

## session21\_assign\_pca.R

Data Set

2. Perform the below given activities:
  - a. Apply PCA to the dataset and show proportion of variance
  - b. Perform PCA using SVD approach
  - c. Show the graphs of PCA components

```
setwd("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session21")
library(readr)
epi_r <- read_csv("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/
session21/epi_r.csv")
View(epi_r)
data<-epi_r
View(data)
head(data, n=10)

# data sets in package
data(package="arules")

# Split data
dt <- split(data$rating, data$arizona)
dt

# Loading arules package
require(arules)
require(arulesViz)

# Convert data to transaction level
dt2 = as(dt,"transactions")
dt2
```

```

summary(dt2)
inspect(dt2)

# Most Frequent Items
itemFrequency(dt2, type = "relative")
itemFrequencyPlot(dt2,topN = 5)

# with support parameters
itemFrequency(dt2, type = "relative")
itemFrequencyPlot(dt2,support= 0.10)

# aggregated data
rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8))
rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8, minlen = 3
))
rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8, maxlen = 4
))
rules
summary(rules)

inspect(rules[1:10]) # to view first 10 rules

#Convert rules into data frame
rules3 = as(rules, "data.frame")
write(rules, "C:/Users/Seshan/Desktop/PCA//rules2.csv", sep=",")

# Show only particular product rules
inspect( subset( rules, subset = rhs %pin% "0" )[1:10])

# Show the top 10 rules
options(digits=2)
inspect(rules[1:10])

```

```

# Get Summary Information

summary(rules)
plot(rules)
plot(rules, method = "graph", interactive = T)

# Sort by Lift
rules<-sort(rules, by="lift", decreasing=TRUE)

# Sort by Lift
rules<-sort(rules, by="lift", decreasing=TRUE)

# Remove Unnecessary Rules
subset.matrix <- is.subset(rules, rules)
subset.matrix[lower.tri(subset.matrix, diag=T)] <- NA
redundant <- colSums(subset.matrix, na.rm=T) >= 1
which(redundant)
rules.pruned <- rules[!redundant]
rules<-rules.pruned
rules

#Clean Rules
rules3$rules=gsub("\\{", "", rules3$rules)
rules3$rules=gsub("\\}", "", rules3$rules)
rules3$rules=gsub("\\", "", rules3$rules)

#Split the rule
library(splitstackshape)

```

```

Rules4=cSplit(rules3, "rules", ">")
names(Rules4)[names(Rules4) == 'rules_1'] <- 'LHS'
Rules5=cSplit(Rules4, "LHS", ",", ",")
Rules6=subset(Rules5, select= -c(rules_2))
names(Rules6)[names(Rules6) == 'rules_3'] <- 'RHS'

# What are customers likely to buy before they purchase "Product A"
rules<-apriori(data=dt, parameter=list(supp=0.001,conf = 0.8),
               appearance = list(default="lhs",rhs="0"),
               control = list(verbose=F))
rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])

# What are customers likely to buy if they purchased "Product A"
rules<-apriori(data=dt, parameter=list(supp=0.001,conf = 0.8),appearance = list(
  default="rhs",lhs="0"),control = list(verbose=F))
rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])

rules
support<-seq(0.01,0.1,0.01)
support
rules_count<-c(472,128,46,26,14, 10, 10,8,8,8)
rules_count
plot(support,rules_count,type = "l",main="Number of rules at different support %",col="darkred",lwd=3)

conf<-seq(0.10,1.0,0.10)
conf

```

```

rules_count<-c(472,231,125,62,15,0,0,0,0,0)
rules_count

plot(conf,rules_count,type = "l",main="Number of rules at different confidence %",col="darkred",lwd=3)

#rules_ec <- eclat(epi_r, parameter = list(supp = 0.05))
#summary(rules_ec)
#sorting out the most relevant rules
rules<-sort(rules, by="confidence", decreasing=TRUE)
inspect(rules[1:5])

rules<-sort(rules, by="lift", decreasing=TRUE)
inspect(rules[1:5])

#####

library(factoextra)

library("factoextra")
data1<-na.exclude(data)
na.omit(data1)
data1.active <- data1[2:100, 2:6]
na.exclude(data1.active)
View(data1.active)
head(data1.active[, 2:5])

#Compute PCA in R using prcomp()
library(factoextra)
res.pca <- prcomp(data1.active, scale = TRUE)
res.pca

```

```

summary(res.pca)

fviz_eig(res.pca)

fviz_pca_ind(res.pca, col.ind = "cos2", # Color by the quality of representat
ion gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"), repel = TRUE      # Av
oid text overlapping)

fviz_pca_var(res.pca, col.var = "contrib", # Color by contributions to the PC
gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"), repel = TRUE      # Avoid
text overlapping)

fviz_pca_biplot(res.pca, repel = TRUE,col.var = "#2E9FDF", # Variables color
col.ind = "#696969" # Individuals color)

library(factoextra)

# Eigenvalues

eig.val <- get_eigenvalue(res.pca)

eig.val

# Results for Variables

res.var <- get_pca_var(res.pca)

res.var$coord          # Coordinates

res.var$contrib         # Contributions to the PCs

res.var$cos2           # Quality of representation


##                                     title rating
## 1                               Lentil, Apple, and Turkey Wrap  2.500
## 2          Boudin Blanc Terrine with Red Onion Confit  4.375
## 3                               Potato and Fennel Soup Hodge  3.750
## 4          Mahi-Mahi in Tomato Olive Sauce  5.000
## 5                               Spinach Noodle Casserole  3.125
## 6                               The Best Blts  4.375
## 7  Ham and Spring Vegetable Salad with Shallot Vinaigrette  4.375
## 8                               Spicy-Sweet Kumquats  3.750
## 9                               Korean Marinated Beef  4.375
## 10 Ham Persillade with Mustard Potato Salad and Mashed Peas  3.750
##  calories protein fat sodium X.cakeweek X.wasteless X22.minute.meals
## 1         426      30  7    559          0          0          0
## 2         403      18 23   1439          0          0          0
## 3         165       6  7    165          0          0          0

```

## 4	NA	NA	NA	NA	0	0	0
## 5	547	20	32	452	0	0	0
## 6	948	19	79	1042	0	0	0
## 7	NA	NA	NA	NA	0	0	0
## 8	NA	NA	NA	NA	0	0	0
## 9	170	7	10	1272	0	0	0
## 10	602	23	41	1696	0	0	0
##	X3.ingredient.recipes			X30.days.of.groceries		advance.prep.required	
## 1			0			0	0
## 2			0			0	0
## 3			0			0	0
## 4			0			0	0
## 5			0			0	0
## 6			0			0	0
## 7			0			0	0
## 8			0			0	0
## 9			0			0	0
## 10			0			0	0
##	alabama	alaska	alcoholic	almond	amaretto	anchovy	anise anniversary
## 1	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0
##	anthony.bourdain		aperitif	appetizer	apple	apple.juice	apricot arizona
## 1			0	0	1	0	0
## 2			0	0	0	0	0
## 3			0	0	0	0	0
## 4			0	0	0	0	0
## 5			0	0	0	0	0
## 6			0	0	0	0	0
## 7			0	0	0	0	0
## 8			0	0	0	0	0
## 9			0	0	0	0	0
## 10			0	0	0	0	0
##	artichoke	arugula	asian.pear	asparagus	aspen	atlanta	australia avocado
## 1	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0
## 7	0	0	0	1	0	0	0
## 8	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0

```
## 9      0      0      0      0      0      0
## 10     0      0      0      0      0      0
```

```
# check data
```

```
##
```

```
title
```

```
## 1
```

```
Lentil, Apple, and Turkey Wrap
```

```
## 2
```

```
Boudin Blanc Terrine with Red Onion Confit
```

```
## 3
```

```
Potato and Fennel Soup Hodge
```

```
## 4
```

```
Mahi-Mahi in Tomato Olive Sauce
```

```
## 5
```

```
Spinach Noodle Casserole
```

```
## 6
```

```
The Best Blts
```

```
## 7
```

```
ing Vegetable Salad with Shallot Vinaigrette
```

Ham and Spr

```
## 8
```

```
Spicy-Sweet Kumquats
```

```
## 9
```

```
Korean Marinated Beef
```

```
## 10
```

```
de with Mustard Potato Salad and Mashed Peas
```

Ham Persilla

```
## 11
```

```
Yams Braised with Cream, Rosemary and Nutmeg
```

```
## 12
```

```
Spicy Noodle Soup
```

```
## 13
```

```
colate Chip Cake With Peanut Butter Frosting
```

Banana-Cho

```
## 14
```

```
Beef Tenderloin with Garlic and Brandy
```

```
## 15
```

```
Peach Mustard
```

```
## 16
```

```
Raw Cream of Spinach Soup
```

```
## 17
```

```
Sweet Buttermilk Spoon Breads
```

```
## 18
```

```
Crisp Braised Pork Shoulder
```

```
## 19
```

```
ella-Topped Peppers with Tomatoes and Garlic
```

Mozzar

```
## 20
```

```
alad with Chive Vinaigrette and Fried Capers
```

Tuna, Asparagus, and New Potato S

```
## 21
```

Asian Pe



ar and Watercress Salad with Sesame Dressing

## 22

"Fried" Chicken

## 23

Fish Fi

llets in Parchment with Asparagus and Orange

## 24

Pancetta and Taleggio Lasagna with Treviso

## 25

Sea Salt-Roasted Pecans

## 26

Garlic Baguette Crumbs

## 27

Cucumber-Basil Egg Salad

## 28

Dried Pear Crisps

## 29

Green Bean, Red Onion, and

Roast Potato Salad with Rosemary Vinaigrette

## 30

Apricot-Cherry Shortcakes

## 31

Asian Steak Topped with Bell Pepper Stir-Fry

## 32

Moroccan-Style Preserved Lemons

## 33

Roaste

d Sweet-Potato Spears with Bacon Vinaigrette

## 34

Deviled Ham

## 35

Fontina Mac with Squash and Sage

## 36

Aztec Chicken

## 37

Pastry Twists with Spiced Sugar-Honey Glaze

## 38

Sauteed Broccoli Rabe

## 39

Grouper with Tomato and Basil

## 40

Better-Than-Pita Grill Bread

## 41

Coconut-Key Lime Sheet Cake

## 42

Baked Hal

ibut with Orzo, Spinach, and Cherry Tomatoes

## 43

Honey Rye

## 44

Purple-Potato and Crab Gratin

## 45

Grilled Beef, JÃ-cama, and Apple Salad

## 46

Pickled Red Onions	
## 47	
Spicy Black Beans and Rice	
## 48	
Herbed Goat Cheese Spread with Mint	
## 49	
Mexican Lime Soup	
## 50	
Citrus Salad with Mint Sugar	
## 51	
Mexican Chile and Mushroom Soup	
## 52	
Peanut Butter-Banana Muffins	
## 53	
Braised Chicken With Artichokes and Olives	
## 54	
Pancetta Roast Chicken with Walnut Stuffing	
## 55	
1977 Coconut Angel Food Cake	
## 56	Collard-and-Prosc
Butter Chicken Roulades Over Watercress Salad	
## 57	
Veal Burgers Stuffed with Mozzarella Cheese	
## 58	
Pumpkin Muffins	
## 59	
Orange Balsamic Glaze	
## 60	Roasted Egg
Plantain and Olive Spread with Pita Bread Chips	
## 61	
Pecan Blue Cheese Crackers	
## 62	Romaine, Grilled Avocado, and Smoked
Smoked Corn Salad with Chipotle-Caesar Dressing	
## 63	Southwest Corn
Green Bread Stuffing with Corn and Green Chilies	
## 64	Coli
Blackberry and Sorghum and Apple Sticky Pudding	
## 65	
Mixed Berry Pavlovas	
## 66	
Orange-Almond Cake with Chocolate Icing	
## 67	
Scarborough Fair Tofu Burger	
## 68	
Italian Vinaigrette	
## 69	White Choc
Chocolate Tartlets with Strawberries and Bananas	
## 70	
Tomato-Infused Bulgur Pilaf with Fresh Basil	
## 71	Roasted Bu

ternut Squash, Rosemary, and Garlic Lasagne	
## 72	
Grilled Roast Beef and Stilton Sandwich	
## 73	Pear-Ha
zelnut Cheesecakes with Pear-Raspberry Sauce	
## 74	
Nut Butter	
## 75	
Cheese Ravioli with Fresh Tomato Sauce	
## 76	
Banana Layer Cake with Cream Cheese Frosting	
## 77	S
outh American-Style JÃ-cama and Orange Salad	
## 78	
Roasted Acorn Squash and Chestnuts	
## 79	
Maple Pumpkin Pots de CrÃ-me	
## 80	
Anadama Rolls with Mixed Seeds	
## 81	Braised Chicken and Rice with
Orange, Saffron, Almond, and Pistachio Syrup	
## 82	
Horseradish Dill Potato Salad	
## 83	
Chicken in Green Pumpkin-Seed Sauce	
## 84	
Jeweled Rice	
## 85	
Braised Brisket with Bourbon-Peach Glaze	
## 86	Gr
illed Pork Chops with Classic Barbecue Sauce	
## 87	
Bacon Crackers	
## 88	
Roast Chicken With Sorghum and Squash	
## 89	
Asparagus with Bacon and Onion	
## 90	
Ricotta Omelets	
## 91	Carrot, Snow Pea,
and Red Pepper Julienne in Honey Vinaigrette	
## 92	
Salmon with Chili-Mango Salsa	
## 93	
Turkey and Pinto Bean Chili	
## 94	
Cucumber-Yogurt Salad with Mint	
## 95	
Lamb Shanks Braised with Anise and Orange	
## 96	

Parsley Mayo	
## 97	
Acini di Pepe Pasta with Garlic and Olives	
## 98	R
oast Beef Salad with Cabbage and Horseradish	
## 99	
Savoy Cabbage and Arugula Salad	
## 100	
Fennel, Beet and Orange Salad with Olives	
## 101	
Shrimp Gazpacho	
## 102	
Grilled Steak Salad with Beets and Scallions	
## 103	
Parsnip and Apple Soup	
## 104	
Short Rib Pot Pie	
## 105	
Stout Floats	
## 106	
Apricot-Pistachio Muffins Baked on the Grill	
## 107	
Garlic Bruschetta	
## 108	
Asian Noodles with Barbecued Duck Confit	
## 109	
Sausage Fennel Stuffing	
## 110	Banan
a Split with Curried Chocolate-Coconut Sauce	
## 111	
Escarole and Cheese Spoon Bread	
## 112	
Honey-Ginger Barbecue Sauce	
## 113	Baked Pea
rs with Rosemary, Gorgonzola Cheese and Port	
## 114	
Kids' Matzoh Pizza	
## 115	
Cranberry, Quince, and Pearl Onion Compote	
## 116	
Chocolate-Mint Shamrock Shake	
## 117	
Tropical Rum Punch	
## 118	Chickpea S
alad Sandwich With Creamy Carrot-Radish Slaw	
## 119	
Blackberry-Raspberry Sauce	
## 120	
Laddie's Sub-Bourbon	
## 121	

