

Project Report

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Project Description

Getting and preparing the data set

importing the library we are going to use

```
library(dplyr)
library(ggplot2)
library(forcats)
library(arules)
library(hrbrthemes)
```

Reading the data and checking the first 10 rows of it

```
dataPath <- readline("Enter the path to the data set : ")
grc <- as_tibble(read.csv(dataPath,stringsAsFactors = FALSE))
# displaying first 10 rows of our data
print(grc,n = 10, width = 80)

## # A tibble: 9,835 x 8
##   items                                count total   rnd customer   age city   paymentType
##   <chr>                                <int> <int> <int> <chr>    <int> <chr>   <chr>
## 1 citrus fruit,semi-finish~         4  1612     9 Maged        60 Hurgh~ Cash
## 2 tropical fruit,yogurt,co~         3   509    12 Eman         23 Aswan Cash
## 3 whole milk                         1  2084     8 Rania         37 Dakah~ Cash
## 4 pip fruit,yogurt,cream c~         4   788     8 Rania         37 Dakah~ Cash
## 5 other vegetables,whole m~         4  1182    14 Magdy        36 Sohag Cash
## 6 whole milk,butter,yogurt~         5  1771     3 Ahmed         30 Giza  Credit
## 7 rolls/buns                         1  2196     7 Huda         39 Gharb~ Cash
## 8 other vegetables,UHT-mil~         5  1657     6 Walaa         29 Cairo Cash
## 9 pot plants                         1  2373     2 Mohamed        25 Alexa~ Credit
## 10 whole milk,cereals                2   343     5 Shimaa        55 Port ~ Cash
## # ... with 9,825 more rows
```

Preparing the data for k-means and apriori algorithms

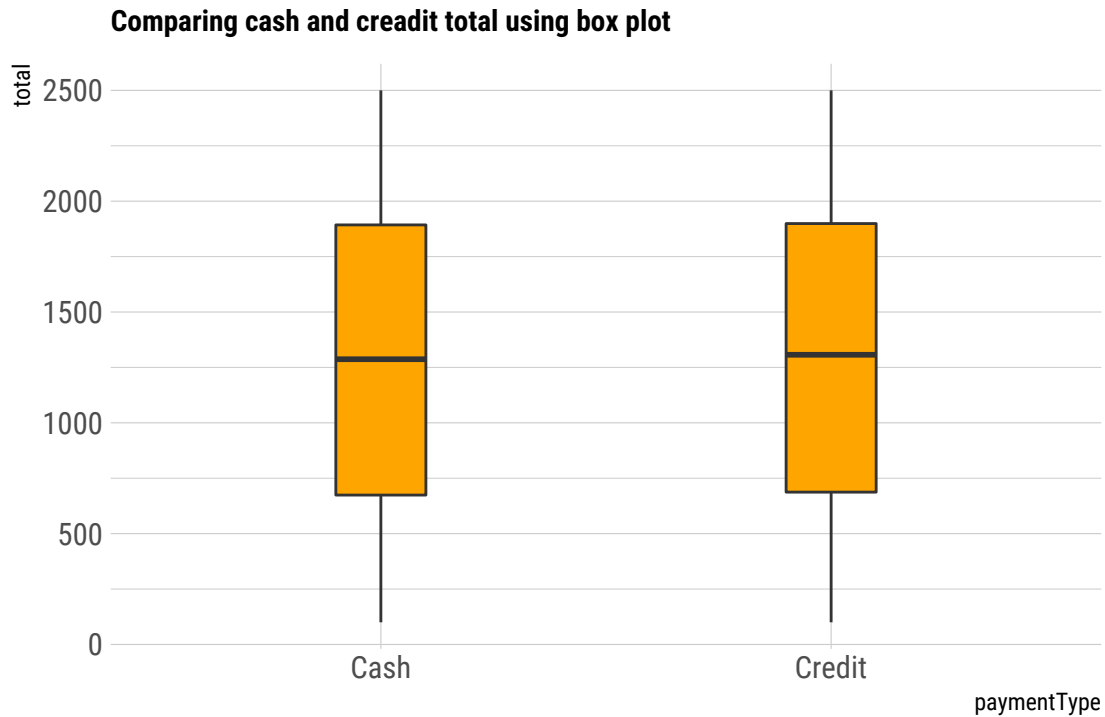
```
# creating a data framing containing customers, their ages and total spending
grc_customers <- grc %>%
  select(customer,age,total) %>%
  group_by(customer)%>%
  mutate(total = sum(total))%>%
  unique()
# making the data frame suitable for k-means
grc_k <- data.frame(grc_customers[,2:3],row.names = grc_customers$customer)
# splitting the items to suitable for apriori algorithms
tdata <- strsplit(as.vector(grc$items), ',')
tdata <- transactions(tdata)
```

