

Data Ingestion and Summary statistics

```
credit_data.raw<-read.csv("BOA.csv")
str(credit_data.raw)
```

```
## 'data.frame': 8950 obs. of 19 variables:
## $ X : int 1 2 3 4 5 6 7 8 9 10 ...
## $ CUST_ID : Factor w/ 8950 levels "C10001","C10002",...: 1 2 3 4 5 6 7 8 9 10 ...
## $ BALANCE : num 40.9 3202.5 2495.1 1666.7 817.7 ...
## $ BALANCE_FREQUENCY : num 0.818 0.909 1 0.636 1 ...
## $ PURCHASES : num 95.4 0 773.2 1499 16 ...
## $ ONEOFF_PURCHASES : num 0 0 773 1499 16 ...
## $ INSTALLMENTS_PURCHASES : num 95.4 0 0 0 0 ...
## $ CASH_ADVANCE : num 0 6443 0 206 0 ...
## $ PURCHASES_FREQUENCY : num 0.1667 0 1 0.0833 0.0833 ...
## $ ONEOFF_PURCHASES_FREQUENCY : num 0 0 1 0.0833 0.0833 ...
## $ PURCHASES_INSTALLMENTS_FREQUENCY : num 0.0833 0 0 0 0 ...
## $ CASH_ADVANCE_FREQUENCY : num 0 0.25 0 0.0833 0 ...
## $ CASH_ADVANCE_TRX : int 0 4 0 1 0 0 0 0 0 0 ...
## $ PURCHASES_TRX : int 2 0 12 1 1 8 64 12 5 3 ...
## $ CREDIT_LIMIT : num 1000 7000 7500 7500 1200 1800 13500 2300 7000 11000 ...
## $ PAYMENTS : num 202 4103 622 0 678 ...
## $ MINIMUM_PAYMENTS : num 140 1072 627 0 245 ...
## $ PRC_FULL_PAYMENT : num 0 0.222 0 0 0 ...
## $ TENURE : int 12 12 12 12 12 12 12 12 12 12 ...
```

```
summary(credit_data.raw)
```

```
##           X           CUST_ID           BALANCE           BALANCE_FREQUENCY
## Min.      : 1      C10001 : 1      Min.      : 0.0      Min.      :0.0000
## 1st Qu.:2238      C10002 : 1      1st Qu.: 128.3      1st Qu.:0.8889
## Median :4476      C10003 : 1      Median : 873.4      Median :1.0000
## Mean    :4476      C10004 : 1      Mean    :1564.5      Mean    :0.8773
## 3rd Qu.:6713      C10005 : 1      3rd Qu.:2054.1      3rd Qu.:1.0000
## Max.    :8950      C10006 : 1      Max.    :19043.1      Max.    :1.0000
##           (Other):8944
## PURCHASES      ONEOFF_PURCHASES      INSTALLMENTS_PURCHASES      CASH_ADVANCE
## Min.      : 0.00      Min.      : 0.0      Min.      : 0.0      Min.      : 0.0
## 1st Qu.: 39.63      1st Qu.: 0.0      1st Qu.: 0.0      1st Qu.: 0.0
## Median : 361.28      Median : 38.0      Median : 89.0      Median : 0.0
## Mean    :1003.20      Mean    : 592.4      Mean    : 411.1      Mean    : 978.9
## 3rd Qu.:1110.13      3rd Qu.: 577.4      3rd Qu.: 468.6      3rd Qu.:1113.8
## Max.    :49039.57      Max.    :40761.2      Max.    :22500.0      Max.    :47137.2
##
## PURCHASES_FREQUENCY      ONEOFF_PURCHASES_FREQUENCY
## Min.      :0.00000      Min.      :0.00000
## 1st Qu.:0.08333      1st Qu.:0.00000
## Median :0.50000      Median :0.08333
## Mean    :0.49035      Mean    :0.20246
## 3rd Qu.:0.91667      3rd Qu.:0.30000
## Max.    :1.00000      Max.    :1.00000
##
## PURCHASES_INSTALLMENTS_FREQUENCY      CASH_ADVANCE_FREQUENCY      CASH_ADVANCE_TRX
## Min.      :0.0000      Min.      :0.0000      Min.      : 0.000
```

```
## 1st Qu.:0.0000      1st Qu.:0.0000      1st Qu.: 0.000
## Median :0.1667      Median :0.0000      Median : 0.000
## Mean :0.3644        Mean :0.1351      Mean : 3.249
## 3rd Qu.:0.7500      3rd Qu.:0.2222      3rd Qu.: 4.000
## Max. :1.0000        Max. :1.5000      Max. :123.000
##
## PURCHASES_TRX      CREDIT_LIMIT      PAYMENTS      MINIMUM_PAYMENTS
## Min. : 0.00      Min. : 50      Min. : 0.0      Min. : 0.0
## 1st Qu.: 1.00      1st Qu.: 1600      1st Qu.: 383.3      1st Qu.: 163.0
## Median : 7.00      Median : 3000      Median : 856.9      Median : 289.6
## Mean : 14.71      Mean : 4494      Mean : 1733.1      Mean : 834.0
## 3rd Qu.: 17.00      3rd Qu.: 6500      3rd Qu.: 1901.1      3rd Qu.: 788.7
## Max. :358.00      Max. :30000      Max. :50721.5      Max. :76406.2
##
## PRC_FULL_PAYMENT      TENURE
## Min. :0.0000      Min. : 6.00
## 1st Qu.:0.0000      1st Qu.:12.00
## Median :0.0000      Median :12.00
## Mean :0.1537      Mean :11.52
## 3rd Qu.:0.1429      3rd Qu.:12.00
## Max. :1.0000      Max. :12.00
##
```

#Data Exploration and Cleaning

```
library(psych)
library(dplyr)
```

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

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

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

```
#Remove identity columns - cust id and x from the table
credit_data<-credit_data.raw[c(-1,-2)]
describe(credit_data)
```

```
##              vars      n    mean      sd median trimmed
## BALANCE              1 8950 1564.47 2081.53  873.39 1128.17
## BALANCE_FREQUENCY    2 8950    0.88   0.24    1.00    0.94
## PURCHASES            3 8950 1003.20 2136.63  361.28  583.10
## ONEOFF_PURCHASES     4 8950  592.44 1659.89   38.00  263.82
## INSTALLMENTS_PURCHASES 5 8950  411.07  904.34   89.00  223.03
## CASH_ADVANCE         6 8950  978.87 2097.16    0.00  494.90
## PURCHASES_FREQUENCY  7 8950    0.49   0.40    0.50   0.49
## ONEOFF_PURCHASES_FREQUENCY 8 8950    0.20   0.30    0.08   0.14
## PURCHASES_INSTALLMENTS_FREQUENCY 9 8950    0.36   0.40    0.17   0.33
```

```
## CASH_ADVANCE_FREQUENCY      10 8950      0.14      0.20      0.00      0.09
## CASH_ADVANCE_TRX             11 8950      3.25      6.82      0.00      1.73
## PURCHASES_TRX                12 8950     14.71     24.86      7.00      9.43
## CREDIT_LIMIT                 13 8950  4494.45  3638.61  3000.00  3927.21
## PAYMENTS                     14 8950  1733.14  2895.06   856.90  1152.89
## MINIMUM_PAYMENTS             15 8950   833.98  2335.99   289.63   463.58
## PRC_FULL_PAYMENT             16 8950      0.15      0.29      0.00      0.08
## TENURE                       17 8950     11.52      1.34     12.00     11.92
##                               mad min      max      range      skew      kurtosis
## BALANCE                      1185.88    0 19043.14 19043.14    2.39      7.67
## BALANCE_FREQUENCY            0.00    0      1.00      1.00   -2.02      3.09
## PURCHASES                    535.63    0 49039.57 49039.57    8.14     111.30
## ONEOFF_PURCHASES             56.34    0 40761.25 40761.25   10.04     164.06
## INSTALLMENTS_PURCHASES      131.95    0 22500.00 22500.00    7.30      96.50
## CASH_ADVANCE                 0.00    0 47137.21 47137.21    5.16      52.86
## PURCHASES_FREQUENCY          0.62    0      1.00      1.00    0.06     -1.64
## ONEOFF_PURCHASES_FREQUENCY   0.12    0      1.00      1.00    1.54      1.16
## PURCHASES_INSTALLMENTS_FREQUENCY 0.25    0      1.00      1.00    0.51     -1.40
## CASH_ADVANCE_FREQUENCY        0.00    0      1.50      1.50    1.83      3.33
## CASH_ADVANCE_TRX             0.00    0     123.00   123.00    5.72      61.60
## PURCHASES_TRX                10.38    0     358.00   358.00    4.63      34.76
## CREDIT_LIMIT                 2668.68   50 30000.00 29950.00    1.52       2.83
## PAYMENTS                     861.91    0 50721.48 50721.48    5.91      54.73
## MINIMUM_PAYMENTS             279.80    0 76406.21 76406.21   13.80     292.13
## PRC_FULL_PAYMENT             0.00    0      1.00      1.00    1.94       2.43
## TENURE                       0.00    6     12.00      6.00   -2.94      7.69
##                               se
## BALANCE                      22.00
## BALANCE_FREQUENCY            0.00
## PURCHASES                    22.58
## ONEOFF_PURCHASES             17.55
## INSTALLMENTS_PURCHASES       9.56
## CASH_ADVANCE                 22.17
## PURCHASES_FREQUENCY          0.00
## ONEOFF_PURCHASES_FREQUENCY   0.00
## PURCHASES_INSTALLMENTS_FREQUENCY 0.00
## CASH_ADVANCE_FREQUENCY        0.00
## CASH_ADVANCE_TRX             0.07
## PURCHASES_TRX                0.26
## CREDIT_LIMIT                 38.46
## PAYMENTS                     30.60
## MINIMUM_PAYMENTS             24.69
## PRC_FULL_PAYMENT             0.00
## TENURE                       0.01
```

```
#credit_data<-credit_data[(credit_data[,3]>0),]

# To check for null value in the data
sapply(credit_data, function(x) sum(is.na(x)))
```

```
##                               BALANCE                               BALANCE_FREQUENCY
##                               0                               0
##                               PURCHASES                               ONEOFF_PURCHASES
##                               0                               0
```

```
##          INSTALLMENTS_PURCHASES          CASH_ADVANCE
##                      0                      0
##          PURCHASES_FREQUENCY    ONEOFF_PURCHASES_FREQUENCY
##                      0                      0
## PURCHASES_INSTALLMENTS_FREQUENCY    CASH_ADVANCE_FREQUENCY
##                      0                      0
##          CASH_ADVANCE_TRX          PURCHASES_TRX
##                      0                      0
##          CREDIT_LIMIT          PAYMENTS
##                      0                      0
##          MINIMUM_PAYMENTS    PRC_FULL_PAYMENT
##                      0                      0
##          TENURE
##                      0
```

```
#Plot graph for each variable for distribution
```