Red Cabbage and Onions

## 122

Roast Cod with Potatoes, Onions, and Olives

## 123

Spicy Tomato Sauce

## 124

Cod Ca

nnelloni with Swiss Chard and Roasted Pepper

## 125

Swiss Chard with Roasted Pepper

## 126

Chocolate Almond Butter

## 127

Pastry Dough

## 128

Roasted Bell P

epper Halves Stuffed with Bulgur and Spinach

## 129

Spicy Sesame

Noodles with Chopped Peanuts and Thai Basil

## 130

Potato Gratin with Goat Cheese and Garlic

## 131

Country Sausage and Sage Dressing

## 132

Cherry Lime Virgin Rickeys

## 133

Buttermilk-Spinach Spaetzle

## 134

Radishes with Burrata

## 135

Winter Squash SoufflÃ©

## 136

Blueberry Streusel Cake

## 137

Low-Fat Chicken Stock

## 138

Honey Mustard Sauce

## 139

Rosemary and Lemon Pinto Beans

## 140

Asian Dipping Sauce

## 141

Shrimp and Green Onion Pancakes

## 142

Gnocchi with Tomato, Basil, and Olives

## 143

Mustard-Ginger Shrimp Canapes

## 144

Rumbrosia

## 145

Roasted Root Vegetables

## 146

# Thai Vegetables

## 147

Sage-Roasted T

urkey with Caramelized Onions and Sage Gravy

##	rating	calories	protein	fat	sodium	X.cakeweek	X.wasteless
## 1	2.500	426	30	7	559	0	0
## 2	4.375	403	18	23	1439	0	0
## 3	3.750	165	6	7	165	0	0
## 4	5.000	NA	NA	NA	NA	0	0
## 5	3.125	547	20	32	452	0	0
## 6	4.375	948	19	79	1042	0	0
## 7	4.375	NA	NA	NA	NA	0	0
## 8	3.750	NA	NA	NA	NA	0	0
## 9	4.375	170	7	10	1272	0	0
## 10	3.750	602	23	41	1696	0	0
## 11	3.750	256	4	5	30	0	0
## 12	4.375	NA	NA	NA	NA	0	0
## 13	4.375	766	12	48	439	0	0
## 14	4.375	174	11	12	176	0	0
## 15	3.125	134	4	3	1394	0	0
## 16	4.375	382	5	31	977	0	0
## 17	1.875	146	4	5	160	0	0
## 18	4.375	890	59	68	1027	0	0
## 19	5.000	107	5	7	344	0	0
## 20	5.000	421	10	33	383	0	0
## 21	4.375	345	11	19	423	0	0
## 22	3.750	NA	NA	NA	NA	0	0
## 23	3.750	NA	NA	NA	NA	0	0
## 24	3.750	NA	NA	NA	NA	0	0
## 25	3.750	279	3	30	206	0	0
## 26	0.000	95	1	7	103	0	0
## 27	3.750	215	6	20	250	0	0
## 28	2.500	14	0	0	0	0	0
## 29	4.375	351	6	19	79	0	0
## 30	4.375	311	5	5	226	0	0
## 31	4.375	NA	NA	NA	NA	0	0
## 32	5.000	NA	NA	NA	NA	0	0
## 33	4.375	376	7	18	604	0	0
## 34	3.125	185	10	13	765	0	0
## 35	5.000	NA	NA	NA	NA	0	0
## 36	3.750	625	39	44	1248	0	0
## 37	0.000	NA	NA	NA	NA	0	0
## 38	4.375	107	4	10	329	0	0
## 39	4.375	336	44	16	413	0	0
## 40	2.500	145	3	6	208	0	0
## 41	4.375	483	5	35	100	0	0
## 42	4.375	634	44	31	181	0	0
## 43	0.000	NA	NA	NA	NA	0	0
## 44	3.750	NA	NA	NA	NA	0	0
## 45	4.375	NA	NA	NA	NA	0	0
## 46	4.375	90	2	0	881	0	0

## 47	3.750	202	19	8	815	0	0
##							
## 101	0	0	0	0	0	0	0
## 102	0	0	0	0	1	0	0
## 103	0	0	0	0	0	0	0
## 104	0	0	0	0	1	0	0
## 105	0	1	0	0	0	0	0
## 106	0	0	0	0	0	0	0
## 107	0	0	0	0	0	0	0
## 108	0	0	0	0	0	0	0
## 109	0	0	0	0	0	0	0
## 110	0	1	0	0	0	0	0
## 111	0	0	0	0	0	0	0
## 112	0	0	0	0	0	0	0
## 113	0	1	0	0	0	0	0
## 114	0	0	0	0	0	0	0
## 115	0	0	0	0	0	0	0
## 116	0	0	0	0	0	0	0
## 117	0	0	0	0	0	0	0
## 118	0	0	0	0	0	0	0
## 119	0	1	0	0	0	0	0
## 120	0	0	0	0	0	0	0
## 121	0	0	0	0	0	0	0
## 122	0	0	0	0	0	0	0
## 123	0	0	0	0	0	0	0
## 124	0	0	0	0	0	0	0
## 125	0	0	0	0	0	0	0
## 126	0	0	0	0	0	0	0
## 127	0	0	0	0	0	0	0
## 128	0	0	0	0	0	0	0
## 129	0	0	0	0	1	0	0
## 130	0	0	0	0	0	0	0
## 131	0	0	0	0	0	0	0
## 132	0	0	0	0	0	0	0
## 133	0	0	0	0	0	0	0
## 134	0	0	0	0	0	0	0
## 135	0	0	0	0	0	0	0
## 136	0	1	0	0	0	0	0
## 137	0	0	0	0	0	0	0
## 138	0	0	0	0	0	0	0
## 139	0	0	0	0	0	0	0
## 140	0	0	0	0	0	0	0
## 141	0	0	0	0	1	0	0
## 142	0	0	0	0	1	0	0
## 143	0	0	0	0	0	0	0
## 144	0	0	0	0	0	0	0
## 145	0	0	0	0	0	0	0
## 146	0	0	0	0	0	0	0
## 147	0	0	0	0	1	0	0
##	dorie.greenspan	double.boiler	dried.fruit	drink	drinks	duck	easter

## 1	0	0	0	0	0	0	0
## 2	0	0	1	0	0	0	0
## 3	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	1
## 8	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0
## 12	0	0	0	0	0	0	0
## 13	0	0	0	0	0	0	0
## 14	0	0	0	0	0	0	0
## 15	0	0	0	0	0	0	0
## 16	0	0	0	0	0	0	0
## 17	0	0	0	0	0	0	0
## 18	0	0	0	0	0	0	0
## 19	0	0	0	0	0	0	0
## 20	0	0	0	0	0	0	0
## 21	0	0	0	0	0	0	0
## 116	0	0					
## 117	0	0					
## 118	0	0					
## 119	0	0					
## 120	0	0					
## 121	0	0					
## 122	0	0					
## 123	0	0					
## 124	0	0					
## 125	0	0					
## 126	0	0					
## 127	0	0					
## 128	0	0					
## 129	0	0					
## 130	0	0					
## 131	0	0					
## 132	0	0					
## 133	0	0					
## 134	0	0					
## 135	0	0					
## 136	0	0					
## 137	0	0					
## 138	0	0					
## 139	0	0					
## 140	0	0					
## 141	0	0					
## 142	0	0					
## 143	0	0					
## 144	0	0					



```
## 145          0      0
## 146          0      0
## 147          0      1
## [ reached getOption("max.print") -- omitted 19905 rows ]
```

```
head(data, n=10)
```

```
##                                     title rating
## 1                               Lentil, Apple, and Turkey Wrap 2.500
## 2                   Boudin Blanc Terrine with Red Onion Confit 4.375
## 3                   Potato and Fennel Soup Hodge              3.750
## 4                   Mahi-Mahi in Tomato Olive Sauce           5.000
## 5                   Spinach Noodle Casserole                  3.125
## 6                               The Best Blts                 4.375
## 7   Ham and Spring Vegetable Salad with Shallot Vinaigrette 4.375
## 8                               Spicy-Sweet Kumquats          3.750
## 9                               Korean Marinated Beef         4.375
## 10  Ham Persillade with Mustard Potato Salad and Mashed Peas 3.750
##      calories protein fat sodium X.cakeweek X.wasteless X22.minute.meals
## 1      426      30   7   559          0          0          0
## 2      403      18  23  1439          0          0          0
## 3      165       6   7   165          0          0          0
## 4       NA      NA  NA    NA          0          0          0
## 5      547      20  32   452          0          0          0
## 6      948      19  79  1042          0          0          0
## 7       NA      NA  NA    NA          0          0          0
## 8       NA      NA  NA    NA          0          0          0
## 9      170       7  10   1272         0          0          0
## 10     602      23  41   1696         0          0          0
##      X3.ingredient.recipes X30.days.of.groceries advance.prep.required

## [12541] 4.375 4.375 3.125 4.375 3.750 5.000 4.375 4.375 4.375 3.750 4.375
## [12552] 0.000 4.375 5.000 3.750 4.375 4.375 0.000 4.375 3.125 3.750 5.000
## [12563] 3.750 3.750 4.375 4.375 5.000 2.500 3.125 5.000 3.750 3.125 4.375
## [12574] 4.375 3.750 5.000 5.000 3.750 2.500 4.375 5.000 4.375 3.125 4.375
## [12585] 3.750 4.375 5.000 4.375 4.375 4.375 2.500 3.750 3.125 4.375 5.000
## [12596] 3.125 3.750 0.000 4.375 3.750 4.375 0.000 4.375 3.750 0.000 5.000
## [12607] 4.375 4.375 1.250 4.375 0.000 3.750 3.750 4.375 0.000 0.000 0.000
## [12618] 4.375 0.000 4.375 3.750 4.375 4.375 3.750 5.000 5.000 4.375 3.750
## [12629] 3.750 3.750 3.750 4.375 3.750 0.000 3.750 3.750 0.000 4.375 4.375
## [12640] 3.750 1.875 3.750 4.375 5.000 3.750 3.750 3.125 4.375 2.500 4.375
## [12651] 3.750 3.750 4.375 3.125 3.750 4.375 3.750 4.375 2.500 4.375 4.375
## [12662] 4.375 4.375 3.750 3.750 5.000 3.750 4.375 2.500 0.000 0.000 4.375
## [12673] 3.125 5.000 5.000 4.375 5.000 4.375 4.375 4.375 3.750 0.000 0.000
## [12684] 4.375 0.000 4.375 4.375 3.750 3.125 4.375 4.375 4.375 3.750 4.375
## [12695] 4.375 4.375 4.375 5.000 4.375 0.000 4.375 0.000 4.375 5.000 4.375
## [12706] 4.375 5.000 4.375 5.000 3.750 4.375 5.000 3.750 4.375 4.375 5.000
## [12717] 3.750 1.250 4.375 4.375 5.000 4.375 3.750 4.375 3.750 3.750 5.000
```

## [12728] 4.375 3.125 5.000 5.000 5.000 1.875 3.750 3.750 3.750 0.000 4.375  
## [12739] 4.375 4.375 3.750 3.750 3.125 1.250 4.375 4.375 3.750 4.375 3.750  
## [12750] 3.750 4.375 3.750 3.750 4.375 4.375 4.375 2.500 3.125 4.375 4.375  
## [12761] 5.000 4.375 4.375 5.000 0.000 5.000 4.375 3.750 4.375 4.375 3.125  
## [12772] 3.750 3.125 3.750 3.750 4.375 0.000 5.000 5.000 3.750 4.375 1.250  
## [12783] 4.375 4.375 3.750 3.750 4.375 4.375 3.750 4.375 3.125 4.375 5.000  
## [12794] 5.000 3.125 3.750 3.750 3.750 5.000 4.375 5.000 5.000 5.000 3.750  
## [12805] 2.500 5.000 0.000 3.125 3.750 4.375 4.375 4.375 5.000 4.375 4.375  
## [12816] 3.750 5.000 4.375 0.000 5.000 3.750 5.000 4.375 4.375 1.250 4.375  
## [12827] 4.375 4.375 5.000 4.375 3.750 4.375 4.375 5.000 4.375 5.000 3.750  
## [12838] 2.500 3.750 4.375 0.000 5.000 4.375 4.375 3.750 3.750 4.375 3.750  
## [12849] 4.375 4.375 4.375 3.750 0.000 0.000 4.375 3.750 3.750 3.750 5.000  
## [12860] 3.750 4.375 4.375 0.000 4.375 0.000 4.375 0.000 4.375 0.000 3.125  
## [12871] 4.375 5.000 4.375 3.750 5.000 4.375 3.125 3.750 2.500 4.375 5.000  
## [12882] 3.750 3.750 5.000 0.000 1.250 3.750 4.375 3.750 4.375 3.750 5.000  
## [12893] 3.750 4.375 3.125 3.750 4.375 3.750 4.375 4.375 4.375 3.750 0.000  
## [12904] 4.375 3.750 4.375 4.375 4.375 3.750 5.000 3.750 4.375 3.125 3.125  
## [12915] 0.000 5.000 4.375 0.000 3.750 4.375 3.750 4.375 4.375 0.000 3.750  
## [12926] 3.125 3.125 3.750 4.375 5.000 0.000 4.375 3.750 0.000 5.000 4.375  
## [12937] 2.500 4.375 3.750 4.375 4.375 4.375 0.000 4.375 3.750 4.375 3.750  
## [12948] 4.375 3.750 5.000 4.375 4.375 3.125 4.375 4.375 0.000 4.375 3.750  
## [12959] 0.000 3.750 4.375 3.125 3.750 0.000 4.375 5.000 3.125 5.000 3.750  
## [12970] 3.125 3.125 5.000 4.375 4.375 0.000 3.125 0.000 3.750 3.750 4.375  
## [12981] 5.000 4.375 3.750 4.375 5.000 4.375 3.125 3.750 5.000 3.750 0.000  
## [12992] 0.000 5.000 5.000 4.375 4.375 4.375 0.000 5.000 3.750 0.000 3.750  
## [13003] 0.000 0.000 4.375 4.375 4.375 4.375 3.750 4.375 0.000 5.000 4.375  
## [13014] 4.375 4.375 3.750 4.375 4.375 0.000 4.375 4.375 0.000 4.375 3.750  
## [13025] 4.375 3.125 0.000 4.375 3.750 3.750 0.000 5.000 3.750 0.000 5.000  
## [13036] 5.000 4.375 3.125 3.125 4.375 3.125 4.375 5.000 4.375 3.750 3.750  
## [13047] 5.000 0.000 3.750 3.750 5.000 4.375 4.375 3.750 3.750 4.375 0.000  
## [13058] 3.750 3.750 3.125 5.000 3.125 4.375 4.375 3.750 3.750 3.750 3.750  
## [13069] 4.375 3.750 4.375 3.125 0.000 5.000 0.000 4.375 4.375 0.000 4.375  
## [13080] 0.000 3.750 2.500 4.375 4.375 4.375 0.000 5.000 3.125 2.500 4.375  
## [13091] 3.125 3.750 3.750 0.000 4.375 3.750 0.000 5.000 3.750 3.750 4.375  
## [13102] 4.375 0.000 3.125 3.125 5.000 3.750 4.375 4.375 3.750 4.375 3.750  
## [13113] 3.750 3.750 3.750 4.375 4.375 4.375 4.375 5.000 3.750 3.750 3.750  
## [13124] 3.750 0.000 3.750 4.375 0.000 3.750 4.375 4.375 4.375 3.750 4.375  
## [13135] 0.000 4.375 0.000 3.750 5.000 4.375 3.125 4.375 0.000 4.375 4.375  
## [13146] 4.375 4.375 4.375 4.375 4.375 3.750 5.000 4.375 3.750 3.750 3.750  
## [13157] 5.000 5.000 0.000 3.750 4.375 5.000 4.375 4.375 3.125 4.375 3.125  
## [13168] 4.375 5.000 3.750 5.000 5.000 4.375 5.000 5.000 3.125 4.375 4.375  
## [13179] 2.500 0.000 2.500 4.375 4.375 4.375 4.375 4.375 4.375 4.375 4.375  
## [13190] 3.750 3.750 2.500 4.375 4.375 3.750 3.125 4.375 0.000 4.375 1.875  
## [13201] 4.375 4.375 4.375 5.000 3.750 4.375 4.375 4.375 3.125 5.000 3.750  
## [13212] 4.375 3.750 0.000 4.375 4.375 5.000 0.000 0.000 4.375 5.000 0.000  
## [13223] 4.375 0.000 4.375 3.125 0.000 3.750 3.750 4.375 2.500 4.375 4.375  
## [13234] 4.375 4.375 3.750 3.750 3.750 3.125 5.000 4.375 4.375 0.000 0.000  
## [13245] 3.750 3.125 4.375 4.375 4.375 3.750 3.125 3.125 0.000 4.375 4.375  
## [13256] 4.375 0.000 3.125 3.750 0.000 5.000 3.750 4.375 3.750 4.375 1.250  
## [13267] 0.000 5.000 4.375 4.375 3.750 4.375 2.500 3.125 3.750 4.375 4.375

```

## [13278] 5.000 3.750 5.000 4.375 1.875 4.375 5.000 4.375 3.125 3.750 0.000
## [13289] 3.750 4.375 4.375 3.125 5.000 4.375 4.375 3.125 3.125 4.375 4.375
## [19900] 4.375 4.375 3.125 4.375 3.750 4.375 3.750 4.375 5.000 4.375 3.750
## [19911] 4.375 4.375 0.000 4.375 3.125 3.750 4.375 5.000 3.750 4.375 4.375
## [19922] 4.375 4.375 4.375 3.125 4.375 3.750 3.125 4.375 4.375 4.375 0.000
## [19933] 0.000 3.750 3.750 3.750 3.125 5.000 0.000 4.375 4.375 1.250 0.000
## [19944] 0.000 5.000 4.375 4.375 3.750 3.125 3.750 3.750 3.750 3.750 3.750
## [19955] 4.375 4.375 5.000 0.000 0.000 4.375 0.000 3.750 4.375 4.375 3.750
## [19966] 3.750 3.125 4.375 0.000 3.750 3.750 3.125 4.375 4.375 4.375 0.000
## [19977] 5.000 4.375 4.375 3.750 5.000 4.375 3.750 4.375 4.375 3.750 3.750
## [19988] 3.750 0.000 4.375 5.000 5.000 0.000 4.375 2.500 2.500 3.750 4.375
## [19999] 0.000 4.375 0.000 3.750 5.000 5.000 3.750 3.750 4.375 4.375 3.125
## [20010] 4.375 5.000 0.000 3.750 5.000 4.375 3.125 4.375 4.375 5.000 4.375
## [20021] 3.750 3.750 3.750 5.000 4.375 5.000 4.375 3.750 5.000 0.000 3.125
## [20032] 3.125 4.375 2.500 2.500 5.000 3.750 3.750 3.750 3.125 4.375 4.375
## [20043] 4.375 4.375
##
## `$1`
## [1] 3.750 3.750 4.375 4.375 3.750 4.375 4.375 5.000

# Loading arules package
require(arules)

## Loading required package: arules

## Loading required package: Matrix

##
## Attaching package: 'arules'

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

require(arulesViz)

## Loading required package: arulesViz

## Loading required package: grid

# Convert data to transaction level
dt2 = as(dt,"transactions")

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

dt2

## transactions in sparse format with
## 2 transactions (rows) and
## 8 items (columns)

summary(dt2)