Visualizing our Data

Comparison between cash and credit total spending using box plot

```
boxplot_cashCredit <- ggplot(
  grc,
  aes(x = paymentType, y = total, fill = paymentType)
) +
  geom_boxplot(width = .2,
               fill = "orange",
               outlier.color = "orange",
               outlier.size = 2)+
  theme_ipsum_rc() +
  theme(
    legend.position="none",
    plot.title = element_text(size=11)
  ) +
```

```
ggtitle("Comparing cash and creadit total using box plot")
boxplot_cashCreadit
```



Observations

After brief moments of seeing the figure, it is quite easy to see that people nearly equally pay with Cash or credit money.

Compare each age and sum of total spending.

Before visualizing

let's create a contingency table using the data prepared and look at a table containing customers, their ages and their total individual spending

```
table(grc_customers$age)
```

```
##
## 22 23 25 29 30 35 36 37 39 50 55 60
##  2  1  1  1  1  1  1  2  1  1  2  1
```

```
arrange(grc_customers,desc(total))
```

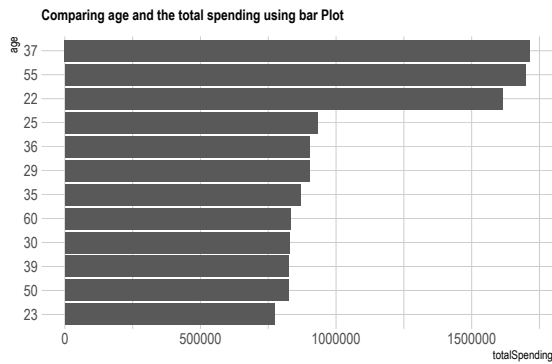
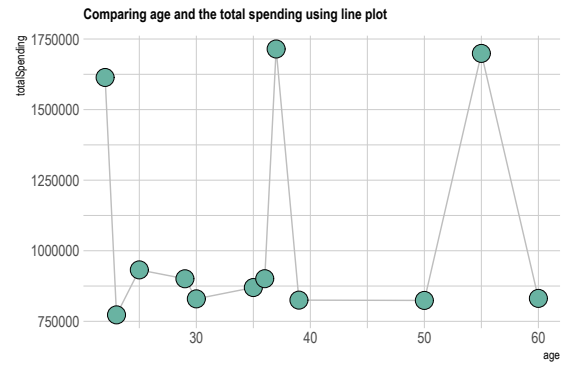
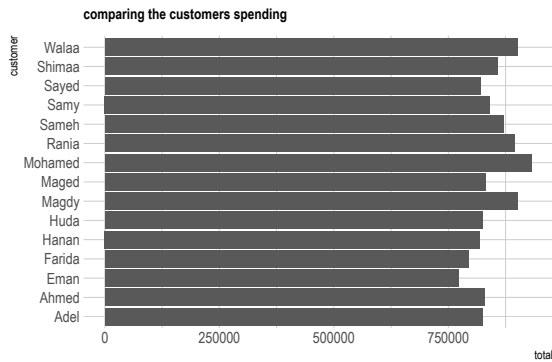
```
## # A tibble: 15 x 3
## # Groups:   customer [15]
##   customer  age  total
##   <chr>    <int> <int>
## 1 Mohamed    25 932250
```

```
## 2 Magdy      36 901010
## 3 Walaa      29 900797
## 4 Rania      37 893789
## 5 Sameh      35 869668
## 6 Shimaa     55 857901
## 7 Samy       55 841167
## 8 Maged      60 831272
## 9 Ahmed      30 829587
## 10 Huda      39 825147
## 11 Adel      50 824064
## 12 Sayed     37 820900
## 13 Hanan     22 819231
## 14 Farida    22 794570
## 15 Eman      23 772871
```