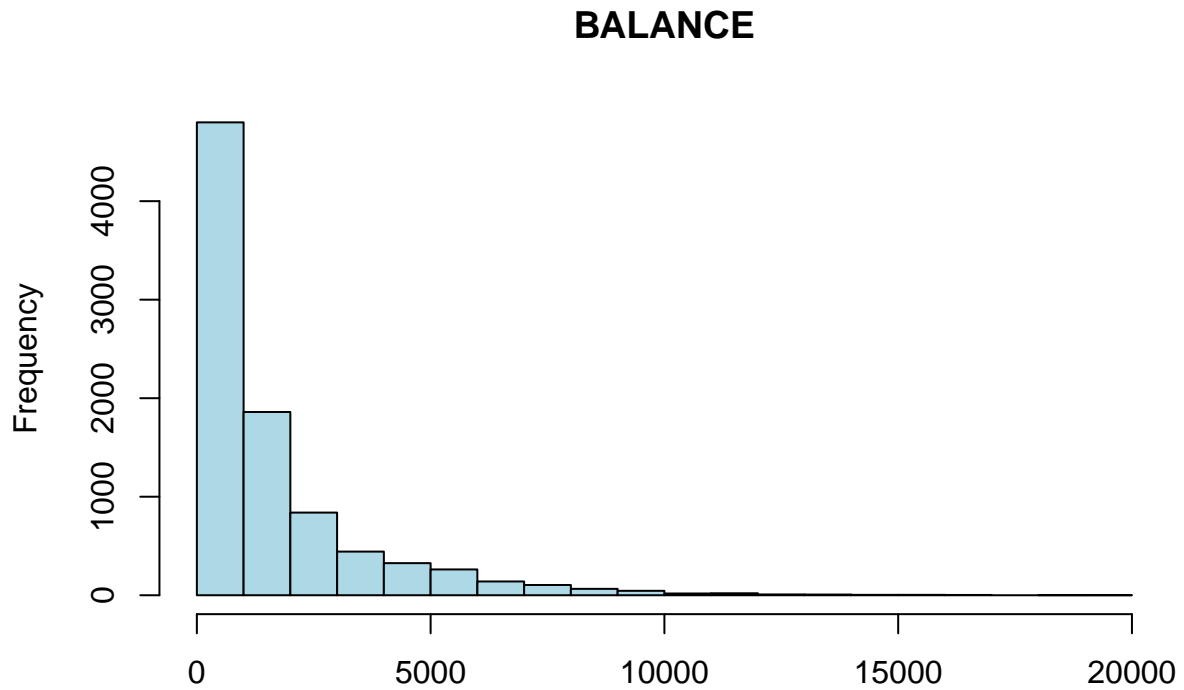
```
for ( i in seq(1,length( credit_data ),1) ) hist(credit_data[,i],ylab="Frequency",xlab = "",type="l",ma
```

```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

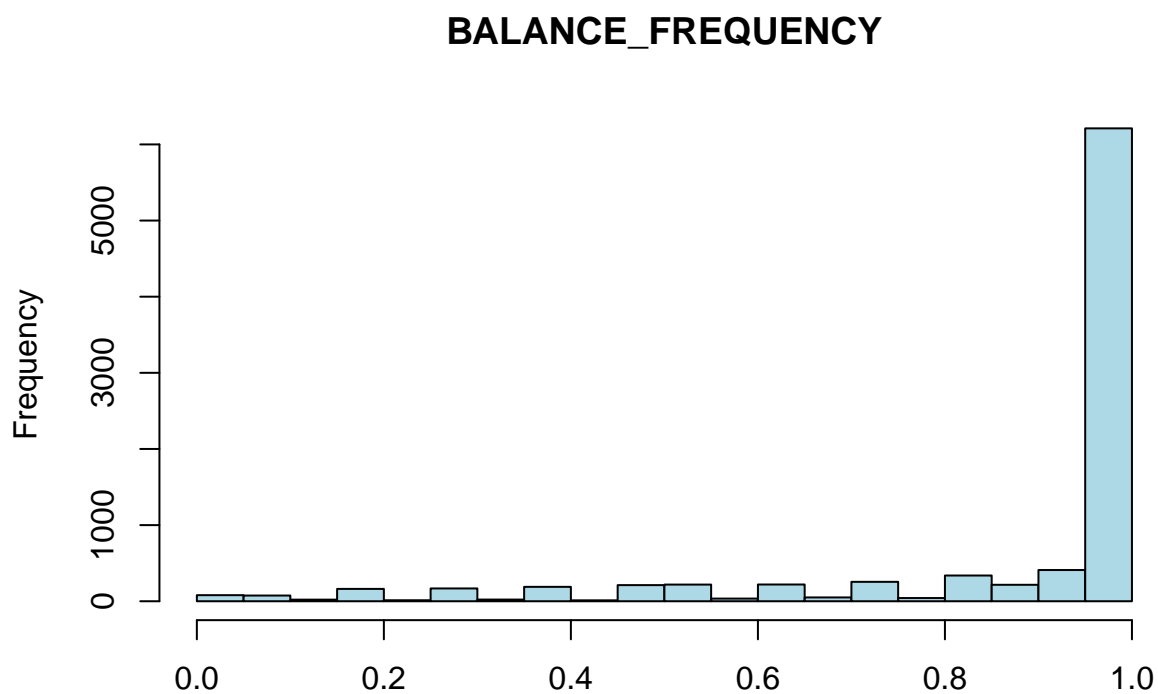


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```



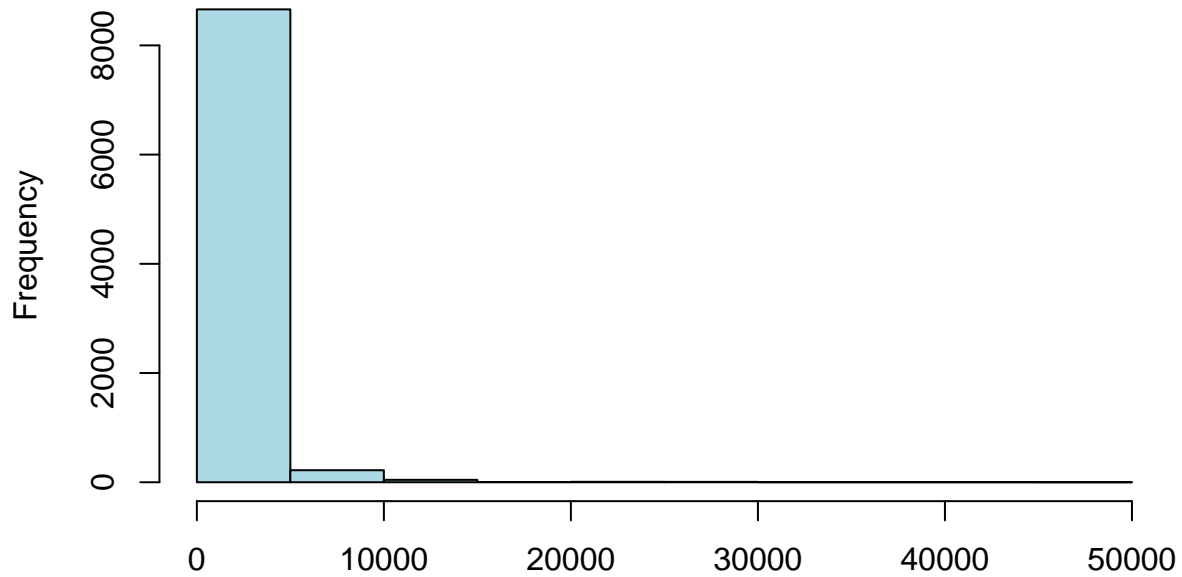
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

PURCHASES

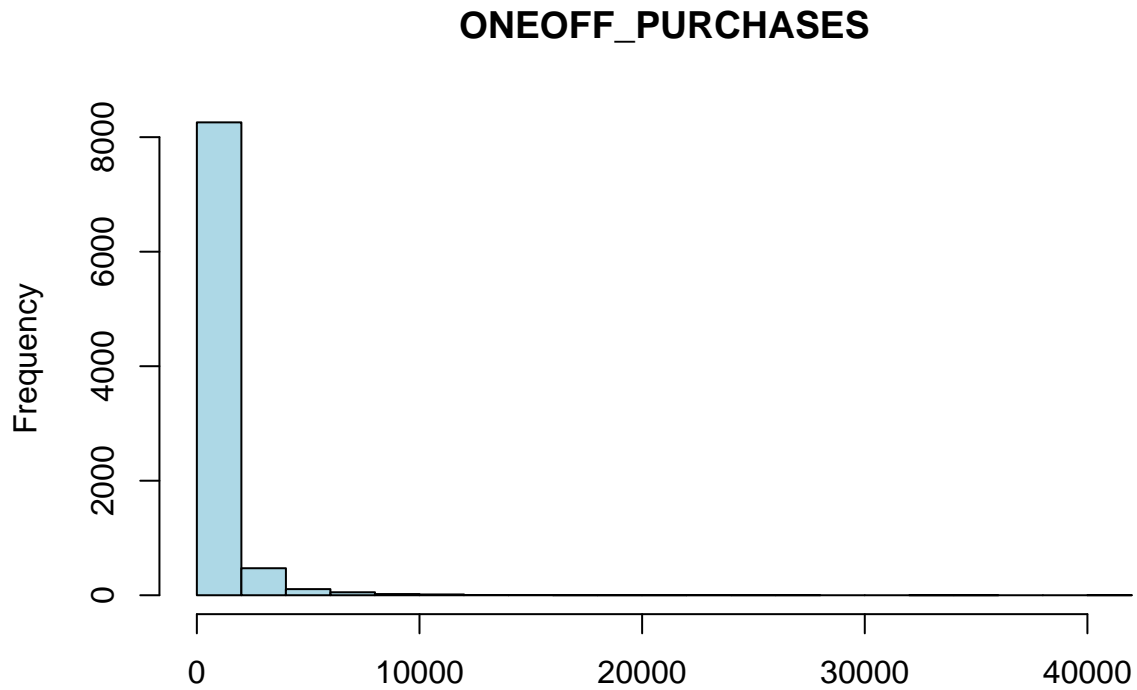


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```



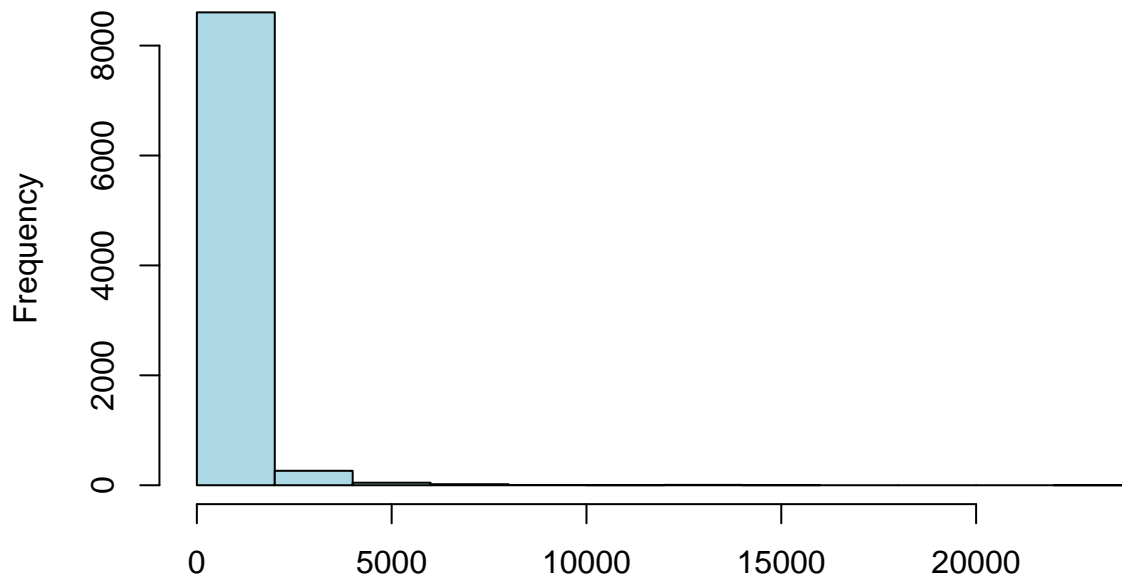
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```


INSTALLMENTS_PURCHASES



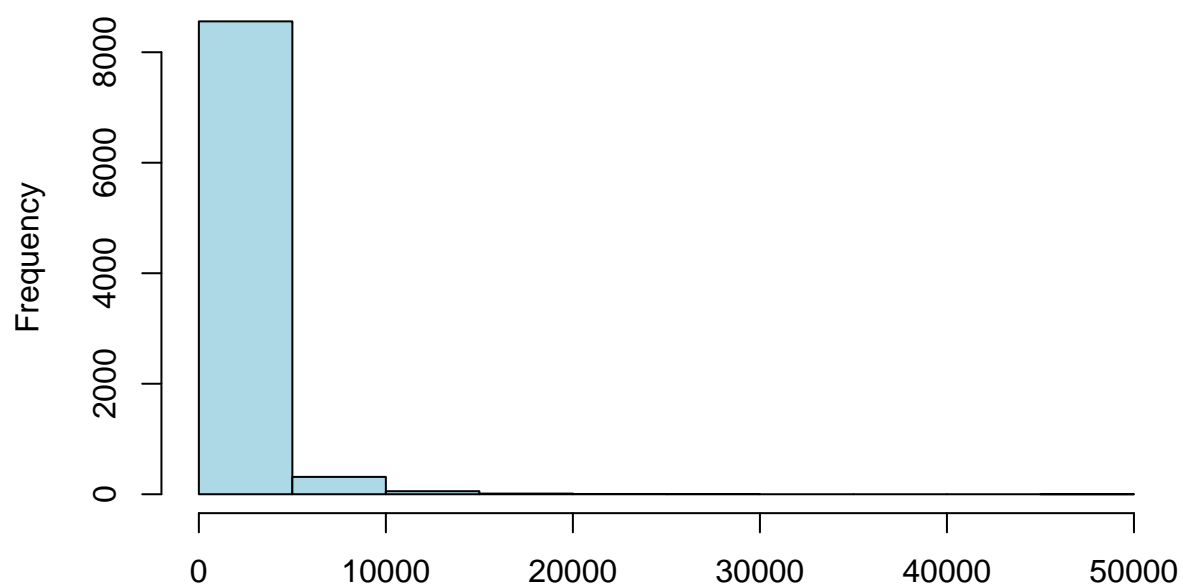
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

CASH_ADVANCE



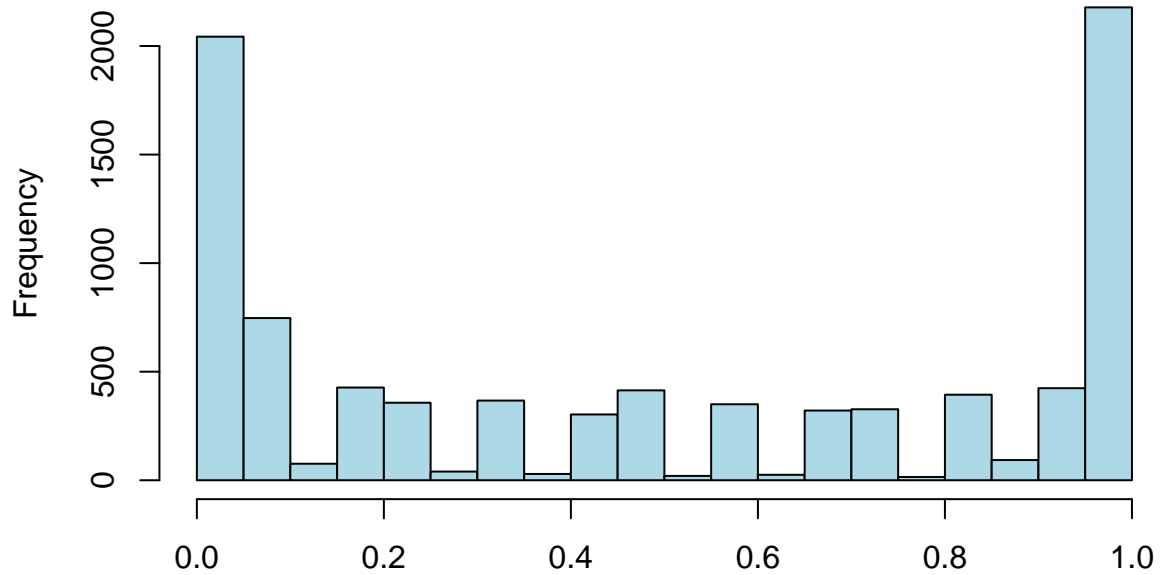
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

PURCHASES_FREQUENCY



