("#66CC6680", "#9999CC80")
## nodeCol = c("#EE0000FF", "#EE0303FF", "#EE0606FF", "#EE0909FF", "#EE0C0CFF", "#E
E0F0FFF", "#EE1212FF", "#EE1515FF", "#EE1818FF", "#EE1B1BFF", "#EE1E1EFF", "#EE2222F
F", "#EE2525FF", "#EE2828FF", "#EE2B2BFF", "#EE2E2EFF", "#EE3131FF", "#EE3434FF", "#EE
3737FF", "#EE3A3AFF", "#EE3D3DFF", "#EE4040FF", "#EE4444FF", "#EE4747FF", "#EE4A4AFF",
"#EE4D4DFF", "#EE5050FF", "#EE5353FF", "#EE5656FF", "#EE5959FF", "#EE5C5CFF", "#EE5F5F
FF", "#EE6262FF", "#EE6666FF", "#EE6969FF", "#EE6C6CFF", "#EE6F6FFF", "#EE7272FF", "#E
E7575FF", "#EE7878FF", "#EE7B7BFF", "#EE7E7EFF", "#EE8181FF", "#EE8484FF", "#EE8888F
F", "#EE8B8BFF", "#EE8E8EFF", "#EE9191FF", "#EE9494FF", "#EE9797FF", "#EE9999FF", "#EE
9B9BFF", "#EE9D9DFF", "#EE9F9FFF", "#EEA0A0FF", "#EEA2A2FF", "#EEA4A4FF", "#EEA5A5FF",
"#EEA7A7FF", "#EEA9A9FF", "#EEABABFF", "#EEACACFF", "#EEAEA EFF", "#EEB0B0FF", "#EEB1B1
FF", "#EEB3B3FF", "#EEB5B5FF", "#EEB7B7FF", "#EEB8B8FF", "#EEBABAFF", "#EEBCBCFF", "#E
EBDBDFF", "#EEBFBBBB", "#EEC1C1FF", "#EEC3C3FF", "#EEC4C4FF", "#EEC6C6FF", "#EEC8C8F
F", "#EEC9C9FF", "#EECBCBFF", "#EECD CDF", "#EECF CFFF", "#EED0D0FF", "#EED2D2FF", "#E
ED4D4FF", "#EED5D5FF", "#EED7D7FF", "#EED9D9FF", "#EEDBDBFF", "#EEDC CDF", "#EED EDEF
F", "#EEE0E0FF", "#EEE1E1FF", "#EEE3E3FF", "#EEE5E5FF", "#EEE7E7FF", "#EEE8E8FF", "#EE
EAEAFF", "#EECECEFF", "#EEEEEEFF")
## edgeCol = c("#474747FF", "#494949FF", "#4B4B4BFF", "#4D4D4DFF", "#4F4F4FFF", "#5
15151FF", "#535353FF", "#555555FF", "#575757FF", "#595959FF", "#5B5B5BFF", "#5E5E5EF
F", "#606060FF", "#626262FF", "#646464FF", "#666666FF", "#686868FF", "#6A6A6AFF", "#6C
6C6CFF", "#6E6E6EFF", "#707070FF", "#727272FF", "#747474FF", "#767676FF", "#787878FF",
"#7A7A7AFF", "#7C7C7CFF", "#7E7E7EFF", "#808080FF", "#828282FF", "#848484FF", "#868686
FF", "#888888FF", "#8A8A8AFF", "#8C8C8CFF", "#8D8D8DFF", "#8F8F8FFF", "#919191FF", "#9
39393FF", "#959595FF", "#979797FF", "#999999FF", "#9A9A9AFF", "#9C9C9CFF", "#9E9E9EF
F", "#A0A0A0FF", "#A2A2A2FF", "#A3A3A3FF", "#A5A5A5FF", "#A7A7A7FF", "#A9A9A9FF", "#AA
AAAAFF", "#ACACACFF", "#AEAEAEFF", "#AFAFAFFF", "#B1B1B1FF", "#B3B3B3FF", "#B4B4B4FF",
"#B6B6B6FF", "#B7B7B7FF", "#B9B9B9FF", "#BBBBBBFF", "#BCBCBCFF", "#BEBEBEFF", "#BFBFBF
FF", "#C1C1C1FF", "#C2C2C2FF", "#C3C3C4FF", "#C5C5C5FF", "#C6C6C6FF", "#C8C8C8FF", "#C
9C9C9FF", "#CACACAFF", "#CCCCCFF", "#CDCDCDF", "#CECECEFF", "#CFCFCFFF", "#D1D1D1F
F", "#D2D2D2FF", "#D3D3D3FF", "#D4D4D4FF", "#D5D5D5FF", "#D6D6D6FF", "#D7D7D7FF", "#D
8D8D8FF", "#D9D9D9FF", "#DADADAFF", "#DBDBDBFF", "#DCDC CDF", "#DDDDDDFF", "#DEDEDEF
F", "#DEDEDEFF", "#DFDFDFFF", "#E0E0E0FF", "#E0E0E0FF", "#E1E1E1FF", "#E1E1E1FF", "#E2
E2E2FF", "#E2E2E2FF", "#E2E2E2FF")
## alpha = 0.5
## cex = 1
## itemLabels = TRUE
## labelCol = #000000B3
## measureLabels = FALSE
## precision = 3
## layout = NULL
## layoutParams = list()
## arrowSize = 0.5
## engine = igraph
## plot = TRUE
## plot_options = list()

```



```
## max    = 100
## verbose = FALSE
```

Graph for 7 rules

size: support (0.001 - 0.011)
color: lift (1 - 7.479)

