

# MAX 503 Exam Project | Elizabeth and Rosemary | 5-1-23

1 a

```
data(iris)
str(iris)
```

```
## 'data.frame':    150 obs. of  5 variables:
## $ Sepal.Length: num  5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
## $ Sepal.Width : num  3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
## $ Petal.Length: num  1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
## $ Petal.Width : num  0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
## $ Species      : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...
```

b The steps in classification are (1) collect data with predictors and outcome, (2) divide the observations into training and test cases, the training data set is used to train the model, while the test data set used to evaluate the model's performance, (3) use training cases to fit a model predicting the outcomes, (4) confirm that the model works well for test cases, also referred to evaluating the performance using metrics such as accuracy, precision, recall, and F1 score to assess how the model performs on unseen data, and (5) apply the model to new data to obtain predictions which involves using an untrained model to classify new instances based on their input variables.

c

```
species.summ <- function(data, groups) {
  aggregate(data, list(groups), function(x) mean(as.numeric(x)))
}
```

d

```
# make it repeatable
set.seed(04625)

# train on 65% of data. Hold 35% for testing
train.prop <- 0.65
train.cases <- sample(nrow(iris), nrow(iris)*train.prop)

species.df.train <- iris[ train.cases, ]
species.df.test  <- iris[-train.cases, ]
```

e Naive Bayes classification method uses training data in order to learn the probability of class membership as a function of each predictor variable considered independently. Basically, it computes conditional probabilities in the training data.

f

```
library(e1071)
```

```
## Warning: package 'e1071' was built under R version 4.2.3
```

```
(species.nb <- naiveBayes(Species~.,data=species.df.train))
```

```
##  
## Naive Bayes Classifier for Discrete Predictors  
##  
## Call:  
## naiveBayes.default(x = X, y = Y, laplace = laplace)  
##  
## A-priori probabilities:  
## Y  
##      setosa versicolor  virginica  
## 0.3092784  0.3402062  0.3505155  
##  
## Conditional probabilities:  
##      Sepal.Length  
## Y      [,1]      [,2]  
## setosa  5.053333 0.3626848  
## versicolor 5.903030 0.5462857  
## virginica 6.591176 0.6215170  
##  
##      Sepal.Width  
## Y      [,1]      [,2]  
## setosa  3.543333 0.3654787  
## versicolor 2.800000 0.3344772  
## virginica 3.002941 0.3019987  
##  
##      Petal.Length  
## Y      [,1]      [,2]  
## setosa  1.476667 0.1654322  
## versicolor 4.260606 0.5302944  
## virginica 5.526471 0.5478934  
##  
##      Petal.Width  
## Y      [,1]      [,2]  
## setosa  0.2466667 0.1008014  
## versicolor 1.3151515 0.2237761  
## virginica 2.0529412 0.2537420
```

```
# predicting species membership  
species.nb.class<-predict(species.nb,species.df.train)  
  
prop.table(table(species.nb.class))
```

```
## species.nb.class  
##      setosa versicolor  virginica  
## 0.3092784  0.3298969  0.3608247
```

This means that in the train data prediction, 30.93% are setosa, 32.99% are versicolor, and 36.08% are virginica.

g

```
mean(species.df.train$Species==species.nb.class)
```

```
## [1] 0.9690722
```

```
library(mclust)
```

```
## Warning: package 'mclust' was built under R version 4.2.3
```

```
## Package 'mclust' version 6.0.0
```

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

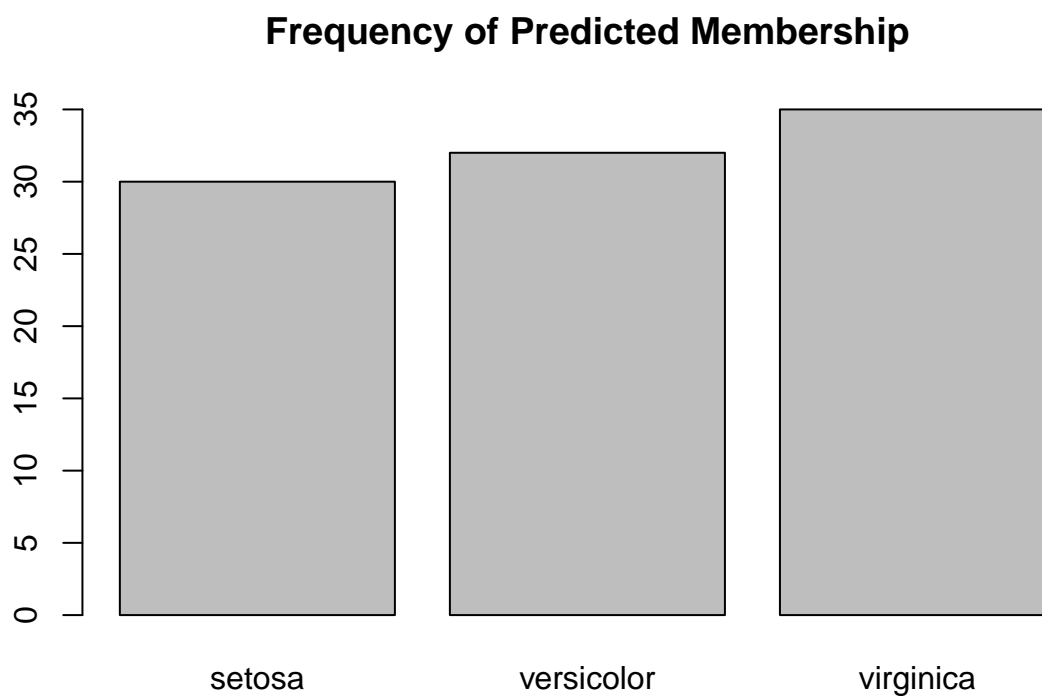
```
adjustedRandIndex(species.nb.class,species.df.train$Species)
```

```
## [1] 0.9063881
```

The results are very high for both as they are in the 90th percent. This means there was excellent data recovery and the two data clusters match almost perfectly.

h

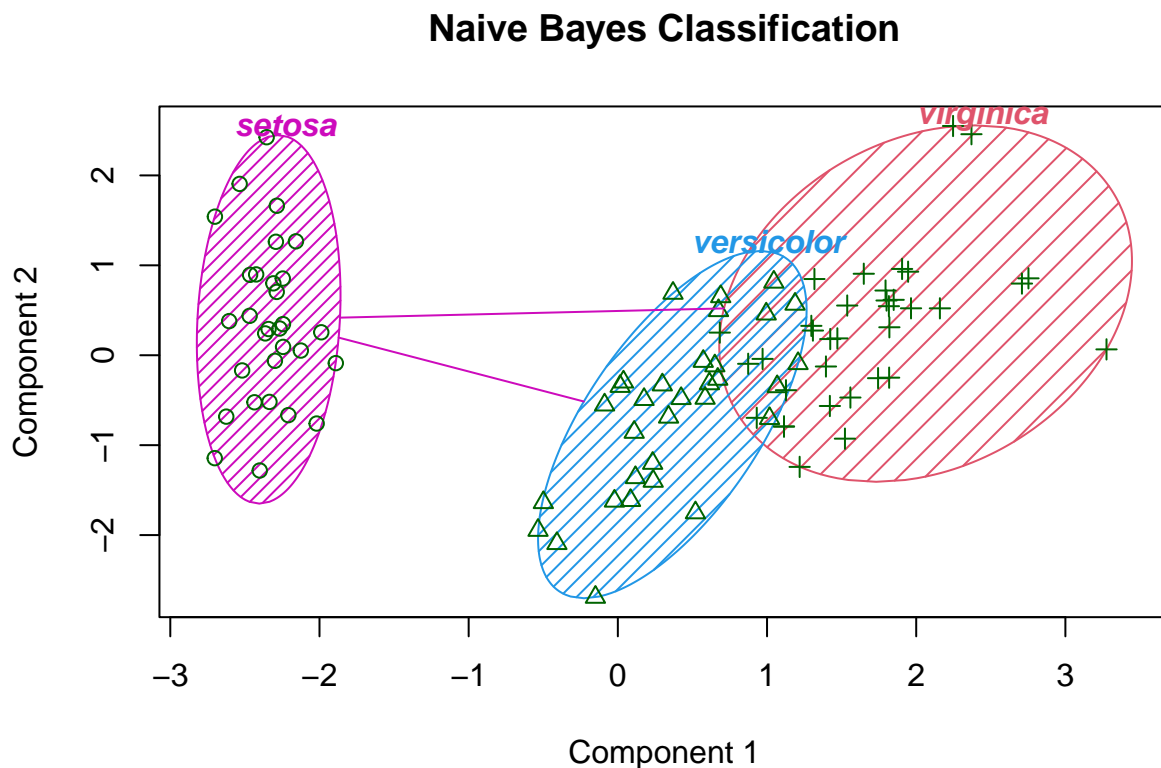
```
barplot(table(species.nb.class),  
        main = "Frequency of Predicted Membership")
```



They are all very close together, with virginica having the highest count.

i

```
library(cluster)
clusplot(species.df.train[, -5],
         species.nb.class,
         color = TRUE,
         shade = TRUE,
         labels = 4,
         main = "Naive Bayes Classification")
```



These two components explain 96.06 % of the point variability.

j1

```
# raw correct proportions
species.nb.class.test<-predict(species.nb,species.df.test)
mean(species.df.test$Species == species.nb.class.test)
```

```
## [1] 0.9433962
```

The predicted membership is very close to the known species as there is a 94% agreement between predicted and actual species membership.

j2

```
library(mclust)
adjustedRandIndex(species.nb.class.test, species.df.test$Species)
```

```
## [1] 0.8518591
```

The adjusted random index yields 85% which isn't as high but it's a better indicator and is still high.

**j3**

```
#confusion matrix
table(species.nb.class.test, species.df.test$Species)
```

```
##
## species.nb.class.test setosa versicolor virginica
##          setosa      20          0          0
##          versicolor  0         17          3
##          virginica   0          0         13
```

Setosa was predicted correctly 20 out of 20 times. Versicolor was predicted 17 out of 17 times. However, virginica was predicted correctly 13 out of 16 times, which is an 81.25% success rate.

**j4**

```
species.summ(species.df.test,species.nb.class.test)
```

```
##      Group.1 Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1      setosa   4.935000    3.255000    1.440000    0.245000     1.00
## 2 versicolor   5.950000    2.670000    4.375000    1.375000     2.15
## 3 virginica    6.792308    3.023077    5.738462    2.069231     3.00
```

```
species.summ(species.df.test,species.df.test$Species)
```

```
##      Group.1 Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1      setosa   4.93500    3.255000    1.440000    0.245000         1
## 2 versicolor   6.00000    2.711765    4.258824    1.347059         2
## 3 virginica    6.58125    2.912500    5.606250    1.968750         3
```

**j5**

```
mean(species.df.test$Species == species.nb.class.test)
```

```
## [1] 0.9433962
```

This is very high as it means there is a 94% agreement between actual and predicted values in the test data.