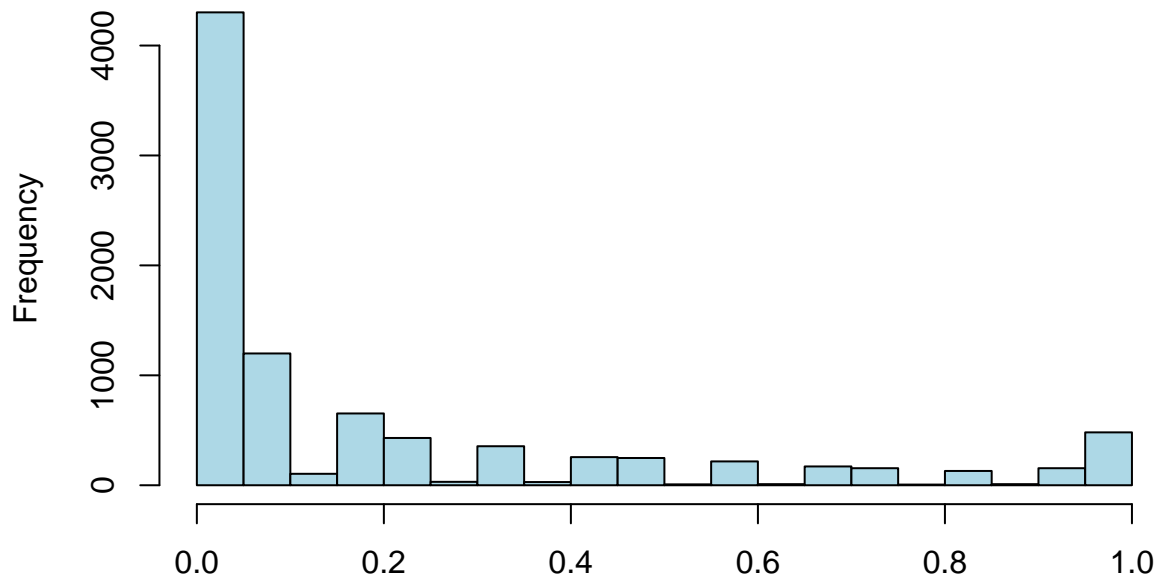
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

ONEOFF_PURCHASES_FREQUENCY

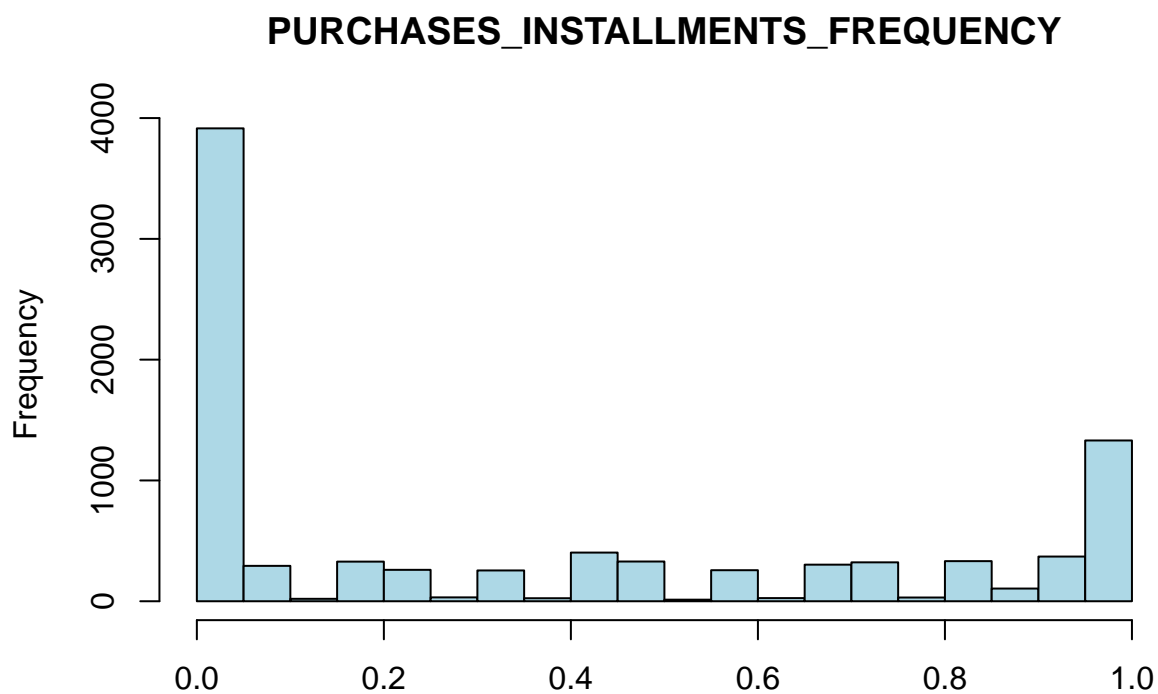


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```



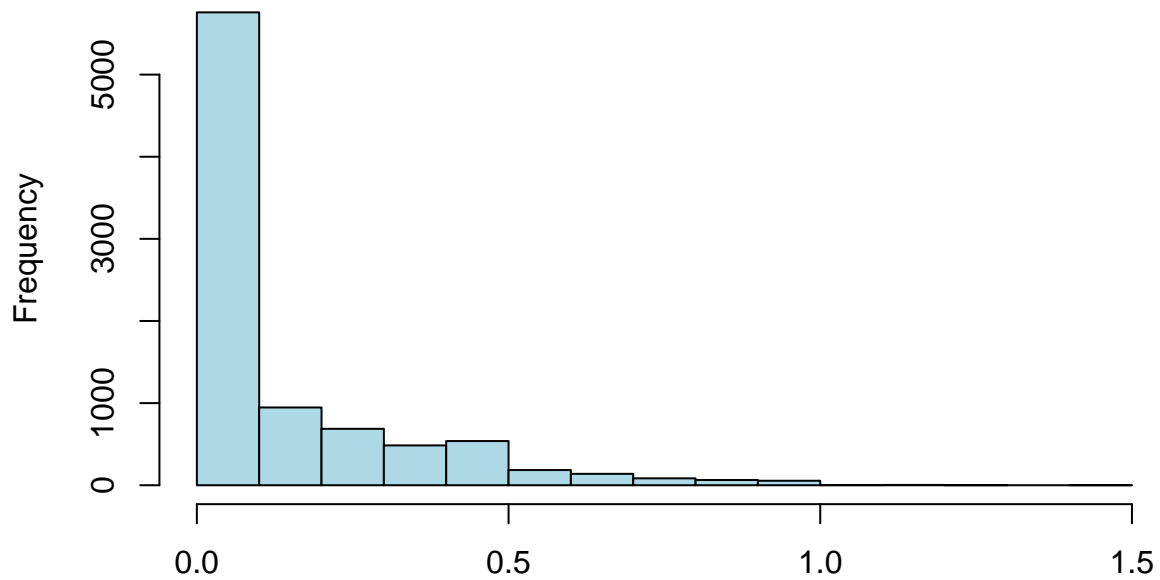
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

CASH_ADVANCE_FREQUENCY

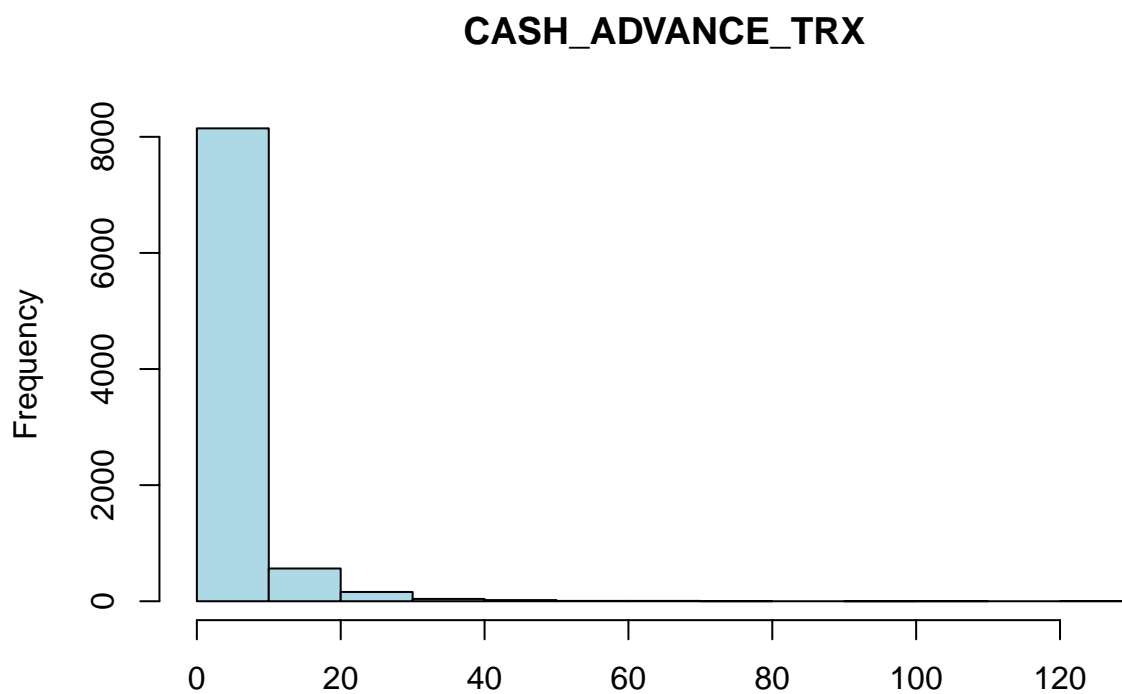


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```



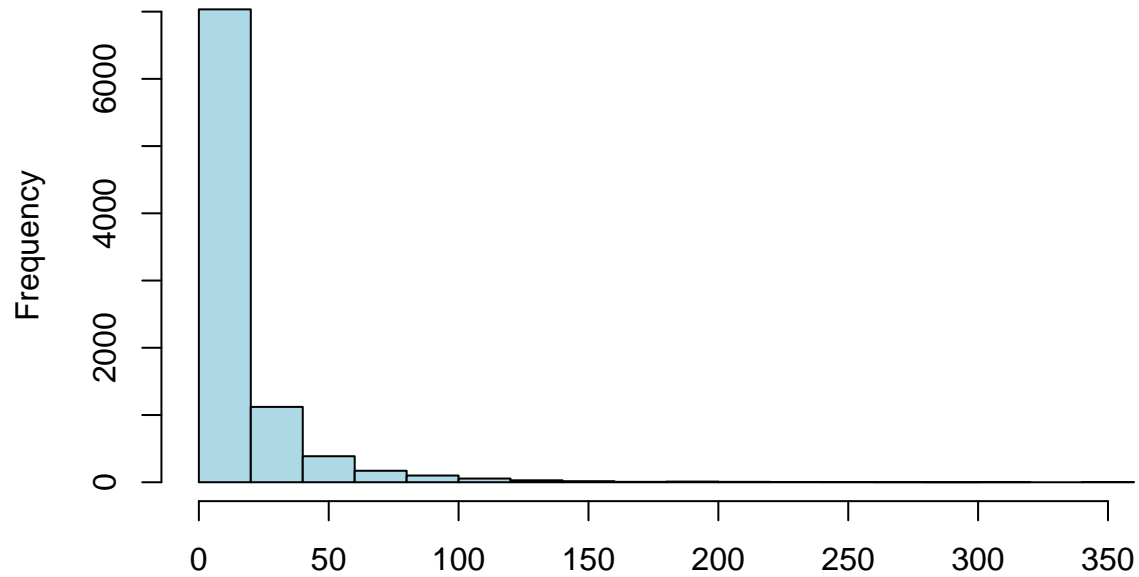
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

PURCHASES_TRX

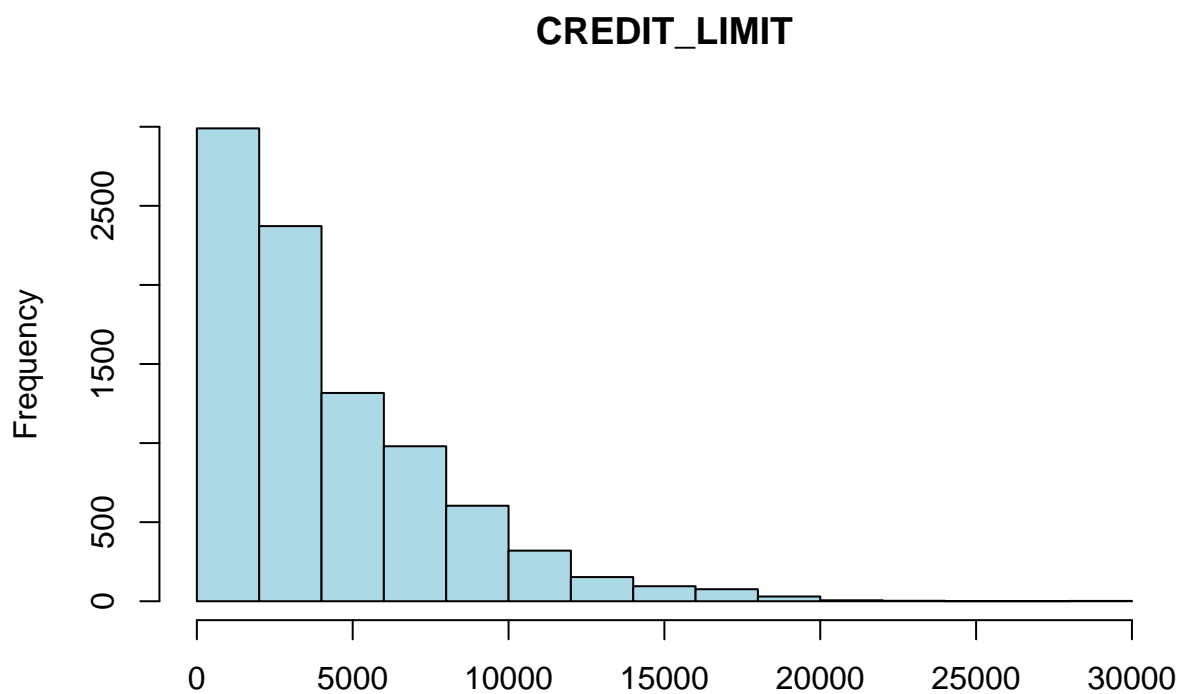


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

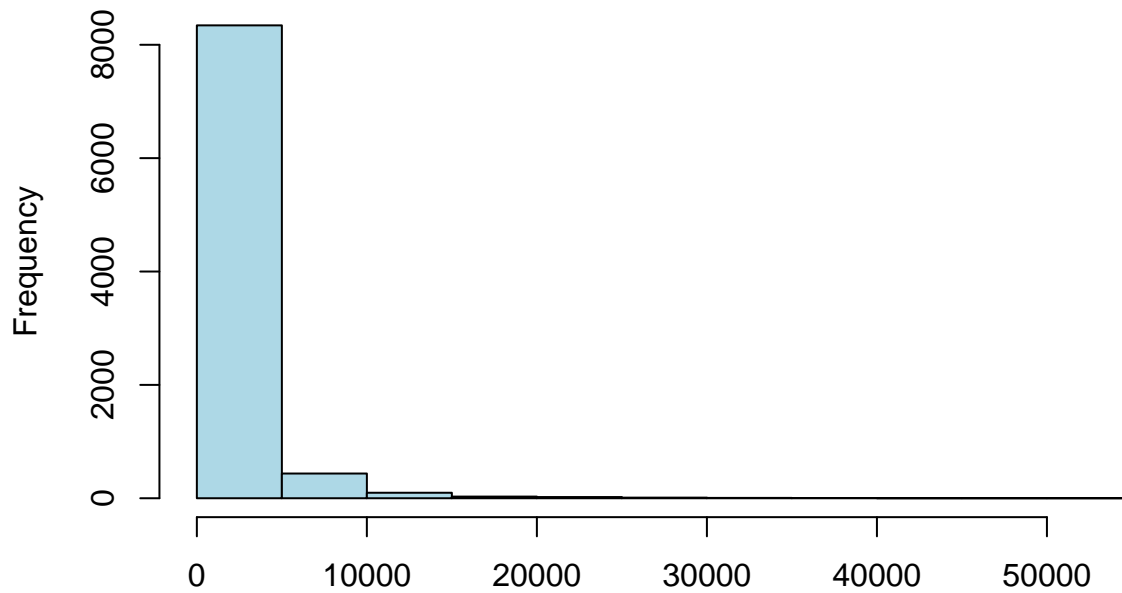
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

PAYMENTS

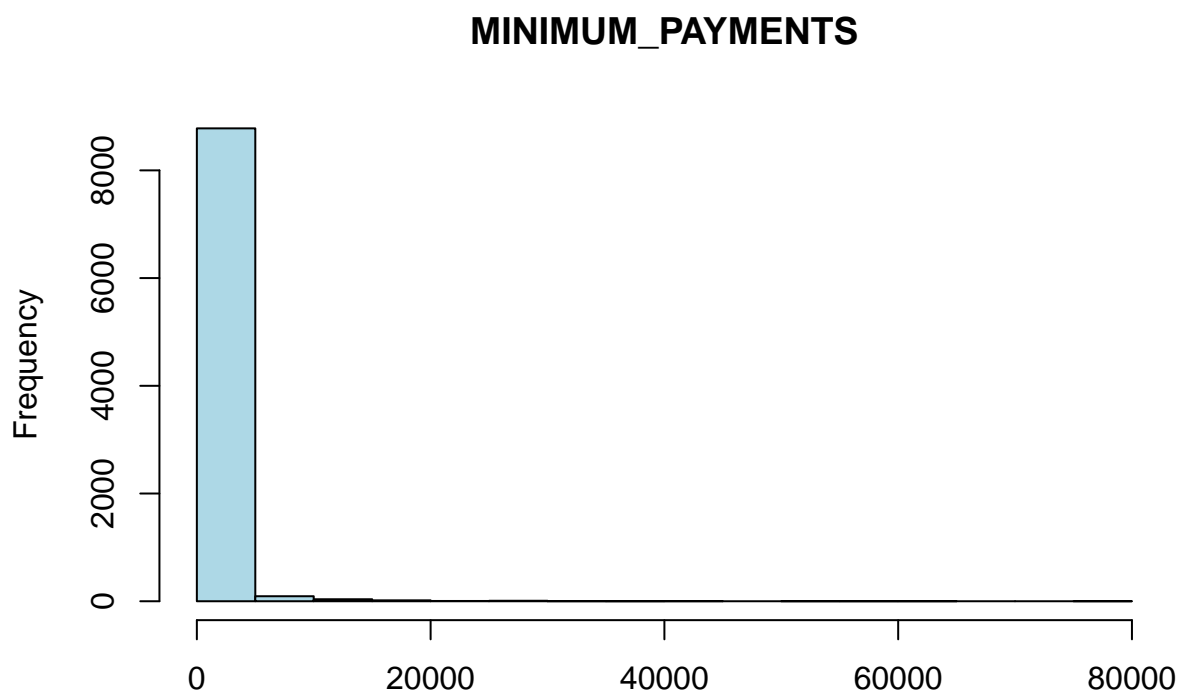


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

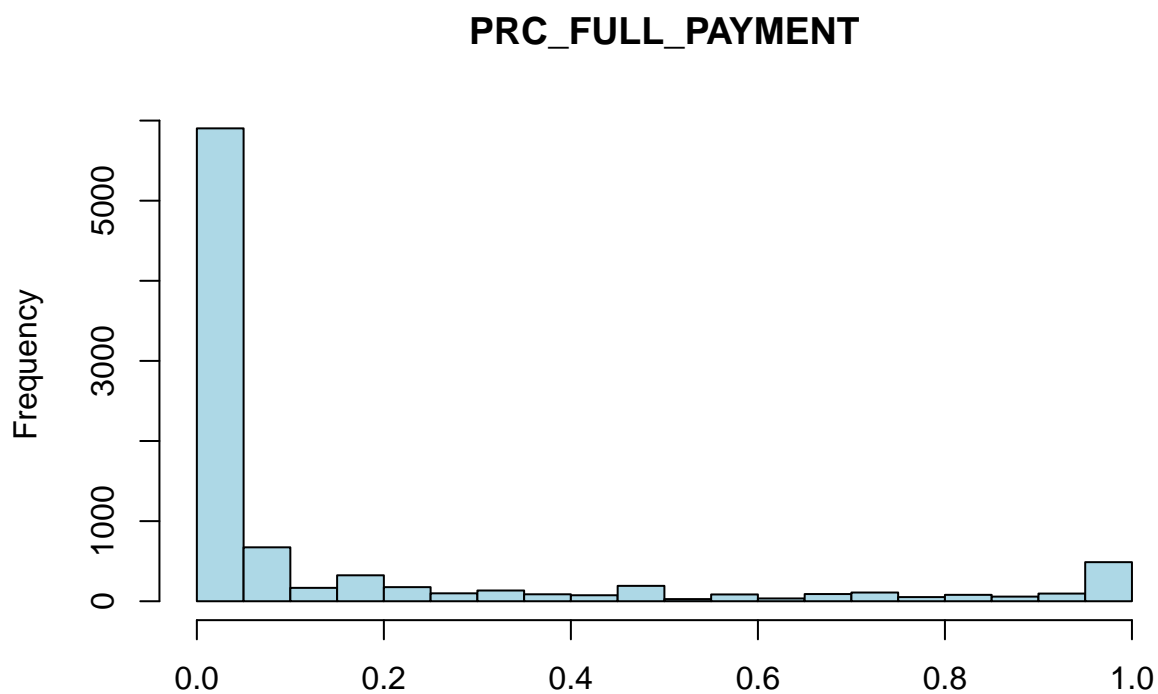