```

```

## transactions as itemMatrix in sparse format with
## 2 rows (elements/itemsets/transactions) and
## 8 columns (items) and a density of 0.6875
##
## most frequent items:
##      3.75  4.375      5      0      1.25 (Other)
##        2      2      2      1      1      3
##
## element (itemset/transaction) length distribution:
## sizes
## 3 8
## 1 1
##
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      3.00   4.25   5.50   5.50   6.75   8.00
##
## includes extended item information - examples:
## labels
## 1      0
## 2   1.25
## 3   1.875
##
## includes extended transaction information - examples:
## transactionID
## 1      0
## 2      1

inspect(dt2)

##      items                                transactionID
## [1] {0,1.25,1.875,2.5,3.125,3.75,4.375,5} 0
## [2] {3.75,4.375,5}                        1

# Most Frequent Items
itemFrequency(dt2, type = "relative")

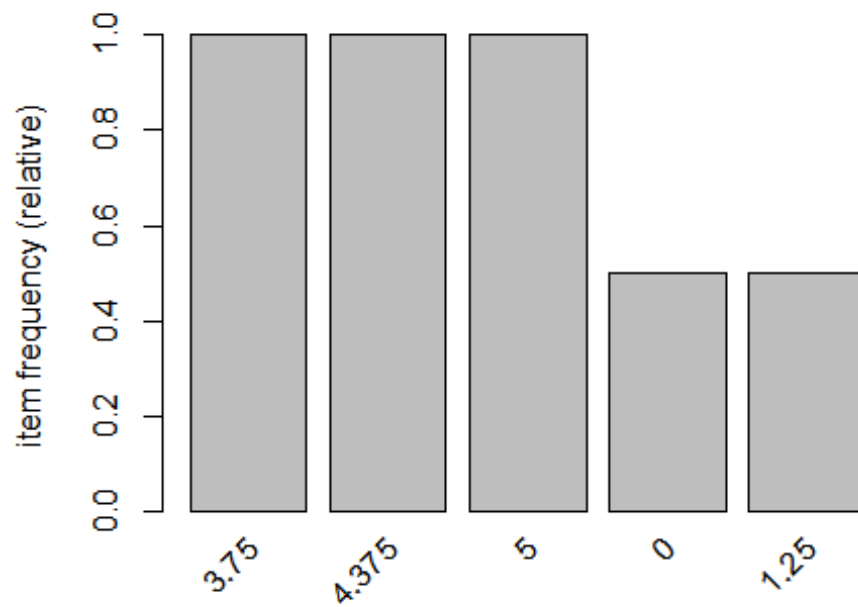
##      0  1.25 1.875   2.5 3.125   3.75 4.375      5
##    0.5  0.5  0.5   0.5  0.5   1.0  1.0   1.0

```

```

itemFrequencyPlot(dt2,topN = 5)

```

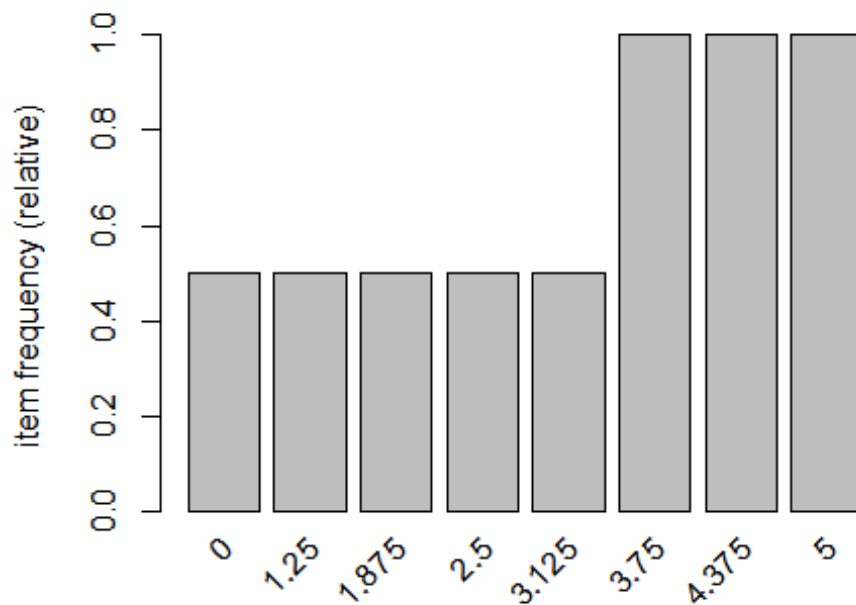


*# with support parameters*

```
itemFrequency(dt2, type = "relative")
```

```
##      0  1.25 1.875   2.5 3.125  3.75 4.375    5
##    0.5  0.5  0.5   0.5  0.5   1.0  1.0   1.0
```

```
itemFrequencyPlot(dt2, support= 0.10)
```



*# aggregated data*

```
rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8))
```

```
## Apriori
```

```
##
```

```
## Parameter specification:
```

```
## confidence minval smax arem aval originalSupport maxtime support minlen
##          0.8    0.1    1 none FALSE                TRUE         5   0.005    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: 0
```

```
##
```

```
## set item appearances ...[0 item(s)] done [0.00s].
```

```
## set transactions ...[8 item(s), 2 transaction(s)] done [0.00s].
```

```
## sorting and recoding items ... [8 item(s)] done [0.00s].
```

```
## creating transaction tree ... done [0.00s].
```

```
## checking subsets of size 1 2 3 4 5 6 7 8 done [0.00s].
```

```
## writing ... [984 rule(s)] done [0.00s].
```

```
## creating S4 object ... done [0.00s].
```

```
rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8, minlen = 3
))
```

```

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##          0.8    0.1    1 none FALSE                TRUE         5   0.005      3
## 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: 0
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[8 item(s), 2 transaction(s)] done [0.00s].
## sorting and recoding items ... [8 item(s)] done [0.00s].
## creating transaction tree ... done [0.00s].
## checking subsets of size 1 2 3 4 5 6 7 8 done [0.00s].
## writing ... [940 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].

rules = apriori(dt2, parameter=list(support=0.005, confidence=0.8, maxlen = 4
))

## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##          0.8    0.1    1 none FALSE                TRUE         5   0.005      1
## maxlen target  ext
##           4  rules FALSE
##
## Algorithmic control:
## filter tree heap memopt load sort verbose
##    0.1 TRUE TRUE  FALSE TRUE     2     TRUE
##
## Absolute minimum support count: 0
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[8 item(s), 2 transaction(s)] done [0.00s].
## sorting and recoding items ... [8 item(s)] done [0.00s].
## creating transaction tree ... done [0.00s].
## checking subsets of size 1 2 3 4

## Warning in apriori(dt2, parameter = list(support = 0.005, confidence =
## 0.8, : Mining stopped (maxlen reached). Only patterns up to a length of 4
## returned!

```

```

## done [0.00s].
## writing ... [472 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].

rules

## set of 472 rules

summary(rules)

## set of 472 rules
##
## rule length distribution (lhs + rhs):sizes
##   1   2   3   4
##   3  41 153 275
##
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   1.000   3.000   4.000   3.483   4.000   4.000
##
## summary of quality measures:
##      support      confidence      lift      count
##   Min.   :0.5000   Min.   :1   Min.   :1.000   Min.   :1.000
##   1st Qu.:0.5000   1st Qu.:1   1st Qu.:1.000   1st Qu.:1.000
##   Median :0.5000   Median :1   Median :2.000   Median :1.000
##   Mean   :0.5127   Mean   :1   Mean   :1.593   Mean   :1.025
##   3rd Qu.:0.5000   3rd Qu.:1   3rd Qu.:2.000   3rd Qu.:1.000
##   Max.   :1.0000   Max.   :1   Max.   :2.000   Max.   :2.000
##
## mining info:
## data ntransactions support confidence
##   dt2                2   0.005        0.8

inspect(rules[1:10]) # to view first 10 rules

##      lhs      rhs      support confidence lift count
## [1] {}      => {3.75} 1.0      1          1      2
## [2] {}      => {4.375} 1.0      1          1      2
## [3] {}      => {5}     1.0      1          1      2
## [4] {0}     => {1.25} 0.5      1          2      1
## [5] {1.25} => {0}     0.5      1          2      1
## [6] {0}     => {1.875} 0.5      1          2      1
## [7] {1.875} => {0}     0.5      1          2      1
## [8] {0}     => {2.5}   0.5      1          2      1
## [9] {2.5}   => {0}     0.5      1          2      1
## [10] {0}    => {3.125} 0.5      1          2      1

#Convert rules into data frame
rules3 = as(rules, "data.frame")
write(rules, "C:/Users/Seshan/Desktop/PCA//rules2.csv", sep=",")

```



```
# Show only particular product rules
```

```
inspect( subset( rules, subset = rhs %pin% "0" )[1:10])
```

##	lhs	rhs	support	confidence	lift	count
## [1]	{1.25}	=> {0}	0.5	1	2	1
## [2]	{1.875}	=> {0}	0.5	1	2	1
## [3]	{2.5}	=> {0}	0.5	1	2	1
## [4]	{3.125}	=> {0}	0.5	1	2	1
## [5]	{1.25,1.875}	=> {0}	0.5	1	2	1
## [6]	{1.25,2.5}	=> {0}	0.5	1	2	1
## [7]	{1.25,3.125}	=> {0}	0.5	1	2	1
## [8]	{1.25,3.75}	=> {0}	0.5	1	2	1
## [9]	{1.25,4.375}	=> {0}	0.5	1	2	1
## [10]	{1.25,5}	=> {0}	0.5	1	2	1

```
# Show the top 10 rules
```

```
options(digits=2)
```

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

##	lhs	rhs	support	confidence	lift	count
## [1]	{}	=> {3.75}	1.0	1	1	2
## [2]	{}	=> {4.375}	1.0	1	1	2
## [3]	{}	=> {5}	1.0	1	1	2
## [4]	{0}	=> {1.25}	0.5	1	2	1
## [5]	{1.25}	=> {0}	0.5	1	2	1
## [6]	{0}	=> {1.875}	0.5	1	2	1
## [7]	{1.875}	=> {0}	0.5	1	2	1
## [8]	{0}	=> {2.5}	0.5	1	2	1
## [9]	{2.5}	=> {0}	0.5	1	2	1
## [10]	{0}	=> {3.125}	0.5	1	2	1

```
# Get Summary Information
```

```
summary(rules)
```

```
## set of 472 rules
```

```
##
```

```
## rule length distribution (lhs + rhs):sizes
```

```
## 1 2 3 4
```

```
## 3 41 153 275
```

```
##
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

```
## 1.0 3.0 4.0 3.5 4.0 4.0
```

```
##
```

```
## summary of quality measures:
```

```
## support confidence lift count
```

```
## Min. :0.50 Min. :1 Min. :1.00 Min. :1.00
```

```
## 1st Qu.:0.50 1st Qu.:1 1st Qu.:1.00 1st Qu.:1.00
```

```
## Median :0.50 Median :1 Median :2.00 Median :1.00
```

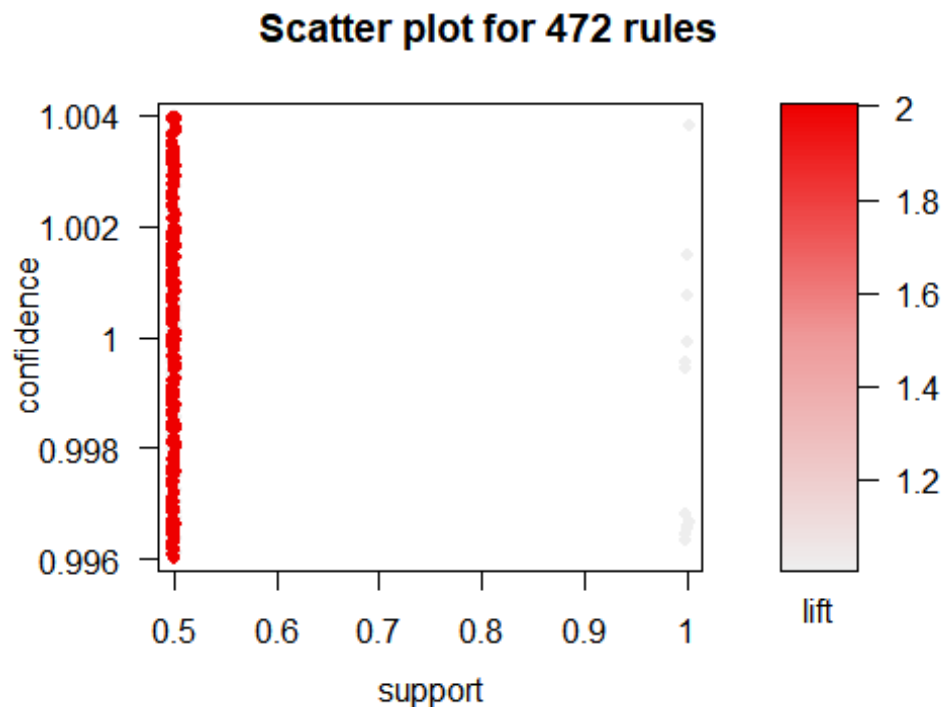
```
## Mean :0.51 Mean :1 Mean :1.59 Mean :1.03
```

```
## 3rd Qu.:0.50 3rd Qu.:1 3rd Qu.:2.00 3rd Qu.:1.00
```

```
## Max. :1.00 Max. :1 Max. :2.00 Max. :2.00
##
## mining info:
## data ntransactions support confidence
## dt2 2 0.005 0.8

plot(rules)