**k**

```
predict(species.nb,species.df.test,type = "raw")
```

##		setosa	versicolor	virginica
##	[1,]	1.000000e+00	1.382240e-13	1.228564e-26
##	[2,]	1.000000e+00	1.180140e-13	1.359806e-26
##	[3,]	1.000000e+00	3.709144e-14	9.233836e-27
##	[4,]	1.000000e+00	4.759290e-14	3.833709e-27
##	[5,]	1.000000e+00	8.195480e-14	2.387935e-26
##	[6,]	1.000000e+00	2.865063e-14	1.231276e-27
##	[7,]	1.000000e+00	6.323988e-12	2.099984e-23
##	[8,]	1.000000e+00	1.458226e-15	4.987077e-29
##	[9,]	1.000000e+00	2.767393e-14	6.082124e-27
##	[10,]	1.000000e+00	2.448785e-11	4.031202e-23
##	[11,]	1.000000e+00	1.883127e-13	2.844692e-26
##	[12,]	1.000000e+00	2.345325e-14	1.573437e-27
##	[13,]	1.000000e+00	2.793732e-14	7.080749e-27
##	[14,]	1.000000e+00	4.427492e-14	1.822954e-27
##	[15,]	1.000000e+00	1.667236e-11	2.451299e-25
##	[16,]	1.000000e+00	4.147900e-08	3.126962e-19
##	[17,]	1.000000e+00	5.494669e-10	3.362963e-21
##	[18,]	1.000000e+00	9.828742e-13	1.552929e-25
##	[19,]	1.000000e+00	2.608771e-14	2.444105e-27
##	[20,]	1.000000e+00	2.843786e-14	4.426494e-27
##	[21,]	3.415948e-76	9.999860e-01	1.399747e-05
##	[22,]	1.263282e-114	9.549057e-01	4.509432e-02
##	[23,]	4.734906e-105	9.931738e-01	6.826196e-03
##	[24,]	1.766340e-75	9.999390e-01	6.098779e-05
##	[25,]	1.252706e-93	9.961365e-01	3.863490e-03
##	[26,]	3.952617e-100	9.832728e-01	1.672721e-02
##	[27,]	5.072023e-110	9.970336e-01	2.966354e-03
##	[28,]	6.177117e-130	9.500520e-01	4.994795e-02
##	[29,]	5.758707e-100	9.888023e-01	1.119766e-02
##	[30,]	1.540545e-45	9.999998e-01	2.370173e-07
##	[31,]	1.098686e-67	9.999849e-01	1.513397e-05
##	[32,]	4.250388e-106	9.937909e-01	6.209060e-03
##	[33,]	9.326279e-96	9.996879e-01	3.120651e-04
##	[34,]	6.851686e-72	9.999822e-01	1.779191e-05
##	[35,]	6.819023e-84	9.998833e-01	1.166708e-04
##	[36,]	1.715284e-89	9.991936e-01	8.064462e-04
##	[37,]	1.252491e-79	9.998865e-01	1.134839e-04
##	[38,]	2.346792e-233	4.337145e-06	9.999957e-01
##	[39,]	1.162732e-117	9.865850e-01	1.341502e-02
##	[40,]	6.516021e-244	2.255931e-05	9.999774e-01
##	[41,]	6.045580e-205	2.876828e-03	9.971232e-01
##	[42,]	1.837789e-283	7.414914e-10	1.000000e+00
##	[43,]	5.544553e-173	1.436212e-03	9.985638e-01
##	[44,]	4.927745e-207	3.798356e-05	9.999620e-01
##	[45,]	5.478560e-135	9.827120e-01	1.728801e-02
##	[46,]	8.805394e-237	3.995475e-07	9.999996e-01
##	[47,]	1.503253e-292	7.654235e-08	9.999999e-01
##	[48,]	2.642143e-147	1.878444e-01	8.121556e-01
##	[49,]	6.229840e-237	1.329534e-05	9.999867e-01
##	[50,]	5.937513e-165	7.635349e-01	2.364651e-01
##	[51,]	1.815179e-235	3.967125e-07	9.999996e-01
##	[52,]	7.357992e-250	2.060379e-07	9.999998e-01
##	[53,]	1.560667e-156	9.928176e-02	9.007182e-01

l

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 4.2.3
```

```
## randomForest 4.7-1.1
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
set.seed(98040)
```

```
species.rf <- randomForest(Species~.,data=species.df.train,ntree=1500)
```

m

```
species.rf
```

```
##
```

```
## Call:
```

```
## randomForest(formula = Species ~ ., data = species.df.train,      ntree = 1500)
```

```
##              Type of random forest: classification
```

```
##              Number of trees: 1500
```

```
## No. of variables tried at each split: 2
```

```
##
```

```
##              OOB estimate of  error rate: 6.19%
```

```
## Confusion matrix:
```

```
##              setosa versicolor virginica class.error
```

```
## setosa              30              0              0 0.00000000
```

```
## versicolor          0              30              3 0.09090909
```

```
## virginica           0              3              31 0.08823529
```

It was correct 94% as the error rate is about 6%. For setosa it was correct 100% and this was the best. It was also nearly perfect for versicolor and virginica, with versicolor being slightly worse.

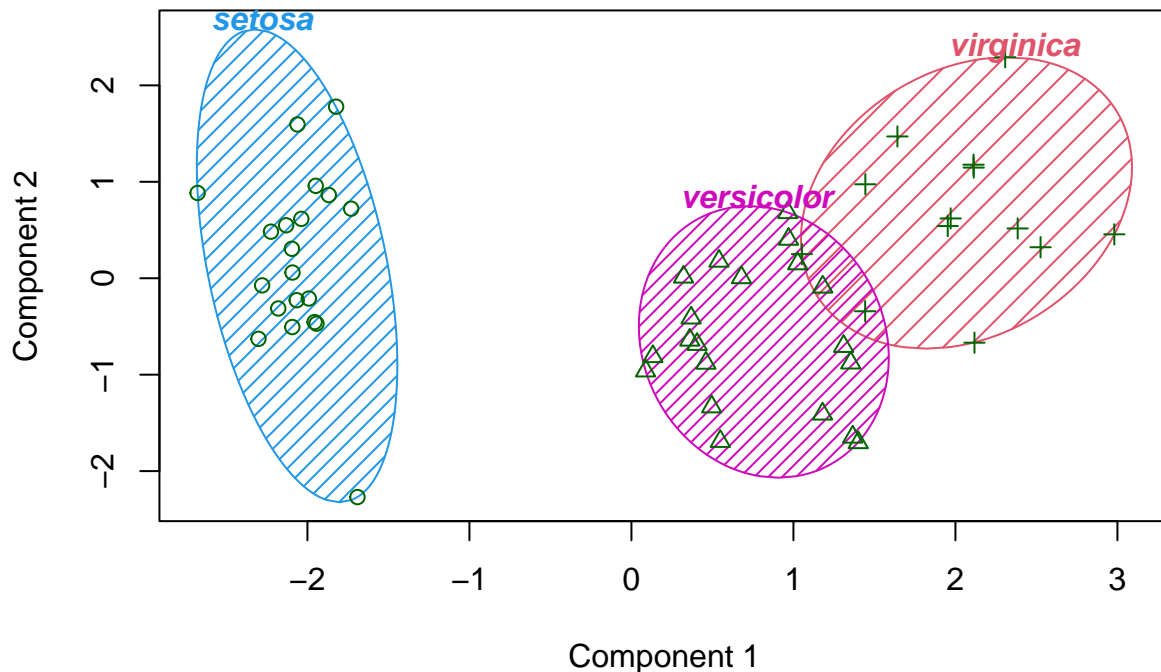
n

```
species.rf.class <- predict(species.rf, species.df.test)
```

```
library(cluster)
```

```
clusplot(species.df.test[, -5], species.rf.class, color = TRUE, shade = TRUE, labels = 4, lines = 0, main
```

## Random Forest Classes, test data



These two components explain 95.72 % of the point variability.

o1

```
# proposed speciesments
species.summ(species.df.test, species.rf.class)
```

##	Group.1	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	setosa	4.935000	3.255000	1.440000	0.245000	1.00
## 2	versicolor	5.950000	2.670000	4.375000	1.375000	2.15
## 3	virginica	6.792308	3.023077	5.738462	2.069231	3.00

```
# actual speciesments
species.summ(species.df.test, species.df.test$Species)
```

##	Group.1	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	setosa	4.93500	3.255000	1.440000	0.245000	1
## 2	versicolor	6.00000	2.711765	4.258824	1.347059	2
## 3	virginica	6.58125	2.912500	5.606250	1.968750	3

o2

```
mean(species.df.test$Species == species.rf.class)
```

```
## [1] 0.9433962
```



This is very good as it means there is 94% agreement between actual and predicted species membership.

**o3**

```
# confusion matrix
table(species.df.test$Species, species.rf.class)
```

```
##           species.rf.class
##           setosa versicolor virginica
## setosa           20           0           0
## versicolor        0           17           0
## virginica         0           3           13
```

Setosa was predicted correctly 20 out of 20 times. Versicolor was predicted 17 out of 20 times with a success rate of 85%. Virginica was predicted correctly 13 out of 13 times.

**p**

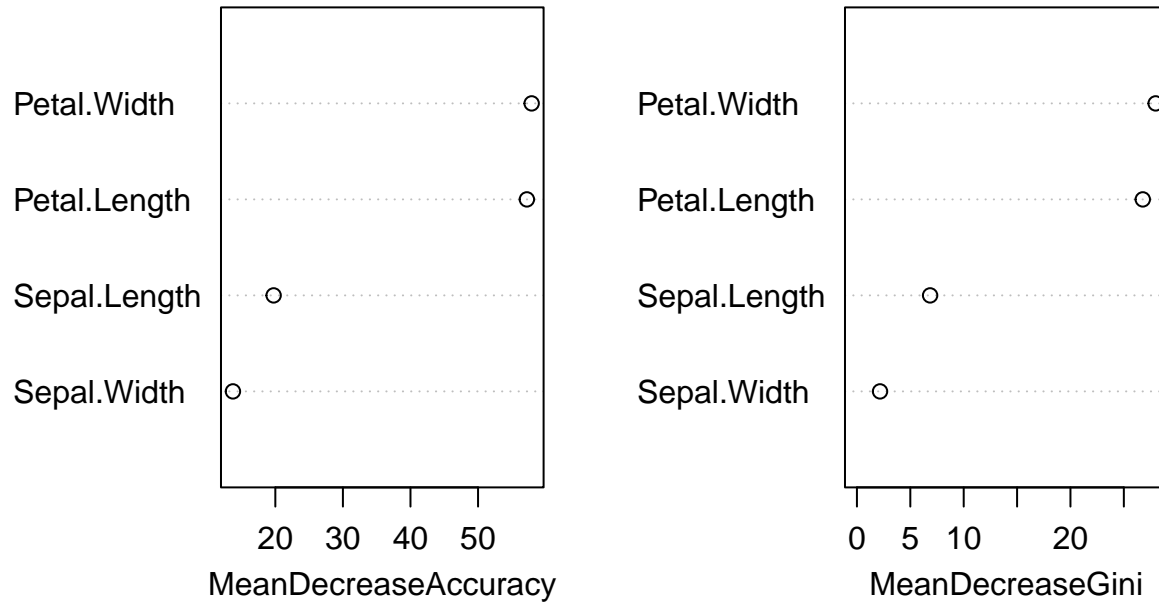
```
set.seed(98040)
species.rf<-randomForest(Species~., data = species.df.train, ntree = 1500, importance=TRUE)
importance(species.rf)
```

```
##           setosa versicolor virginica MeanDecreaseAccuracy
## Sepal.Length 12.33893   9.133094 18.955426           19.74545
## Sepal.Width  10.39294   9.014322  9.292696           13.72235
## Petal.Length 39.89980  45.997779 45.347488           57.22617
## Petal.Width  37.92793  45.763906 54.267123           57.93071
##           MeanDecreaseGini
## Sepal.Length      6.855285
## Sepal.Width       2.164326
## Petal.Length     26.772326
## Petal.Width      27.978400
```

**q**

```
varImpPlot(species.rf, main = "Variable importance by speciesment")
```

## Variable importance by species



The most important variables in this data set are petal width and petal length.

**2** Note: the output has been limited to the first 100 lines for the sake of the document length.