```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```



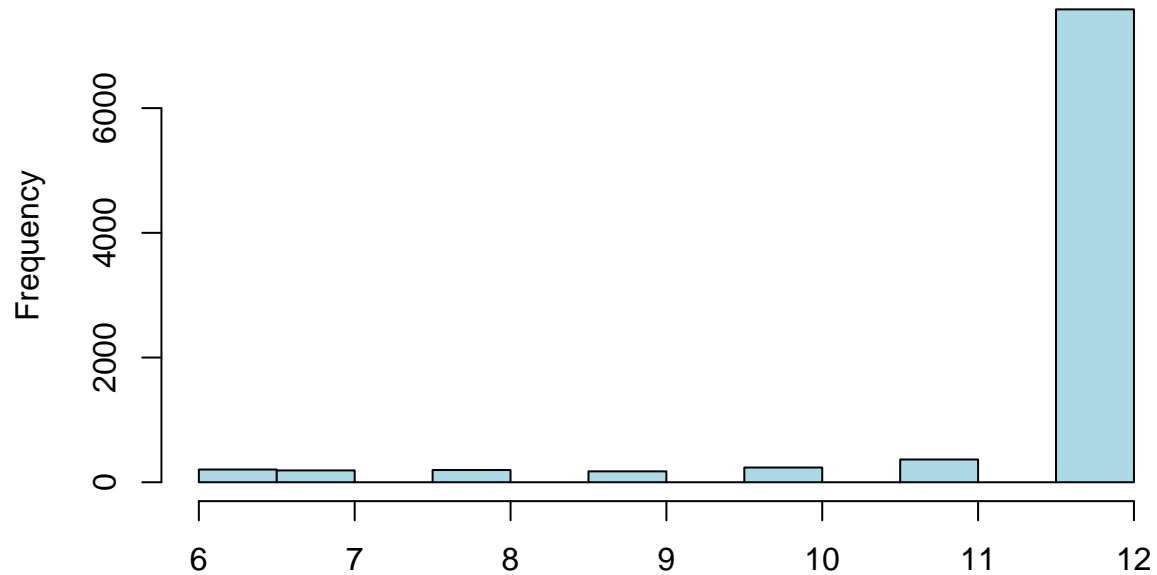
```
## Warning in plot.window(xlim, ylim, "", ...): graphical parameter "type" is  
## obsolete
```

```
## Warning in title(main = main, sub = sub, xlab = xlab, ylab = ylab, ...):  
## graphical parameter "type" is obsolete
```

```
## Warning in axis(1, ...): graphical parameter "type" is obsolete
```

```
## Warning in axis(2, ...): graphical parameter "type" is obsolete
```

TENURE



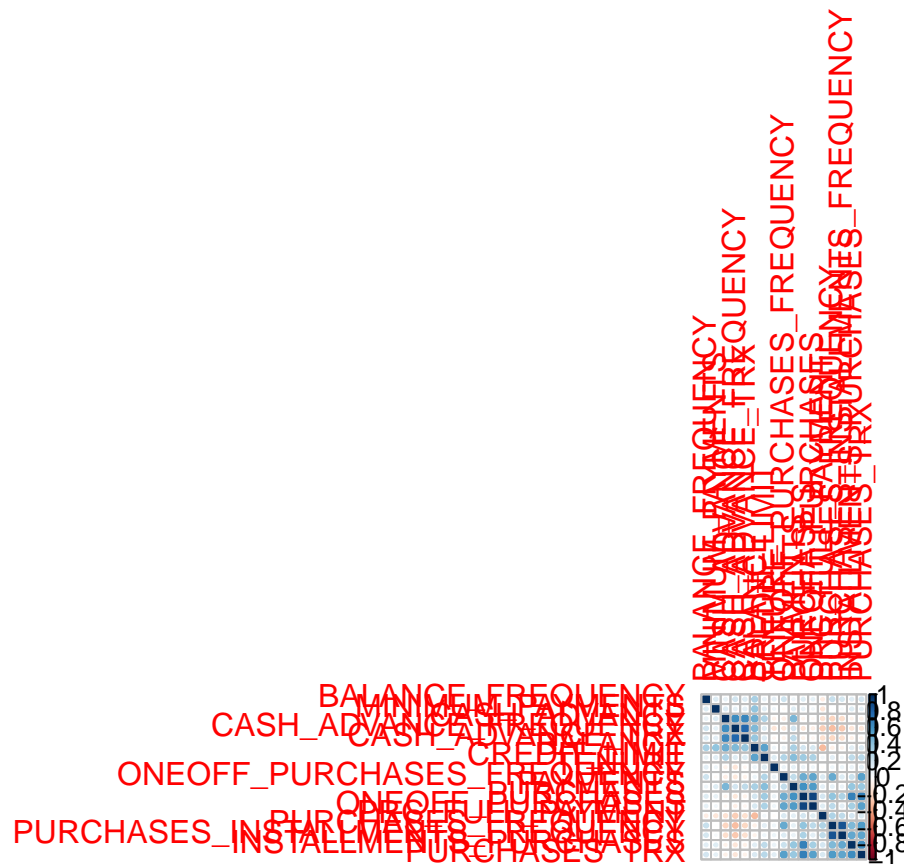
```
#Function to calculate the mean of each variable for each group/class
data.summ <- function(data, groups) {
  aggregate(data, list(groups), function(x) mean(as.numeric(x)))
}
```

```
#rescales the data for normalization to avoid biasness in the model
credit_data_scale<-scale(credit_data)
```

```
library(corrplot)
```

```
## corrplot 0.84 loaded
```

```
#Checking for correlation between variables
corrplot(cor(credit_data_scale [ , ]), order="hclust")
```



```
# creating principal components that will one of the input for the models
credit_data_pc<-prcomp(credit_data_scale)
```

```
library(cluster)
#rescales the data to avoid over proportion and then computes distance

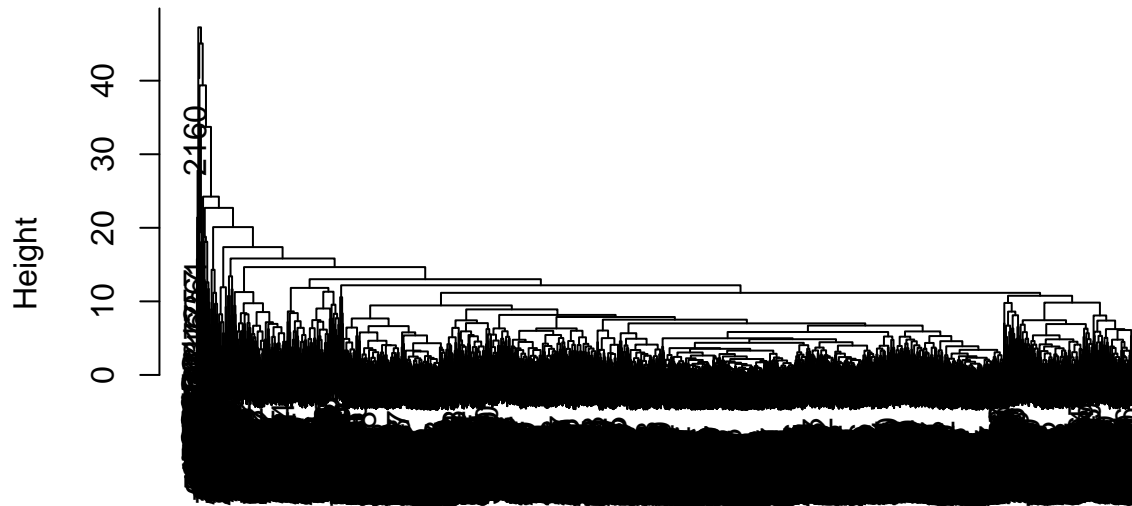
seg.dist<- daisy(credit_data_scale)

#Sample of distance matrix
as.matrix(seg.dist)[1:5,1:5]
```

```
##           1           2           3           4           5
## 1 0.0000000 4.393692 4.616890 2.455220 0.9694075
## 2 4.3936916 0.000000 5.630916 3.980908 4.1914567
## 3 4.6168904 5.630916 0.000000 4.242227 4.3428945
## 4 2.4552203 3.980908 4.242227 0.000000 2.6590634
## 5 0.9694075 4.191457 4.342894 2.659063 0.0000000
```

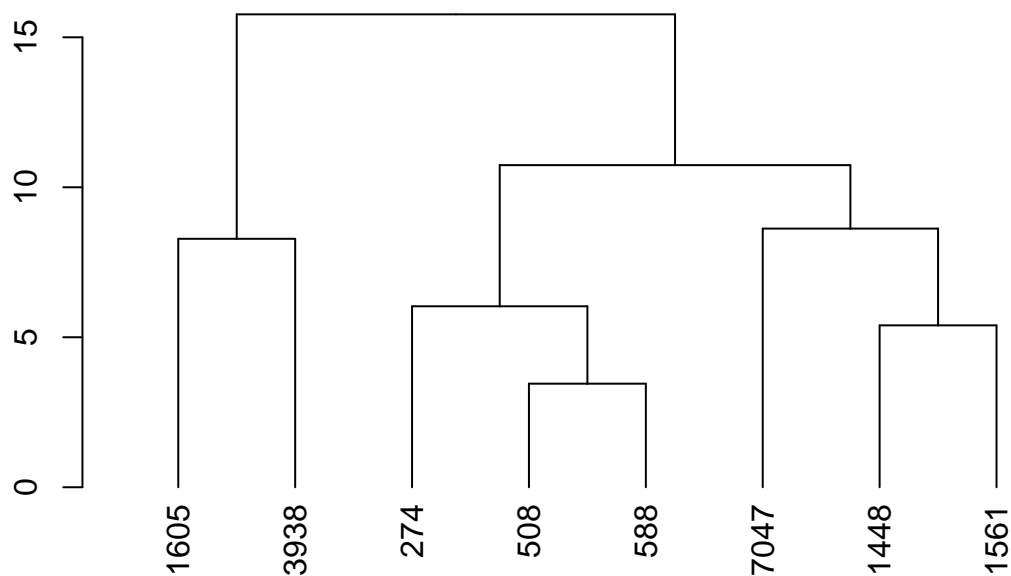
```
#using hclust to form a hierarchical tree by using distance matrix
seg.hc <- hclust(seg.dist, method="complete")
plot(seg.hc)
```

Cluster Dendrogram



```
seg.dist  
hclust (*, "complete")
```

```
plot(cut(as.dendrogram(seg.hc), h=20)$lower[[1]]) #we cut from the tree at a height of 20 for sam
```



```
credit_data.raw[c(508, 588), ]
```

```
##      X CUST_ID  BALANCE BALANCE_FREQUENCY PURCHASES ONEOFF_PURCHASES
## 508 508  C10529 2643.343                1 26402.39      22257.39
## 588 588  C10611 2492.726                1 27957.68      23032.97
##      INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 508                4145.00                0                1
## 588                4924.71                0                1
##      ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 508                1                0.333333
## 588                1                0.916667
##      CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS
## 508                0                0                114      16500 24529.28
## 588                0                0                70       9000 27255.01
##      MINIMUM_PAYMENTS PRC_FULL_PAYMENT TENURE
## 508        534.0323                1      12
## 588        537.3727                1      12
```

```
cor(cophenetic(seg.hc), seg.dist) #cpcc shows how well the clustering model reflects the distance matrix
```

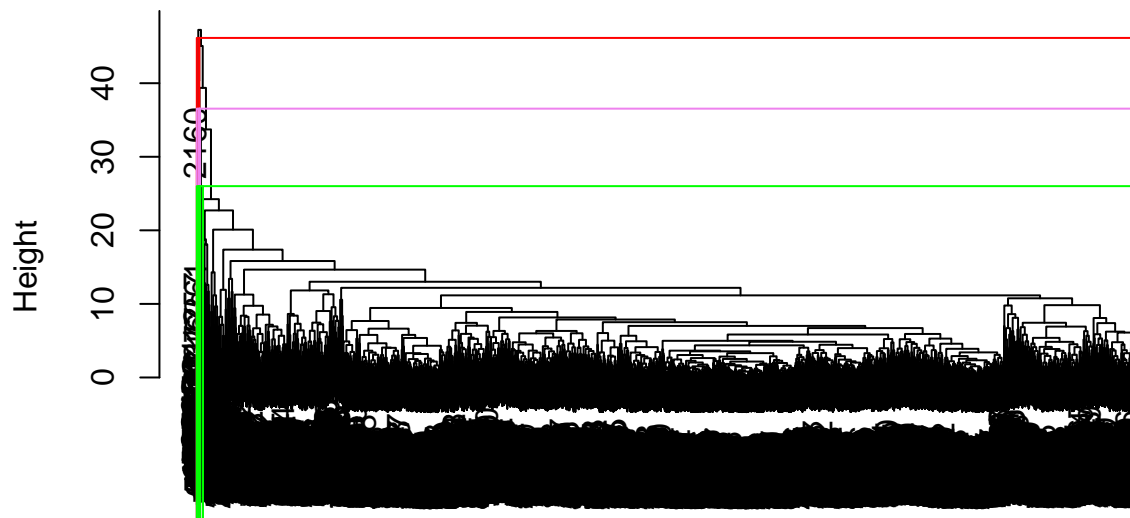
```
## [1] 0.8149523
```

```
# Plot to check on different cluster sizes
plot(seg.hc)
```



```
rect.hclust(seg.hc, k=2, border="red")
rect.hclust(seg.hc, k=4, border="violet")
rect.hclust(seg.hc, k=6, border="green")
```

Cluster Dendrogram



```
seg.dist
hclust (*, "complete")
```

```
#install.packages("factoextra")
library(factoextra)
```

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

```
##
```

```
## Attaching package: 'ggplot2'
```

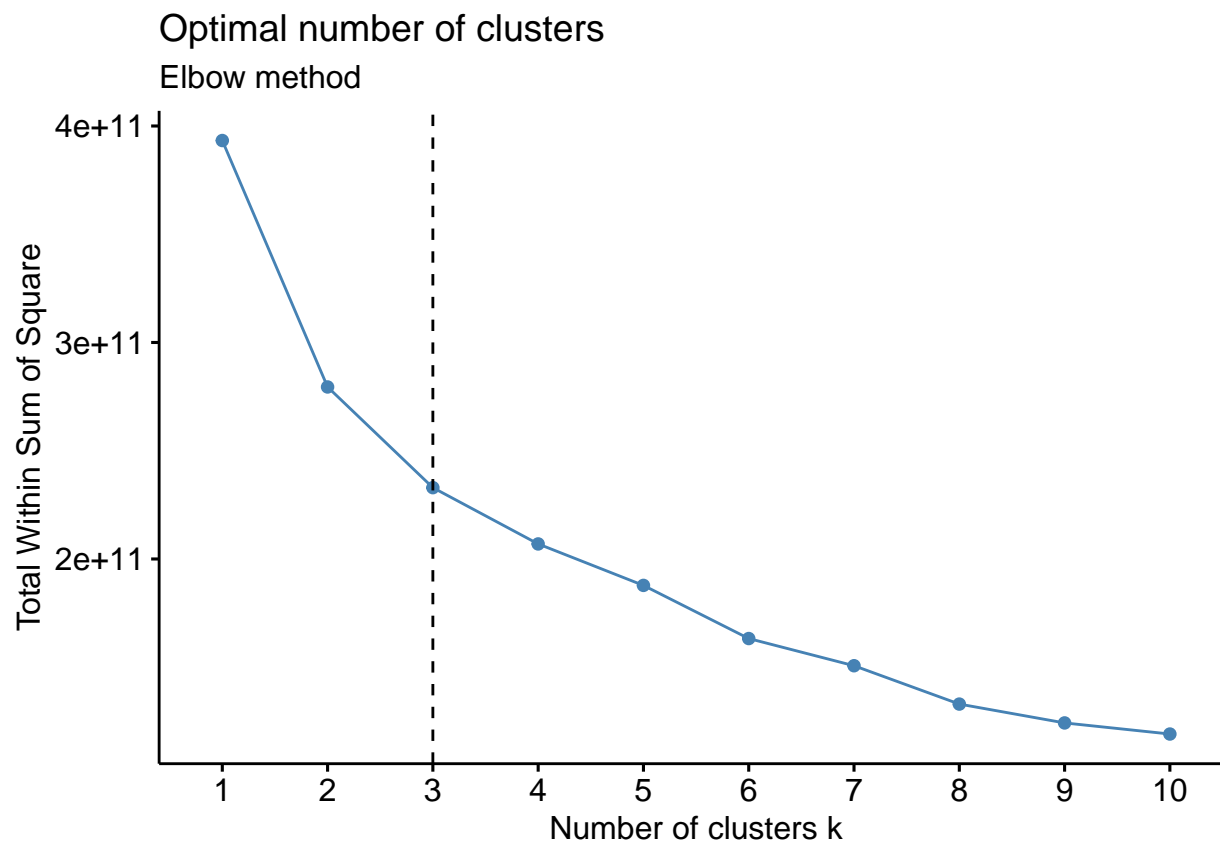
```
## The following objects are masked from 'package:psych':
```

```
##
```

```
## %+%, alpha
```