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



```
plot(rules, method = "graph", interactive = T)

## Warning in plot.rules(rules, method = "graph", interactive = T): The
## parameter interactive is deprecated. Use engine='interactive' instead.

## Warning: plot: Too many rules supplied. Only plotting the best 100 rules
## using 'support' (change control parameter max if needed)

# Sort by Lift
rules<-sort(rules, by="lift", decreasing=TRUE)

# Sort by Lift
rules<-sort(rules, by="lift", decreasing=TRUE)

# Remove Unnecessary Rules
subset.matrix <- is.subset(rules, rules)
subset.matrix[lower.tri(subset.matrix, diag=T)] <- NA
```

```
## Warning in `[<-`(`*tmp*`, as.vector(i), value = NA): x[.] <- val: x is
## "ngTMatrix", val not in {TRUE, FALSE} is coerced; NA |--> TRUE.
```

```
redundant <- colSums(subset.matrix, na.rm=T) >= 1
which(redundant)
```

```
##          {0,1.25}          {0,1.25}          {0,1.875}
##              1              2              3
##          {0,1.875}          {0,2.5}          {0,2.5}
##              4              5              6
##          {0,3.125}          {0,3.125}          {1.25,1.875}
##              7              8              9
##          {1.25,1.875}          {1.25,2.5}          {1.25,2.5}
##             10             11             12
##          {1.25,3.125}          {1.25,3.125}          {1.875,2.5}
##             13             14             15
##          {1.875,2.5}          {1.875,3.125}          {1.875,3.125}
##             16             17             18
##          {2.5,3.125}          {2.5,3.125}          {0,1.25,1.875}
##             19             20             21
##          {0,1.25,1.875}          {0,1.25,1.875}          {0,1.25,2.5}
##             22             23             24
##          {0,1.25,2.5}          {0,1.25,2.5}          {0,1.25,3.125}
##             25             26             27
##          {0,1.25,3.125}          {0,1.25,3.125}          {0,1.25,3.75}
##             28             29             30
##          {0,1.25,3.75}          {0,1.25,4.375}          {0,1.25,4.375}
##             31             32             33
##          {0,1.25,5}          {0,1.25,5}          {0,1.875,2.5}
##             34             35             36
##          {0,1.875,2.5}          {0,1.875,2.5}          {0,1.875,3.125}
##             37             38             39
##          {0,1.875,3.125}          {0,1.875,3.125}          {0,1.875,3.75}
##             40             41             42
##          {0,1.875,3.75}          {0,1.875,4.375}          {0,1.875,4.375}
##             43             44             45
##          {0,1.875,5}          {0,1.875,5}          {0,2.5,3.125}
##             46             47             48
##          {0,2.5,3.125}          {0,2.5,3.125}          {0,2.5,3.75}
##             49             50             51
##          {0,2.5,3.75}          {0,2.5,4.375}          {0,2.5,4.375}
##             52             53             54
##          {0,2.5,5}          {0,2.5,5}          {0,3.125,3.75}
##             55             56             57
##          {0,3.125,3.75}          {0,3.125,4.375}          {0,3.125,4.375}
##             58             59             60
##          {0,3.125,5}          {0,3.125,5}          {1.25,1.875,2.5}
##             61             62             63
##          {1.25,1.875,2.5}          {1.25,1.875,2.5}          {1.25,1.875,3.125}
##             64             65             66
```

##	{1.25,1.875,3.125}	{1.25,1.875,3.125}	{1.25,1.875,3.75}
##	67	68	69
##	{1.25,1.875,3.75}	{1.25,1.875,4.375}	{1.25,1.875,4.375}
##	70	71	72
##	{1.25,1.875,5}	{1.25,1.875,5}	{1.25,2.5,3.125}
##	73	74	75
##	{1.25,2.5,3.125}	{1.25,2.5,3.125}	{1.25,2.5,3.75}
##	76	77	78
##	{1.25,2.5,3.75}	{1.25,2.5,4.375}	{1.25,2.5,4.375}
##	79	80	81
##	{1.25,2.5,5}	{1.25,2.5,5}	{1.25,3.125,3.75}
##	82	83	84
##	{1.25,3.125,3.75}	{1.25,3.125,4.375}	{1.25,3.125,4.375}
##	85	86	87
##	{1.25,3.125,5}	{1.25,3.125,5}	{1.875,2.5,3.125}
##	88	89	90
##	{1.875,2.5,3.125}	{1.875,2.5,3.125}	{1.875,2.5,3.75}
##	91	92	93
##	{1.875,2.5,3.75}	{1.875,2.5,4.375}	{1.875,2.5,4.375}
##	94	95	96
##	{1.875,2.5,5}	{1.875,2.5,5}	{1.875,3.125,3.75}
##	97	98	99
##	{1.875,3.125,3.75}	{1.875,3.125,4.375}	{1.875,3.125,4.375}
##	100	101	102
##	{1.875,3.125,5}	{1.875,3.125,5}	{2.5,3.125,3.75}
##	103	104	105
##	{2.5,3.125,3.75}	{2.5,3.125,4.375}	{2.5,3.125,4.375}
##	106	107	108
##	{2.5,3.125,5}	{2.5,3.125,5}	{0,1.25,1.875,2.5}
##	109	110	111
##	{0,1.25,1.875,2.5}	{0,1.25,1.875,2.5}	{0,1.25,1.875,2.5}
##	112	113	114
##	{0,1.25,1.875,3.125}	{0,1.25,1.875,3.125}	{0,1.25,1.875,3.125}
##	115	116	117
##	{0,1.25,1.875,3.125}	{0,1.25,1.875,3.75}	{0,1.25,1.875,3.75}
##	118	119	120
##	{0,1.25,1.875,3.75}	{0,1.25,1.875,4.375}	{0,1.25,1.875,4.375}
##	121	122	123
##	{0,1.25,1.875,4.375}	{0,1.25,1.875,5}	{0,1.25,1.875,5}
##	124	125	126
##	{0,1.25,1.875,5}	{0,1.25,2.5,3.125}	{0,1.25,2.5,3.125}
##	127	128	129
##	{0,1.25,2.5,3.125}	{0,1.25,2.5,3.125}	{0,1.25,2.5,3.75}
##	130	131	132
##	{0,1.25,2.5,3.75}	{0,1.25,2.5,3.75}	{0,1.25,2.5,4.375}
##	133	134	135
##	{0,1.25,2.5,4.375}	{0,1.25,2.5,4.375}	{0,1.25,2.5,5}
##	136	137	138
##	{0,1.25,2.5,5}	{0,1.25,2.5,5}	{0,1.25,3.125,3.75}
##	139	140	141

##	{0,1.25,3.125,3.75}	{0,1.25,3.125,3.75}	{0,1.25,3.125,4.375}
##	142	143	144
##	{0,1.25,3.125,4.375}	{0,1.25,3.125,4.375}	{0,1.25,3.125,5}
##	145	146	147
##	{0,1.25,3.125,5}	{0,1.25,3.125,5}	{0,1.25,3.75,4.375}
##	148	149	150
##	{0,1.25,3.75,4.375}	{0,1.25,3.75,5}	{0,1.25,3.75,5}
##	151	152	153
##	{0,1.25,4.375,5}	{0,1.25,4.375,5}	{0,1.875,2.5,3.125}
##	154	155	156
##	{0,1.875,2.5,3.125}	{0,1.875,2.5,3.125}	{0,1.875,2.5,3.125}
##	157	158	159
##	{0,1.875,2.5,3.75}	{0,1.875,2.5,3.75}	{0,1.875,2.5,3.75}
##	160	161	162
##	{0,1.875,2.5,4.375}	{0,1.875,2.5,4.375}	{0,1.875,2.5,4.375}
##	163	164	165
##	{0,1.875,2.5,5}	{0,1.875,2.5,5}	{0,1.875,2.5,5}
##	166	167	168
##	{0,1.875,3.125,3.75}	{0,1.875,3.125,3.75}	{0,1.875,3.125,3.75}
##	169	170	171
##	{0,1.875,3.125,4.375}	{0,1.875,3.125,4.375}	{0,1.875,3.125,4.375}
##	172	173	174
##	{0,1.875,3.125,5}	{0,1.875,3.125,5}	{0,1.875,3.125,5}
##	175	176	177
##	{0,1.875,3.75,4.375}	{0,1.875,3.75,4.375}	{0,1.875,3.75,5}
##	178	179	180
##	{0,1.875,3.75,5}	{0,1.875,4.375,5}	{0,1.875,4.375,5}
##	181	182	183
##	{0,2.5,3.125,3.75}	{0,2.5,3.125,3.75}	{0,2.5,3.125,3.75}
##	184	185	186
##	{0,2.5,3.125,4.375}	{0,2.5,3.125,4.375}	{0,2.5,3.125,4.375}
##	187	188	189
##	{0,2.5,3.125,5}	{0,2.5,3.125,5}	{0,2.5,3.125,5}
##	190	191	192
##	{0,2.5,3.75,4.375}	{0,2.5,3.75,4.375}	{0,2.5,3.75,5}
##	193	194	195
##	{0,2.5,3.75,5}	{0,2.5,4.375,5}	{0,2.5,4.375,5}
##	196	197	198
##	{0,3.125,3.75,4.375}	{0,3.125,3.75,4.375}	{0,3.125,3.75,5}
##	199	200	201
##	{0,3.125,3.75,5}	{0,3.125,4.375,5}	{0,3.125,4.375,5}
##	202	203	204
##	{1.25,1.875,2.5,3.125}	{1.25,1.875,2.5,3.125}	{1.25,1.875,2.5,3.125}
##	205	206	207
##	{1.25,1.875,2.5,3.125}	{1.25,1.875,2.5,3.75}	{1.25,1.875,2.5,3.75}
##	208	209	210
##	{1.25,1.875,2.5,3.75}	{1.25,1.875,2.5,4.375}	{1.25,1.875,2.5,4.375}
##	211	212	213
##	{1.25,1.875,2.5,4.375}	{1.25,1.875,2.5,5}	{1.25,1.875,2.5,5}
##	214	215	216

##	{1.25,1.875,2.5,5}	{1.25,1.875,3.125,3.75}	{1.25,1.875,3.125,3.75}
##	217	218	219
##	{1.25,1.875,3.125,3.75}	{1.25,1.875,3.125,4.375}	{1.25,1.875,3.125,4.375}
##	220	221	222
##	{1.25,1.875,3.125,4.375}	{1.25,1.875,3.125,5}	{1.25,1.875,3.125,5}
##	223	224	225
##	{1.25,1.875,3.125,5}	{1.25,1.875,3.75,4.375}	{1.25,1.875,3.75,4.375}
##	226	227	228
##	{1.25,1.875,3.75,5}	{1.25,1.875,3.75,5}	{1.25,1.875,4.375,5}
##	229	230	231
##	{1.25,1.875,4.375,5}	{1.25,2.5,3.125,3.75}	{1.25,2.5,3.125,3.75}
##	232	233	234
##	{1.25,2.5,3.125,3.75}	{1.25,2.5,3.125,4.375}	{1.25,2.5,3.125,4.375}
##	235	236	237
##	{1.25,2.5,3.125,4.375}	{1.25,2.5,3.125,5}	{1.25,2.5,3.125,5}
##	238	239	240
##	{1.25,2.5,3.125,5}	{1.25,2.5,3.75,4.375}	{1.25,2.5,3.75,4.375}
##	241	242	243
##	{1.25,2.5,3.75,5}	{1.25,2.5,3.75,5}	{1.25,2.5,4.375,5}
##	244	245	246
##	{1.25,2.5,4.375,5}	{1.25,3.125,3.75,4.375}	{1.25,3.125,3.75,4.375}
##	247	248	249
##	{1.25,3.125,3.75,5}	{1.25,3.125,3.75,5}	{1.25,3.125,4.375,5}
##	250	251	252
##	{1.25,3.125,4.375,5}	{1.875,2.5,3.125,3.75}	{1.875,2.5,3.125,3.75}
##	253	254	255
##	{1.875,2.5,3.125,3.75}	{1.875,2.5,3.125,4.375}	{1.875,2.5,3.125,4.375}
##	256	257	258
##	{1.875,2.5,3.125,4.375}	{1.875,2.5,3.125,5}	{1.875,2.5,3.125,5}
##	259	260	261
##	{1.875,2.5,3.125,5}	{1.875,2.5,3.75,4.375}	{1.875,2.5,3.75,4.375}
##	262	263	264
##	{1.875,2.5,3.75,5}	{1.875,2.5,3.75,5}	{1.875,2.5,4.375,5}
##	265	266	267
##	{1.875,2.5,4.375,5}	{1.875,3.125,3.75,4.375}	{1.875,3.125,3.75,4.375}
##	268	269	270
##	{1.875,3.125,3.75,5}	{1.875,3.125,3.75,5}	{1.875,3.125,4.375,5}
##	271	272	273
##	{1.875,3.125,4.375,5}	{2.5,3.125,3.75,4.375}	{2.5,3.125,3.75,4.375}
##	274	275	276
##	{2.5,3.125,3.75,5}	{2.5,3.125,3.75,5}	{2.5,3.125,4.375,5}
##	277	278	279
##	{2.5,3.125,4.375,5}	{3.75}	{4.375}
##	280	281	282
##	{5}	{0,3.75}	{0,4.375}
##	283	284	285
##	{0,5}	{1.25,3.75}	{1.25,4.375}
##	286	287	288
##	{1.25,5}	{1.875,3.75}	{1.875,4.375}
##	289	290	291

##	{1.875,5}	{2.5,3.75}	{2.5,4.375}
##	292	293	294
##	{2.5,5}	{3.125,3.75}	{3.125,4.375}
##	295	296	297
##	{3.125,5}	{3.75,4.375}	{3.75,4.375}
##	298	299	300
##	{3.75,5}	{3.75,5}	{4.375,5}
##	301	302	303
##	{4.375,5}	{0,1.25,3.75}	{0,1.25,4.375}
##	304	305	306
##	{0,1.25,5}	{0,1.875,3.75}	{0,1.875,4.375}
##	307	308	309
##	{0,1.875,5}	{0,2.5,3.75}	{0,2.5,4.375}
##	310	311	312
##	{0,2.5,5}	{0,3.125,3.75}	{0,3.125,4.375}
##	313	314	315
##	{0,3.125,5}	{0,3.75,4.375}	{0,3.75,4.375}
##	316	317	318
##	{0,3.75,5}	{0,3.75,5}	{0,4.375,5}
##	319	320	321
##	{0,4.375,5}	{1.25,1.875,3.75}	{1.25,1.875,4.375}
##	322	323	324
##	{1.25,1.875,5}	{1.25,2.5,3.75}	{1.25,2.5,4.375}
##	325	326	327
##	{1.25,2.5,5}	{1.25,3.125,3.75}	{1.25,3.125,4.375}
##	328	329	330
##	{1.25,3.125,5}	{1.25,3.75,4.375}	{1.25,3.75,4.375}
##	331	332	333
##	{1.25,3.75,5}	{1.25,3.75,5}	{1.25,4.375,5}
##	334	335	336
##	{1.25,4.375,5}	{1.875,2.5,3.75}	{1.875,2.5,4.375}
##	337	338	339
##	{1.875,2.5,5}	{1.875,3.125,3.75}	{1.875,3.125,4.375}
##	340	341	342
##	{1.875,3.125,5}	{1.875,3.75,4.375}	{1.875,3.75,4.375}
##	343	344	345
##	{1.875,3.75,5}	{1.875,3.75,5}	{1.875,4.375,5}
##	346	347	348
##	{1.875,4.375,5}	{2.5,3.125,3.75}	{2.5,3.125,4.375}
##	349	350	351
##	{2.5,3.125,5}	{2.5,3.75,4.375}	{2.5,3.75,4.375}
##	352	353	354
##	{2.5,3.75,5}	{2.5,3.75,5}	{2.5,4.375,5}
##	355	356	357
##	{2.5,4.375,5}	{3.125,3.75,4.375}	{3.125,3.75,4.375}
##	358	359	360
##	{3.125,3.75,5}	{3.125,3.75,5}	{3.125,4.375,5}
##	361	362	363
##	{3.125,4.375,5}	{3.75,4.375,5}	{3.75,4.375,5}
##	364	365	366

