

Association Rules Project

Stacy

2022-04-03

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.5    v purrr   0.3.4
## v tibble  3.1.6    v dplyr   1.0.8
## v tidyr   1.2.0    v stringr 1.4.0
## v readr   2.1.2    v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(cluster)
library(factoextra)
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(ggplot2)
library(dendextend)
```

```
##
## -----
## Welcome to dendextend version 1.15.2
## Type citation('dendextend') for how to cite the package.
##
## Type browseVignettes(package = 'dendextend') for the package vignette.
## The github page is: https://github.com/talgalili/dendextend/
##
## Suggestions and bug-reports can be submitted at: https://github.com/talgalili/dendextend/issues
## You may ask questions at stackoverflow, use the r and dendextend tags:
##   https://stackoverflow.com/questions/tagged/dendextend
##
## To suppress this message use: suppressPackageStartupMessages(library(dendextend))
## -----
##
## Attaching package: 'dendextend'
```

```
## The following object is masked from 'package:stats':  
##  
##      cutree
```

```
library(tidyverse)  
library(magrittr)
```

```
##  
## Attaching package: 'magrittr'
```

```
## The following object is masked from 'package:purrr':  
##  
##      set_names
```

```
## The following object is masked from 'package:tidyr':  
##  
##      extract
```

```
library(plotly)
```

```
##  
## Attaching package: 'plotly'
```

```
## The following object is masked from 'package:ggplot2':  
##  
##      last_plot
```

```
## The following object is masked from 'package:stats':  
##  
##      filter
```

```
## The following object is masked from 'package:graphics':  
##  
##      layout
```

```
library(psych)
```

```
##  
## Attaching package: 'psych'
```

```
## The following objects are masked from 'package:ggplot2':  
##  
##      %+%, alpha
```

```
library(numDeriv)  
library(arules)
```

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

```
##
## Attaching package: 'Matrix'

## The following objects are masked from 'package:tidyr':
##
##   expand, pack, unpack

##
## Attaching package: 'arules'

## The following object is masked from 'package:dplyr':
##
##   recode

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

```
library(caret)
```

```
## Loading required package: lattice

##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
##   lift
```

```
library(moments)
library(gridExtra)
```

```
##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':
##
##   combine
```

```
library(arulesViz)
library(relaimpo)
```

```
## Loading required package: MASS

##
## Attaching package: 'MASS'

## The following object is masked from 'package:plotly':
##
##   select
```

```

## The following object is masked from 'package:dplyr':
##
##      select

## Loading required package: boot

##
## Attaching package: 'boot'

## The following object is masked from 'package:lattice':
##
##      melanoma

## The following object is masked from 'package:psych':
##
##      logit

## Loading required package: survey

## Loading required package: grid

## Loading required package: survival

##
## Attaching package: 'survival'

## The following object is masked from 'package:boot':
##
##      aml

## The following object is masked from 'package:caret':
##
##      cluster

##
## Attaching package: 'survey'

## The following object is masked from 'package:graphics':
##
##      dotchart

## Loading required package: mitools

## This is the global version of package relaimpo.

## If you are a non-US user, a version with the interesting additional metric pmvd is available

## from Ulrike Groempings web site at prof.beuth-hochschule.de/groemping.

```

```
path <-"http://bit.ly/SupermarketDatasetII"
```

```
supermarket<-read.transactions(path, sep = ",")
```