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

```
fviz_nbclust(credit_data, kmeans, method = "wss") +
geom_vline(xintercept = 3, linetype = 2)+
labs(subtitle = "Elbow method")
```



```
#Based on above we cut the tree for 3 groups
seg.hc.segment <- cutree(seg.hc, k=3) # membership vector for 3 groups
table(seg.hc.segment)
```

```
## seg.hc.segment
##      1      2      3
## 8926   23      1
```

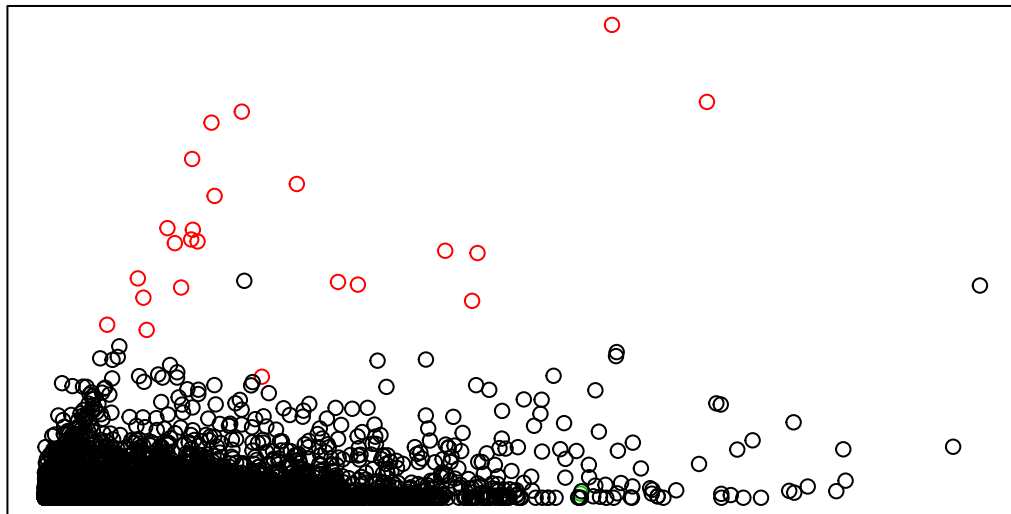
```
data.summ(credit_data, seg.hc.segment)
```

```
##   Group.1  BALANCE BALANCE_FREQUENCY PURCHASES ONEOFF_PURCHASES
## 1      1  1555.059         0.8770538   934.9797         536.2514
## 2      2  4812.383         0.9561265  27505.3396        22417.4522
## 3      3 10905.054         1.0000000   431.9300         133.5000
##   INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1              399.0293      972.0536         0.4892715
## 2              5087.8874     1617.7861         0.9050724
## 3              298.4300     47137.2118         0.5833330
##   ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 1              0.2007931              0.3635351
## 2              0.8463769              0.7086956
## 3              0.2500000              0.5000000
##   CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS
## 1              0.13523683         3.237060      14.42718      4463.11  1660.922
## 2              0.06159417         2.608696     124.13043     16000.00 28138.985
```

```
## 3          1.00000000      123.000000      21.00000      19600.00 39048.598
##  MINIMUM_PAYMENTS PRC_FULL_PAYMENT    TENURE
## 1           828.9243          0.1527534 11.51624
## 2          2599.0909          0.5334321 11.91304
## 3          5394.1737          0.0000000 12.00000
```

```
plot(jitter(as.numeric(credit_data$PURCHASES)) ~ jitter(as.numeric(credit_data$BALANCE))),
```

col=seg

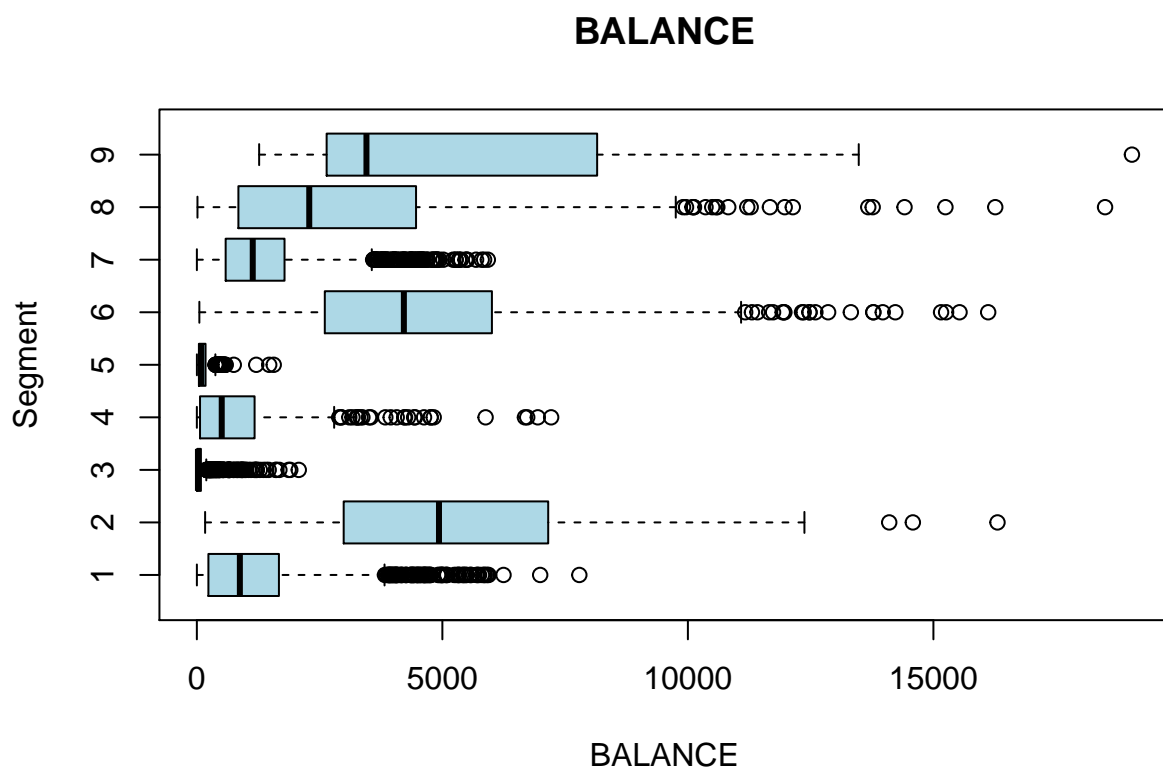


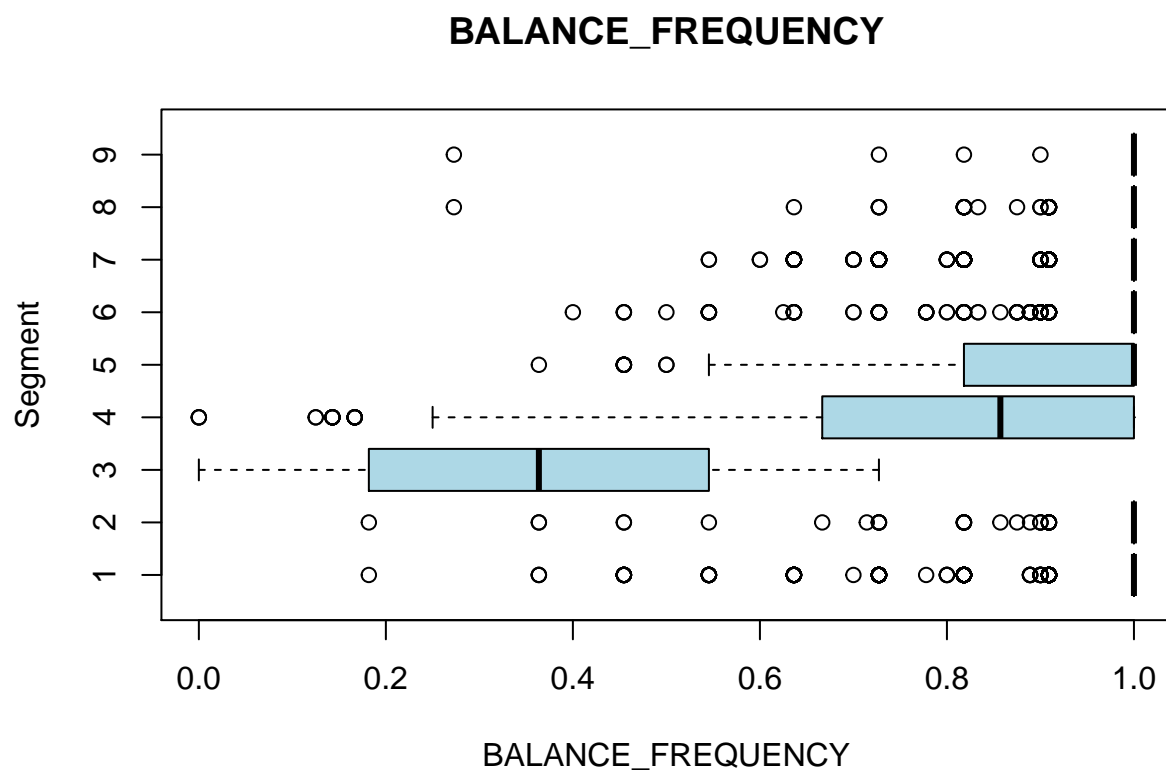
```
#k-Means method
# number of k group was selected after multiple iteration and also referring the mclust
seg.k <- kmeans(credit_data_scale, centers=9)
data.summ(credit_data, seg.k$cluster)
```

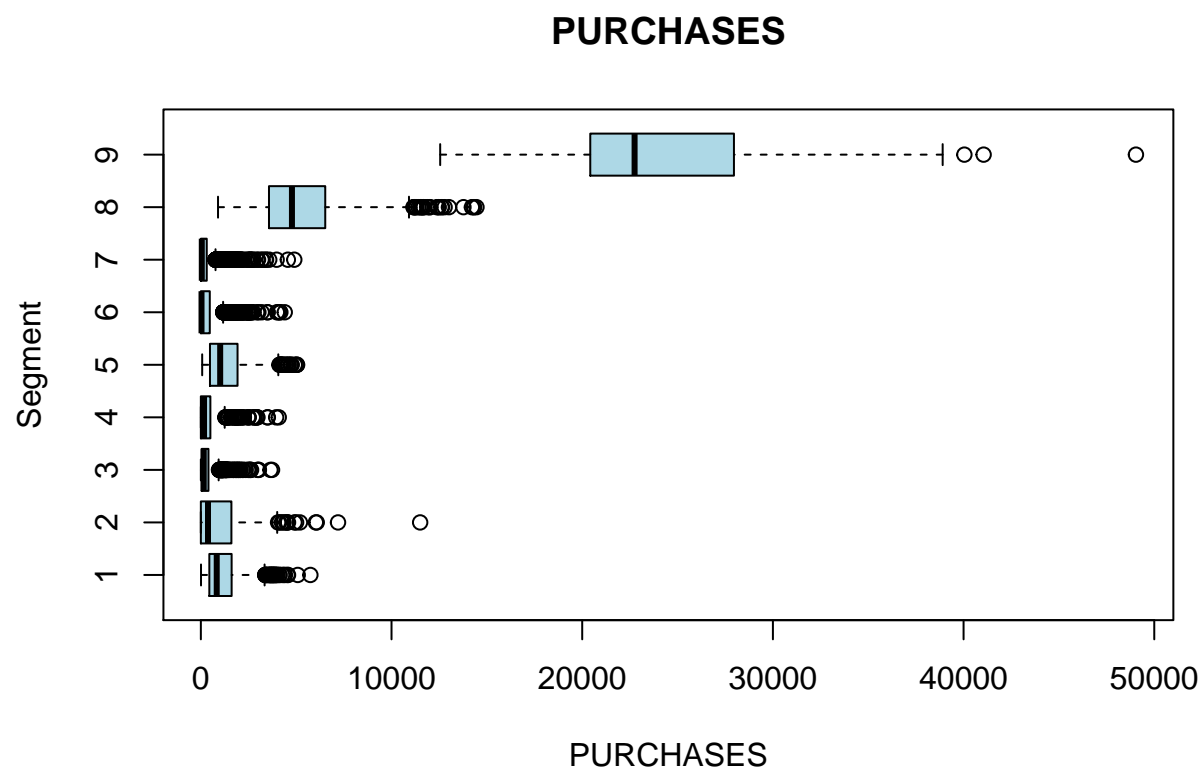
```
##  Group.1  BALANCE BALANCE_FREQUENCY PURCHASES ONEOFF_PURCHASES
## 1      1 1201.6917          0.9707153 1105.8495          512.9371
## 2      2 5260.6708          0.9524638 1134.5458          665.1059
## 3      3 106.9111          0.3523941  315.0669          190.2356
## 4      4 851.6887          0.7942606  385.8174          247.9074
## 5      5 129.2708          0.9111572 1319.5333          616.3755
## 6      6 4553.9713          0.9758161  366.0156          240.7264
## 7      7 1340.2645          0.9708072  249.3716          205.6638
## 8      8 3127.3023          0.9893151 5399.2597          3463.3085
## 9      9 5519.2375          0.9557994 25421.9383         18707.4228
##  INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1              593.29866      288.70767          0.8757008
```