It becomes clear that there are more people aged (22,37 and 55) than the other age groups

```
customersBarPlot<- grc_customers %>%
  ggplot(
    aes(x = customer, y = total)
  ) +
  geom_col() +
  coord_flip()+
  theme_ipsum() +
  theme(
    plot.title = element_text(size=11)
  )+
  ggtitle("comparing the customers spending")
grc_age <- select(grc,age,total)
grc_age <- grc_age %>%
  group_by(age) %>%
  summarise(totalSpending = sum(total))
linePlotAgeSum <- ggplot(
  grc_age,
  aes(x = age, y = totalSpending)) +
  geom_line( color="gray") +
  geom_point(shape=21, color="black", fill="#69b3a2", size=6) +
  theme_ipsum() +
  theme(
    plot.title = element_text(size=11)
  )+
  ggtitle("Comparing age and the total spending using line plot")
grc_age <- mutate(grc_age,age = fct_reorder(as.factor(age),totalSpending))
barPlotAgeSum<-ggplot(
  grc_age,
  aes(x = age, y = totalSpending)) +
  geom_col() +
  coord_flip() +
  theme_ipsum() +
  theme(
    plot.title = element_text(size=11)
  )+
  ggtitle("Comparing age and the total spending using bar Plot")
customersBarPlot
linePlotAgeSum
```

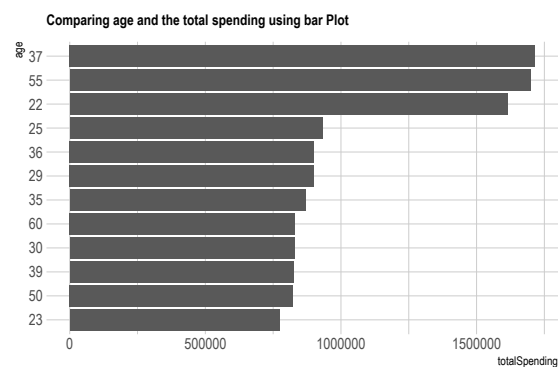
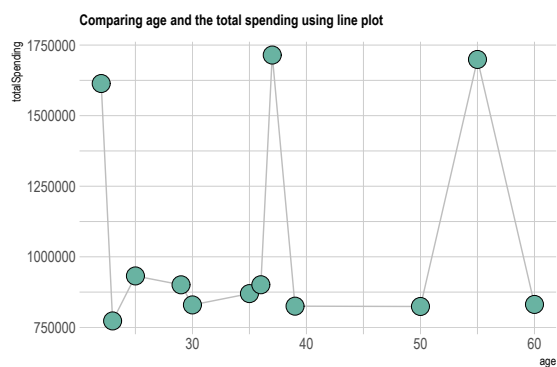
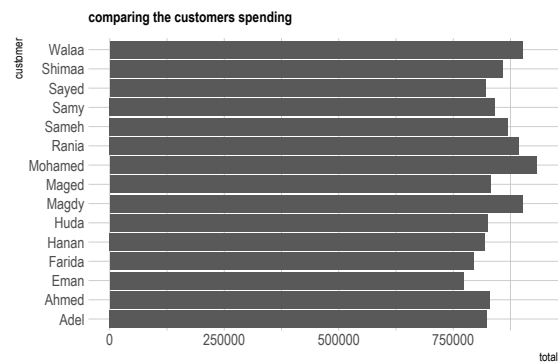
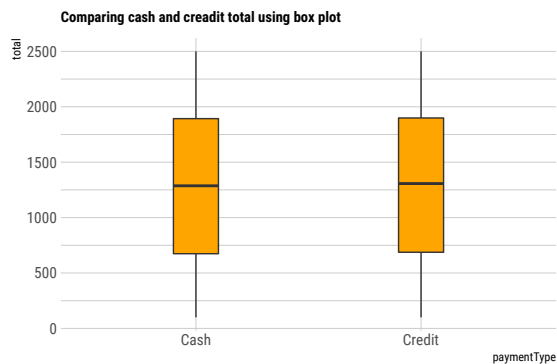
barPlotAgeSum