```
# upload data
boa.df <- read.csv('boa.csv')

# dimensions of the data set
dim(boa.df)
```

```
## [1] 8950  19
```

```
# summarize the data set
summary(boa.df)
```

```
##          X          CUST_ID          BALANCE          BALANCE_FREQUENCY
## Min.      : 1    Length:8950    Min.      : 0.0    Min.      :0.0000
## 1st Qu.:2238    Class :character 1st Qu.: 128.3    1st Qu.:0.8889
## Median :4476    Mode  :character Median : 873.4    Median :1.0000
## Mean   :4476                                Mean  :1564.5    Mean   :0.8773
## 3rd Qu.:6713                                3rd Qu.:2054.1    3rd Qu.:1.0000
## Max.   :8950                                Max.   :19043.1   Max.   :1.0000
## 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
```

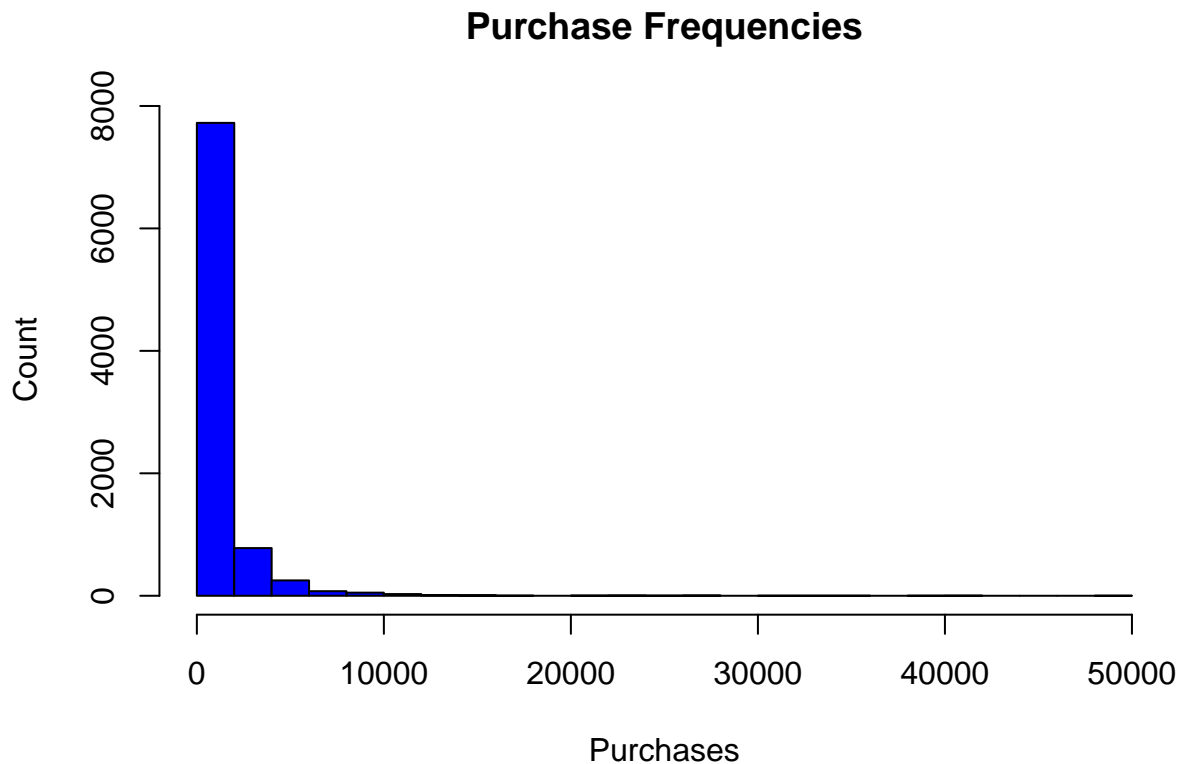
```
## Mean      : 1003.20    Mean      : 592.4    Mean      : 411.1    Mean      : 978.9
## 3rd Qu.: 1110.13    3rd Qu.: 577.4    3rd Qu.: 468.6    3rd Qu.: 1113.8
## Max.      :49039.57    Max.      :40761.2    Max.      :22500.0    Max.      :47137.2
## PURCHASES_FREQUENCY ONEOFF_PURCHASES_FREQUENCY
## 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
```

```
# structure of the data set
str(boa.df)
```

```
## 'data.frame': 8950 obs. of 19 variables:
## $ X : int 1 2 3 4 5 6 7 8 9 10 ...
## $ CUST_ID : chr "C10001" "C10002" "C10003" "C10004" ...
## $ 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 ...
```

```
# purchases histogram
hist(boa.df$PURCHASES, main = "Purchase Frequencies",
     xlab="Purchases", ylab="Count",
     breaks = 30,
     col = "blue")
```



```
# remove customer id and purchase columns to look at which clusters would emerge when we remove purchases
boa.df.new <- boa.df[, -c(2,5)]
str(boa.df.new)
```

```
## 'data.frame': 8950 obs. of 17 variables:
## $ X : int 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 ...
## $ 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 ...
## $ 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 ...
```

```
# function to look at mean values by group
```

```
boa.summ <- function(data, groups) { aggregate(data, list(groups), function(x) mean(as.numeric(x)))}
```