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

```
head(supermarket)
```

```
## transactions in sparse format with  
## 6 transactions (rows) and  
## 119 items (columns)
```

```
colnames(supermarket)
```

```
##      [1] "almonds"           "antioxydant juice"  "asparagus"  
##      [4] "avocado"           "babies food"        "bacon"  
##      [7] "barbecue sauce"    "black tea"          "blueberries"  
##     [10] "body spray"        "bramble"            "brownies"  
##     [13] "bug spray"         "burger sauce"       "burgers"  
##     [16] "butter"            "cake"               "candy bars"  
##     [19] "carrots"           "cauliflower"        "cereals"  
##     [22] "champagne"         "chicken"            "chili"  
##     [25] "chocolate"         "chocolate bread"    "chutney"  
##     [28] "cider"             "clothes accessories" "cookies"  
##     [31] "cooking oil"       "corn"               "cottage cheese"  
##     [34] "cream"             "dessert wine"       "eggplant"  
##     [37] "eggs"              "energy bar"         "energy drink"  
##     [40] "escalope"          "extra dark chocolate" "flax seed"  
##     [43] "french fries"      "french wine"        "fresh bread"  
##     [46] "fresh tuna"        "fromage blanc"      "frozen smoothie"  
##     [49] "frozen vegetables" "gluten free bar"    "grated cheese"  
##     [52] "green beans"       "green grapes"       "green tea"  
##     [55] "ground beef"       "gums"               "ham"  
##     [58] "hand protein bar"  "herb & pepper"      "honey"  
##     [61] "hot dogs"          "ketchup"            "light cream"  
##     [64] "light mayo"        "low fat yogurt"     "magazines"  
##     [67] "mashed potato"     "mayonnaise"         "meatballs"  
##     [70] "melons"            "milk"               "mineral water"  
##     [73] "mint"              "mint green tea"     "muffins"  
##     [76] "mushroom cream sauce" "napkins"            "nonfat milk"  
##     [79] "oatmeal"           "oil"                "olive oil"  
##     [82] "pancakes"          "parmesan cheese"    "pasta"  
##     [85] "pepper"            "pet food"           "pickles"  
##     [88] "protein bar"       "red wine"           "rice"  
##     [91] "salad"             "salmon"             "salt"  
##     [94] "sandwich"          "shallot"            "shampoo"  
##     [97] "shrimp"            "soda"               "soup"  
##    [100] "spaghetti"         "sparkling water"    "spinach"  
##    [103] "strawberries"      "strong cheese"      "tea"  
##    [106] "tomato juice"      "tomato sauce"       "tomatoes"  
##    [109] "toothpaste"        "turkey"             "vegetables mix"
```

```
## [112] "water spray"          "white wine"          "whole weat flour"
## [115] "whole wheat pasta"    "whole wheat rice"    "yams"
## [118] "yogurt cake"          "zucchini"
```

```
class(supermarket)
```

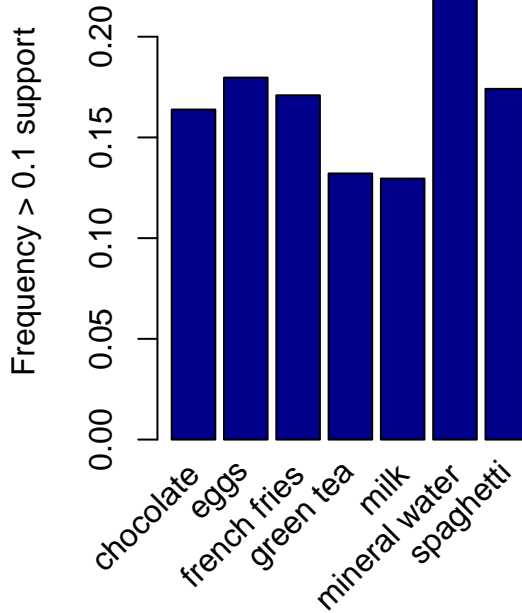
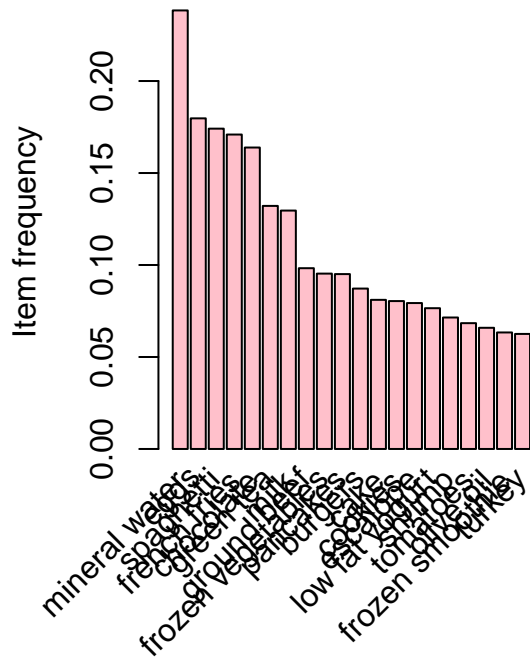
```
## [1] "transactions"
## attr(,"package")
## [1] "arules"
```

```
summary(supermarket)
```

```
## 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
```

```
# Plotting item frequency considering the top 20 items
par(mfcol=c(1,2))
itemFrequencyPlot(supermarket,topN=20,col="pink",
                  ylab="Item frequency",
                  main=" Item Frequency Plots")
itemFrequencyPlot(supermarket,support=0.112,col="darkblue",
                  ylab="Frequency > 0.1 support")
```

