

# markass.R

*spoor*

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```
# Association Rules for Market Basket Analysis (R)

library(arules) # association rules

## Loading required package: Matrix
##
## Attaching package: 'arules'
## The following objects are masked from 'package:base':
##
##      abbreviate, write
library(arulesViz) # data visualization of association rules

## Loading required package: grid
library(RColorBrewer) # color palettes for plots

data(Groceries) # grocery transactions object from arules package

# show the dimensions of the transactions object
print(dim(Groceries))

## [1] 9835 169
print(dim(Groceries)[1]) # 9835 market baskets for shopping trips

## [1] 9835
print(dim(Groceries)[2]) # 169 initial store items

## [1] 169
# examine frequency for each item with support greater than 0.025
pdf(file="fig_market_basket_initial_item_support.pdf",
    width = 8.5, height = 11)
itemFrequencyPlot(Groceries, support = 0.025, cex.names=0.8, xlim = c(0,0.3),
    type = "relative", horiz = TRUE, col = "dark red", las = 1,
    xlab = paste("Proportion of Market Baskets Containing Item",
        "\n(Item Relative Frequency or Support)"))
dev.off()

## pdf
## 2
# explore possibilities for combining similar items
print(head(itemInfo(Groceries)))

##           labels level2           level1
## 1 frankfurter sausage meat and sausage
## 2      sausage sausage meat and sausage
## 3      liver loaf sausage meat and sausage
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## 4          ham sausage meat and sausage
## 5          meat sausage meat and sausage
## 6 finished products sausage meat and sausage

print(levels(itemInfo(Groceries)[["level1"]])) # 10 levels... too few

## [1] "canned food"          "detergent"          "drinks"
## [4] "fresh products"        "fruit and vegetables" "meat and sausage"
## [7] "non-food"              "perfumery"          "processed food"
## [10] "snacks and candies"

print(levels(itemInfo(Groceries)[["level2"]])) # 55 distinct levels

## [1] "baby food"              "bags"
## [3] "bakery improver"        "bathroom cleaner"
## [5] "beef"                  "beer"
## [7] "bread and backed goods" "candy"
## [9] "canned fish"           "canned fruit/vegetables"
## [11] "cheese"                "chewing gum"
## [13] "chocolate"             "cleaner"
## [15] "coffee"               "condiments"
## [17] "cosmetics"             "dairy produce"
## [19] "delicatessen"          "dental care"
## [21] "detergent/softener"    "eggs"
## [23] "fish"                  "frozen foods"
## [25] "fruit"                 "games/books/hobby"
## [27] "garden"                "hair care"
## [29] "hard drinks"           "health food"
## [31] "jam/sweet spreads"     "long-life bakery products"
## [33] "meat spreads"          "non-alc. drinks"
## [35] "non-food house keeping products" "non-food kitchen"
## [37] "packaged fruit/vegetables" "perfumery"
## [39] "personal hygiene"      "pet food/care"
## [41] "pork"                  "poultry"
## [43] "pudding powder"        "sausage"
## [45] "seasonal products"     "shelf-stable dairy"
## [47] "snacks"                "soap"
## [49] "soups/sauces"          "staple foods"
## [51] "sweetener"             "tea/cocoa drinks"
## [53] "vegetables"            "vinegar/oils"
## [55] "wine"

# aggregate items using the 55 level2 levels for food categories
# to create a more meaningful set of items
groceries <- aggregate(Groceries, itemInfo(Groceries)[["level2"]])

print(dim(groceries)[1]) # 9835 market baskets for shopping trips

## [1] 9835

print(dim(groceries)[2]) # 55 final store items (categories)

## [1] 55

pdf(file="fig_market_basket_final_item_support.pdf", width = 8.5, height = 11)
itemFrequencyPlot(groceries, support = 0.025, cex.names=1.0, xlim = c(0,0.5),
  type = "relative", horiz = TRUE, col = "blue", las = 1,

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xlab = paste("Proportion of Market Baskets Containing Item",
"\n(Item Relative Frequency or Support)")
dev.off()

## pdf
## 2

# obtain large set of association rules for items by category and all shoppers
# this is done by setting very low criteria for support and confidence
first.rules <- apriori(groceries,
  parameter = list(support = 0.001, confidence = 0.05))

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
## 0.05 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: 9
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[55 item(s), 9835 transaction(s)] done [0.01s].
## sorting and recoding items ... [54 item(s)] done [0.00s].
## creating transaction tree ... done [0.01s].
## checking subsets of size 1 2 3 4 5 6 7 8 done [0.03s].
## writing ... [69921 rule(s)] done [0.04s].
## creating S4 object ... done [0.06s].