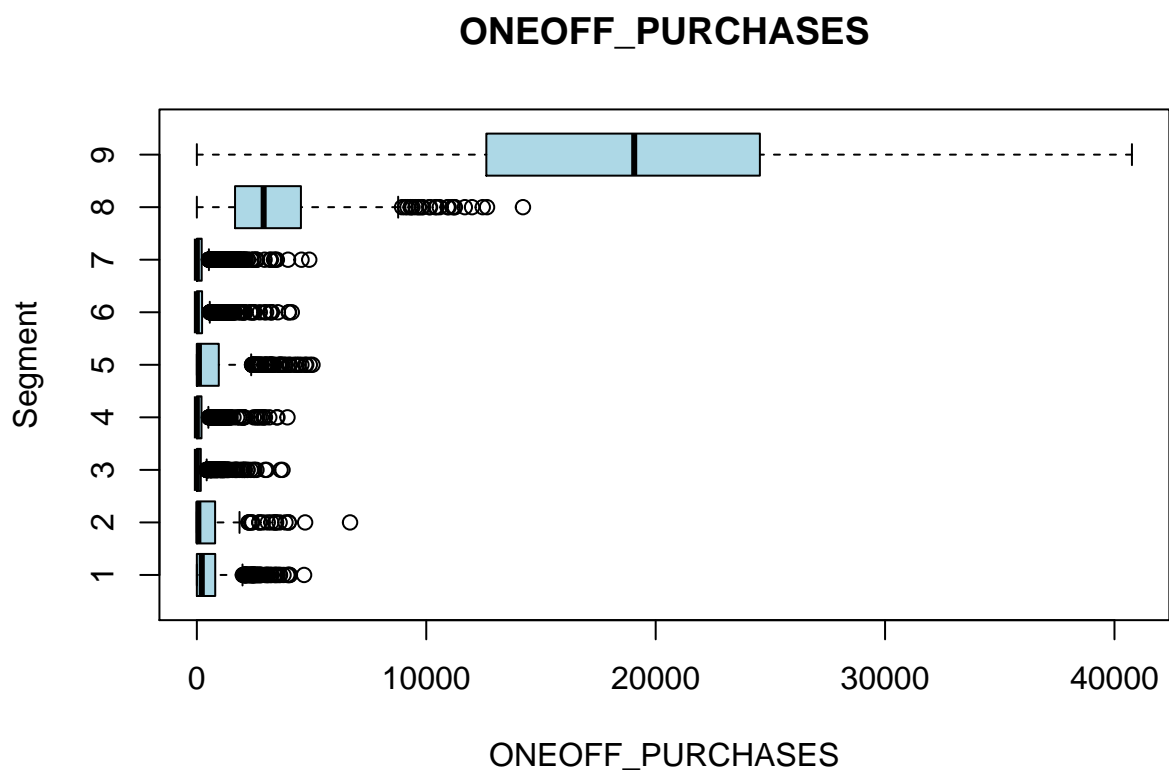
## 2	469.43988	10098.83863	0.4129682		
## 3	125.09922	321.05182	0.2586243		
## 4	138.39780	1108.57517	0.4136024		
## 5	703.25420	37.76055	0.8601567		
## 6	125.38726	3454.24888	0.2391794		
## 7	43.91036	609.35475	0.1493029		
## 8	1937.04611	565.33769	0.9550183		
## 9	6714.51552	1401.09676	0.9074712		
##	ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY				
## 1	0.27980488	0.71137786			
## 2	0.21602255	0.30573573			
## 3	0.06553617	0.18609265			
## 4	0.12202090	0.27318235			
## 5	0.31444365	0.65863298			
## 6	0.11463645	0.14389397			
## 7	0.08996138	0.05837768			
## 8	0.72087600	0.79305663			
## 9	0.76609200	0.74597700			
##	CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS				
## 1	0.05513745	1.0725264	21.305476	3848.964	1279.8600
## 2	0.64488215	29.4848485	16.339394	9354.848	10643.1679
## 3	0.03287452	0.7193141	4.097473	3798.795	1133.8009
## 4	0.19492899	3.2048780	5.126829	2415.471	594.2495
## 5	0.00651996	0.1213434	20.752979	5044.344	1430.4664
## 6	0.44137626	11.5699132	5.438766	7292.001	2334.1259
## 7	0.13281853	2.4686860	2.715514	3082.610	929.3490
## 8	0.07724783	1.8394161	75.447080	8708.029	4968.3909
## 9	0.05172410	2.1379310	142.482759	15624.138	25353.2578
##	MINIMUM_PAYMENTS PRC_FULL_PAYMENT TENURE				
## 1	858.3048	0.05698105	11.901057		
## 2	2194.0887	0.10679539	11.575758		
## 3	131.1111	0.18977085	11.784296		
## 4	326.1832	0.14496709	7.222764		
## 5	183.6771	0.78976752	11.771398		
## 6	2122.6107	0.02097680	11.659595		
## 7	663.0414	0.02311058	11.889889		
## 8	1598.7898	0.21663564	11.941606		
## 9	3539.1300	0.48915883	11.931034		

```
for ( i in seq(1,length(credit_data ),1) ) boxplot(credit_data[,i] ~ seg.k$cluster, main=names(credit_d
```