##	{3.75,4.375,5}	{0,1.25,1.875,3.75}	{0,1.25,1.875,4.375}
##	367	368	369
##	{0,1.25,1.875,5}	{0,1.25,2.5,3.75}	{0,1.25,2.5,4.375}
##	370	371	372
##	{0,1.25,2.5,5}	{0,1.25,3.125,3.75}	{0,1.25,3.125,4.375}
##	373	374	375
##	{0,1.25,3.125,5}	{0,1.25,3.75,4.375}	{0,1.25,3.75,4.375}
##	376	377	378
##	{0,1.25,3.75,5}	{0,1.25,3.75,5}	{0,1.25,4.375,5}
##	379	380	381
##	{0,1.25,4.375,5}	{0,1.875,2.5,3.75}	{0,1.875,2.5,4.375}
##	382	383	384
##	{0,1.875,2.5,5}	{0,1.875,3.125,3.75}	{0,1.875,3.125,4.375}
##	385	386	387
##	{0,1.875,3.125,5}	{0,1.875,3.75,4.375}	{0,1.875,3.75,4.375}
##	388	389	390
##	{0,1.875,3.75,5}	{0,1.875,3.75,5}	{0,1.875,4.375,5}
##	391	392	393
##	{0,1.875,4.375,5}	{0,2.5,3.125,3.75}	{0,2.5,3.125,4.375}
##	394	395	396
##	{0,2.5,3.125,5}	{0,2.5,3.75,4.375}	{0,2.5,3.75,4.375}
##	397	398	399
##	{0,2.5,3.75,5}	{0,2.5,3.75,5}	{0,2.5,4.375,5}
##	400	401	402
##	{0,2.5,4.375,5}	{0,3.125,3.75,4.375}	{0,3.125,3.75,4.375}
##	403	404	405
##	{0,3.125,3.75,5}	{0,3.125,3.75,5}	{0,3.125,4.375,5}
##	406	407	408
##	{0,3.125,4.375,5}	{0,3.75,4.375,5}	{0,3.75,4.375,5}
##	409	410	411
##	{0,3.75,4.375,5}	{1.25,1.875,2.5,3.75}	{1.25,1.875,2.5,4.375}
##	412	413	414
##	{1.25,1.875,2.5,5}	{1.25,1.875,3.125,3.75}	{1.25,1.875,3.125,4.375}
##	415	416	417
##	{1.25,1.875,3.125,5}	{1.25,1.875,3.75,4.375}	{1.25,1.875,3.75,4.375}
##	418	419	420
##	{1.25,1.875,3.75,5}	{1.25,1.875,3.75,5}	{1.25,1.875,4.375,5}
##	421	422	423
##	{1.25,1.875,4.375,5}	{1.25,2.5,3.125,3.75}	{1.25,2.5,3.125,4.375}
##	424	425	426
##	{1.25,2.5,3.125,5}	{1.25,2.5,3.75,4.375}	{1.25,2.5,3.75,4.375}
##	427	428	429
##	{1.25,2.5,3.75,5}	{1.25,2.5,3.75,5}	{1.25,2.5,4.375,5}
##	430	431	432
##	{1.25,2.5,4.375,5}	{1.25,3.125,3.75,4.375}	{1.25,3.125,3.75,4.375}
##	433	434	435
##	{1.25,3.125,3.75,5}	{1.25,3.125,3.75,5}	{1.25,3.125,4.375,5}
##	436	437	438
##	{1.25,3.125,4.375,5}	{1.25,3.75,4.375,5}	{1.25,3.75,4.375,5}
##	439	440	441



```

##      {1.25,3.75,4.375,5}   {1.875,2.5,3.125,3.75} {1.875,2.5,3.125,4.375}
##                                442                                443                                444
##      {1.875,2.5,3.125,5}   {1.875,2.5,3.75,4.375} {1.875,2.5,3.75,4.375}
##                                445                                446                                447
##      {1.875,2.5,3.75,5}           {1.875,2.5,3.75,5}           {1.875,2.5,4.375,5}
##                                448                                449                                450
##      {1.875,2.5,4.375,5} {1.875,3.125,3.75,4.375} {1.875,3.125,3.75,4.375}
##                                451                                452                                453
##      {1.875,3.125,3.75,5}           {1.875,3.125,3.75,5}           {1.875,3.125,4.375,5}
##                                454                                455                                456
##      {1.875,3.125,4.375,5}           {1.875,3.75,4.375,5}           {1.875,3.75,4.375,5}
##                                457                                458                                459
##      {1.875,3.75,4.375,5}           {2.5,3.125,3.75,4.375}           {2.5,3.125,3.75,4.375}
##                                460                                461                                462
##      {2.5,3.125,3.75,5}           {2.5,3.125,3.75,5}           {2.5,3.125,4.375,5}
##                                463                                464                                465
##      {2.5,3.125,4.375,5}           {2.5,3.75,4.375,5}           {2.5,3.75,4.375,5}
##                                466                                467                                468
##      {2.5,3.75,4.375,5}           {3.125,3.75,4.375,5}           {3.125,3.75,4.375,5}
##                                469                                470                                471
##      {3.125,3.75,4.375,5}
##                                472

```

```

rules.pruned <- rules[!redundant]
rules<-rules.pruned
rules

```

```

## set of 0 rules

```

#### *#Clean Rules*

```

rules3$rules=gsub("\\{", "", rules3$rules)
rules3$rules=gsub("\\}", "", rules3$rules)
rules3$rules=gsub("\\", "", rules3$rules)

```

#### *#Split the rule*

```

library(splitstackshape)
Rules4=cSplit(rules3, "rules", ">")
names(Rules4)[names(Rules4) == 'rules_1'] <- 'LHS'
Rules5=cSplit(Rules4, "LHS", ",")
Rules6=subset(Rules5, select= -c(rules_2))
names(Rules6)[names(Rules6) == 'rules_3'] <- 'RHS'

```

#### *# What are customers likely to buy before they purchase "Product A"*

```

rules<-apriori(data=dt, parameter=list(supp=0.001,conf = 0.8),
               appearance = list(default="lhs",rhs="0"),
               control = list(verbose=F))

```

```

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

```

```

rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])

```

```

##      lhs      rhs support confidence lift count
## [1] {1.25}    => {0} 0.5      1          2      1
## [2] {1.875}   => {0} 0.5      1          2      1
## [3] {2.5}     => {0} 0.5      1          2      1
## [4] {3.125}   => {0} 0.5      1          2      1
## [5] {1.25,1.875} => {0} 0.5      1          2      1

# What are customers likely to buy if they purchased "Product A"
rules<-apriori(data=dt, parameter=list(supp=0.001,conf = 0.8),
               appearance = list(default="rhs",lhs="0"),
               control = list(verbose=F))

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

rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])

##      lhs      rhs      support confidence lift count
## [1] {} => {3.75} 1.0          1          1      2
## [2] {} => {4.375} 1.0          1          1      2
## [3] {} => {5} 1.0            1          1      2
## [4] {0} => {1.25} 0.5          1          2      1
## [5] {0} => {1.875} 0.5          1          2      1

rules

## set of 10 rules

support<-seq(0.01,0.1,0.01)
support

## [1] 0.01 0.02 0.03 0.04 0.05 0.06 0.07 0.08 0.09 0.10

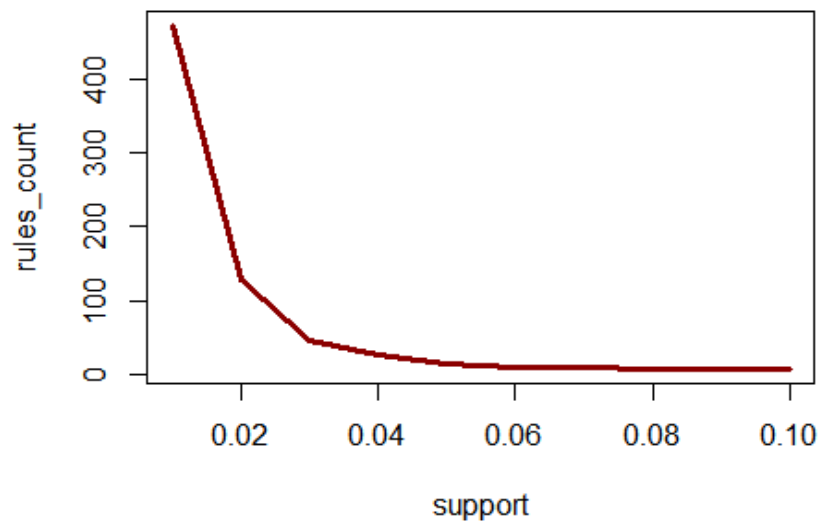
rules_count<-c(472,128,46,26,14, 10, 10,8,8,8)
rules_count

## [1] 472 128 46 26 14 10 10 8 8 8

plot(support,rules_count,type = "l",main="Number of rules at different support %",
     col="darkred",lwd=3)

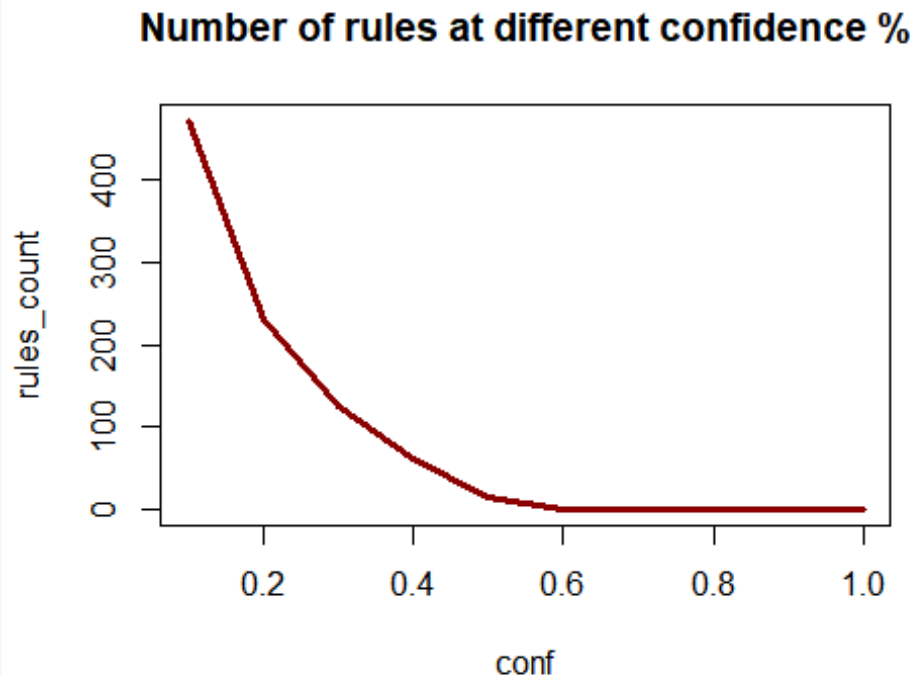
```

**Number of rules at different support %**



```
conf<-seq(0.10,1.0,0.10)
conf
## [1] 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1.0
rules_count<-c(472,231,125,62,15,0,0,0,0,0)
rules_count
## [1] 472 231 125 62 15 0 0 0 0 0
plot(conf,rules_count,type = "l",main="Number of rules at different confidence %",
```

```
col="darkred",lwd=3)
```



```
#rules_ec <- eclat(epi_r, parameter = list(supp = 0.05))
```

```
#summary(rules_ec)
```

```
#sorting out the most relevant rules
```

```
rules<-sort(rules, by="confidence", decreasing=TRUE)
```

```
inspect(rules[1:5])
```

```
##      lhs    rhs      support confidence lift count
## [1] {} => {3.75}  1.0         1          1      2
## [2] {} => {4.375} 1.0         1          1      2
## [3] {} => {5}      1.0         1          1      2
## [4] {0} => {1.25}  0.5         1          2      1
## [5] {0} => {1.875} 0.5         1          2      1
```

```
rules<-sort(rules, by="lift", decreasing=TRUE)
```

```
inspect(rules[1:5])
```

```
##      lhs    rhs      support confidence lift count
## [1] {0} => {1.25}  0.5         1          2      1
## [2] {0} => {1.875} 0.5         1          2      1
## [3] {0} => {2.5}    0.5         1          2      1
## [4] {0} => {3.125} 0.5         1          2      1
## [5] {}  => {3.75}  1.0         1          1      2
```

```
#####
```

```
library(factoextra)
```

```
## Loading required package: ggplot2
```

```
## Welcome! Related Books: `Practical Guide To Cluster Analysis in R` at https://goo.gl/13EFCZ
```

```
library("factoextra")
```

```
data1<-na.exclude(data)
```

```
na.omit(data1)
```

```
##
```

```
title
```

```
## 1
```

```
Lentil, Apple, and Turkey Wrap
```

```
## 2
```

```
Boudin Blanc Terrine with Red Onion Confit
```

```
## 3
```

```
Potato and Fennel Soup Hodge
```

```
## 5
```

```
Spinach Noodle Casserole
```

```
## 6
```

```
The Best Blts
```

```
## 9
```

```
Korean Marinated Beef
```

```
## 10
```

Ham Persilla

```
de with Mustard Potato Salad and Mashed Peas
```

```
## 11
```

```
Yams Braised with Cream, Rosemary and Nutmeg
```

```
## 13
```

Banana-Cho

```
colate Chip Cake With Peanut Butter Frosting
```

```
## 14
```

```
Beef Tenderloin with Garlic and Brandy
```

```
## 15
```

```
Peach Mustard
```

```
## 16
```

```
Raw Cream of Spinach Soup
```

```
## 17
```

```
Sweet Buttermilk Spoon Breads
```

```
## 18
```

```
Crisp Braised Pork Shoulder
```

```
## 19
```

Mozzar

```
ella-Topped Peppers with Tomatoes and Garlic
```

```
## 20
```

Tuna, Asparagus, and New Potato S

```
alad with Chive Vinaigrette and Fried Capers
```

```
## 21
```

Asian Pe

```
ar and Watercress Salad with Sesame Dressing
```

```
## 25
```

```
Sea Salt-Roasted Pecans
```

```
## 26
```

Garlic Baguette Crumbs	
## 27	
Cucumber-Basil Egg Salad	
## 28	
Dried Pear Crisps	
## 29	Green Bean, Red Onion, and
Roast Potato Salad with Rosemary Vinaigrette	
## 30	
Apricot-Cherry Shortcakes	
## 33	Roaste
d Sweet-Potato Spears with Bacon Vinaigrette	
## 34	
Deviled Ham	
## 36	
Aztec Chicken	
## 38	
Sauteed Broccoli Rabe	
## 39	
Grouper with Tomato and Basil	
## 40	
Better-Than-Pita Grill Bread	
## 41	
Coconut-Key Lime Sheet Cake	
## 42	Baked Hal
ibut with Orzo, Spinach, and Cherry Tomatoes	
## 46	
Pickled Red Onions	
## 47	
Spicy Black Beans and Rice	
## 49	
Mexican Lime Soup	
## 50	
Citrus Salad with Mint Sugar	
## 51	
Mexican Chile and Mushroom Soup	
## 52	
Peanut Butter-Banana Muffins	
## 54	
Pancetta Roast Chicken with Walnut Stuffing	
## 55	
1977 Coconut Angel Food Cake	
## 57	
Veal Burgers Stuffed with Mozzarella Cheese	
## 58	
Pumpkin Muffins	
## 59	
Orange Balsamic Glaze	
## 60	Roasted Egg
plant and Olive Spread with Pita Bread Chips	
## 61	

Pecan Blue Cheese Crackers	
## 62	Romaine, Grilled Avocado, and Smoky Corn Salad with Chipotle-Caesar Dressing
## 63	Southwest Corn Bread Stuffing with Corn and Green Chilies
## 64	Coli
## 65	on Perry's Sorghum and Apple Sticky Pudding
## 67	Mixed Berry Pavlovas
## 68	Scarborough Fair Tofu Burger
## 69	Italian Vinaigrette
## 70	White Choc
## 71	olate Tartlets with Strawberries and Bananas
## 72	Tomato-Infused Bulgur Pilaf with Fresh Basil
## 73	Roasted Bu
## 74	ternut Squash, Rosemary, and Garlic Lasagne
## 75	Grilled Roast Beef and Stilton Sandwich
## 76	Pear-Ha
## 77	zelnut Cheesecakes with Pear-Raspberry Sauce
## 78	Nut Butter
## 79	Cheese Ravioli with Fresh Tomato Sauce
## 80	Banana Layer Cake with Cream Cheese Frosting
## 81	S
## 82	outh American-Style JÃ-cama and Orange Salad
## 83	Roasted Acorn Squash and Chestnuts
## 84	Maple Pumpkin Pots de CrÃ-me
## 85	Braised Chicken and Rice with
## 86	Orange, Saffron, Almond, and Pistachio Syrup
## 87	Horseradish Dill Potato Salad
## 88	Chicken in Green Pumpkin-Seed Sauce
## 89	Jeweled Rice
## 90	Braised Brisket with Bourbon-Peach Glaze
## 91	Gr
## 92	illed Pork Chops with Classic Barbecue Sauce
## 93	Roast Chicken With Sorghum and Squash
## 94	