print(summary(first.rules)) # yields 69,921 rules... too many

## set of 69921 rules
##
## rule length distribution (lhs + rhs):sizes
## 1 2 3 4 5 6 7 8
## 21 1205 10467 23895 22560 9888 1813 72
##
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 1.000 4.000 4.000 4.502 5.000 8.000
##
## summary of quality measures:
## support confidence lift
## Min. :0.001017 Min. :0.0500 Min. : 0.4475
## 1st Qu.:0.001118 1st Qu.:0.2110 1st Qu.: 1.8315
## Median :0.001525 Median :0.4231 Median : 2.2573
## Mean :0.002488 Mean :0.4364 Mean : 2.5382
## 3rd Qu.:0.002339 3rd Qu.:0.6269 3rd Qu.: 2.9662
## Max. :0.443010 Max. :1.0000 Max. :16.1760
##
## mining info:

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##      data ntransactions support confidence
## groceries      9835    0.001      0.05

# select association rules using thresholds for support and confidence
second.rules <- apriori(groceries,
  parameter = list(support = 0.025, confidence = 0.05))

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##      0.05    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: 245
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[55 item(s), 9835 transaction(s)] done [0.01s].
## sorting and recoding items ... [32 item(s)] done [0.00s].
## creating transaction tree ... done [0.01s].
## checking subsets of size 1 2 3 4 done [0.00s].
## writing ... [344 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].

print(summary(second.rules)) # yields 344 rules

## set of 344 rules
##
## rule length distribution (lhs + rhs):sizes
##   1   2   3   4
## 21 162 129  32
##
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.0    2.0    2.0    2.5    3.0    4.0
##
## summary of quality measures:
##      support      confidence      lift
## Min.   :0.02542 Min.   :0.05043 Min.   :0.6669
## 1st Qu.:0.03030 1st Qu.:0.18202 1st Qu.:1.2498
## Median :0.03854 Median :0.39522 Median :1.4770
## Mean   :0.05276 Mean   :0.37658 Mean   :1.4831
## 3rd Qu.:0.05236 3rd Qu.:0.51271 3rd Qu.:1.7094
## Max.   :0.44301 Max.   :0.79841 Max.   :2.4073
##
## mining info:
##      data ntransactions support confidence
## groceries      9835    0.025      0.05

# data visualization of association rules in scatter plot
pdf(file="fig_market_basket_rules.pdf", width = 8.5, height = 8.5)
plot(second.rules,

```

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control=list(jitter=2, col = rev(brewer.pal(9, "Greens")[4:9])),
shading = "lift")
dev.off()

```

```

## pdf
## 2

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# grouped matrix of rules
pdf(file="fig_market_basket_rules_matrix.pdf", width = 8.5, height = 8.5)
plot(second.rules, method="grouped",
      control=list(col = rev(brewer.pal(9, "Greens")[4:9])))
dev.off()

```

```

## pdf
## 2

```

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# select rules with vegetables in consequent (right-hand-side) item subsets
vegie.rules <- subset(second.rules, subset = rhs %pin% "vegetables")
inspect(vegie.rules) # 41 rules

```

##	lhs	rhs	support	confidence	lift
## [1]	{}	=> {vegetables}	0.27300458	0.2730046	1.0000000
## [2]	{poultry}	=> {vegetables}	0.02897814	0.5745968	2.1047148
## [3]	{pork}	=> {vegetables}	0.03009659	0.5220459	1.9122238
## [4]	{staple foods}	=> {vegetables}	0.02613116	0.5160643	1.8903136
## [5]	{eggs}	=> {vegetables}	0.03141840	0.4951923	1.8138608
## [6]	{games/books/hobby}	=> {vegetables}	0.02785968	0.3145809	1.1522918
## [7]	{long-life bakery products}	=> {vegetables}	0.02907982	0.3492063	1.2791227
## [8]	{perfumery}	=> {vegetables}	0.03213015	0.4056483	1.4858662
## [9]	{beef}	=> {vegetables}	0.04585663	0.5595533	2.0496116
## [10]	{bags}	=> {vegetables}	0.03141840	0.3175745	1.1632571
## [11]	{vinegar/oils}	=> {vegetables}	0.04199288	0.4666667	1.7093731
## [12]	{chocolate}	=> {vegetables}	0.03192679	0.2934579	1.0749195
## [13]	{beer}	=> {vegetables}	0.03406202	0.2189542	0.8020168
## [14]	{frozen foods}	=> {vegetables}	0.04738180	0.4052174	1.4842879
## [15]	{cheese}	=> {vegetables}	0.05531266	0.4365971	1.5992300
## [16]	{sausage}	=> {vegetables}	0.07625826	0.4032258	1.4769929
## [17]	{fruit}	=> {vegetables}	0.10706660	0.4297959	1.5743176
## [18]	{non-alc. drinks}	=> {vegetables}	0.09456024	0.2974097	1.0893944
## [19]	{bread and backed goods}	=> {vegetables}	0.11621759	0.3363743	1.2321198
## [20]	{dairy produce}	=> {vegetables}	0.17041179	0.3846683	1.4090180
## [21]	{beef,				
##	dairy produce}	=> {vegetables}	0.02989324	0.6074380	2.2250104
## [22]	{dairy produce,				
##	vinegar/oils}	=> {vegetables}	0.03141840	0.5355286	1.9616103
## [23]	{dairy produce,				
##	frozen foods}	=> {vegetables}	0.03436706	0.5121212	1.8758704
## [24]	{cheese,				
##	fruit}	=> {vegetables}	0.02674123	0.5197628	1.9038613
## [25]	{bread and backed goods,				
##	cheese}	=> {vegetables}	0.02887646	0.4536741	1.6617821
## [26]	{cheese,				
##	dairy produce}	=> {vegetables}	0.04219624	0.4987981	1.8270686
## [27]	{fruit,				
##	sausage}	=> {vegetables}	0.03426538	0.5290424	1.9378517