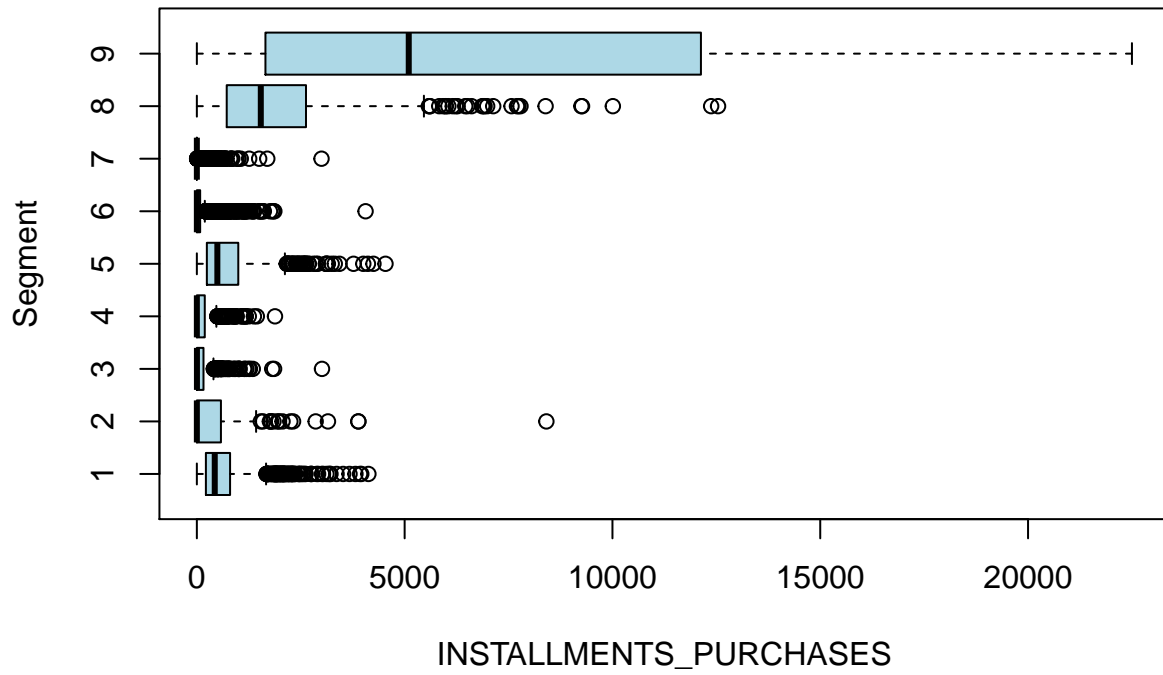


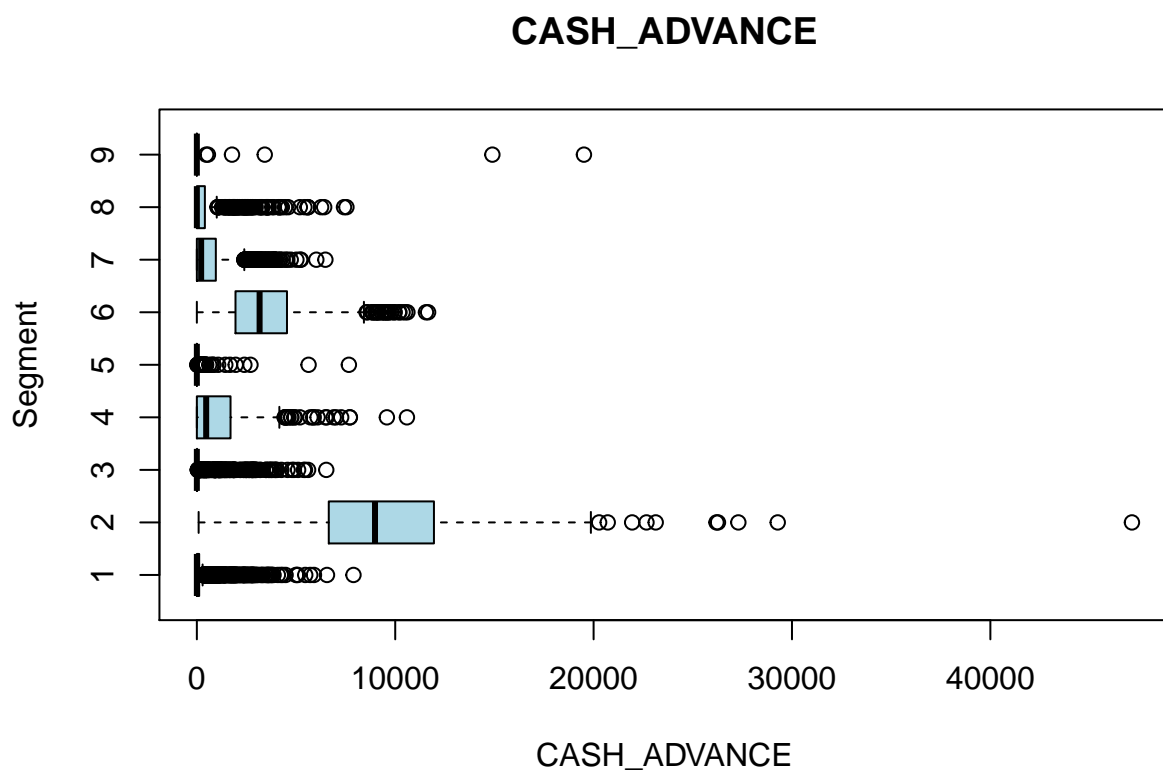




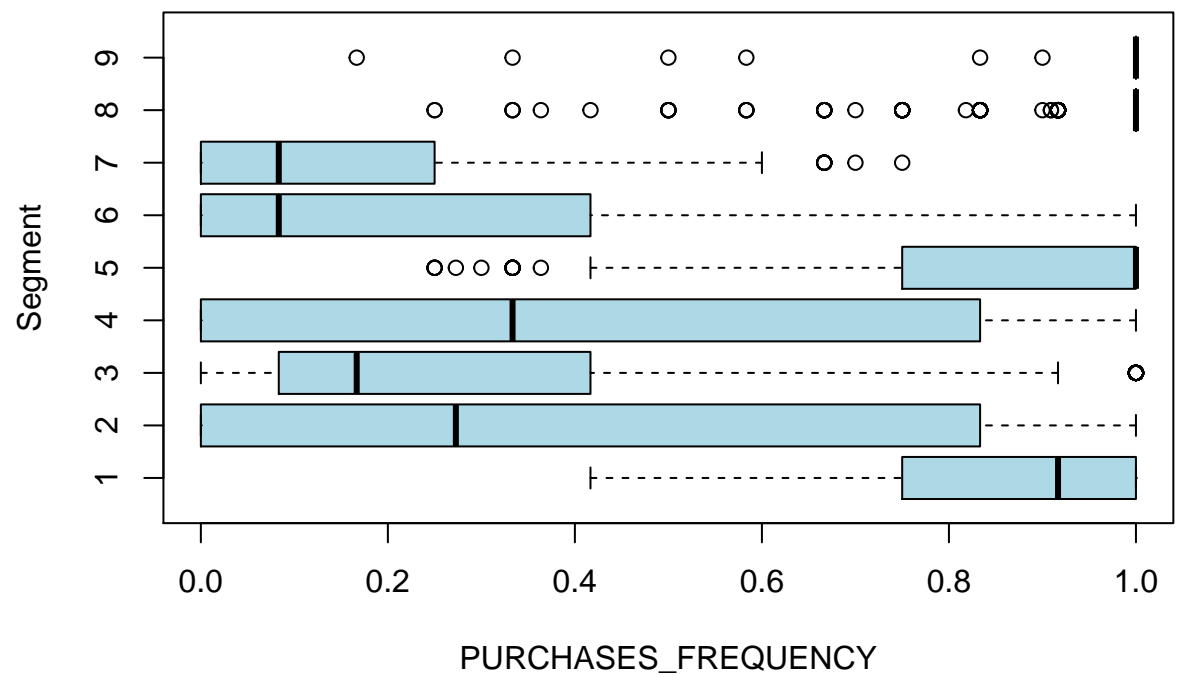


INSTALLMENTS_PURCHASES

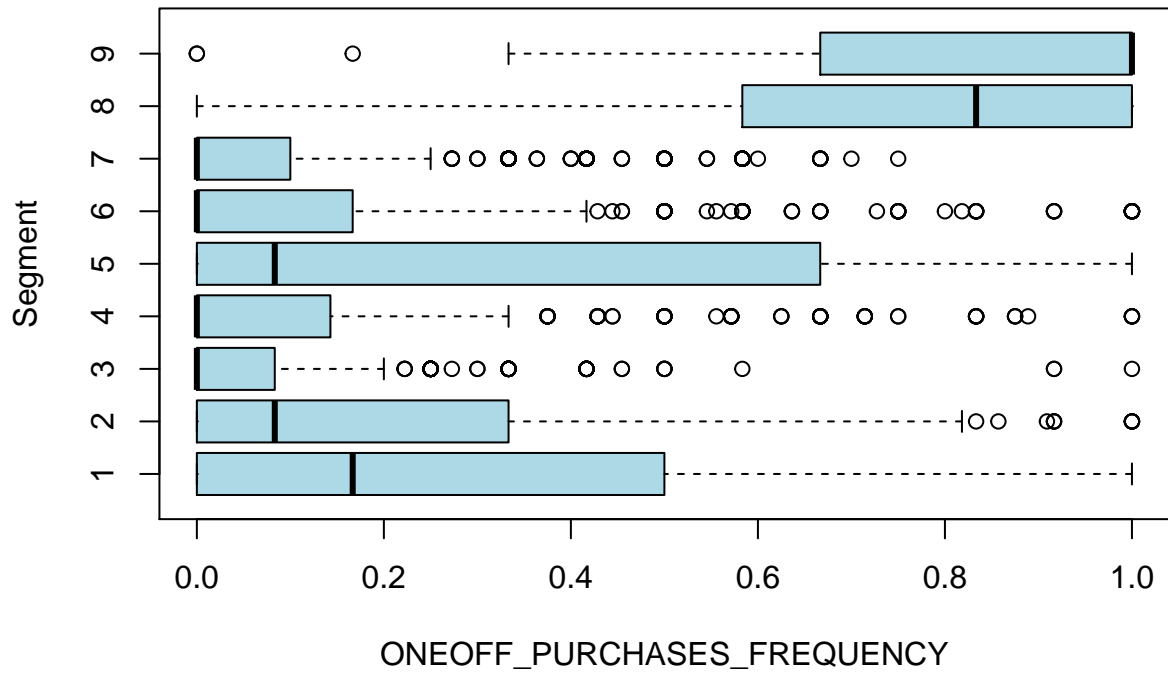




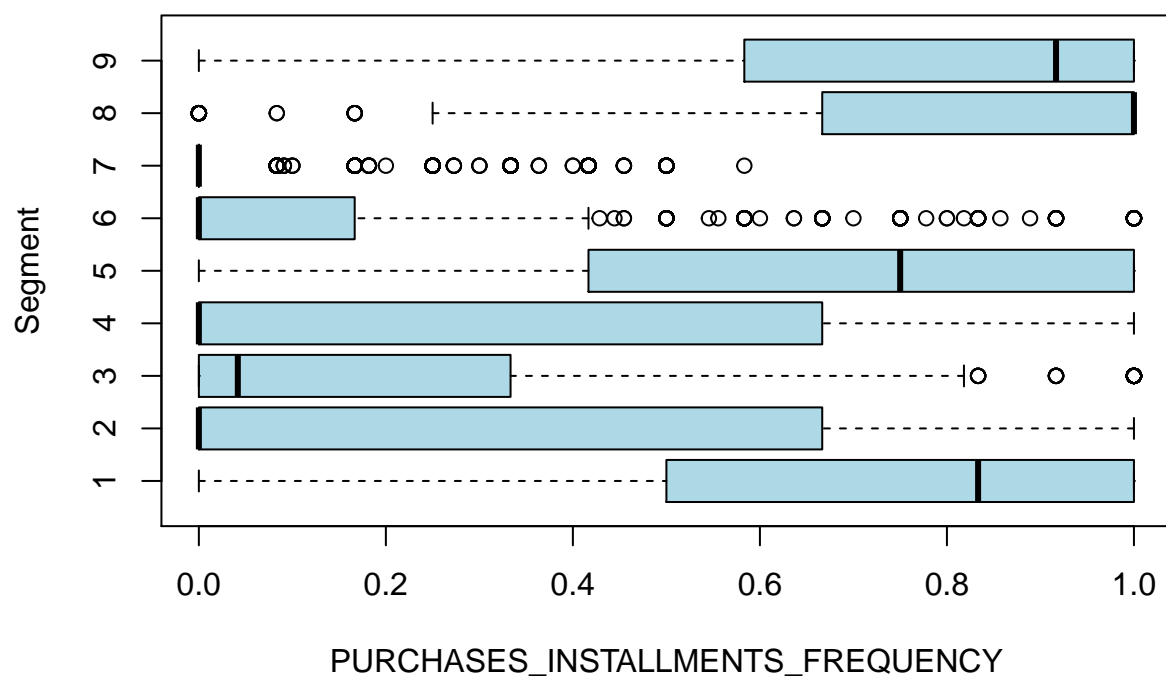
PURCHASES_FREQUENCY



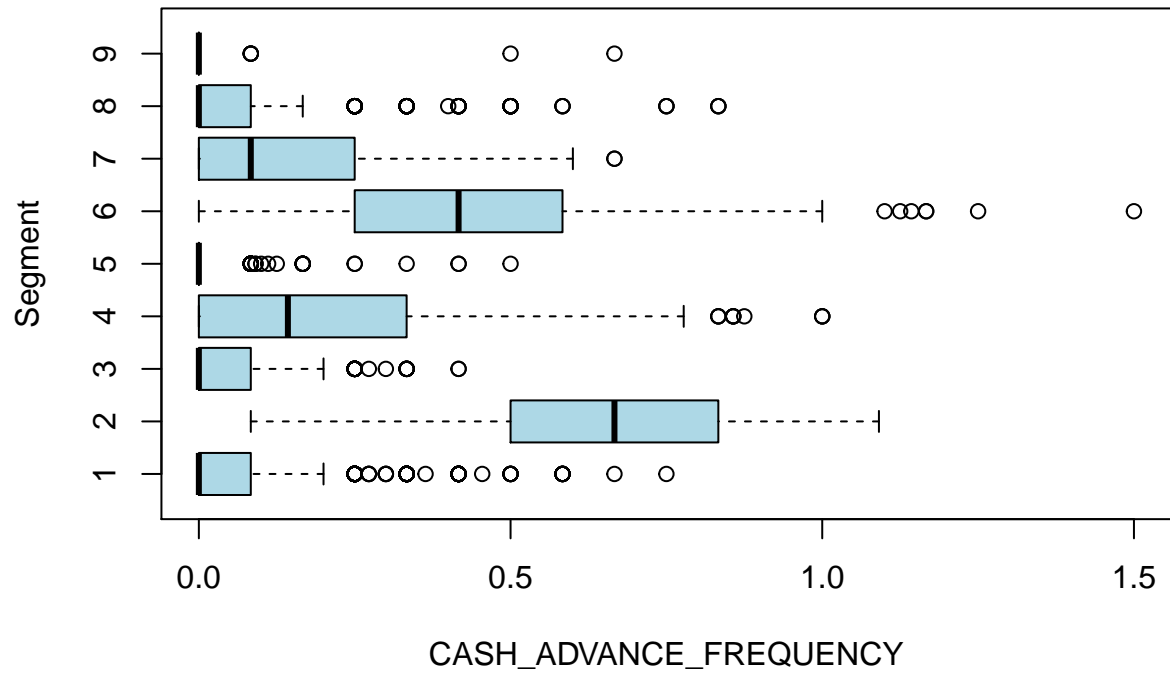
ONEOFF_PURCHASES_FREQUENCY



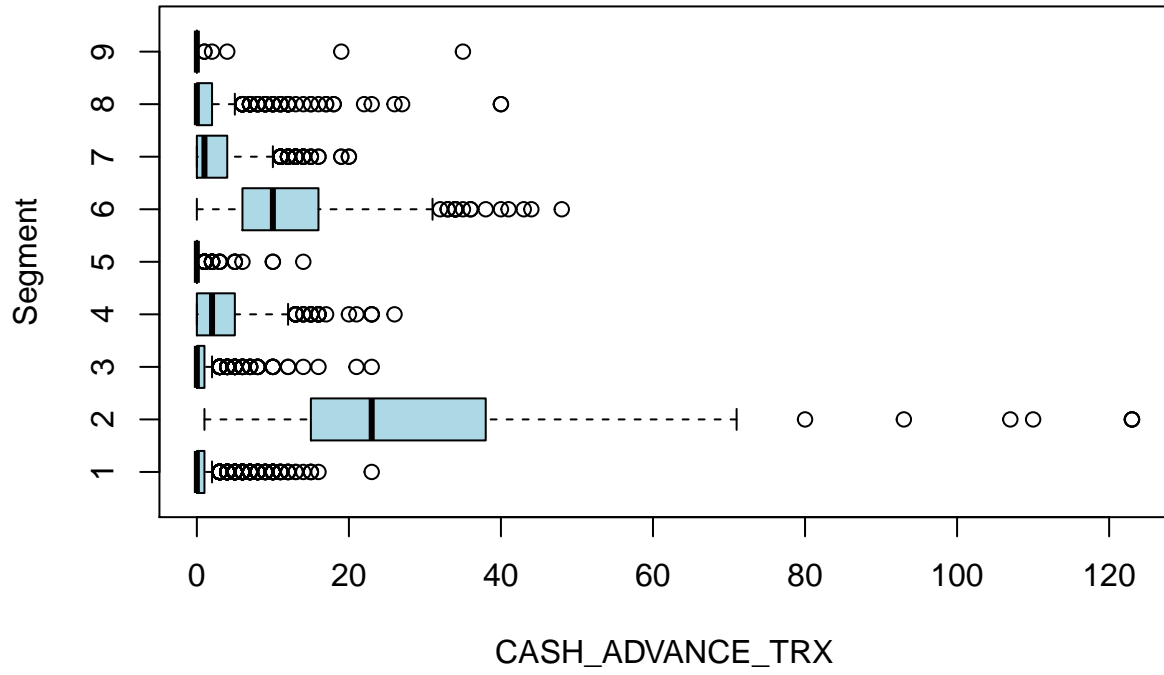
PURCHASES_INSTALLMENTS_FREQUENCY

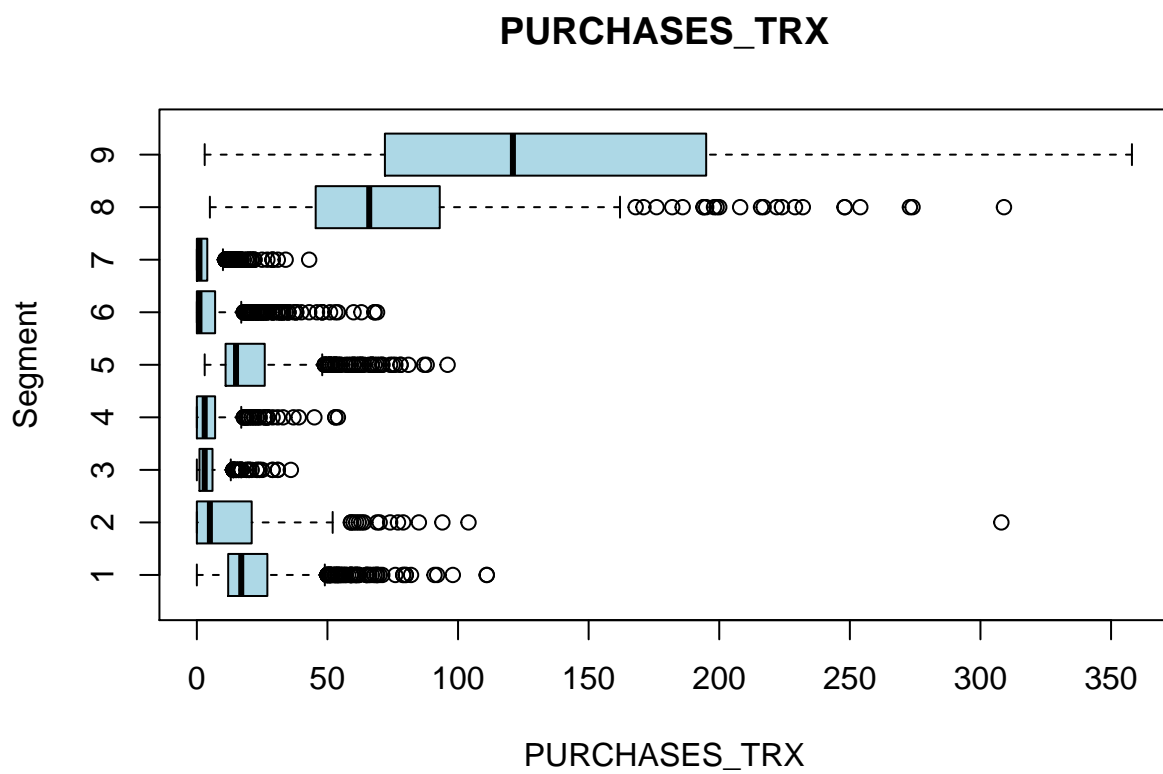


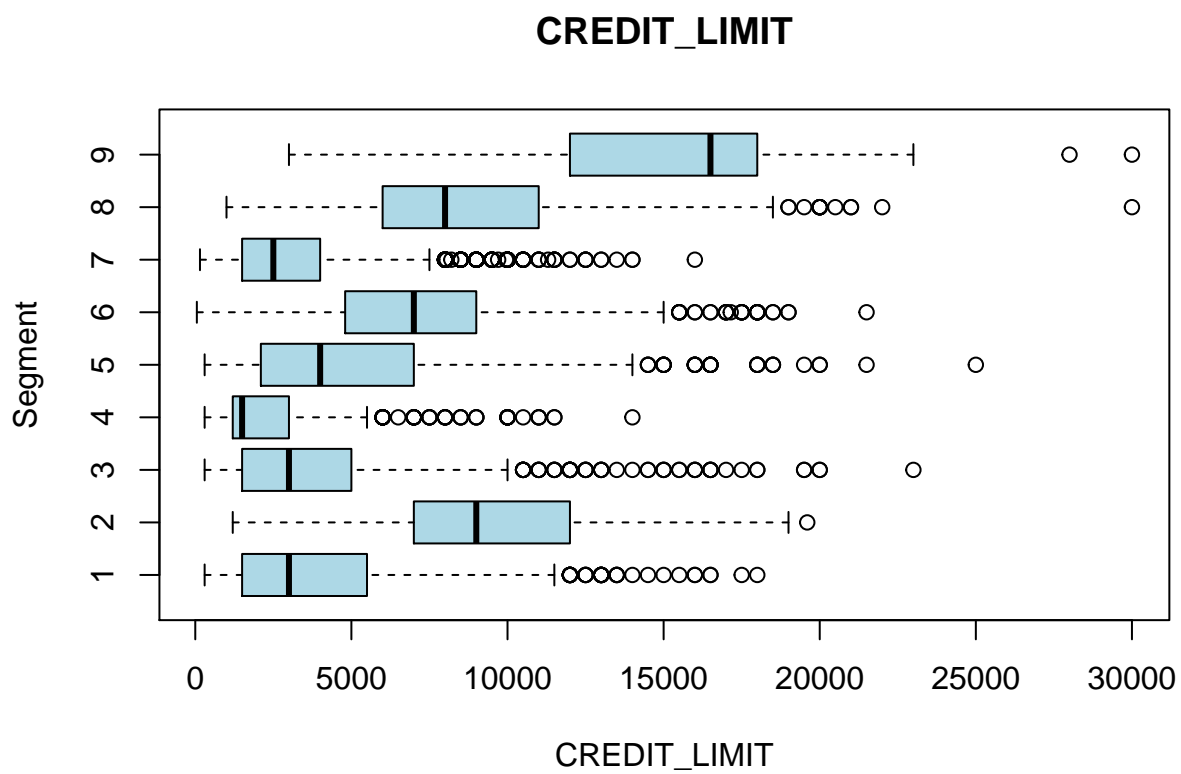
CASH_ADVANCE_FREQUENCY



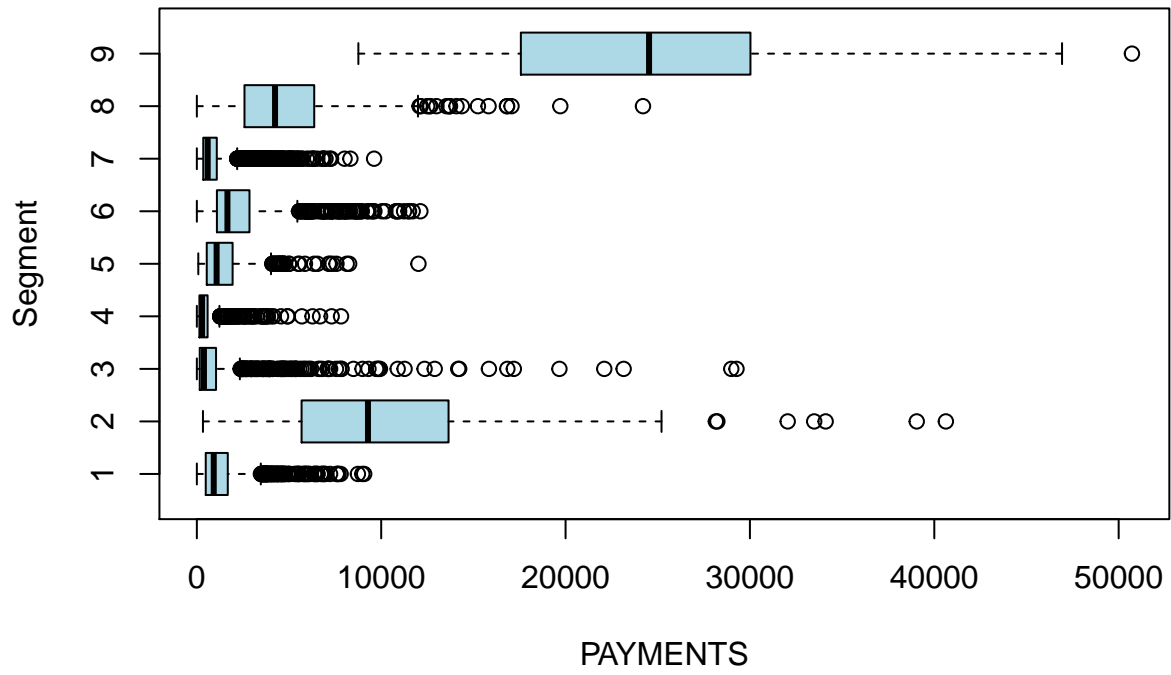
CASH_ADVANCE_TRX

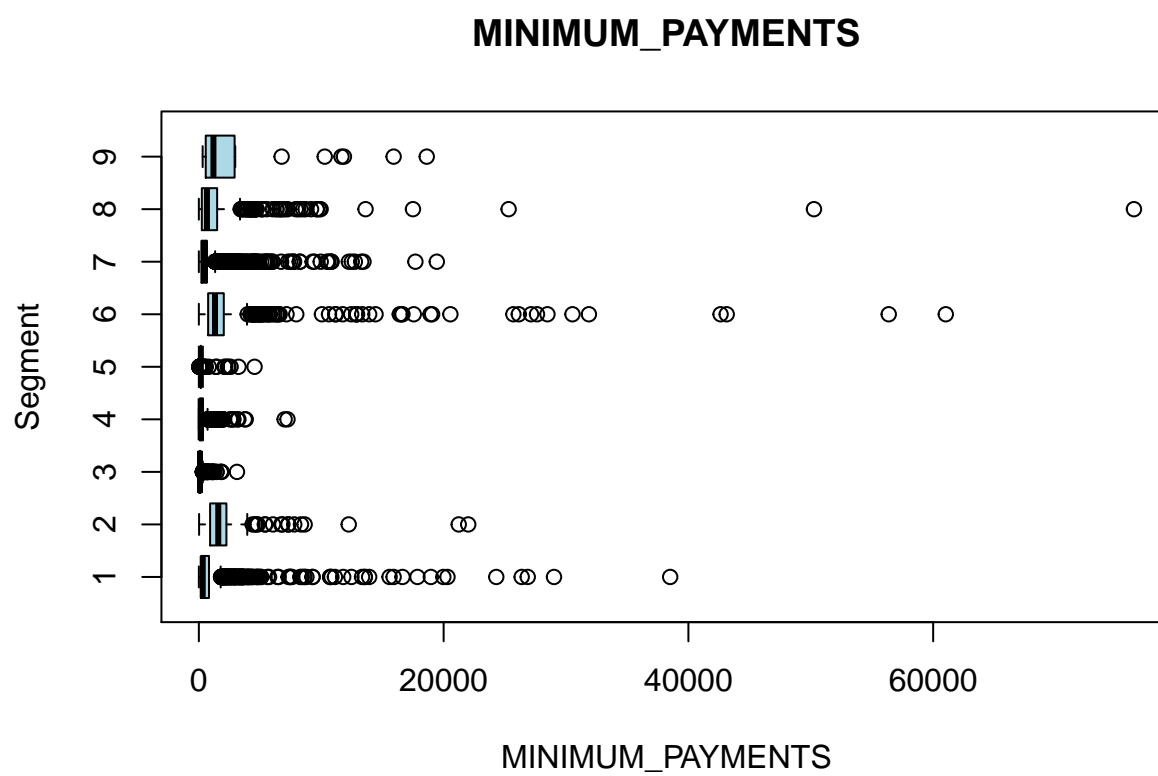


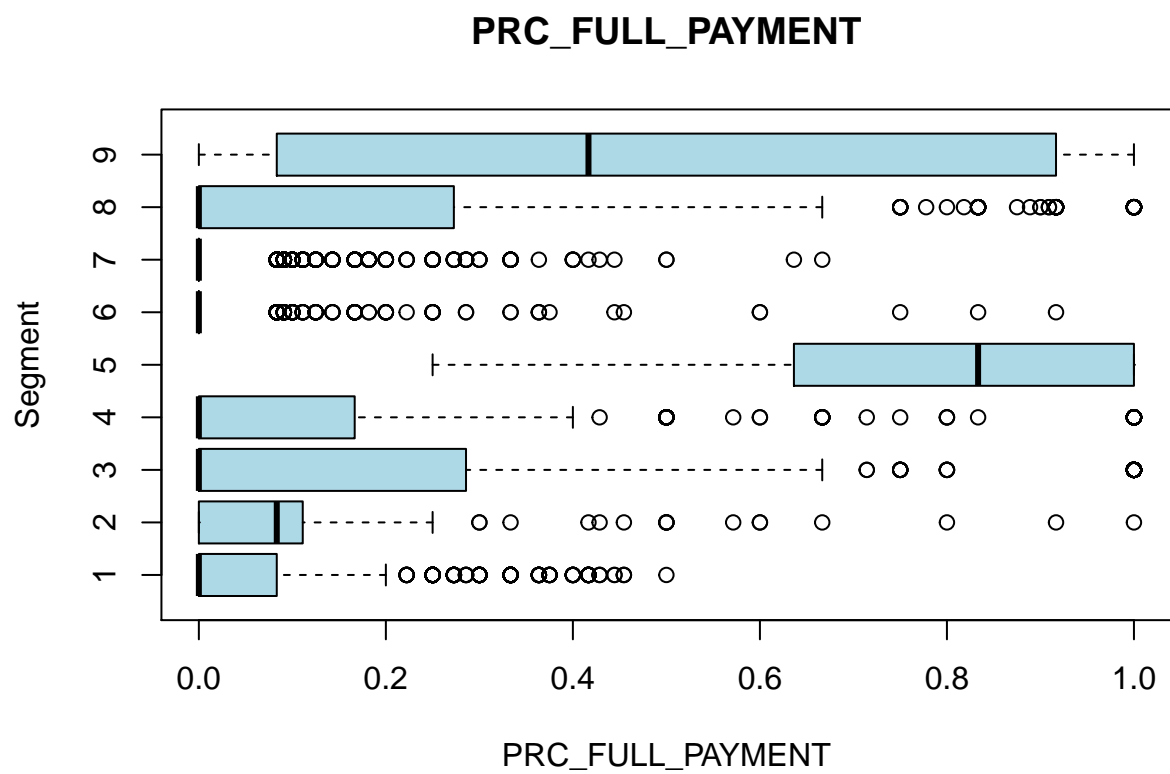


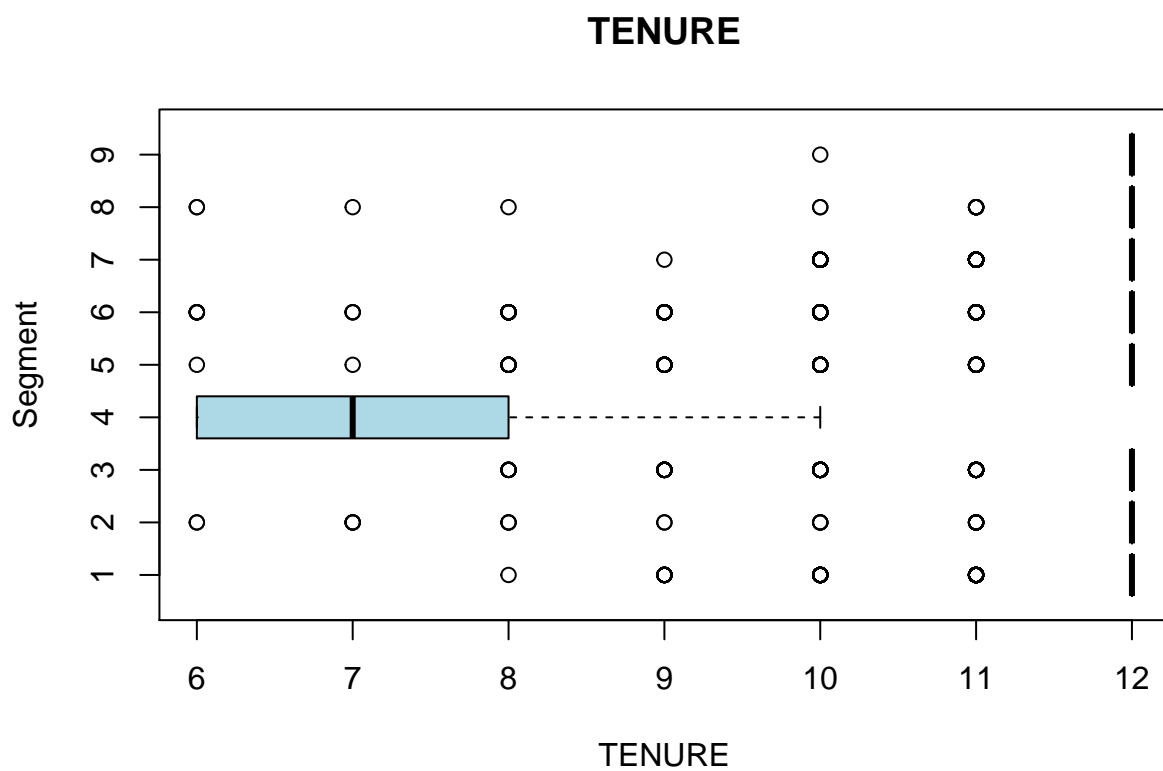


PAYMENTS



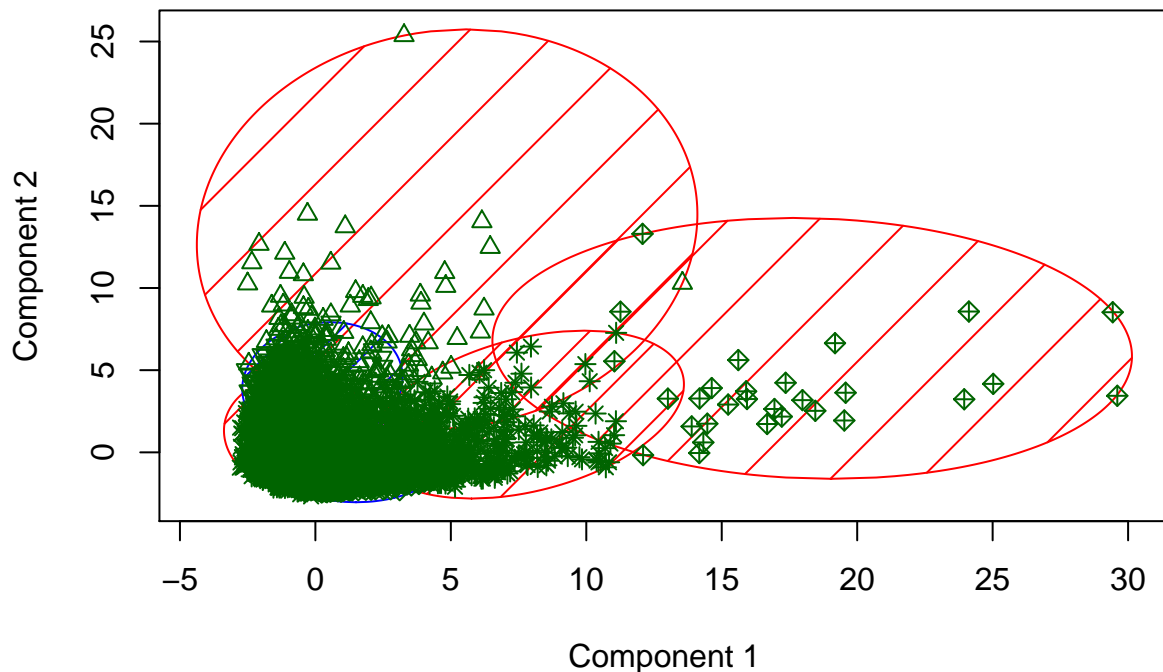






```
# cluster plot to see the observations in the group
clusplot(credit_data, seg.k$cluster, color=TRUE, shade=TRUE, labels=9, lines=0, main="K-means cluster p
```

K-means cluster plot



These two components explain 47.62 % of the point variability.

```
summary(seg.k)
```

```
##           Length Class  Mode
## cluster    8950   -none- numeric
## centers     153   -none- numeric
## totss        1   -none- numeric
## withinss     9   -none- numeric
## tot.withinss 1   -none- numeric
## betweenss    1   -none- numeric
## size         9   -none- numeric
## iter         1   -none- numeric
## ifault       1   -none- numeric
```

```
#Mclust method
```

```
# setting up random seed to remove the biasness of model to select same observations in the sample
set.seed(96743)
#install.packages("mclust")
library(mclust)
```

```
## Package 'mclust' version 5.4.6
## Type 'citation("mclust")' for citing this R package in publications.
```

```
##
## Attaching package: 'mclust'
```

```
## The following object is masked from 'package:psych':
##
##      sim
```

```
#Running mclust on data before scaling
seg.mc_data <- Mclust(credit_data)
#Running mclust on data after scaling
seg.mc_scale <- Mclust(credit_data_scale)
#Running mclust with rincipal component analysis
seg.mc_pc<- Mclust(credit_data_pc$x)
#Running mclust with overridden group=4
seg.mc_scale4 <- Mclust(credit_data_scale, G=4)

#comparing above models for statistical significance
BIC(seg.mc_data,seg.mc_scale,seg.mc_pc,seg.mc_scale4)
```

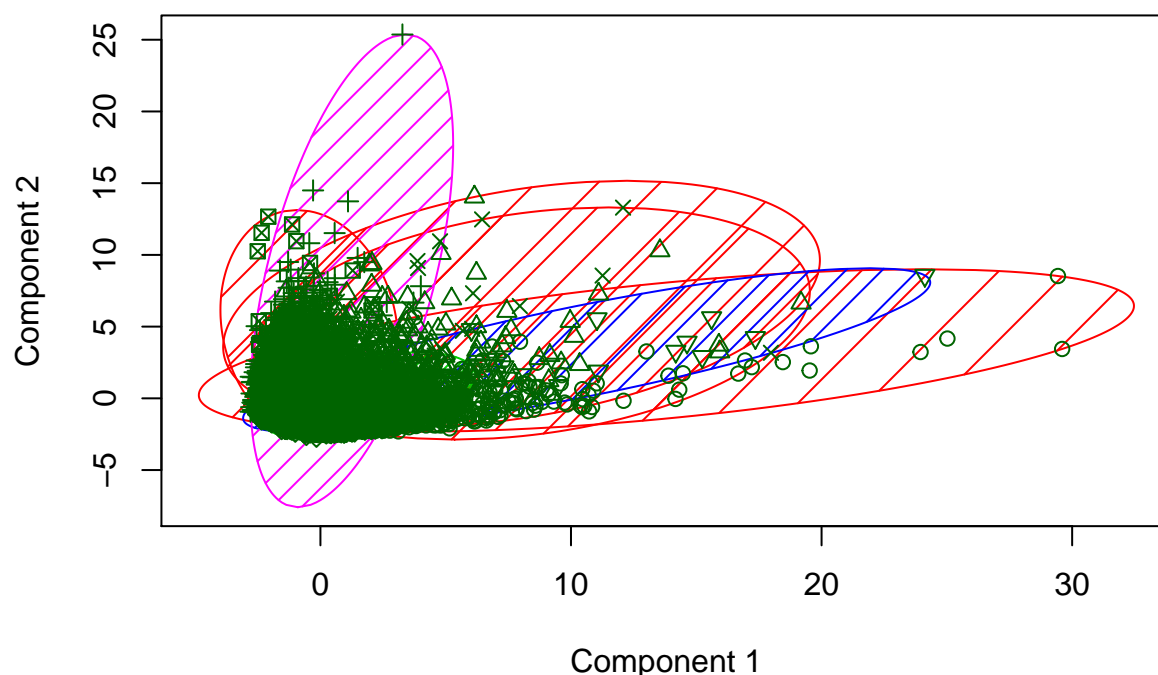
```
##           df      BIC
## seg.mc_data 1402 943225.80
## seg.mc_scale  480  69925.18
## seg.mc_pc    1094  20488.99
## seg.mc_scale4 632 101748.41
```

```
#Summary of the model which has the lowest BIC value