```
head(boa.summ(boa.df.new, boa.df$PURCHASES), 100)
```

##	Group.1	X	BALANCE	BALANCE_FREQUENCY	ONEOFF_PURCHASES
## 1	0.00	4569.760	2151.2029350	0.8825000	0.00000
## 2	0.01	2817.000	1692.2194260	0.6590910	0.01000
## 3	0.05	329.000	850.6301140	1.0000000	0.05000
## 4	0.24	3591.000	1.8694350	0.4545450	0.24000
## 5	0.70	5814.000	212.3061480	0.1818180	0.70000
## 6	1.00	4832.500	2321.6780220	1.0000000	1.00000
## 7	1.40	7002.000	46.7957010	0.2727270	1.40000
## 8	2.00	2045.000	1709.6939560	1.0000000	2.00000
## 9	4.44	3556.000	6744.8456860	1.0000000	0.00000
## 10	4.80	6233.000	7.4569660	0.2727270	0.00000
## 11	4.99	3163.000	793.0052270	1.0000000	4.99000
## 12	6.90	8195.000	0.4372850	0.1818180	6.90000
## 13	7.26	6518.000	644.9339110	0.8333330	0.00000
## 14	8.40	3708.667	2901.6828830	1.0000000	8.40000
## 15	9.28	5929.000	0.0000000	0.0000000	0.00000
## 16	9.68	5260.000	2.6411010	1.0000000	0.00000
## 17	9.87	7640.000	1.6952940	0.0909090	0.00000
## 18	9.90	6643.000	1083.3791710	1.0000000	0.00000
## 19	10.79	7849.000	411.3629690	1.0000000	0.00000
## 20	10.89	7920.000	0.0000000	0.0000000	0.00000
## 21	11.06	615.000	3419.6616760	1.0000000	11.06000
## 22	11.35	4851.000	1074.2584870	1.0000000	0.00000
## 23	11.41	2339.000	3860.2380720	1.0000000	11.41000
## 24	12.00	2794.500	34.2897165	0.6363640	6.00000
## 25	12.22	3932.000	2944.1569960	1.0000000	12.22000
## 26	12.40	6180.000	933.0725190	1.0000000	12.40000
## 27	12.64	2743.000	6784.2582760	1.0000000	0.00000
## 28	12.65	1132.000	0.0000000	0.0000000	0.00000
## 29	13.00	8666.000	1112.9156570	1.0000000	13.00000
## 30	13.66	4756.000	207.2095970	1.0000000	0.00000
## 31	13.69	5911.000	56.0603270	0.1818180	0.00000
## 32	14.28	2593.000	339.4062590	1.0000000	0.00000
## 33	14.72	5553.000	1175.9226280	1.0000000	14.72000
## 34	14.95	321.000	4732.3735400	1.0000000	0.00000
## 35	14.98	3273.000	124.1178380	0.3636360	14.98000
## 36	15.00	7854.500	647.5574050	0.7272725	15.00000
## 37	15.30	7396.000	1.9413280	0.1818180	0.00000
## 38	15.40	1956.000	107.1300200	1.0000000	15.40000
## 39	15.80	5105.000	900.5417510	1.0000000	0.00000
## 40	15.92	5721.000	3881.6795820	1.0000000	15.92000
## 41	16.00	1984.000	410.4281480	0.5909090	16.00000
## 42	16.50	384.000	1.8380840	0.2727270	16.50000

## 43	17.00	6253.000	492.0402230	1.0000000	17.00000
## 44	17.30	6173.000	4198.6180900	0.7272730	17.30000
## 45	17.60	7294.000	1242.8862660	1.0000000	17.60000
## 46	17.70	3169.000	0.6839390	0.1818180	17.70000
## 47	17.80	2589.500	225.7873070	0.4090905	17.80000
## 48	18.00	5487.000	2893.8974870	1.0000000	0.00000
## 49	18.05	859.000	910.0725140	1.0000000	0.00000
## 50	18.35	3909.000	6928.7213770	1.0000000	0.00000
## 51	18.41	600.000	917.7328460	1.0000000	0.00000
## 52	18.77	4402.000	0.0000000	0.0000000	0.00000
## 53	19.08	1582.000	1.3432070	0.1818180	0.00000
## 54	19.25	4630.000	0.0000000	0.0000000	0.00000
## 55	19.50	5774.000	17.2455390	0.6363640	0.00000
## 56	19.75	7544.000	803.0527040	1.0000000	19.75000
## 57	19.88	6869.000	826.6886250	1.0000000	19.88000
## 58	19.90	3161.000	147.8458865	0.6000000	9.95000
## 59	19.91	4107.000	634.7599730	1.0000000	0.00000
## 60	19.96	6713.000	701.0172700	1.0000000	19.96000
## 61	19.98	2064.000	2828.7658220	1.0000000	19.98000
## 62	20.00	4817.333	1199.7706802	0.8232323	16.66667
## 63	20.09	921.000	39.7989930	0.4545450	20.09000
## 64	20.74	7575.000	1.0631140	0.0909090	0.00000
## 65	20.90	5794.500	666.7627825	0.7500000	20.90000
## 66	21.00	3224.000	509.9003170	0.3636360	0.00000
## 67	21.05	4059.000	1.8168140	0.0909090	0.00000
## 68	21.75	3967.000	1033.4987360	1.0000000	0.00000
## 69	21.90	6196.000	556.8574380	1.0000000	21.90000
## 70	21.99	6776.000	12.4233240	0.5454550	0.00000
## 71	22.16	1131.000	6090.5222730	1.0000000	22.16000
## 72	22.48	3406.000	35.2765510	1.0000000	22.48000
## 73	22.50	2241.000	1013.8676360	0.6818180	0.00000
## 74	22.68	2214.000	689.6976440	1.0000000	0.00000
## 75	23.00	2132.500	0.4297235	0.2272725	11.50000
## 76	23.21	1636.000	1621.0353730	1.0000000	0.00000
## 77	23.75	7870.000	183.2858400	1.0000000	23.75000
## 78	23.91	2327.000	1.0594250	0.3636360	23.91000
## 79	24.00	737.000	0.2557100	0.0909090	0.00000
## 80	24.63	5249.000	385.8419970	1.0000000	0.00000
## 81	24.64	587.000	0.3271990	0.4545450	24.64000
## 82	25.00	6273.000	2080.3129910	0.8888890	0.00000
## 83	25.20	3344.000	1396.3858050	1.0000000	25.20000
## 84	25.26	2896.500	1650.7292830	1.0000000	25.26000
## 85	25.50	1167.000	1.0591030	0.1818180	25.50000
## 86	25.69	4795.000	4.1256500	0.3636360	25.69000
## 87	26.09	7364.000	443.5585780	1.0000000	26.09000
## 88	26.12	3183.000	1839.9304600	1.0000000	26.12000
## 89	26.49	4972.000	2.0813570	0.1818180	26.49000
## 90	26.53	8491.000	2.0447450	0.1818180	26.53000
## 91	26.62	919.000	0.8675280	0.2727270	0.00000
## 92	26.80	2992.000	347.5772930	0.8571430	26.80000
## 93	27.00	916.000	1068.4340640	1.0000000	0.00000
## 94	27.05	5123.000	500.2194150	1.0000000	0.00000
## 95	27.16	5033.000	2872.7986140	1.0000000	0.00000
## 96	27.22	8221.000	435.5180680	1.0000000	27.22000

## 97	27.42	4690.000	4.4172730	0.1818180	0.00000
## 98	27.51	3600.000	286.2643560	0.8571430	27.51000
## 99	27.79	8214.000	57.1168110	0.4545450	27.79000
## 100	27.89	4243.000	1868.5237860	1.0000000	27.89000
##	INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY				
## 1		0.04253914	1988.74251	0.0001630788	
## 2		0.00000000	2877.42895	0.0852270000	
## 3		0.00000000	1700.63459	0.0833330000	
## 4		0.00000000	0.00000	0.0833330000	
## 5		0.00000000	1922.59360	0.0833330000	
## 6		0.00000000	710.16094	0.0833330000	
## 7		0.00000000	189.11039	0.1666670000	
## 8		0.00000000	5246.29645	0.0833330000	
## 9		4.44000000	19862.87844	0.2500000000	
## 10		4.80000000	0.00000	0.0833330000	
## 11		0.00000000	0.00000	0.0833330000	
## 12		0.00000000	0.00000	0.0833330000	
## 13		7.26000000	966.74937	0.1666670000	
## 14		0.00000000	18.17314	1.0000000000	
## 15		9.28000000	0.00000	0.2500000000	
## 16		9.68000000	0.00000	0.0833330000	
## 17		9.87000000	0.00000	0.0833330000	
## 18		9.90000000	0.00000	0.1428570000	
## 19		10.79000000	0.00000	0.1666670000	
## 20		10.89000000	0.00000	0.0833330000	
## 21		0.00000000	1261.34778	0.0833330000	
## 22		11.35000000	0.00000	0.0833330000	
## 23		0.00000000	0.00000	0.1666670000	
## 24		6.00000000	83.08013	0.5416665000	
## 25		0.00000000	244.69466	0.0833330000	
## 26		0.00000000	1085.48705	0.0833330000	
## 27		12.64000000	4516.16260	0.0833330000	
## 28		12.65000000	0.00000	0.0833330000	
## 29		0.00000000	256.94406	0.0833330000	
## 30		13.66000000	0.00000	0.0909090000	
## 31		13.69000000	0.00000	0.0833330000	
## 32		14.28000000	0.00000	0.4166670000	
## 33		0.00000000	79.25217	0.0833330000	
## 34		14.95000000	2173.35607	0.0833330000	
## 35		0.00000000	0.00000	0.0833330000	
## 36		0.00000000	104.51269	0.0871210000	
## 37		15.30000000	0.00000	0.0833330000	
## 38		0.00000000	182.66132	0.0833330000	
## 39		15.80000000	0.00000	0.2500000000	
## 40		0.00000000	2183.78246	0.0833330000	
## 41		0.00000000	0.00000	0.0833330000	
## 42		0.00000000	0.00000	0.0833330000	
## 43		0.00000000	171.81466	0.0833330000	
## 44		0.00000000	2845.86846	0.0833330000	
## 45		0.00000000	0.00000	0.0833330000	
## 46		0.00000000	0.00000	0.0833330000	
## 47		0.00000000	1214.90989	0.0833330000	
## 48		18.00000000	3904.51445	0.0833330000	
## 49		18.05000000	19.11884	0.0833330000	

## 50	18.35000000	14836.45141	0.4166670000
## 51	18.41000000	0.00000	0.0833330000
## 52	18.77000000	0.00000	0.0833330000
## 53	19.08000000	0.00000	0.1666670000
## 54	19.25000000	0.00000	0.0833330000
## 55	19.50000000	0.00000	0.0833330000
## 56	0.00000000	0.00000	0.0833330000
## 57	0.00000000	869.24620	0.0833330000
## 58	9.95000000	37.62758	0.1416665000
## 59	19.91000000	0.00000	0.0833330000
## 60	0.00000000	153.59777	0.0833330000
## 61	0.00000000	0.00000	0.0833330000
## 62	3.33333333	275.21390	0.1296293333
## 63	0.00000000	0.00000	0.3333330000
## 64	20.74000000	0.00000	0.1666670000
## 65	0.00000000	0.00000	0.1250000000
## 66	21.00000000	0.00000	0.0833330000
## 67	21.05000000	0.00000	0.0833330000
## 68	21.75000000	1238.74695	0.0833330000
## 69	0.00000000	0.00000	0.1666670000
## 70	21.99000000	0.00000	0.0909090000
## 71	0.00000000	0.00000	0.0833330000
## 72	0.00000000	0.00000	0.0833330000
## 73	22.50000000	0.00000	0.2500000000
## 74	22.68000000	0.00000	0.0833330000
## 75	11.50000000	0.00000	0.0833330000
## 76	23.21000000	0.00000	0.3333330000
## 77	0.00000000	64.00625	0.0833330000
## 78	0.00000000	0.00000	0.3333330000
## 79	24.00000000	0.00000	0.1666670000
## 80	24.63000000	0.00000	0.4166670000
## 81	0.00000000	0.00000	0.0833330000
## 82	25.00000000	2422.88959	0.1111110000
## 83	0.00000000	1920.24988	0.0833330000
## 84	0.00000000	226.40183	0.1250000000
## 85	0.00000000	0.00000	0.0833330000
## 86	0.00000000	0.00000	0.0833330000
## 87	0.00000000	646.93268	0.0833330000
## 88	0.00000000	158.97124	0.0833330000
## 89	0.00000000	0.00000	0.0833330000
## 90	0.00000000	0.00000	0.0833330000
## 91	26.62000000	0.00000	0.3333330000
## 92	0.00000000	470.84647	0.1428570000
## 93	27.00000000	0.00000	0.0833330000
## 94	27.05000000	0.00000	0.0833330000
## 95	27.16000000	2692.77704	0.0833330000
## 96	0.00000000	0.00000	0.0833330000
## 97	27.42000000	0.00000	0.0833330000
## 98	0.00000000	871.19926	0.1428570000
## 99	0.00000000	381.17115	0.0833330000
## 100	0.00000000	47.99887	0.0833330000
##	ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY		
## 1	0.00000000		0.000285388
## 2	0.08522700		0.000000000



## 3	0.08333300	0.000000000
## 4	0.08333300	0.000000000
## 5	0.08333300	0.000000000
## 6	0.08333300	0.000000000
## 7	0.16666700	0.000000000
## 8	0.08333300	0.000000000
## 9	0.00000000	0.166667000
## 10	0.00000000	0.083333000
## 11	0.08333300	0.000000000
## 12	0.08333300	0.000000000
## 13	0.00000000	0.166667000
## 14	1.00000000	0.000000000
## 15	0.00000000	0.250000000
## 16	0.00000000	0.083333000
## 17	0.00000000	0.083333000
## 18	0.00000000	0.142857000
## 19	0.00000000	0.166667000
## 20	0.00000000	0.083333000
## 21	0.08333300	0.000000000
## 22	0.00000000	0.083333000
## 23	0.16666700	0.000000000
## 24	0.04166650	0.458333500
## 25	0.08333300	0.000000000
## 26	0.08333300	0.000000000
## 27	0.00000000	0.083333000
## 28	0.00000000	0.083333000
## 29	0.08333300	0.000000000
## 30	0.00000000	0.090909000
## 31	0.00000000	0.083333000
## 32	0.00000000	0.416667000
## 33	0.08333300	0.000000000
## 34	0.00000000	0.083333000
## 35	0.08333300	0.000000000
## 36	0.08712100	0.000000000
## 37	0.00000000	0.083333000
## 38	0.08333300	0.000000000
## 39	0.00000000	0.250000000
## 40	0.08333300	0.000000000
## 41	0.08333300	0.000000000
## 42	0.08333300	0.000000000
## 43	0.08333300	0.000000000
## 44	0.08333300	0.000000000
## 45	0.08333300	0.000000000
## 46	0.08333300	0.000000000
## 47	0.08333300	0.000000000
## 48	0.00000000	0.083333000
## 49	0.00000000	0.083333000
## 50	0.00000000	0.416667000
## 51	0.00000000	0.083333000
## 52	0.00000000	0.083333000
## 53	0.00000000	0.083333000
## 54	0.00000000	0.083333000
## 55	0.00000000	0.083333000
## 56	0.08333300	0.000000000

## 57	0.08333300	0.00000000		
## 58	0.04166650	0.05000000		
## 59	0.00000000	0.08333300		
## 60	0.08333300	0.00000000		
## 61	0.08333300	0.00000000		
## 62	0.07407383	0.05555550		
## 63	0.33333300	0.00000000		
## 64	0.00000000	0.16666700		
## 65	0.12500000	0.00000000		
## 66	0.00000000	0.08333300		
## 67	0.00000000	0.08333300		
## 68	0.00000000	0.08333300		
## 69	0.16666700	0.00000000		
## 70	0.00000000	0.09090900		
## 71	0.08333300	0.00000000		
## 72	0.08333300	0.00000000		
## 73	0.00000000	0.25000000		
## 74	0.00000000	0.08333300		
## 75	0.04166650	0.04166650		
## 76	0.00000000	0.33333300		
## 77	0.08333300	0.00000000		
## 78	0.33333300	0.00000000		
## 79	0.00000000	0.16666700		
## 80	0.00000000	0.41666700		
## 81	0.08333300	0.00000000		
## 82	0.00000000	0.11111100		
## 83	0.08333300	0.00000000		
## 84	0.12500000	0.00000000		
## 85	0.08333300	0.00000000		
## 86	0.08333300	0.00000000		
## 87	0.08333300	0.00000000		
## 88	0.08333300	0.00000000		
## 89	0.08333300	0.00000000		
## 90	0.08333300	0.00000000		
## 91	0.00000000	0.33333300		
## 92	0.14285700	0.00000000		
## 93	0.00000000	0.08333300		
## 94	0.00000000	0.08333300		
## 95	0.00000000	0.08333300		
## 96	0.08333300	0.00000000		
## 97	0.00000000	0.08333300		
## 98	0.14285700	0.00000000		
## 99	0.08333300	0.00000000		
## 100	0.08333300	0.00000000		
##	CASH_ADVANCE_FREQUENCY	CASH_ADVANCE_TRX	PURCHASES_TRX	CREDIT_LIMIT
## 1	0.27259049	6.2964775	0.001956947	4029.727
## 2	0.10984850	4.2500000	1.000000000	6375.000
## 3	0.33333300	7.0000000	1.000000000	1000.000
## 4	0.00000000	0.0000000	0.000000000	3000.000
## 5	0.16666700	3.0000000	1.000000000	3500.000
## 6	0.12500000	2.0000000	1.000000000	4100.000
## 7	0.08333300	1.0000000	2.000000000	4500.000
## 8	0.25000000	6.0000000	1.000000000	2500.000
## 9	0.41666700	11.0000000	3.000000000	16500.000

## 10	0.00000000	0.00000000	1.000000000	2200.000
## 11	0.00000000	0.00000000	1.000000000	2500.000
## 12	0.00000000	0.00000000	1.000000000	8500.000
## 13	0.16666700	1.00000000	1.000000000	1500.000
## 14	0.02777767	0.6666667	12.000000000	6400.000
## 15	0.00000000	0.00000000	3.000000000	5000.000
## 16	0.00000000	0.00000000	1.000000000	8000.000
## 17	0.00000000	0.00000000	1.000000000	5000.000
## 18	0.00000000	0.00000000	1.000000000	1900.000
## 19	0.00000000	0.00000000	1.000000000	1200.000
## 20	0.00000000	0.00000000	1.000000000	4000.000
## 21	0.08333300	1.00000000	1.000000000	4000.000
## 22	0.00000000	0.00000000	1.000000000	1200.000
## 23	0.00000000	0.00000000	2.000000000	10000.000
## 24	0.04166650	1.50000000	6.500000000	1400.000
## 25	0.16666700	2.00000000	1.000000000	3000.000
## 26	0.16666700	6.00000000	1.000000000	2500.000
## 27	0.41666700	13.00000000	1.000000000	10000.000
## 28	0.00000000	0.00000000	1.000000000	5000.000
## 29	0.25000000	5.00000000	1.000000000	1500.000
## 30	0.00000000	0.00000000	1.000000000	7000.000
## 31	0.00000000	0.00000000	1.000000000	2900.000
## 32	0.00000000	0.00000000	5.000000000	10500.000
## 33	0.25000000	3.00000000	1.000000000	1200.000
## 34	0.41666700	11.00000000	1.000000000	5700.000
## 35	0.00000000	0.00000000	1.000000000	1700.000
## 36	0.04545450	0.50000000	1.000000000	4500.000
## 37	0.00000000	0.00000000	1.000000000	5000.000
## 38	0.33333300	6.00000000	1.000000000	200.000
## 39	0.00000000	0.00000000	3.000000000	7000.000
## 40	0.33333300	9.00000000	1.000000000	5500.000
## 41	0.00000000	0.00000000	1.000000000	3350.000
## 42	0.00000000	0.00000000	1.000000000	6000.000
## 43	0.33333300	4.00000000	1.000000000	600.000
## 44	0.25000000	6.00000000	1.000000000	8954.545
## 45	0.00000000	0.00000000	1.000000000	4000.000
## 46	0.00000000	0.00000000	1.000000000	1500.000
## 47	0.12500000	1.50000000	1.000000000	2500.000
## 48	0.41666700	23.00000000	1.000000000	6000.000
## 49	0.08333300	1.00000000	1.000000000	1000.000
## 50	0.08333300	1.00000000	5.000000000	18000.000
## 51	0.00000000	0.00000000	1.000000000	1000.000
## 52	0.00000000	0.00000000	1.000000000	5000.000
## 53	0.00000000	0.00000000	2.000000000	7000.000
## 54	0.00000000	0.00000000	1.000000000	4000.000
## 55	0.00000000	0.00000000	1.000000000	1500.000
## 56	0.00000000	0.00000000	1.000000000	2000.000
## 57	0.33333300	20.00000000	2.000000000	900.000
## 58	0.08333350	1.00000000	1.500000000	3750.000
## 59	0.00000000	0.00000000	1.000000000	6000.000
## 60	0.25000000	3.00000000	1.000000000	3500.000
## 61	0.00000000	0.00000000	2.000000000	5500.000
## 62	0.15277767	2.1666667	1.500000000	3000.000
## 63	0.00000000	0.00000000	5.000000000	3200.000

## 64	0.00000000	0.00000000	2.000000000	1500.000
## 65	0.00000000	0.00000000	1.000000000	1000.000
## 66	0.00000000	0.00000000	1.000000000	11500.000
## 67	0.00000000	0.00000000	1.000000000	2900.000
## 68	0.33333300	5.00000000	1.000000000	3000.000
## 69	0.00000000	0.00000000	2.000000000	600.000
## 70	0.00000000	0.00000000	1.000000000	4000.000
## 71	0.00000000	0.00000000	1.000000000	9000.000
## 72	0.00000000	0.00000000	1.000000000	7500.000
## 73	0.00000000	0.00000000	3.000000000	6500.000
## 74	0.00000000	0.00000000	1.000000000	11500.000
## 75	0.00000000	0.00000000	1.000000000	4000.000
## 76	0.00000000	0.00000000	4.000000000	2000.000
## 77	0.08333300	2.00000000	1.000000000	500.000
## 78	0.00000000	0.00000000	4.000000000	15500.000
## 79	0.00000000	0.00000000	2.000000000	1500.000
## 80	0.00000000	0.00000000	5.000000000	4500.000
## 81	0.00000000	0.00000000	1.000000000	1800.000
## 82	0.33333300	6.00000000	1.000000000	2500.000
## 83	0.58333300	19.00000000	1.000000000	2500.000
## 84	0.12500000	1.50000000	1.500000000	3050.000
## 85	0.00000000	0.00000000	1.000000000	5500.000
## 86	0.00000000	0.00000000	1.000000000	4500.000
## 87	0.50000000	16.00000000	4.000000000	1200.000
## 88	0.08333300	1.00000000	1.000000000	1950.000
## 89	0.00000000	0.00000000	1.000000000	7500.000
## 90	0.00000000	0.00000000	1.000000000	2500.000
## 91	0.00000000	0.00000000	4.000000000	4500.000
## 92	0.42857100	11.00000000	1.000000000	500.000
## 93	0.00000000	0.00000000	1.000000000	1200.000
## 94	0.00000000	0.00000000	1.000000000	1200.000
## 95	0.41666700	15.00000000	1.000000000	5000.000
## 96	0.00000000	0.00000000	1.000000000	500.000
## 97	0.00000000	0.00000000	1.000000000	1800.000
## 98	0.42857100	3.00000000	1.000000000	500.000
## 99	0.08333300	1.00000000	1.000000000	4500.000
## 100	0.08333300	1.00000000	2.000000000	2000.000

##	PAYMENTS	MINIMUM_PAYMENTS	PRC_FULL_PAYMENT	TENURE
## 1	1653.130103	968.694784	0.04432393	11.31849
## 2	1273.540476	629.643771	0.03125000	11.75000
## 3	1084.281127	367.409536	0.00000000	12.00000
## 4	150.381107	53.294711	0.00000000	12.00000
## 5	0.000000	0.000000	0.00000000	12.00000
## 6	848.476405	604.791575	0.00000000	12.00000
## 7	0.000000	0.000000	0.00000000	12.00000
## 8	4678.254753	464.860442	0.08333300	12.00000
## 9	21440.298660	1350.823356	0.10000000	12.00000
## 10	70.517745	246.213204	0.00000000	12.00000
## 11	1127.028432	223.627537	0.08333300	12.00000
## 12	4.523555	4.763689	0.00000000	12.00000
## 13	431.732367	92.246251	0.00000000	6.00000
## 14	769.002348	1076.353714	0.00000000	12.00000
## 15	0.000000	0.000000	0.00000000	12.00000
## 16	72.282497	17.530337	0.00000000	12.00000

## 17	151.732627	98.617471	0.50000000	12.00000
## 18	186.923852	7243.733403	0.00000000	7.00000
## 19	918.160924	166.281511	0.00000000	12.00000
## 20	0.000000	0.000000	0.00000000	12.00000
## 21	745.984976	1243.564770	0.00000000	12.00000
## 22	421.348577	206.637191	0.00000000	12.00000
## 23	7280.584479	1013.780486	0.08333300	12.00000
## 24	79.733163	65.870176	0.00000000	12.00000
## 25	628.314620	1288.695715	0.00000000	12.00000
## 26	481.589626	238.636185	0.08333300	12.00000
## 27	10813.829330	1661.518089	0.08333300	12.00000
## 28	0.000000	0.000000	0.00000000	12.00000
## 29	396.056561	308.356999	0.00000000	12.00000
## 30	1703.339459	208.367436	0.00000000	11.00000
## 31	2165.403988	219.961065	0.00000000	12.00000
## 32	1263.292815	1486.059666	0.08333300	12.00000
## 33	347.568201	463.992512	0.00000000	12.00000
## 34	1289.284068	1172.289863	0.00000000	12.00000
## 35	607.535166	43.582049	0.00000000	12.00000
## 36	352.719365	261.538020	0.00000000	11.50000
## 37	190.254352	117.892476	0.00000000	12.00000
## 38	255.103798	198.933162	0.00000000	12.00000
## 39	1542.583506	375.127536	0.00000000	12.00000
## 40	1032.183632	1129.747227	0.00000000	12.00000
## 41	667.190047	182.614350	0.50000000	12.00000
## 42	57.595083	67.192500	0.00000000	12.00000
## 43	243.905711	291.892373	0.00000000	12.00000
## 44	10226.601760	1024.126428	0.10000000	12.00000
## 45	589.674674	472.763536	0.00000000	12.00000
## 46	18.825023	10.074393	0.00000000	12.00000
## 47	5253.799504	411.220052	0.50000000	12.00000
## 48	483.729934	720.971058	0.00000000	12.00000
## 49	233.886707	230.806229	0.00000000	12.00000
## 50	22281.700460	1592.560164	0.09090900	12.00000
## 51	1407.291593	424.623595	0.00000000	12.00000
## 52	0.000000	0.000000	0.00000000	12.00000
## 53	170.180973	161.234787	0.00000000	12.00000
## 54	0.000000	0.000000	0.00000000	12.00000
## 55	95.789433	119.880198	0.00000000	12.00000
## 56	402.155365	322.394919	0.00000000	12.00000
## 57	1072.638864	373.912056	0.08333300	12.00000
## 58	132.130910	111.680274	0.25000000	11.00000
## 59	610.491342	246.804668	0.00000000	12.00000
## 60	218.851473	332.899226	0.00000000	12.00000
## 61	593.220384	1159.310924	0.00000000	12.00000
## 62	711.253366	388.624279	0.00000000	11.50000
## 63	0.000000	0.000000	0.00000000	12.00000
## 64	168.489938	98.859855	0.00000000	12.00000
## 65	173.167499	263.360908	0.00000000	9.00000
## 66	23150.571840	1863.225391	0.33333300	12.00000
## 67	243.137299	112.019269	0.00000000	12.00000
## 68	255.779858	489.808320	0.00000000	12.00000
## 69	162.272808	271.112945	0.00000000	12.00000
## 70	0.000000	0.000000	0.00000000	11.00000