Item Frequency Plots



```
# Exploring the frequency of some articles
# some operation in percentage terms of the total transactions
itemFrequency(supermarket[, 1:5], type = "absolute")
```

##	almonds	antioxydant	juice	asparagus	avocado
##	153		67	36	250
##	babies	food			
##	34				

```
round(itemFrequency(supermarket[, 1:5], type = "relative")*100, 2)
```

##	almonds	antioxydant	juice	asparagus	avocado
##	2.04		0.89	0.48	3.33
##	babies	food			
##	0.45				

```
# The first rules
rule1<-apriori(supermarket,parameter = list(support=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].

rule1

## set of 74 rules

plot(rule1,type = "graph",control=list(type="items"))

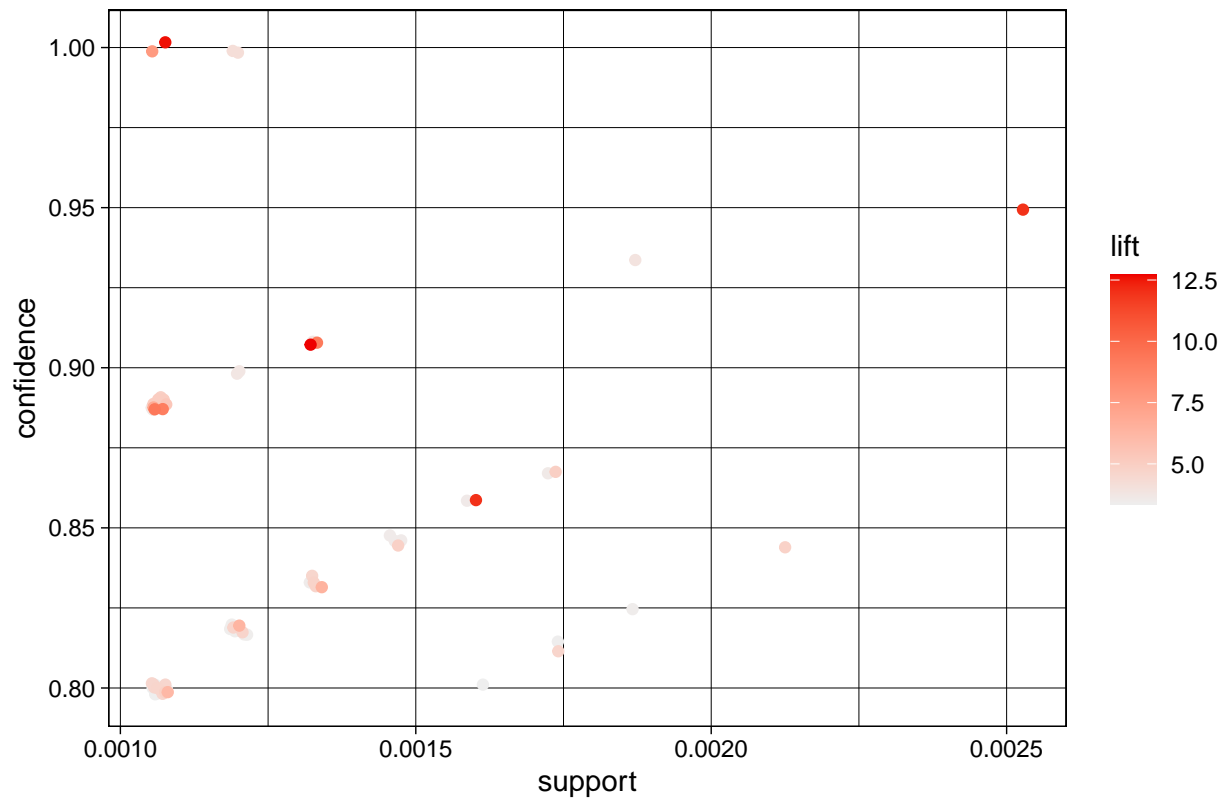
## Warning: Unknown control parameters: type, type

## Available control parameters (with default values):
## main = Scatter plot for 74 rules
## colors = c("#EE0000FF", "#EEEEEEFF")
## jitter = NA
## engine = ggplot2
## verbose = FALSE

## To reduce overplotting, jitter is added! Use jitter = 0 to prevent jitter.