summary(seg.mc_pc)
```

```
## -----
## Gaussian finite mixture model fitted by EM algorithm
## -----
##
## Mclust EEV (ellipsoidal, equal volume and shape) model with 7 components:
##
##   log-likelihood    n    df      BIC      ICL
##   -5267.116 8950 1094 -20488.99 -21333.51
##
## Clustering table:
##    1    2    3    4    5    6    7
##  700 1098 2873  111 2233 1682  253
```

```
# Plot the groups
clusplot(credit_data, seg.mc_pc$class, color=TRUE, shade=TRUE, labels=7, lines=0, main="Model-based clu
```

Model-based cluster plot



These two components explain 47.62 % of the point variability.

```
# Attributes summary for each clusters
data.summ(credit_data,seg.mc_pc$class)
```

##	Group.1	BALANCE	BALANCE_FREQUENCY	PURCHASES	ONEOFF_PURCHASES
## 1	1	917.5157	0.9754546	3945.5567	2588.1762
## 2	2	3475.3192	0.9996688	1354.5822	627.6138
## 3	3	2151.9635	0.9071967	142.0709	140.8687
## 4	4	3188.0582	0.9377920	1854.9192	921.8739
## 5	5	112.2660	0.6646972	667.0962	277.8202
## 6	6	1433.2396	0.9889201	1484.4968	962.3229
## 7	7	1367.6820	0.8419556	509.2766	219.0825
##	INSTALLMENTS_PURCHASES	CASH_ADVANCE	PURCHASES_FREQUENCY		
## 1	1359.739643	370.415471	0.95000002		
## 2	727.575200	1744.441652	0.68237705		
## 3	1.202235	1932.077764	0.07597625		
## 4	933.045315	3365.067779	0.69763367		
## 5	389.289436	18.711782	0.60647614		
## 6	522.187063	1.928592	0.70650921		
## 7	291.447036	2437.942488	0.53778882		
##	ONEOFF_PURCHASES_FREQUENCY	PURCHASES_INSTALLMENTS_FREQUENCY			
## 1	0.76107145		0.665595266		
## 2	0.21964174		0.588342467		
## 3	0.07533894		0.001397257		
## 4	0.12992859		0.613424477		
## 5	0.11029555		0.500269339		
## 6	0.31157441		0.508152290		

## 7	0.14565472	0.418495941		
##	CASH_ADVANCE_FREQUENCY	CASH_ADVANCE_TRX	PURCHASES_TRX	CREDIT_LIMIT PAYMENTS
## 1	0.0592856871	1.098571429	52.777143	6951.110 3869.6294
## 2	0.2563751940	6.462659381	26.101093	5290.665 2340.1520
## 3	0.2623738138	5.939436129	1.414201	4109.281 1665.3710
## 4	0.2622550631	8.117117117	21.216216	4537.838 3386.2320
## 5	0.0064548012	0.075235110	10.670846	3815.964 807.4388
## 6	0.0006080404	0.008323424	19.876932	4591.577 1599.6310
## 7	0.3485931818	12.114624506	9.371542	3939.328 2289.8698
##	MINIMUM_PAYMENTS	PRC_FULL_PAYMENT	TENURE	
## 1	326.9777	0.48639476	12.00000	
## 2	2372.1770	0.01090226	12.00000	
## 3	717.5676	0.02924398	11.32022	
## 4	9344.5296	0.03260013	10.00901	
## 5	129.1959	0.37370189	11.26019	
## 6	673.1520	0.03206320	11.99287	
## 7	438.9982	0.18678225	10.09486	

““