```
## 71    568.781100    30528.432400    0.00000000 12.00000
## 72     68.205028     200.034390    0.00000000 12.00000
## 73   1441.188109     248.194890    0.00000000 12.00000
## 74    906.092735     224.709183    0.00000000 12.00000
## 75      0.000000      0.000000    0.00000000 12.00000
## 76    500.515114     475.627205    0.00000000 12.00000
## 77    584.679478     139.651483    0.08333300 12.00000
## 78     31.845894      22.081988    0.11111100 12.00000
## 79      9.533313      8.842600    0.00000000 12.00000
## 80   1349.938041     203.987924    0.00000000 12.00000
## 81     45.640106      26.849523    0.00000000 12.00000
## 82    540.687217     570.576789    0.00000000  9.00000
## 83    410.617369     503.627101    0.00000000 12.00000
## 84    819.279904     526.338710    0.00000000 12.00000
## 85    563.748018      83.648417    0.50000000 12.00000
## 86    163.352538     128.336305    0.00000000 12.00000
## 87   1218.866837     200.752923    0.08333300 12.00000
## 88    525.947589     627.660656    0.00000000 12.00000
## 89     78.573329      42.095280    0.00000000 12.00000
## 90    155.972776      82.345774    0.00000000 12.00000
## 91     18.336805       8.745383    0.09090900 12.00000
## 92     28.654864     256.522546    0.00000000  7.00000
## 93    219.692369     310.328990    0.00000000 12.00000
## 94    104.687781     300.921255    0.00000000 12.00000
## 95    507.240956    1102.034141    0.00000000 12.00000
## 96    160.536841     220.943631    0.00000000 12.00000
## 97    228.417814     175.836378    1.00000000 12.00000
## 98    733.901545     122.835190    0.25000000  7.00000
## 99   1215.965938      99.950697    1.00000000 12.00000
## 100   440.707307     517.115912    0.00000000 12.00000
```

Compute Distance Matrix