Observations

Since there are more customers aged (22, 37 and 55), it makes sense that the total spending of these groups is higher than the rest. But, looking at the data again we can see that if we look at the customers and their individual spending, we find that customers aged between 20 and 30 -even though there are single individual of these groups- are the highest spending.

Dashboard



Generating of association rules

Brief explanation of Apriori algorithm for generating the rules

Apriori algorithm is an iterative approach for discovering the most frequent item sets. The frequent item sets generated by the algorithm can be used to determine association rules that highlight general trends in the data-set, it is especially useful in the analysis of super-market items in our data set

Implementing the alorithm

Reading both minimum support and minimum confidance from the user

```
min_support <- as.numeric(readline("Enter the minimum Support : "))
min_conf <- as.numeric(readline("Enter the minimum Confidance : "))
```

implementing the algorithm using the built-in function

```
apriori_rules <- apriori(
  tdata,
  parameter = list(supp = min_support, conf = min_conf, minlen = 2))
```

```
## Apriori
##
## Parameter specification:
```

```
## confidence minval smax arem aval originalSupport maxtime support minlen maxlen target ext
##      0.5      0.1      1 none FALSE          TRUE      5    0.001      2     10 rules TRUE
##
## 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 ...[169 item(s), 9835 transaction(s)] done [0.00s].
## sorting and recoding items ... [157 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 ... [5668 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].

# Displaying at most 100 rows of the rules
as_tibble(DATAFRAME(apriori_rules,separate = TRUE, setStart = "", setEnd = "")) %>%
  print(n = 100, width = 90)
```

```
## # A tibble: 5,668 x 7
##   LHS                                RHS      support confidence coverage lift count
##   <fct>                             <fct>      <dbl>      <dbl>      <dbl> <dbl> <int>
## 1 honey                             whole milk  0.00112    0.733  0.00153  2.87   11
## 2 tidbits                           rolls/buns  0.00122    0.522  0.00234  2.84   12
## 3 cocoa drinks                      whole milk  0.00132    0.591  0.00224  2.31   13
## 4 pudding powder                    whole milk  0.00132    0.565  0.00234  2.21   13
## 5 cooking chocolate                 whole milk  0.00132    0.52   0.00254  2.04   13
## 6 cereals                           whole milk  0.00366    0.643  0.00569  2.52   36
## 7 jam                               whole milk  0.00295    0.547  0.00539  2.14   29
## 8 specialty cheese                  other vegeta~ 0.00427    0.5    0.00854  2.58   42
## 9 rice                              other vegeta~ 0.00397    0.52   0.00763  2.69   39
## 10 rice                             whole milk  0.00468    0.613  0.00763  2.40   46
## 11 baking powder                    whole milk  0.00925    0.523  0.0177   2.05   91
## 12 liver loaf,yogurt                whole milk  0.00102    0.667  0.00153  2.61   10
## 13 curd cheese,tropical fruit        other vegeta~ 0.00102    0.667  0.00153  3.45   10
## 14 curd cheese,rolls/buns            whole milk  0.00102    0.625  0.00163  2.45   10
## 15 curd cheese,other vegetables      whole milk  0.00122    0.571  0.00214  2.24   12
## 16 curd cheese,whole milk            other vegeta~ 0.00122    0.522  0.00234  2.70   12
## 17 cleaner,other vegetables          whole milk  0.00102    0.625  0.00163  2.45   10
## 18 liquor,red/blush wine             bottled beer 0.00193    0.905  0.00214  11.2   19
## 19 liquor,soda                      bottled beer 0.00122    0.571  0.00214  7.10   12
## 20 cereals,curd                     whole milk  0.00102    0.909  0.00112  3.56   10
## 21 cereals,root vegetables           whole milk  0.00102    0.769  0.00132  3.01   10
## 22 cereals,yogurt                   whole milk  0.00173    0.810  0.00214  3.17   17
## 23 cereals,other vegetables          whole milk  0.00132    0.65   0.00203  2.54   13
## 24 butter,jam                       whole milk  0.00102    0.833  0.00122  3.26   10
## 25 jam,root vegetables               other vegeta~ 0.00102    0.526  0.00193  2.72   10
## 26 jam,other vegetables              root vegetab~ 0.00102    0.556  0.00183  5.10   10
## 27 jam,root vegetables               whole milk  0.00132    0.684  0.00193  2.68   13
## 28 jam,other vegetables              whole milk  0.00132    0.722  0.00183  2.83   13
## 29 instant coffee,whipped/sour cre~ other vegeta~ 0.00102    0.769  0.00132  3.98   10
## 30 instant coffee,other vegetables   whipped/sour~ 0.00102    0.526  0.00193  7.34   10
## 31 instant coffee,other vegetables   soda        0.00102    0.526  0.00193  3.02   10
```