Asparagus with Bacon and Onion  
## 92  
Salmon with Chili-Mango Salsa  
## 93  
Turkey and Pinto Bean Chili  
## 94  
Cucumber-Yogurt Salad with Mint  
## 95  
Lamb Shanks Braised with Anise and Orange  
## 96  
Parsley Mayo  
## 97  
Acini di Pepe Pasta with Garlic and Olives  
## 98  
Roast Beef Salad with Cabbage and Horseradish  
## 99  
Savoy Cabbage and Arugula Salad  
## 100  
Fennel, Beet and Orange Salad with Olives  
## 101  
Shrimp Gazpacho  
## 103  
Parsnip and Apple Soup  
## 105  
Stout Floats  
## 106  
Apricot-Pistachio Muffins Baked on the Grill  
## 107  
Garlic Bruschetta  
## 108  
Asian Noodles with Barbecued Duck Confit  
## 110  
Milk Split with Curried Chocolate-Coconut Sauce  
## 111  
Escarole and Cheese Spoon Bread  
## 112  
Honey-Ginger Barbecue Sauce  
## 114  
Kids' Matzoh Pizza  
## 115  
Cranberry, Quince, and Pearl Onion Compote  
## 117  
Tropical Rum Punch  
## 118  
Salad Sandwich With Creamy Carrot-Radish Slaw  
## 119  
Blackberry-Raspberry Sauce  
## 120  
Laddie's Sub-Bourbon  
## 121

R

Banan

Chickpea S



Red Cabbage and Onions

## 122

Roast Cod with Potatoes, Onions, and Olives

## 123

Spicy Tomato Sauce

## 125

Swiss Chard with Roasted Pepper

## 126

Chocolate Almond Butter

## 127

Pastry Dough

## 129

Spicy Sesame

Noodles with Chopped Peanuts and Thai Basil

## 130

Potato Gratin with Goat Cheese and Garlic

## 131

Country Sausage and Sage Dressing

## 133

Buttermilk-Spinach Spaetzle

## 134

Radishes with Burrata

## 135

Winter Squash Souffl  

## 136

Blueberry Streusel Cake

## 137

Low-Fat Chicken Stock

## 138

Honey Mustard Sauce

## 139

Rosemary and Lemon Pinto Beans

## 140

Asian Dipping Sauce

## 141

Shrimp and Green Onion Pancakes

## 143

Mustard-Ginger Shrimp Canapes

## 144

Rumbrosia

## 146

Thai Vegetables

## 147

Sage-Roasted T

urkey with Caramelized Onions and Sage Gravy

## 148

Shrimp Cakes with Andouille Sausage

## 149

Creamy Tofu Salad

## 150

Chocola

te-Cherry Ice Cream Pie with Hot Fudge Sauce

## 151

Jalapeño-Cheddar Frittata	
## 152	
Roasted Beets and Citrus with Feta	
## 153	Gree
n Beans with Crisp Shallots, Chile, and Mint	
## 155	
Cranberry Pear Tart with Gingerbread Crust	
## 156	Sauteed Veal w
ith Shrimp, Mushroom, and Brandy Cream Sauce	
## 157	
Lemon Vinaigrette	
## 159	
Cranberry, Shallot, and Dried-Cherry Compote	
## 161	
Peanut Butter Cream Tart	
## 162	
Cheddar Chicken Tenders with Wilted Spinach	
## 163	
Blueberry Cheesecake	
## 165	
Apple Pie with Whisky-Soaked Cherries	
## 166	
Parsleyed Yellow-Potato Salad	
## 167	
Sauteed Fennel and Carrots	
## 168	Gr
illed Garlic-Marinated Skirt Steak with Lime	
## 169	
Miniature Crab Cakes with Tomato Ginger Jam	
## 170	
Egg Sandwich with Green Bean Slaw	
## 171	
Red Wine Brasato with Glazed Root Vegetables	
## 173	
Egg Salad with Lemon and Fennel	
## 175	
Shaved Brussels Sprout and Shallot Sauté	
## 177	
Roasted Carrot and Beet Salad with Feta	
## 178	
Cassata Cake	
## 179	
Baked Beans with Slab Bacon and Breadcrumbs	
## 180	
Grilled Corn with Lime-Cilantro Butter	
## 181	Roasted Winter Squash and Parsnip
s with Maple Syrup Glaze and Marcona Almonds	
## 185	Roasted
Bell Peppers with Basil and Balsamic Vinegar	
## 186	

Homemade Tomato Ketchup

## 188

Char-Grilled

Beef Tenderloin with Three-Herb Chimichurri

## 189

Pork Roast Braised

with Milk and Fresh Herbs (Maiale al Latte )

## 191

Chocolate Pecan Banana Tarts

##	rating	calories	protein	fat	sodium	X.cakeweek	X.wasteless
## 1	2.5	4.3e+02	30	7	5.6e+02	0	0
## 2	4.4	4.0e+02	18	23	1.4e+03	0	0
## 3	3.8	1.6e+02	6	7	1.6e+02	0	0
## 5	3.1	5.5e+02	20	32	4.5e+02	0	0
## 6	4.4	9.5e+02	19	79	1.0e+03	0	0
## 9	4.4	1.7e+02	7	10	1.3e+03	0	0
## 10	3.8	6.0e+02	23	41	1.7e+03	0	0
## 11	3.8	2.6e+02	4	5	3.0e+01	0	0
## 13	4.4	7.7e+02	12	48	4.4e+02	0	0
## 14	4.4	1.7e+02	11	12	1.8e+02	0	0
## 15	3.1	1.3e+02	4	3	1.4e+03	0	0
## 16	4.4	3.8e+02	5	31	9.8e+02	0	0
## 17	1.9	1.5e+02	4	5	1.6e+02	0	0
## 18	4.4	8.9e+02	59	68	1.0e+03	0	0
## 19	5.0	1.1e+02	5	7	3.4e+02	0	0
## 20	5.0	4.2e+02	10	33	3.8e+02	0	0
## 21	4.4	3.4e+02	11	19	4.2e+02	0	0
## 25	3.8	2.8e+02	3	30	2.1e+02	0	0
## 26	0.0	9.5e+01	1	7	1.0e+02	0	0
## 27	3.8	2.2e+02	6	20	2.5e+02	0	0
## 28	2.5	1.4e+01	0	0	0.0e+00	0	0
## 29	4.4	3.5e+02	6	19	7.9e+01	0	0
## 30	4.4	3.1e+02	5	5	2.3e+02	0	0
## 33	4.4	3.8e+02	7	18	6.0e+02	0	0
## 34	3.1	1.8e+02	10	13	7.6e+02	0	0
## 36	3.8	6.2e+02	39	44	1.2e+03	0	0
## 38	4.4	1.1e+02	4	10	3.3e+02	0	0
## 39	4.4	3.4e+02	44	16	4.1e+02	0	0
## 40	2.5	1.4e+02	3	6	2.1e+02	0	0
## 41	4.4	4.8e+02	5	35	1.0e+02	0	0
## 42	4.4	6.3e+02	44	31	1.8e+02	0	0
## 46	4.4	9.0e+01	2	0	8.8e+02	0	0
## 47	3.8	2.0e+02	19	8	8.2e+02	0	0
## 49	4.4	3.4e+02	14	21	1.7e+02	0	0
## 50	4.4	1.9e+02	3	1	4.0e+00	0	0
## 51	3.1	1.7e+02	8	12	5.1e+02	0	0
## 52	3.8	2.8e+02	6	13	2.4e+02	0	0
## 54	5.0	1.2e+03	89	87	5.8e+02	0	0
## 55	3.8	2.7e+02	4	7	1.5e+02	0	0
## 57	4.4	9.0e+02	38	70	1.4e+03	0	0
## 58	4.4	2.2e+02	4	10	2.1e+02	0	0
## 59	3.8	1.9e+02	2	3	7.0e+02	0	0

## 60	3.8	1.8e+02	5	7	1.2e+02	0	0
## 61	3.8	7.0e+01	2	6	6.0e+01	0	0
## 127			0	0	0	0	0
## 129			0	0	0	0	0
## 130			0	0	0	0	0
## 131			0	0	0	0	0
## 133			0	0	0	0	0
## 134			0	0	0	0	0
1	0	0	0	0	0	0	0
## 62	0	0	0	0	0	0	0
## 63	0	0	0	0	0	0	0
## 64	0	0	0	0	0	0	0
## 65	0	0	0	0	0	0	0
## 67	0	0	0	0	0	0	0
## 68	0	0	0	0	0	0	0
## 69	0	0	0	0	0	0	0
## 70	0	0	0	0	0	0	0
## 71	0	0	0	0	0	0	0
## 72	0	0	0	0	0	0	0
## 73	0	0	0	0	0	0	0
## 74	0	0	0	0	0	0	0
## 75	0	0	0	0	0	0	0
## 76	0	0	0	0	0	0	0
## 77	0	0	0	0	0	0	0
## 78	0	0	0	0	0	0	0
## 79	0	0	0	0	0	0	0
## 81	1	0	0	0	0	0	0
## 82	0	0	0	0	0	0	0
## 83	0	0	0	0	0	0	0
## 84	1	0	0	0	0	0	0
## 85	0	0	0	0	0	0	0
## 86	0	0	0	0	0	0	0
## 88	0	0	0	0	0	0	0
## 89	0	0	0	0	0	0	0
## 92	0	0	0	0	0	0	0
## 93	0	0	0	0	0	0	0
## 94	0	0	0	0	0	0	0
## 95	0	0	0	0	0	0	0
## 96	0	0	0	0	0	0	0
## 97	0	0	0	0	0	0	0
## 98	0	0	0	0	0	0	0
## 99	0	0	0	0	0	0	0
## 100	0	0	0	0	0	0	0
## 101	0	0	0	0	0	0	0
## 103	0	0	0	0	0	0	0
## 105	0	0	0	0	0	0	0
## 106	1	0	0	0	0	0	0
## 107	0	0	0	0	0	0	0
## 108	0	0	0	0	0	0	0

```
## 110      0      0      0      0      0      0      0
## 111      0      0      0      0      0      0      0
## 112      0      0      0      0      0      0      0
## 114      0      0      0      0      0      0      0
```

```
## 191      0      0
## [ reached getOption("max.print") -- omitted 15717 rows ]
```

```
data1.active <- data1[2:100, 2:6]
na.exclude(data1.active)
```

```
##      rating calories protein fat sodium
## 2      4.4      403      18  23   1439
## 3      3.8      165       6   7    165
## 5      3.1      547      20  32   452
## 6      4.4      948      19  79  1042
## 9      4.4      170       7  10  1272
## 10     3.8      602      23  41  1696
## 11     3.8      256       4   5    30
## 13     4.4      766      12  48   439
## 14     4.4      174      11  12   176
## 15     3.1      134       4   3  1394
## 16     4.4      382       5  31   977
## 17     1.9      146       4   5   160
## 18     4.4      890      59  68  1027
## 19     5.0      107       5   7   344
## 20     5.0      421      10  33   383
## 21     4.4      345      11  19   423
## 25     3.8      279       3  30   206
## 26     0.0       95       1   7   103
## 27     3.8      215       6  20   250
## 28     2.5       14       0   0     0
## 29     4.4      351       6  19    79
## 30     4.4      311       5   5   226
## 33     4.4      376       7  18   604
## 34     3.1      185      10  13   765
## 36     3.8      625      39  44  1248
## 38     4.4      107       4  10   329
## 39     4.4      336      44  16   413
## 40     2.5      145       3   6   208
## 41     4.4      483       5  35   100
## 42     4.4      634      44  31   181
## 46     4.4       90       2   0   881
## 47     3.8      202      19   8   815
## 49     4.4      338      14  21   174
## 50     4.4      191       3   1     4
## 51     3.1      166       8  12   508
## 52     3.8      275       6  13   242
## 54     5.0     1203      89  87   583
## 55     3.8      266       4   7   148
```

## 57	4.4	904	38	70	1413
## 58	4.4	223	4	10	211
## 59	3.8	194	2	3	697
## 60	3.8	177	5	7	116
## 61	3.8	70	2	6	60
## 62	4.4	368	6	32	112
## 63	5.0	293	7	15	565
## 64	0.0	523	8	19	694
## 65	3.8	252	4	7	89
## 67	3.1	224	21	12	340
## 68	3.8	185	0	20	155
## 69	4.4	830	9	59	148
## 70	0.0	195	7	5	469
## 71	4.4	684	21	42	637
## 72	4.4	641	39	37	907
## 73	3.8	538	8	36	231
## 74	3.8	264	7	24	84
## 75	3.8	365	19	14	599
## 76	4.4	926	9	56	569
## 77	3.1	26	1	0	364
## 78	3.8	230	2	7	9
## 79	4.4	157	3	8	51
## 81	4.4	1172	54	73	220
## 82	4.4	298	6	12	199
## 83	4.4	682	36	57	909
## 84	5.0	517	7	18	20
## 85	4.4	856	45	54	1797
## 86	4.4	599	48	28	1038
## 88	0.0	1143	63	77	311
## 89	4.4	129	4	11	146
## 92	4.4	571	36	37	106
## 93	4.4	508	45	17	826
## 94	3.8	62	2	3	603
## 95	3.1	1118	92	70	1226
## 96	0.0	306	0	34	302
## 97	4.4	209	5	8	77
## 98	3.1	645	22	52	324
## 99	3.1	126	3	9	46
## 100	4.4	133	2	8	91
## 101	4.4	285	19	7	635
## 103	4.4	310	3	25	89
## 105	3.1	248	3	10	73
## 106	5.0	247	5	10	185
## 107	3.8	201	4	11	210
## 108	5.0	519	14	25	1237
## 110	3.8	1076	11	73	150
## 111	4.4	338	15	18	515
## 112	4.4	298	1	0	1430
## 114	3.1	280	13	18	587
## 115	3.8	171	1	0	6

```
## 117    5.0      230      1  0      26
## 118    3.8      959     21 60     1541
## 119    4.4      129      1  0        2
## 120    2.5      107      0  0         0
## 121    3.8       86       3  1      339
## 122    3.8      679     55 36      333
## 123    0.0      112       2  7        12
## 125    3.8      123       5  7      605
## 126    0.0      273       2 28         5
## 127    5.0      234       3 16         99
## 129    4.4      724     21 28     1130
```

```
View(data1.active)
head(data1.active[, 2:5])
```

```
##      calories protein fat sodium
## 2          403      18  23   1439
## 3          165       6   7    165
## 5          547      20  32    452
## 6          948      19  79   1042
## 9          170       7  10   1272
## 10         602      23  41   1696
```