```
# distance between observations
dist(rbind(c(1,2,3), c(2,3,2)))
```

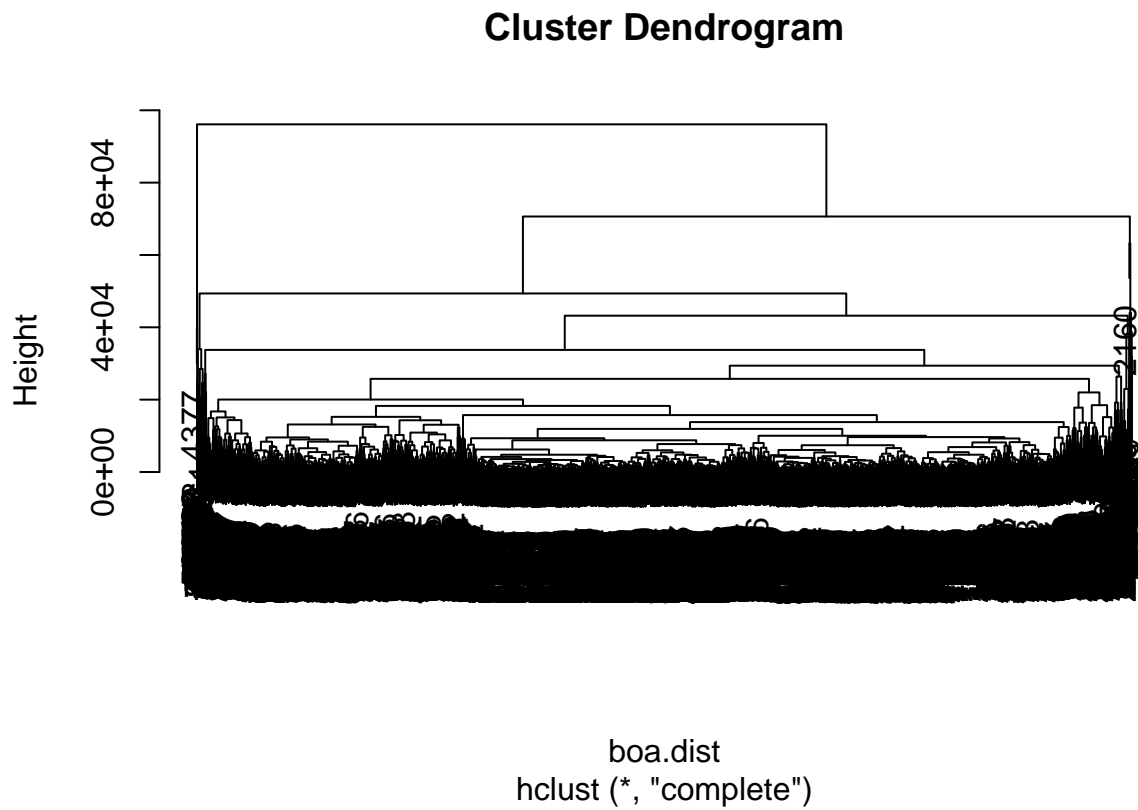
```
##           1
## 2 1.732051
```

```
# get the distance between data types using daisy which combines all datatypes
library(cluster)
boa.dist<-daisy(boa.df.new)
as.matrix(boa.dist)[1:4,1:4]
```

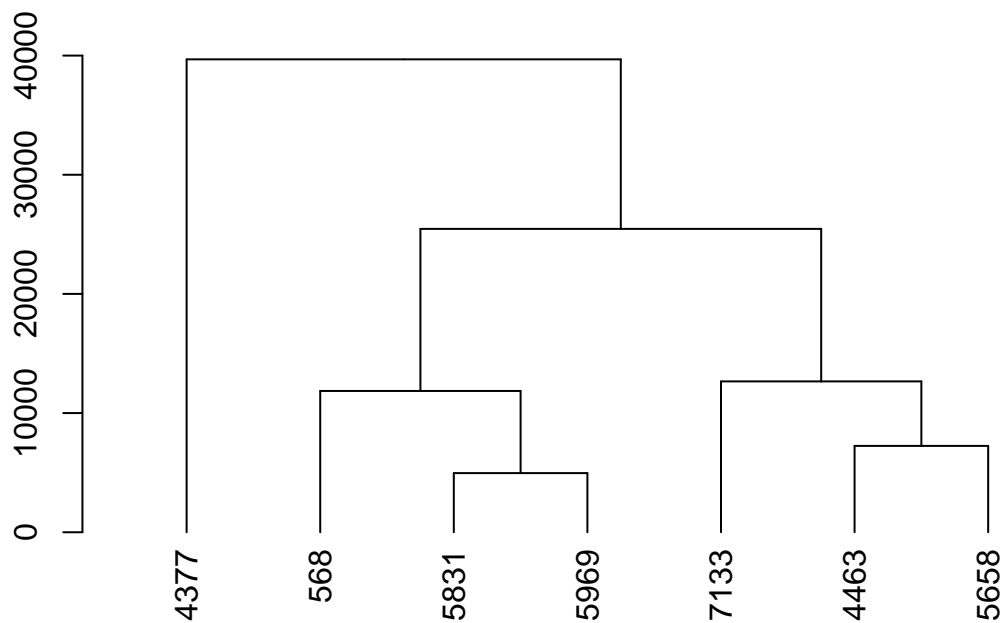
```
##           1           2           3           4
## 1      0.000 10178.701 7021.031 6873.993
## 2 10178.701      0.000 7427.988 7857.647
## 3 7021.031 7427.988      0.000 1426.931
## 4 6873.993 7857.647 1426.931      0.000
```

H Clust

```
# combine closest neighbors into larger groups for hierarchical clustering
boa.hc<-hclust(boa.dist,method="complete")
plot(boa.hc)
```



```
# cut the tree for better visualization
plot(cut(as.dendrogram(boa.hc),h=4e+04)$lower[[1]])
```



```
# comparing observations in branches
boa.df[c(5831,5969),] # similar
```

```
##           X CUST_ID  BALANCE BALANCE_FREQUENCY PURCHASES ONEOFF_PURCHASES
## 5831 5831  C15993 8038.893                1         0             0
## 5969 5969  C16134 6022.224                1         0             0
##      INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 5831                0      197.9156                0
## 5969                0      4111.4656                0
##      ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 5831                0                0
## 5969                0                0
##      CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT
## 5831                0.166667                8         0         9000
## 5969                0.166667                6         0         7000
##      PAYMENTS MINIMUM_PAYMENTS PRC_FULL_PAYMENT TENURE
## 5831 3569.183        43132.73                0      12
## 5969 4560.776        42629.55                0      12
```

```
boa.df[c(568,7133),] # not similar differ in the last 4 variables
```

```
##           X CUST_ID  BALANCE BALANCE_FREQUENCY PURCHASES ONEOFF_PURCHASES
## 568  568  C10591 3457.086                1      2448.6             0
## 7133 7133  C17325 9024.812                1         0.0             0
##      INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
```



```
## 568          2448.6          0.000          1
## 7133          0.0          2750.442          0
##      ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 568          0          1
## 7133          0          0
##      CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT
## 568          0.000000          0          12          1700
## 7133          0.083333          2          0          9000
##      PAYMENTS MINIMUM_PAYMENTS PRC_FULL_PAYMENT TENURE
## 568 227.5145          38512.12          0          12
## 7133 302.4842          61031.62          0          12
```

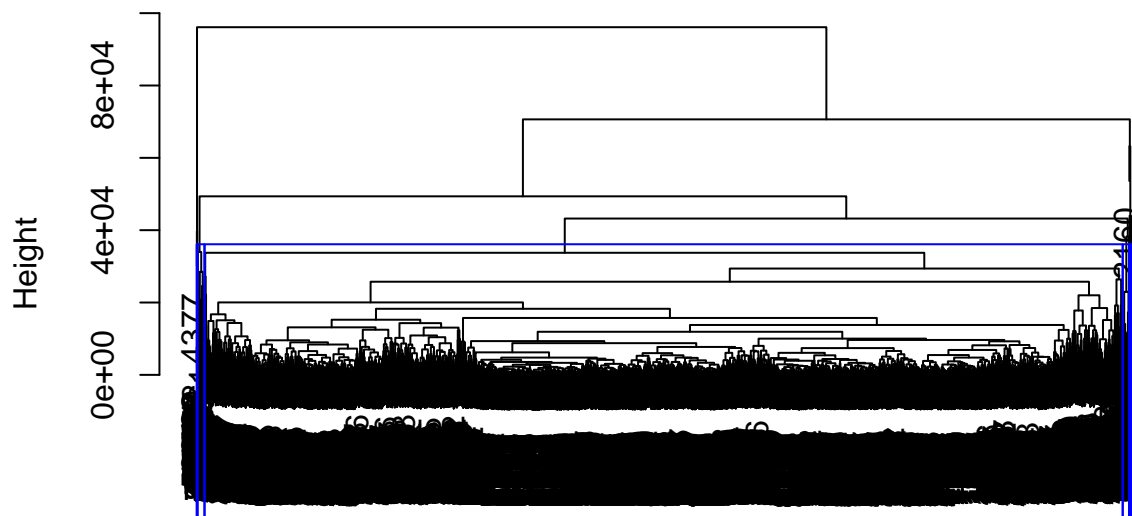
```
# check how well clustering model reflects distance matrix
cor(cophenetic(boa.hc),boa.dist)
```

```
## [1] 0.8598062
```

```
plot(boa.hc)
# At 85% the hierarchical tree represents the distance between the trees well.