##	32	instant coffee,other vegetables	whole milk	0.00102	0.526	0.00193	2.06	10
##	33	vinegar,yogurt	other vegeta~	0.00102	0.556	0.00183	2.87	10
##	34	other vegetables,vinegar	whole milk	0.00132	0.542	0.00244	2.12	13
##	35	vinegar,whole milk	other vegeta~	0.00132	0.5	0.00264	2.58	13
##	36	popcorn,salty snack	soda	0.00122	0.545	0.00224	3.13	12
##	37	popcorn,soda	salty snack	0.00122	0.632	0.00193	16.7	12
##	38	candles,tropical fruit	whole milk	0.00102	0.667	0.00153	2.61	10
##	39	candles,root vegetables	other vegeta~	0.00102	0.588	0.00173	3.04	10
##	40	bottled beer,soups	whole milk	0.00112	0.917	0.00122	3.59	11
##	41	soups,whipped/sour cream	other vegeta~	0.00102	0.667	0.00153	3.45	10
##	42	pip fruit,soups	other vegeta~	0.00102	0.714	0.00142	3.69	10
##	43	root vegetables,soups	other vegeta~	0.00122	0.706	0.00173	3.65	12
##	44	root vegetables,soups	whole milk	0.00112	0.647	0.00173	2.53	11
##	45	other vegetables,soups	whole milk	0.00183	0.581	0.00315	2.27	18
##	46	soups,whole milk	other vegeta~	0.00183	0.621	0.00295	3.21	18
##	47	dog food,tropical fruit	yogurt	0.00102	0.625	0.00163	4.48	10
##	48	dog food,yogurt	tropical fru~	0.00102	0.526	0.00193	5.02	10
##	49	dog food,tropical fruit	whole milk	0.00102	0.625	0.00163	2.45	10
##	50	dog food,root vegetables	whole milk	0.00102	0.714	0.00142	2.80	10
##	51	dog food,yogurt	whole milk	0.00122	0.632	0.00193	2.47	12
##	52	dog food,other vegetables	whole milk	0.00112	0.5	0.00224	1.96	11
##	53	Instant food products,soda	hamburger me~	0.00122	0.632	0.00193	19.0	12
##	54	hamburger meat,Instant food pro~	whole milk	0.00153	0.5	0.00305	1.96	15
##	55	Instant food products,whole milk	hamburger me~	0.00153	0.5	0.00305	15.0	15
##	56	Instant food products,root vege~	other vegeta~	0.00102	0.526	0.00193	2.72	10
##	57	Instant food products,yogurt	whole milk	0.00112	0.786	0.00142	3.08	11
##	58	Instant food products,other veg~	whole milk	0.00153	0.556	0.00275	2.17	15
##	59	Instant food products,whole milk	other vegeta~	0.00153	0.5	0.00305	2.58	15
##	60	specialty cheese,whipped/sour c~	other vegeta~	0.00102	0.714	0.00142	3.69	10
##	61	citrus fruit,specialty cheese	whole milk	0.00102	0.714	0.00142	2.80	10
##	62	bottled water,specialty cheese	other vegeta~	0.00102	0.588	0.00173	3.04	10
##	63	root vegetables,specialty cheese	yogurt	0.00112	0.524	0.00214	3.75	11
##	64	root vegetables,specialty cheese	other vegeta~	0.00142	0.667	0.00214	3.45	14
##	65	root vegetables,specialty cheese	whole milk	0.00132	0.619	0.00214	2.42	13
##	66	specialty cheese,yogurt	other vegeta~	0.00163	0.571	0.00285	2.95	16
##	67	specialty cheese,yogurt	whole milk	0.00203	0.714	0.00285	2.80	20
##	68	specialty cheese,whole milk	yogurt	0.00203	0.541	0.00376	3.87	20
##	69	other vegetables,specialty chee~	whole milk	0.00224	0.524	0.00427	2.05	22
##	70	specialty cheese,whole milk	other vegeta~	0.00224	0.595	0.00376	3.07	22
##	71	butter,flower (seeds)	whole milk	0.00102	0.625	0.00163	2.45	10
##	72	flower (seeds),root vegetables	other vegeta~	0.00112	0.579	0.00193	2.99	11
##	73	flower (seeds),root vegetables	whole milk	0.00102	0.526	0.00193	2.06	10
##	74	frozen potato products,fruit/ve~	whole milk	0.00112	0.688	0.00163	2.69	11
##	75	frozen potato products,yogurt	whole milk	0.00153	0.682	0.00224	2.67	15
##	76	frozen potato products,rolls/bu~	whole milk	0.00112	0.55	0.00203	2.15	11
##	77	frozen potato products,other ve~	whole milk	0.00183	0.692	0.00264	2.71	18
##	78	frozen potato products,whole mi~	other vegeta~	0.00183	0.529	0.00346	2.74	18
##	79	house keeping products,napkins	whole milk	0.00132	0.812	0.00163	3.18	13
##	80	house keeping products,whipped/~	whole milk	0.00122	0.923	0.00132	3.61	12
##	81	house keeping products,sausage	whole milk	0.00112	0.688	0.00163	2.69	11
##	82	house keeping products,tropical~	whole milk	0.00112	0.688	0.00163	2.69	11
##	83	house keeping products,root veg~	whole milk	0.00132	0.765	0.00173	2.99	13
##	84	house keeping products,yogurt	other vegeta~	0.00102	0.556	0.00183	2.87	10
##	85	house keeping products,yogurt	whole milk	0.00102	0.556	0.00183	2.17	10