```


Scatter plot for 74 rules



```
inspect(rule1[1:10])
```

##	lhs	rhs	support	confidence
## [1]	{frozen smoothie, spinach}	=> {mineral water}	0.001066524	0.8888889
## [2]	{bacon, pancakes}	=> {spaghetti}	0.001733102	0.8125000
## [3]	{nonfat milk, turkey}	=> {mineral water}	0.001199840	0.8181818
## [4]	{ground beef, nonfat milk}	=> {mineral water}	0.001599787	0.8571429
## [5]	{mushroom cream sauce, pasta}	=> {escalope}	0.002532996	0.9500000
## [6]	{milk, pasta}	=> {shrimp}	0.001599787	0.8571429
## [7]	{cooking oil, fromage blanc}	=> {mineral water}	0.001199840	0.8181818
## [8]	{black tea, salmon}	=> {mineral water}	0.001066524	0.8000000
## [9]	{black tea, frozen smoothie}	=> {milk}	0.001199840	0.8181818
## [10]	{red wine, tomato sauce}	=> {chocolate}	0.001066524	0.8000000

##	coverage	lift	count
## [1]	0.001199840	3.729058	8
## [2]	0.002133049	4.666587	13
## [3]	0.001466471	3.432428	9
## [4]	0.001866418	3.595877	12
## [5]	0.002666311	11.976387	19
## [6]	0.001866418	11.995203	12
## [7]	0.001466471	3.432428	9
## [8]	0.001333156	3.356152	8
## [9]	0.001466471	6.313973	9
## [10]	0.001333156	4.882669	8

```

sorting1<-sort(rule1,by="confidence",decreasing = TRUE)
inspect(sorting1[1:10])

```

	lhs	rhs	support	confidence	coverage	lift	count
## [1]	{french fries, mushroom cream sauce, pasta}	=> {escalope}	0.001066524	1.0000000	0.001066524	12.606723	8
## [2]	{ground beef, light cream, olive oil}	=> {mineral water}	0.001199840	1.0000000	0.001199840	4.195190	9
## [3]	{cake, meatballs, mineral water}	=> {milk}	0.001066524	1.0000000	0.001066524	7.717078	8
## [4]	{cake, olive oil, shrimp}	=> {mineral water}	0.001199840	1.0000000	0.001199840	4.195190	9
## [5]	{mushroom cream sauce, pasta}	=> {escalope}	0.002532996	0.9500000	0.002666311	11.976387	19
## [6]	{red wine, soup}	=> {mineral water}	0.001866418	0.9333333	0.001999733	3.915511	14
## [7]	{eggs, mineral water, pasta}	=> {shrimp}	0.001333156	0.9090909	0.001466471	12.722185	10
## [8]	{herb & pepper, mineral water, rice}	=> {ground beef}	0.001333156	0.9090909	0.001466471	9.252498	10
## [9]	{ground beef, pancakes, whole wheat rice}	=> {mineral water}	0.001333156	0.9090909	0.001466471	3.813809	10
## [10]	{frozen vegetables, milk, spaghetti, turkey}	=> {mineral water}	0.001199840	0.9000000	0.001333156	3.775671	9

```

# Confidence level for Rule1 is 80%.
#{X} is also called antecedent or left-hand-side (LHS) and
# {Y} is called consequent or right-hand-side (RHS).
#Support is an indication of how frequently a set of items appear in baskets.
# Confidence is an indication of how often the support-rule has been found to be true.
#Lift is a measure of association using both support and confidence.