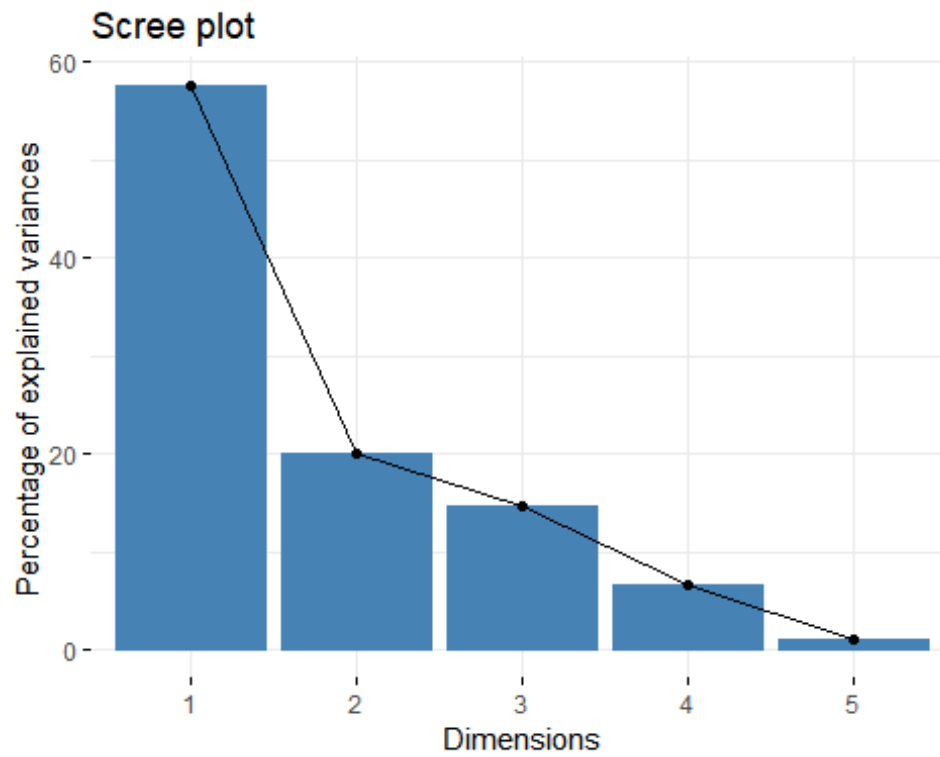
```
#Compute PCA in R using prcomp()
library(factoextra)
res.pca <- prcomp(data1.active, scale = TRUE)
res.pca
```

```
## Standard deviations (1, .., p=5):
## [1] 1.70 1.00 0.86 0.58 0.22
##
## Rotation (n x k) = (5 x 5):
##      PC1    PC2    PC3    PC4    PC5
## rating  -0.11  0.943  0.311  0.0031 -0.038
## calories -0.56 -0.105  0.208 -0.2732  0.746
## protein  -0.51 -0.085  0.052  0.8484 -0.099
## fat       -0.54 -0.169  0.241 -0.4344 -0.657
## sodium   -0.35  0.253 -0.894 -0.1299 -0.023
```

```
summary(res.pca)
```

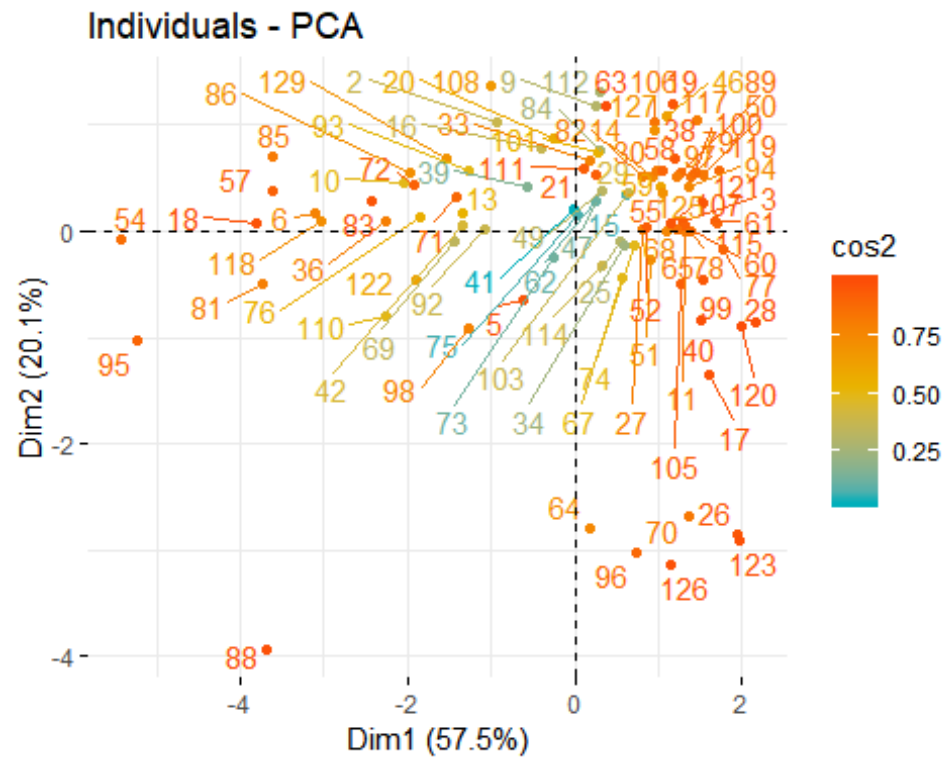
```
## Importance of components:
##      PC1    PC2    PC3    PC4    PC5
## Standard deviation    1.696 1.002 0.856 0.579 0.2243
## Proportion of Variance 0.575 0.201 0.147 0.067 0.0101
## Cumulative Proportion 0.575 0.776 0.923 0.990 1.0000
```

```
fviz_eig(res.pca)
```

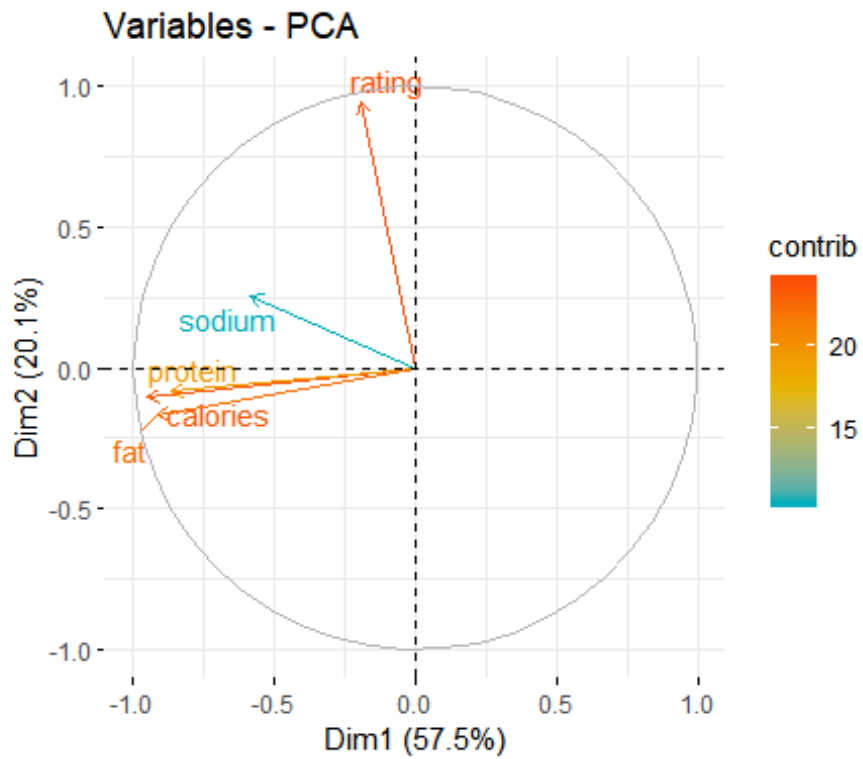


```
fviz_pca_ind(res.pca,  
  col.ind = "cos2", # Color by the quality of representation  
  gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),  
  repel = TRUE      # Avoid text overlapping  
)
```

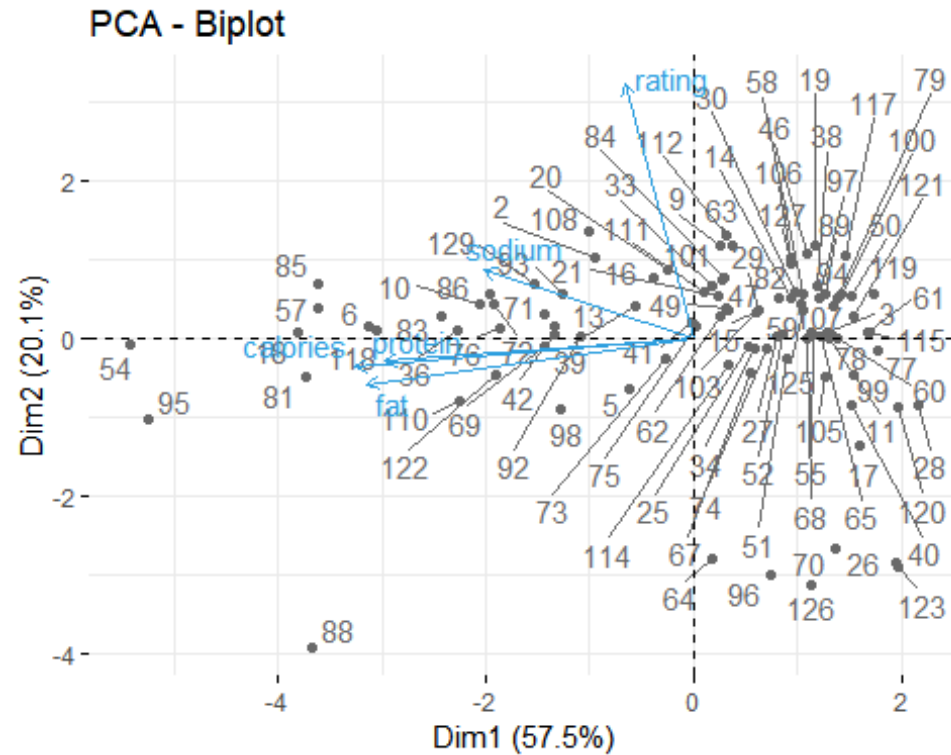




```
fviz_pca_var(res.pca,
  col.var = "contrib", # Color by contributions to the PC
  gradient.cols = c("#00AFBB", "#E7B800", "#FC4E07"),
  repel = TRUE        # Avoid text overlapping
)
```



```
fviz_pca_biplot(res.pca, repel = TRUE,  
  col.var = "#2E9FDF", # Variables color  
  col.ind = "#696969" # Individuals color  
)
```



```
library(factoextra)
# Eigenvalues
eig.val <- get_eigenvalue(res.pca)
eig.val

##          eigenvalue variance.percent cumulative.variance.percent
## Dim.1          2.88           57.5                58
## Dim.2          1.00           20.1                78
## Dim.3          0.73           14.7                92
## Dim.4          0.33            6.7                99
## Dim.5          0.05            1.0               100

# Results for Variables
res.var <- get_pca_var(res.pca)
res.var$coord          # Coordinates

##          Dim.1 Dim.2 Dim.3 Dim.4 Dim.5
## rating    -0.19  0.945  0.267  0.0018 -0.0084
## calories  -0.95 -0.105  0.178 -0.1581  0.1674
## protein   -0.87 -0.085  0.045  0.4911 -0.0223
## fat       -0.92 -0.169  0.207 -0.2514 -0.1473
## sodium    -0.59  0.254 -0.765 -0.0752 -0.0051

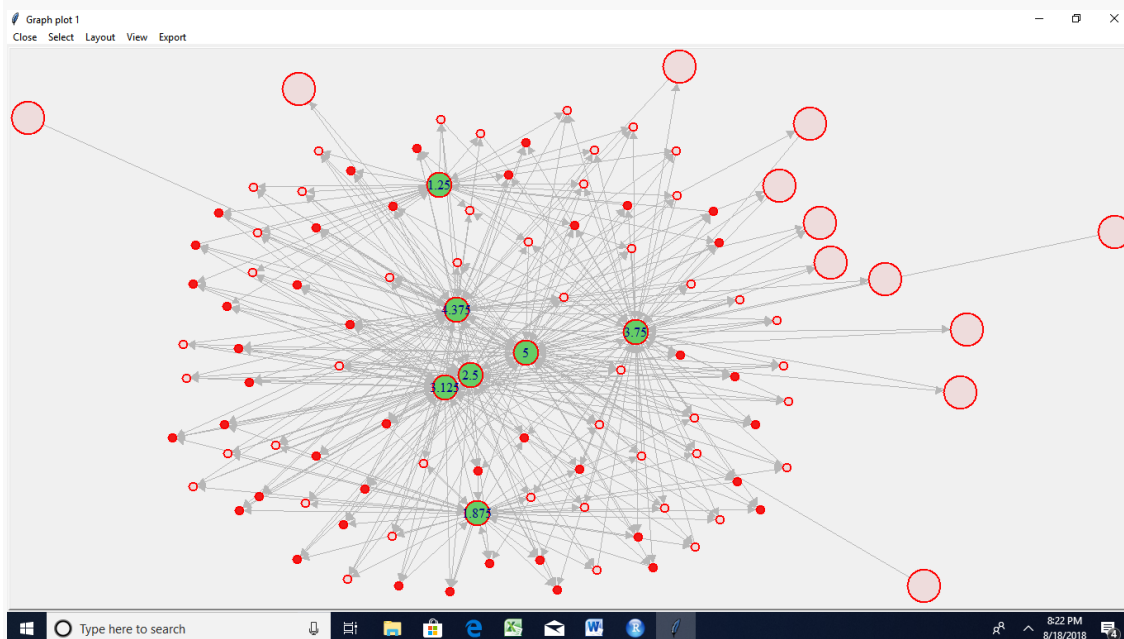
res.var$contrib          # Contributions to the PCs

##          Dim.1 Dim.2 Dim.3 Dim.4 Dim.5
## rating      1.3 88.89  9.70 9.9e-04 0.141
## calories   31.4  1.10  4.32 7.5e+00 55.692
```

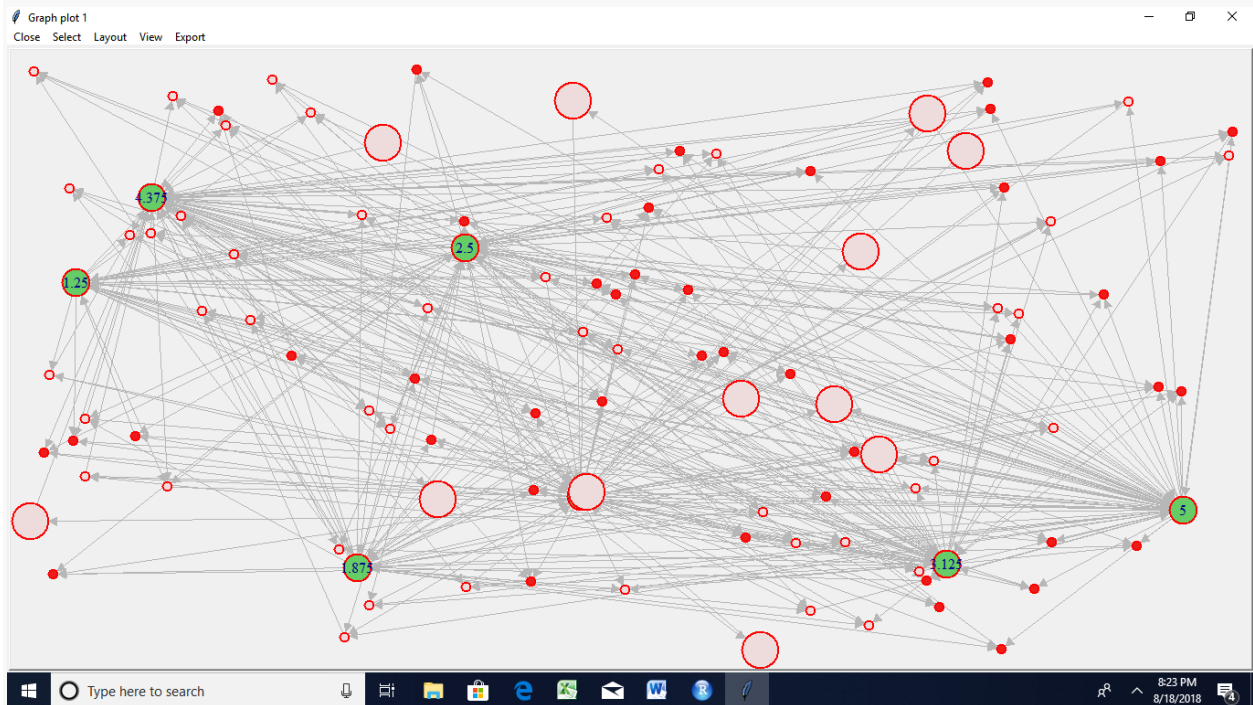
```
## protein    26.0  0.72  0.28 7.2e+01  0.986
## fat        29.3  2.86  5.82 1.9e+01 43.130
## sodium     12.0  6.42 79.89 1.7e+00  0.052
```

```
res.var$cos2          # Quality of representation
```