# getting k groups from tree
rect.hclust(boa.hc,k=10,border = "blue")
```

## Cluster Dendrogram



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

```
# getting purchase membership from hclust
boa.hc.purchases<-cutree(boa.hc,k=10)
table(boa.hc.purchases)
```

```
## boa.hc.purchases
##      1      2      3      4      5      6      7      8      9     10
## 8787   61   67      2   15      5      6      5      1      1
```

```
# comparing segments
boa.summ(boa.df.new, boa.hc.purchases)
```

```
##      Group.1      X      BALANCE BALANCE_FREQUENCY ONEOFF_PURCHASES
## 1          1 4493.627 1512.198          0.8765615          533.8092
## 2          2 4131.869 3172.547          0.9832754          148.9170
## 3          3 3314.060 4438.527          0.8247157          1005.3907
## 4          4 1973.000 16571.695          1.0000000          6426.0900
## 5          5 1771.600 4281.864          0.9933333          19257.4787
## 6          6 2728.600 6879.309          0.9636364          35158.2400
## 7          7 4937.000 6024.526          1.0000000           0.0000
## 8          8 3162.600 6814.389          0.8545454          11529.8200
## 9          9 2160.000 10905.054          1.0000000          133.5000
## 10         10 4377.000 10571.411          1.0000000           0.0000
##      INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1          392.5290          893.1400          0.4901471
## 2          384.4325          965.0994          0.4176602
## 3          868.9455         11043.2705          0.4005621
## 4          7076.7700         10356.3350          1.0000000
## 5          6683.5627          149.3933          0.9933333
## 6          3023.4520          111.6334          0.8833332
## 7          710.3350          1176.6373          0.5000000
## 8          1281.8620          7279.2503          0.7000000
## 9          298.4300          47137.2118          0.5833330
## 10         7739.4800           0.0000          1.0000000
##      ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 1          0.20161626          0.3633034
## 2          0.04284644          0.3831347
## 3          0.23229630          0.3263068
## 4          0.87500000          0.6666665
## 5          0.94222227          0.8033333
## 6          0.83333340          0.6500000
## 7          0.00000000          0.5000000
## 8          0.38333340          0.6000000
## 9          0.25000000          0.5000000
## 10         0.00000000          1.0000000
##      CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS
## 1          0.13309075          3.1159668          14.30044          4413.722 1532.039
## 2          0.11421585          3.2786885          12.50820          3468.033 1054.004
## 3          0.43564643         18.7910448          26.37313         10859.701 17546.531
## 4          0.41666650         18.0000000         125.00000         17250.000 21718.454
## 5          0.01111107          0.3333333         124.00000         15066.667 23169.875
## 6          0.01666660          0.2000000         166.00000         19900.000 35553.331
## 7          0.06944450          2.6666667          32.50000          6200.000 2083.188
## 8          0.28333340         11.4000000          53.00000         15600.000 37341.715
```

## 9	1.00000000	123.0000000	21.00000	19600.000	39048.598
## 10	0.00000000	0.0000000	44.00000	8000.000	2688.447
##	MINIMUM_PAYMENTS	PRC_FULL_PAYMENT	TENURE		
## 1	666.1803	0.153682508	11.51269		
## 2	16785.8559	0.004098344	11.85246		
## 3	1610.8447	0.187596597	11.59701		
## 4	19928.0393	0.000000000	12.00000		
## 5	1212.7009	0.584595933	11.86667		
## 6	6445.0784	0.566666600	12.00000		
## 7	48656.1372	0.000000000	12.00000		
## 8	2910.4627	0.183333400	12.00000		
## 9	5394.1737	0.000000000	12.00000		
## 10	76406.2075	0.000000000	12.00000		