```

```

# Minimizing support thershold alittle bit
rule2<-apriori(supermarket,parameter =list(support=0.001,conf=0.75))

```

```

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##      0.75    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 ... [110 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].
```

```
# Viewing the rules extracted from these
inspect(rule2[1:10])
```

##	lhs	rhs	support	confidence
## [1]	{frozen smoothie, spinach}	=> {mineral water}	0.001066524	0.8888889
## [2]	{blueberries, eggs}	=> {mineral water}	0.001599787	0.7500000
## [3]	{bacon, pancakes}	=> {spaghetti}	0.001733102	0.8125000
## [4]	{nonfat milk, turkey}	=> {mineral water}	0.001199840	0.8181818
## [5]	{ground beef, nonfat milk}	=> {mineral water}	0.001599787	0.8571429
## [6]	{barbecue sauce, chocolate}	=> {mineral water}	0.001333156	0.7692308
## [7]	{mushroom cream sauce, pasta}	=> {escalope}	0.002532996	0.9500000
## [8]	{milk, pasta}	=> {shrimp}	0.001599787	0.8571429
## [9]	{mineral water, pasta}	=> {shrimp}	0.001599787	0.7500000
## [10]	{cooking oil, fromage blanc}	=> {mineral water}	0.001199840	0.8181818

##	coverage	lift	count
## [1]	0.001199840	3.729058	8
## [2]	0.002133049	3.146393	12
## [3]	0.002133049	4.666587	13
## [4]	0.001466471	3.432428	9
## [5]	0.001866418	3.595877	12
## [6]	0.001733102	3.227069	10
## [7]	0.002666311	11.976387	19
## [8]	0.001866418	11.995203	12
## [9]	0.002133049	10.495802	12
## [10]	0.001466471	3.432428	9

```
sorting<-sort(rule2,by="confidence",decreasing = TRUE)
inspect(sorting[1:10])
```

##	lhs	rhs	support	confidence	coverage	lift	count
## [1]	{french fries, mushroom cream sauce, pasta}	=> {escalope}	0.001066524	1.0000000	0.001066524	12.606723	8
## [2]	{ground beef, light cream, olive oil}	=> {mineral water}	0.001199840	1.0000000	0.001199840	4.195190	9
## [3]	{cake, meatballs, mineral water}	=> {milk}	0.001066524	1.0000000	0.001066524	7.717078	8

```
## [4] {cake,
##      olive oil,
##      shrimp}      => {mineral water} 0.001199840 1.0000000 0.001199840 4.195190 9
## [5] {mushroom cream sauce,
##      pasta}      => {escalope} 0.002532996 0.9500000 0.002666311 11.976387 19
## [6] {red wine,
##      soup}      => {mineral water} 0.001866418 0.9333333 0.001999733 3.915511 14
## [7] {eggs,
##      mineral water,
##      pasta}      => {shrimp} 0.001333156 0.9090909 0.001466471 12.722185 10
## [8] {herb & pepper,
##      mineral water,
##      rice}      => {ground beef} 0.001333156 0.9090909 0.001466471 9.252498 10
## [9] {ground beef,
##      pancakes,
##      whole wheat rice} => {mineral water} 0.001333156 0.9090909 0.001466471 3.813809 10
## [10] {frozen vegetables,
##      milk,
##      spaghetti,
##      turkey}      => {mineral water} 0.001199840 0.9000000 0.001333156 3.775671 9
```

```
# Getting items that are bought after milk is bought
milk<-subset(rule2,subset=lhs %pin% "milk")
# Sorting items by their confidence level
sorted_milk<-sort(milk,by="confidence",decreasing = TRUE)
# Viewing the top 10 items
inspect(sorted_milk[1:10])
```