```
## [28] {non-alc. drinks,
##       sausage}          => {vegetables} 0.03029995 0.4156206 1.5223944
## [29] {bread and backed goods,
##       sausage}          => {vegetables} 0.04382308 0.4229637 1.5492916
## [30] {dairy produce,
##       sausage}          => {vegetables} 0.05266904 0.4905303 1.7967842
## [31] {fruit,
##       non-alc. drinks}  => {vegetables} 0.04361973 0.4657980 1.7061914
## [32] {bread and backed goods,
##       fruit}            => {vegetables} 0.05124555 0.4763705 1.7449177
## [33] {dairy produce,
##       fruit}            => {vegetables} 0.07869853 0.5032510 1.8433793
## [34] {bread and backed goods,
##       non-alc. drinks}  => {vegetables} 0.04636502 0.3731588 1.3668590
## [35] {dairy produce,
##       non-alc. drinks}  => {vegetables} 0.06446365 0.4243641 1.5544213
## [36] {bread and backed goods,
##       dairy produce}    => {vegetables} 0.08195221 0.4366197 1.5993128
## [37] {dairy produce,
##       fruit,
##       sausage}          => {vegetables} 0.02714794 0.5741935 2.1032378
## [38] {bread and backed goods,
##       dairy produce,
##       sausage}          => {vegetables} 0.03284189 0.5135135 1.8809704
## [39] {dairy produce,
##       fruit,
##       non-alc. drinks}  => {vegetables} 0.03304525 0.5183413 1.8986543
## [40] {bread and backed goods,
##       dairy produce,
##       fruit}            => {vegetables} 0.04077275 0.5276316 1.9326840
## [41] {bread and backed goods,
##       dairy produce,
##       non-alc. drinks}  => {vegetables} 0.03345196 0.4627286 1.6949480
```

```
# sort by lift and identify the top 10 rules
```

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

```
##      lhs                rhs          support confidence    lift
## [1] {beef,
##      dairy produce}    => {vegetables} 0.02989324 0.6074380 2.225010
## [2] {poultry}          => {vegetables} 0.02897814 0.5745968 2.104715
## [3] {dairy produce,
##      fruit,
##      sausage}          => {vegetables} 0.02714794 0.5741935 2.103238
## [4] {beef}              => {vegetables} 0.04585663 0.5595533 2.049612
## [5] {dairy produce,
##      vinegar/oils}     => {vegetables} 0.03141840 0.5355286 1.961610
## [6] {fruit,
##      sausage}          => {vegetables} 0.03426538 0.5290424 1.937852
## [7] {bread and backed goods,
##      dairy produce,
##      fruit}            => {vegetables} 0.04077275 0.5276316 1.932684
## [8] {pork}              => {vegetables} 0.03009659 0.5220459 1.912224
## [9] {cheese,
```

```
##      fruit}                                => {vegetables} 0.02674123  0.5197628 1.903861
## [10] {dairy produce,
##      fruit,
##      non-alc. drinks}                      => {vegetables} 0.03304525  0.5183413 1.898654
```

```
pdf(file="fig_market_basket_farmer_rules.pdf", width = 11, height = 8.5)
plot(top.vegie.rules, method="graph",
     control=list(type="items"),
     shading = "lift")
dev.off()
```

```
## pdf
## 2
```

```
# Suggestions for the student:
# Suppose your client is someone other than the local farmer,
# a meat producer/butcher, dairy, or brewer perhaps.
# Determine association rules relevant to your client's products
# guided by the market basket model. What recommendations
# would you make about future marketplace actions?
```