## 86	house keeping products,rolls/bu~	whole milk	0.00112	0.647	0.00173	2.53	11
## 87	house keeping products,other ve~	whole milk	0.00173	0.630	0.00275	2.46	17
## 88	sweet spreads,white bread	whole milk	0.00102	0.667	0.00153	2.61	10
## 89	pastry,sweet spreads	whole milk	0.00102	0.909	0.00112	3.56	10
## 90	root vegetables,sweet spreads	whole milk	0.00122	0.6	0.00203	2.35	12
## 91	sweet spreads,yogurt	soda	0.00122	0.545	0.00224	3.13	12
## 92	soda,sweet spreads	whole milk	0.00142	0.56	0.00254	2.19	14
## 93	sweet spreads,yogurt	whole milk	0.00122	0.545	0.00224	2.13	12
## 94	curd,turkey	other vegeta~	0.00122	0.8	0.00153	4.13	12
## 95	curd,turkey	whole milk	0.00102	0.667	0.00153	2.61	10
## 96	butter,turkey	whole milk	0.00102	0.667	0.00153	2.61	10
## 97	turkey,whipped/sour cream	other vegeta~	0.00102	0.769	0.00132	3.98	10
## 98	pastry,turkey	whole milk	0.00112	0.579	0.00193	2.27	11
## 99	bottled water,turkey	whole milk	0.00122	0.75	0.00163	2.94	12
## 100	tropical fruit,turkey	root vegetab~	0.00153	0.577	0.00264	5.29	15
## #	... with 5,568 more rows						