	lhs	rhs	support	confidence	coverage	lift	count
## [1]	{frozen vegetables, milk, spaghetti, turkey}	=> {mineral water}	0.001199840	0.9000000	0.001333156	3.775671	9
## [2]	{cake, meatballs, milk}	=> {mineral water}	0.001066524	0.8888889	0.001199840	3.729058	8
## [3]	{burgers, milk, salmon}	=> {spaghetti}	0.001066524	0.8888889	0.001199840	5.105326	8
## [4]	{chocolate, ground beef, milk, mineral water, spaghetti}	=> {frozen vegetables}	0.001066524	0.8888889	0.001199840	9.325253	8
## [5]	{ground beef, nonfat milk}	=> {mineral water}	0.001599787	0.8571429	0.001866418	3.595877	12
## [6]	{milk, pasta}	=> {shrimp}	0.001599787	0.8571429	0.001866418	11.995203	12
## [7]	{frozen vegetables, milk, shrimp, spaghetti}	=> {mineral water}	0.001466471	0.8461538	0.001733102	3.549776	11
## [8]	{nonfat milk, turkey}	=> {mineral water}	0.001199840	0.8181818	0.001466471	3.432428	9

```
## [9] {french fries,
##      herb & pepper,
##      milk}          => {mineral water}      0.001199840  0.8181818 0.001466471  3.432428      9
## [10] {frozen vegetables,
##       milk,
##       olive oil,
##       soup}          => {mineral water}      0.001199840  0.8181818 0.001466471  3.432428      9
```

```
# Getting items purchased before shrimp
shrimp<-subset(rule2,subset=rhs %pin% "shrimp")
# Sorting items by their confidence level
sorted_shrimp<-sort(shrimp,by="confidence",decreasing = TRUE)
# Viewing the top 10 items
inspect(sorted_shrimp[1:3])
```

```
##      lhs                                rhs      support      confidence coverage
## [1] {eggs, mineral water, pasta} => {shrimp} 0.001333156 0.9090909 0.001466471
## [2] {milk, pasta}                => {shrimp} 0.001599787 0.8571429 0.001866418
## [3] {mineral water, pasta}       => {shrimp} 0.001599787 0.7500000 0.002133049
##      lift      count
## [1] 12.72218 10
## [2] 11.99520 12
## [3] 10.49580 12
```

```
# Getting items purchased before escalope
escalope<-subset(rule2,subset=rhs %pin% "escalope")
# Sorting items by their confidence level
sorted_escalope<-sort(escalope,by="confidence",decreasing = TRUE)
# Viewing the top 10 items
inspect(sorted_escalope[1:2])
```

```
##      lhs                                rhs      support confidence      coverage      lift count
## [1] {french fries,
##      mushroom cream sauce,
##      pasta}          => {escalope} 0.001066524      1.00 0.001066524 12.60672      8
## [2] {mushroom cream sauce,
##      pasta}          => {escalope} 0.002532996      0.95 0.002666311 11.97639     19
```

```
# Getting items purchased before water
water<-subset(rule2,subset=rhs %pin% "mineral water")
# Sorting items by their confidence level
sorted_water <-sort(water,by="confidence",decreasing = TRUE)
# Viewing the top 10 items
inspect(sorted_water[1:10])
```

```
##      lhs                                rhs      support confidence      coverage      lift count
## [1] {ground beef,
##      light cream,
##      olive oil}          => {mineral water} 0.001199840 1.0000000 0.001199840 4.195190      9
## [2] {cake,
##      olive oil,
##      shrimp}            => {mineral water} 0.001199840 1.0000000 0.001199840 4.195190      9
```

```

## [3] {red wine,
##      soup}          => {mineral water} 0.001866418 0.9333333 0.001999733 3.915511 14
## [4] {ground beef,
##      pancakes,
##      whole wheat rice} => {mineral water} 0.001333156 0.9090909 0.001466471 3.813809 10
## [5] {frozen vegetables,
##      milk,
##      spaghetti,
##      turkey}        => {mineral water} 0.001199840 0.9000000 0.001333156 3.775671 9
## [6] {chocolate,
##      frozen vegetables,
##      olive oil,
##      shrimp}        => {mineral water} 0.001199840 0.9000000 0.001333156 3.775671 9
## [7] {frozen smoothie,
##      spinach}        => {mineral water} 0.001066524 0.8888889 0.001199840 3.729058 8
## [8] {cake,
##      meatballs,
##      milk}          => {mineral water} 0.001066524 0.8888889 0.001199840 3.729058 8
## [9] {cake,
##      olive oil,
##      whole wheat pasta} => {mineral water} 0.001066524 0.8888889 0.001199840 3.729058 8
## [10] {brownies,
##       eggs,
##       ground beef}    => {mineral water} 0.001066524 0.8888889 0.001199840 3.729058 8

```

```

# Mineral water is the most bought item.
# Support is an indication of how frequently a set of items appear in baskets.
# Confidence is an indication of how often the support-rule has been
# found to be true.
# Lift is a measure of association using both support and confidence.
# The results indicate what is bought alongside the product:
##

```