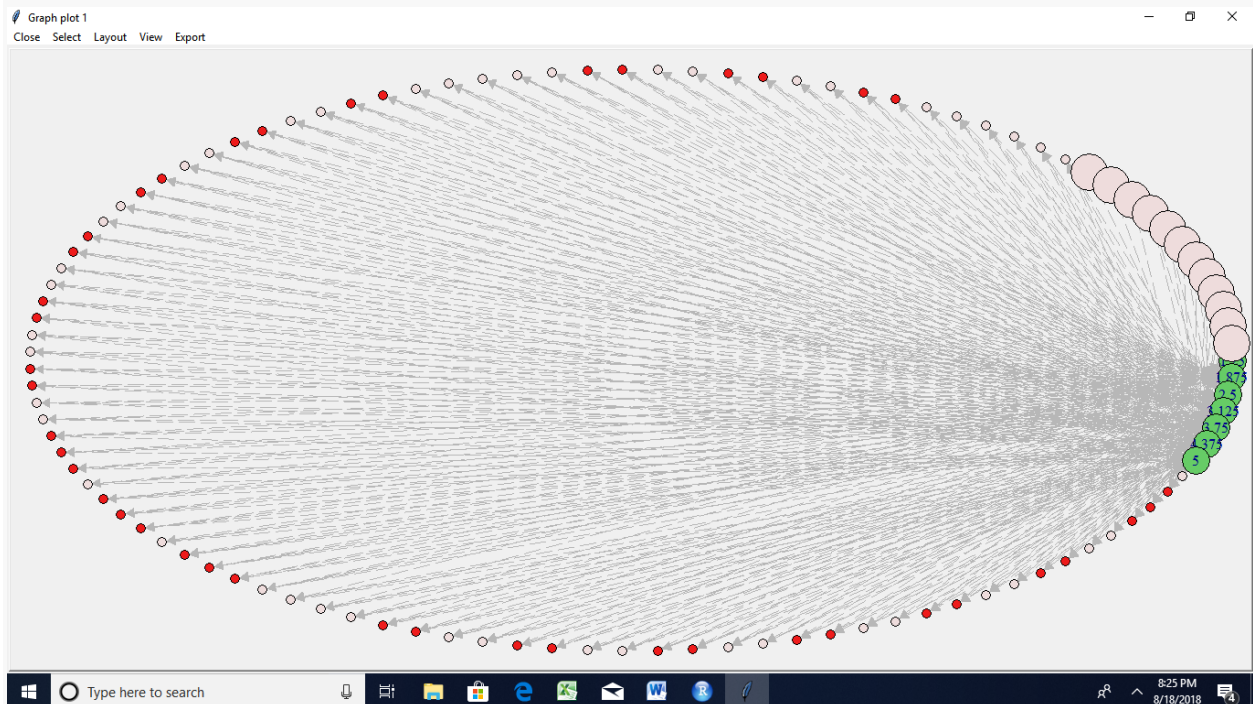
```
##          Dim.1 Dim.2 Dim.3 Dim.4 Dim.5
## rating    0.037 0.8922 0.071 3.3e-06 7.1e-05
## calories  0.904 0.0111 0.032 2.5e-02 2.8e-02
## protein   0.749 0.0073 0.002 2.4e-01 5.0e-04
## fat       0.844 0.0287 0.043 6.3e-02 2.2e-02
## sodium    0.344 0.0645 0.586 5.7e-03 2.6e-05
```

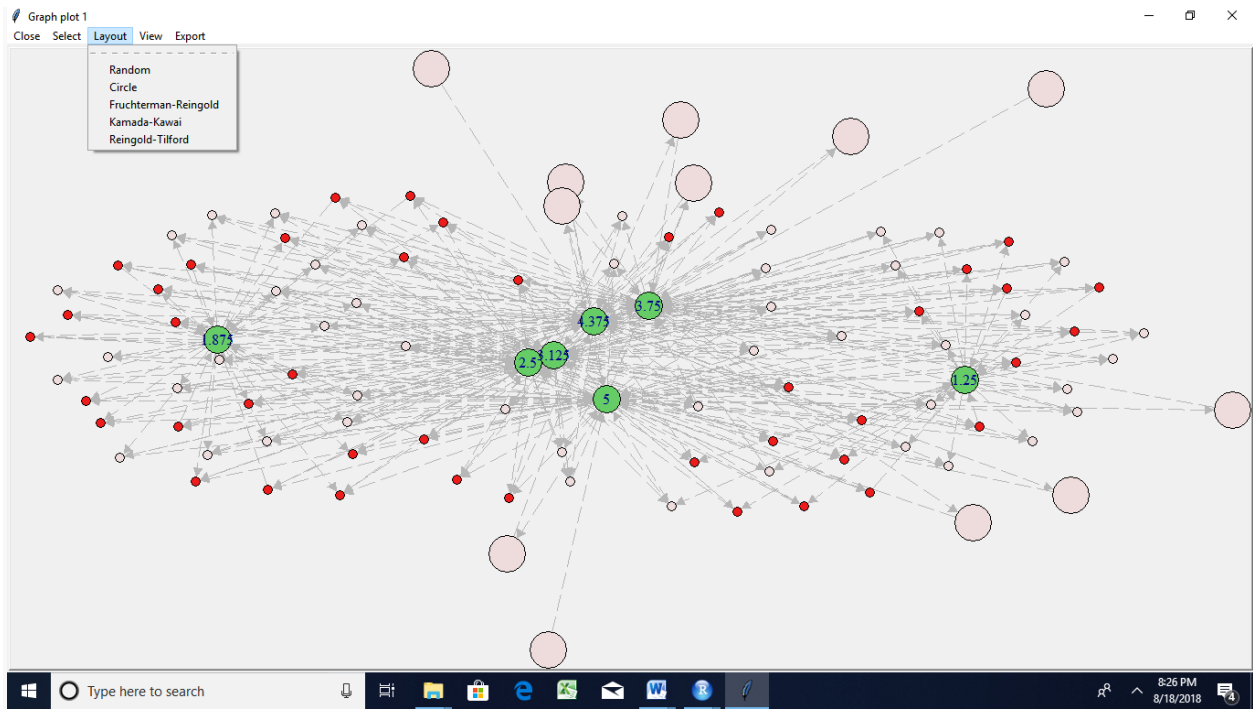


## Random type

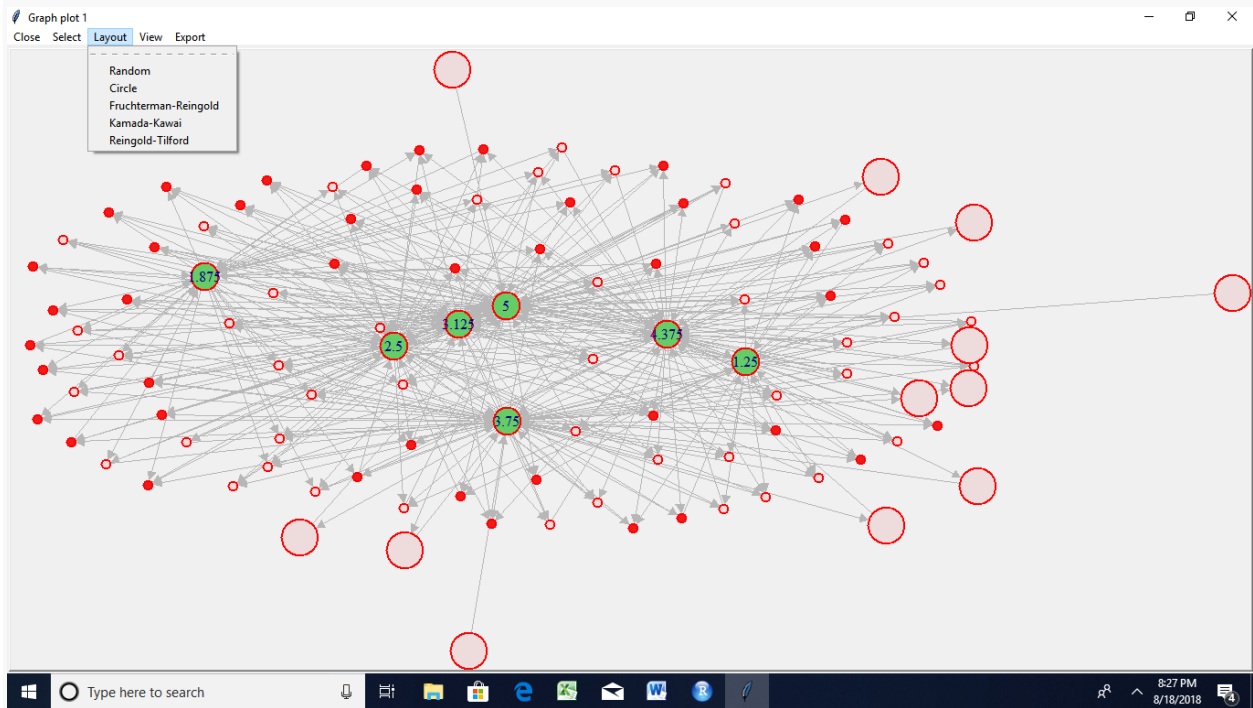


## Circle

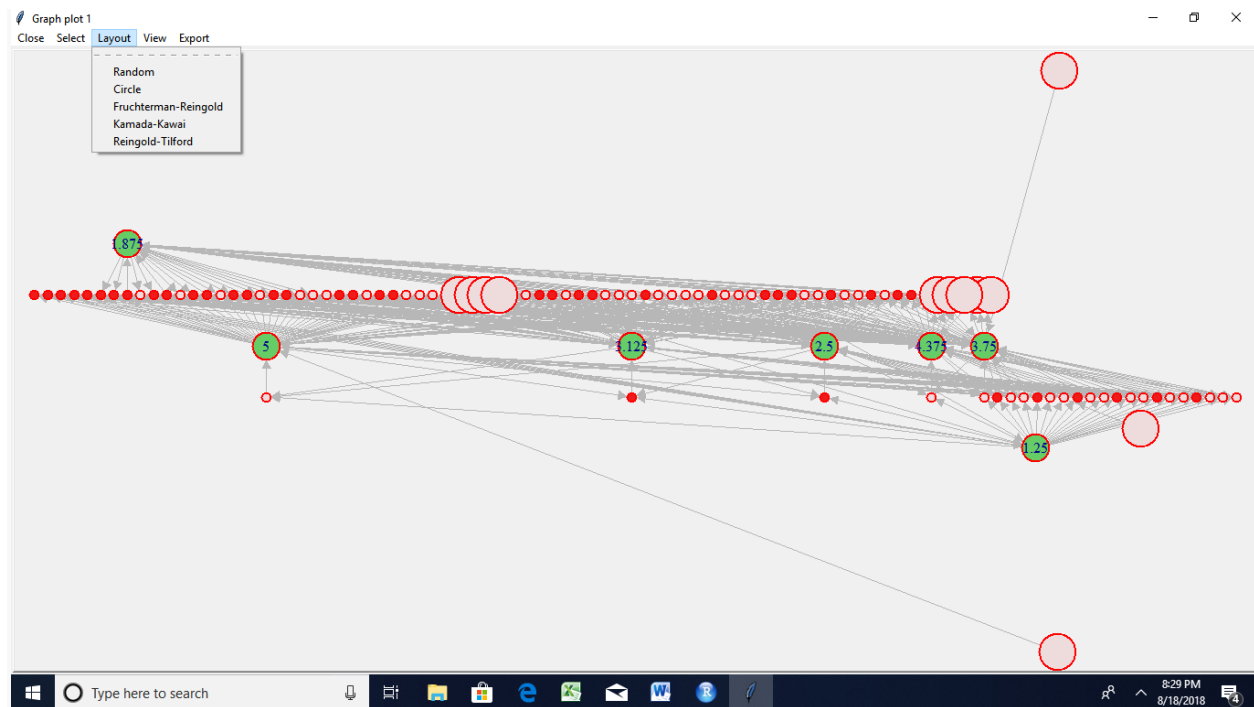




## Kamada kawai



## Reingold



## Session 21 Assignment 2<sup>nd</sup> additional exercise

### session21\_pci.R

Seshan

Thu Aug 16 15:46:09 2018

```
setwd("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/session21")
library(readr)
epi_r <- read_csv("C:/Users/Seshan/Desktop/sv R related/acadgild/assignments/
session21/epi_r.csv")
View(epi_r)
data<-epi_r
View(data)

a <- aggregate(data[, -1], by=list(data[,1]), paste, collapse=",")
a$combined <- apply(a[,2:ncol(a)], 1, paste, collapse=",")
a$combined <- gsub(",NA", "", a$combined) ## this column contains the totality
of all ingredients for a cuisine

cuisines <- as.data.frame(table(data[,1])) ## Number of recipes for each cuis
ine
freq <- lapply(lapply(strsplit(a$combined, ","), table), as.data.frame) ## Fre
quency of ingredients
```

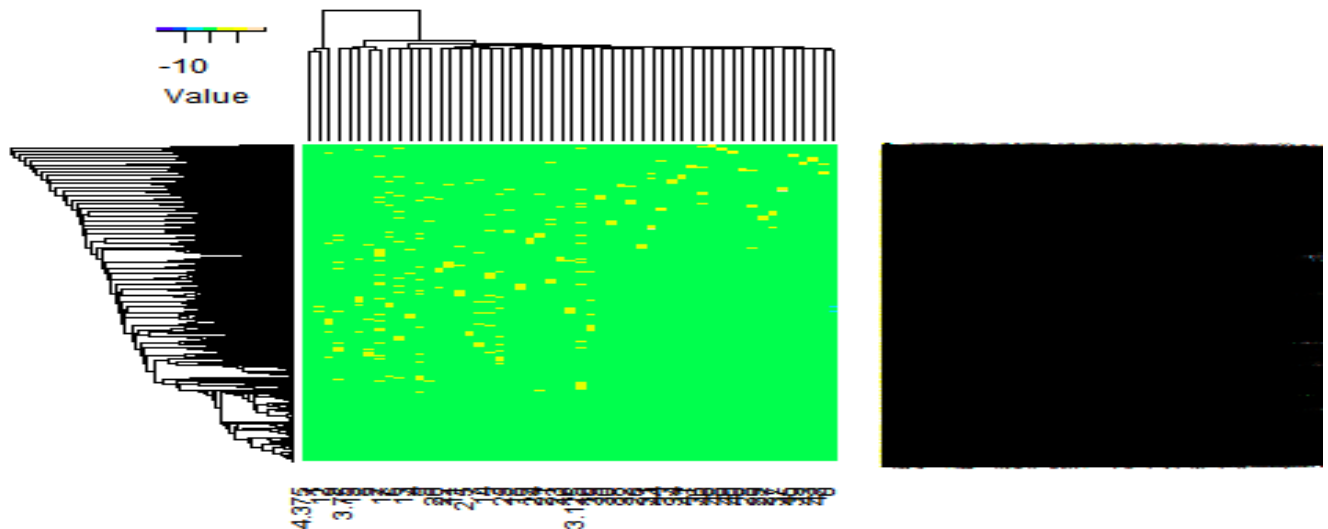
```

names(freq) <- a[,1]
prop <- lapply(seq_along(freq), function(i) { colnames(freq[[i]])[2] <- names
(freq)[i] freq[[i]][,2] <- freq[[i]][,2]/cuisines[i,2] ## proportion (normali
zed frequency) freq[[i]]})
names(prop) <- a[,1] ## this is a list of 26 elements, one for each cuisine
final <- Reduce(function(...) merge(..., all=TRUE, by="Var1"), prop)
row.names(final) <- final[,1]
final <- final[,-1]
final[is.na(final)] <- 0 ## If ingredient missing in all recipes, proportion
set to zero
final <- t(final) ## proportion matrix
s <- sort(apply(final, 2, sd), decreasing=TRUE)
## Selecting ingredients with maximum variation in frequency among cuisines a
nd
## Using standardized proportions for final analysis
final_imp <- scale(subset(final, select=names(which(s > 0.1))))
## heatmap
library(gplots) ##
## Attaching package: 'gplots'

## The following object is masked from 'package:stats':
##
##      lowess

heatmap.2(final_imp, trace="none", margins = c(6,11), col=topo.colors(7),
          key=TRUE, key.title=NA, keysize=1.2, density.info="none")

```



```

## PCA and biplot
p <- princomp(final_imp)
biplot(p, pc.biplot=TRUE, col=c("black", "red"), cex=c(0.9, 0.8), xlim=c(-2.5, 2.5),
       xlab="PC1, 39.7% explained variance", ylab="PC2, 24.5% explained variance")

```