```
head(boa.summ(boa.df.new, boa.df$PURCHASES), 100)
```

##	Group.1	X	BALANCE	BALANCE_FREQUENCY	ONEOFF_PURCHASES
## 1	0.00	4569.760	2151.2029350	0.8825000	0.00000
## 2	0.01	2817.000	1692.2194260	0.6590910	0.01000
## 3	0.05	329.000	850.6301140	1.0000000	0.05000
## 4	0.24	3591.000	1.8694350	0.4545450	0.24000
## 5	0.70	5814.000	212.3061480	0.1818180	0.70000
## 6	1.00	4832.500	2321.6780220	1.0000000	1.00000
## 7	1.40	7002.000	46.7957010	0.2727270	1.40000
## 8	2.00	2045.000	1709.6939560	1.0000000	2.00000
## 9	4.44	3556.000	6744.8456860	1.0000000	0.00000
## 10	4.80	6233.000	7.4569660	0.2727270	0.00000
## 11	4.99	3163.000	793.0052270	1.0000000	4.99000
## 12	6.90	8195.000	0.4372850	0.1818180	6.90000
## 13	7.26	6518.000	644.9339110	0.8333330	0.00000
## 14	8.40	3708.667	2901.6828830	1.0000000	8.40000
## 15	9.28	5929.000	0.0000000	0.0000000	0.00000
## 16	9.68	5260.000	2.6411010	1.0000000	0.00000
## 17	9.87	7640.000	1.6952940	0.0909090	0.00000
## 18	9.90	6643.000	1083.3791710	1.0000000	0.00000
## 19	10.79	7849.000	411.3629690	1.0000000	0.00000
## 20	10.89	7920.000	0.0000000	0.0000000	0.00000
## 21	11.06	615.000	3419.6616760	1.0000000	11.06000
## 22	11.35	4851.000	1074.2584870	1.0000000	0.00000
## 23	11.41	2339.000	3860.2380720	1.0000000	11.41000
## 24	12.00	2794.500	34.2897165	0.6363640	6.00000
## 25	12.22	3932.000	2944.1569960	1.0000000	12.22000
## 26	12.40	6180.000	933.0725190	1.0000000	12.40000
## 27	12.64	2743.000	6784.2582760	1.0000000	0.00000
## 28	12.65	1132.000	0.0000000	0.0000000	0.00000
## 29	13.00	8666.000	1112.9156570	1.0000000	13.00000
## 30	13.66	4756.000	207.2095970	1.0000000	0.00000
## 31	13.69	5911.000	56.0603270	0.1818180	0.00000
## 32	14.28	2593.000	339.4062590	1.0000000	0.00000
## 33	14.72	5553.000	1175.9226280	1.0000000	14.72000
## 34	14.95	321.000	4732.3735400	1.0000000	0.00000
## 35	14.98	3273.000	124.1178380	0.3636360	14.98000
## 36	15.00	7854.500	647.5574050	0.7272725	15.00000
## 37	15.30	7396.000	1.9413280	0.1818180	0.00000

## 38	15.40	1956.000	107.1300200	1.0000000	15.40000
## 39	15.80	5105.000	900.5417510	1.0000000	0.00000
## 40	15.92	5721.000	3881.6795820	1.0000000	15.92000
## 41	16.00	1984.000	410.4281480	0.5909090	16.00000
## 42	16.50	384.000	1.8380840	0.2727270	16.50000
## 43	17.00	6253.000	492.0402230	1.0000000	17.00000
## 44	17.30	6173.000	4198.6180900	0.7272730	17.30000
## 45	17.60	7294.000	1242.8862660	1.0000000	17.60000
## 46	17.70	3169.000	0.6839390	0.1818180	17.70000
## 47	17.80	2589.500	225.7873070	0.4090905	17.80000
## 48	18.00	5487.000	2893.8974870	1.0000000	0.00000
## 49	18.05	859.000	910.0725140	1.0000000	0.00000
## 50	18.35	3909.000	6928.7213770	1.0000000	0.00000
## 51	18.41	600.000	917.7328460	1.0000000	0.00000
## 52	18.77	4402.000	0.0000000	0.0000000	0.00000
## 53	19.08	1582.000	1.3432070	0.1818180	0.00000
## 54	19.25	4630.000	0.0000000	0.0000000	0.00000
## 55	19.50	5774.000	17.2455390	0.6363640	0.00000
## 56	19.75	7544.000	803.0527040	1.0000000	19.75000
## 57	19.88	6869.000	826.6886250	1.0000000	19.88000
## 58	19.90	3161.000	147.8458865	0.6000000	9.95000
## 59	19.91	4107.000	634.7599730	1.0000000	0.00000
## 60	19.96	6713.000	701.0172700	1.0000000	19.96000
## 61	19.98	2064.000	2828.7658220	1.0000000	19.98000
## 62	20.00	4817.333	1199.7706802	0.8232323	16.66667
## 63	20.09	921.000	39.7989930	0.4545450	20.09000
## 64	20.74	7575.000	1.0631140	0.0909090	0.00000
## 65	20.90	5794.500	666.7627825	0.7500000	20.90000
## 66	21.00	3224.000	509.9003170	0.3636360	0.00000
## 67	21.05	4059.000	1.8168140	0.0909090	0.00000
## 68	21.75	3967.000	1033.4987360	1.0000000	0.00000
## 69	21.90	6196.000	556.8574380	1.0000000	21.90000
## 70	21.99	6776.000	12.4233240	0.5454550	0.00000
## 71	22.16	1131.000	6090.5222730	1.0000000	22.16000
## 72	22.48	3406.000	35.2765510	1.0000000	22.48000
## 73	22.50	2241.000	1013.8676360	0.6818180	0.00000
## 74	22.68	2214.000	689.6976440	1.0000000	0.00000
## 75	23.00	2132.500	0.4297235	0.2272725	11.50000
## 76	23.21	1636.000	1621.0353730	1.0000000	0.00000
## 77	23.75	7870.000	183.2858400	1.0000000	23.75000
## 78	23.91	2327.000	1.0594250	0.3636360	23.91000
## 79	24.00	737.000	0.2557100	0.0909090	0.00000
## 80	24.63	5249.000	385.8419970	1.0000000	0.00000
## 81	24.64	587.000	0.3271990	0.4545450	24.64000
## 82	25.00	6273.000	2080.3129910	0.8888890	0.00000
## 83	25.20	3344.000	1396.3858050	1.0000000	25.20000
## 84	25.26	2896.500	1650.7292830	1.0000000	25.26000
## 85	25.50	1167.000	1.0591030	0.1818180	25.50000
## 86	25.69	4795.000	4.1256500	0.3636360	25.69000
## 87	26.09	7364.000	443.5585780	1.0000000	26.09000
## 88	26.12	3183.000	1839.9304600	1.0000000	26.12000
## 89	26.49	4972.000	2.0813570	0.1818180	26.49000
## 90	26.53	8491.000	2.0447450	0.1818180	26.53000
## 91	26.62	919.000	0.8675280	0.2727270	0.00000

## 92	26.80	2992.000	347.5772930	0.8571430	26.80000
## 93	27.00	916.000	1068.4340640	1.0000000	0.00000
## 94	27.05	5123.000	500.2194150	1.0000000	0.00000
## 95	27.16	5033.000	2872.7986140	1.0000000	0.00000
## 96	27.22	8221.000	435.5180680	1.0000000	27.22000
## 97	27.42	4690.000	4.4172730	0.1818180	0.00000
## 98	27.51	3600.000	286.2643560	0.8571430	27.51000
## 99	27.79	8214.000	57.1168110	0.4545450	27.79000
## 100	27.89	4243.000	1868.5237860	1.0000000	27.89000
##	INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY				
## 1		0.04253914	1988.74251	0.0001630788	
## 2		0.00000000	2877.42895	0.0852270000	
## 3		0.00000000	1700.63459	0.0833330000	
## 4		0.00000000	0.00000	0.0833330000	
## 5		0.00000000	1922.59360	0.0833330000	
## 6		0.00000000	710.16094	0.0833330000	
## 7		0.00000000	189.11039	0.1666670000	
## 8		0.00000000	5246.29645	0.0833330000	
## 9		4.44000000	19862.87844	0.2500000000	
## 10		4.80000000	0.00000	0.0833330000	
## 11		0.00000000	0.00000	0.0833330000	
## 12		0.00000000	0.00000	0.0833330000	
## 13		7.26000000	966.74937	0.1666670000	
## 14		0.00000000	18.17314	1.0000000000	
## 15		9.28000000	0.00000	0.2500000000	
## 16		9.68000000	0.00000	0.0833330000	
## 17		9.87000000	0.00000	0.0833330000	
## 18		9.90000000	0.00000	0.1428570000	
## 19		10.79000000	0.00000	0.1666670000	
## 20		10.89000000	0.00000	0.0833330000	
## 21		0.00000000	1261.34778	0.0833330000	
## 22		11.35000000	0.00000	0.0833330000	
## 23		0.00000000	0.00000	0.1666670000	
## 24		6.00000000	83.08013	0.5416665000	
## 25		0.00000000	244.69466	0.0833330000	
## 26		0.00000000	1085.48705	0.0833330000	
## 27		12.64000000	4516.16260	0.0833330000	
## 28		12.65000000	0.00000	0.0833330000	
## 29		0.00000000	256.94406	0.0833330000	
## 30		13.66000000	0.00000	0.0909090000	
## 31		13.69000000	0.00000	0.0833330000	
## 32		14.28000000	0.00000	0.4166670000	
## 33		0.00000000	79.25217	0.0833330000	
## 34		14.95000000	2173.35607	0.0833330000	
## 35		0.00000000	0.00000	0.0833330000	
## 36		0.00000000	104.51269	0.0871210000	
## 37		15.30000000	0.00000	0.0833330000	
## 38		0.00000000	182.66132	0.0833330000	
## 39		15.80000000	0.00000	0.2500000000	
## 40		0.00000000	2183.78246	0.0833330000	
## 41		0.00000000	0.00000	0.0833330000	
## 42		0.00000000	0.00000	0.0833330000	
## 43		0.00000000	171.81466	0.0833330000	
## 44		0.00000000	2845.86846	0.0833330000	

## 45	0.00000000	0.00000	0.0833330000
## 46	0.00000000	0.00000	0.0833330000
## 47	0.00000000	1214.90989	0.0833330000
## 48	18.00000000	3904.51445	0.0833330000
## 49	18.05000000	19.11884	0.0833330000
## 50	18.35000000	14836.45141	0.4166670000
## 51	18.41000000	0.00000	0.0833330000
## 52	18.77000000	0.00000	0.0833330000
## 53	19.08000000	0.00000	0.1666670000
## 54	19.25000000	0.00000	0.0833330000
## 55	19.50000000	0.00000	0.0833330000
## 56	0.00000000	0.00000	0.0833330000
## 57	0.00000000	869.24620	0.0833330000
## 58	9.95000000	37.62758	0.1416665000
## 59	19.91000000	0.00000	0.0833330000
## 60	0.00000000	153.59777	0.0833330000
## 61	0.00000000	0.00000	0.0833330000
## 62	3.33333333	275.21390	0.1296293333
## 63	0.00000000	0.00000	0.3333330000
## 64	20.74000000	0.00000	0.1666670000
## 65	0.00000000	0.00000	0.1250000000
## 66	21.00000000	0.00000	0.0833330000
## 67	21.05000000	0.00000	0.0833330000
## 68	21.75000000	1238.74695	0.0833330000
## 69	0.00000000	0.00000	0.1666670000
## 70	21.99000000	0.00000	0.0909090000
## 71	0.00000000	0.00000	0.0833330000
## 72	0.00000000	0.00000	0.0833330000
## 73	22.50000000	0.00000	0.2500000000
## 74	22.68000000	0.00000	0.0833330000
## 75	11.50000000	0.00000	0.0833330000
## 76	23.21000000	0.00000	0.3333330000
## 77	0.00000000	64.00625	0.0833330000
## 78	0.00000000	0.00000	0.3333330000
## 79	24.00000000	0.00000	0.1666670000
## 80	24.63000000	0.00000	0.4166670000
## 81	0.00000000	0.00000	0.0833330000
## 82	25.00000000	2422.88959	0.1111110000
## 83	0.00000000	1920.24988	0.0833330000
## 84	0.00000000	226.40183	0.1250000000
## 85	0.00000000	0.00000	0.0833330000
## 86	0.00000000	0.00000	0.0833330000
## 87	0.00000000	646.93268	0.0833330000
## 88	0.00000000	158.97124	0.0833330000
## 89	0.00000000	0.00000	0.0833330000
## 90	0.00000000	0.00000	0.0833330000
## 91	26.62000000	0.00000	0.3333330000
## 92	0.00000000	470.84647	0.1428570000
## 93	27.00000000	0.00000	0.0833330000
## 94	27.05000000	0.00000	0.0833330000
## 95	27.16000000	2692.77704	0.0833330000
## 96	0.00000000	0.00000	0.0833330000
## 97	27.42000000	0.00000	0.0833330000
## 98	0.00000000	871.19926	0.1428570000

## 99	0.00000000	381.17115	0.0833330000
## 100	0.00000000	47.99887	0.0833330000
##	ONEOFF_PURCHASES_FREQUENCY	PURCHASES_INSTALLMENTS_FREQUENCY	
## 1	0.00000000		0.000285388
## 2	0.08522700		0.000000000
## 3	0.08333300		0.000000000
## 4	0.08333300		0.000000000
## 5	0.08333300		0.000000000
## 6	0.08333300		0.000000000
## 7	0.16666700		0.000000000
## 8	0.08333300		0.000000000
## 9	0.00000000		0.166667000
## 10	0.00000000		0.083333000
## 11	0.08333300		0.000000000
## 12	0.08333300		0.000000000
## 13	0.00000000		0.166667000
## 14	1.00000000		0.000000000
## 15	0.00000000		0.250000000
## 16	0.00000000		0.083333000
## 17	0.00000000		0.083333000
## 18	0.00000000		0.142857000
## 19	0.00000000		0.166667000
## 20	0.00000000		0.083333000
## 21	0.08333300		0.000000000
## 22	0.00000000		0.083333000
## 23	0.16666700		0.000000000
## 24	0.04166650		0.458333500
## 25	0.08333300		0.000000000
## 26	0.08333300		0.000000000
## 27	0.00000000		0.083333000
## 28	0.00000000		0.083333000
## 29	0.08333300		0.000000000
## 30	0.00000000		0.090909000
## 31	0.00000000		0.083333000
## 32	0.00000000		0.416667000
## 33	0.08333300		0.000000000
## 34	0.00000000		0.083333000
## 35	0.08333300		0.000000000
## 36	0.08712100		0.000000000
## 37	0.00000000		0.083333000
## 38	0.08333300		0.000000000
## 39	0.00000000		0.250000000
## 40	0.08333300		0.000000000
## 41	0.08333300		0.000000000
## 42	0.08333300		0.000000000
## 43	0.08333300		0.000000000
## 44	0.08333300		0.000000000
## 45	0.08333300		0.000000000
## 46	0.08333300		0.000000000
## 47	0.08333300		0.000000000
## 48	0.00000000		0.083333000
## 49	0.00000000		0.083333000
## 50	0.00000000		0.416667000
## 51	0.00000000		0.083333000

## 52	0.00000000	0.083333000		
## 53	0.00000000	0.083333000		
## 54	0.00000000	0.083333000		
## 55	0.00000000	0.083333000		
## 56	0.08333300	0.000000000		
## 57	0.08333300	0.000000000		
## 58	0.04166650	0.050000000		
## 59	0.00000000	0.083333000		
## 60	0.08333300	0.000000000		
## 61	0.08333300	0.000000000		
## 62	0.07407383	0.055555500		
## 63	0.33333300	0.000000000		
## 64	0.00000000	0.166667000		
## 65	0.12500000	0.000000000		
## 66	0.00000000	0.083333000		
## 67	0.00000000	0.083333000		
## 68	0.00000000	0.083333000		
## 69	0.16666700	0.000000000		
## 70	0.00000000	0.090909000		
## 71	0.08333300	0.000000000		
## 72	0.08333300	0.000000000		
## 73	0.00000000	0.250000000		
## 74	0.00000000	0.083333000		
## 75	0.04166650	0.041666500		
## 76	0.00000000	0.333333000		
## 77	0.08333300	0.000000000		
## 78	0.33333300	0.000000000		
## 79	0.00000000	0.166667000		
## 80	0.00000000	0.416667000		
## 81	0.08333300	0.000000000		
## 82	0.00000000	0.111111000		
## 83	0.08333300	0.000000000		
## 84	0.12500000	0.000000000		
## 85	0.08333300	0.000000000		
## 86	0.08333300	0.000000000		
## 87	0.08333300	0.000000000		
## 88	0.08333300	0.000000000		
## 89	0.08333300	0.000000000		
## 90	0.08333300	0.000000000		
## 91	0.00000000	0.333333000		
## 92	0.14285700	0.000000000		
## 93	0.00000000	0.083333000		
## 94	0.00000000	0.083333000		
## 95	0.00000000	0.083333000		
## 96	0.08333300	0.000000000		
## 97	0.00000000	0.083333000		
## 98	0.14285700	0.000000000		
## 99	0.08333300	0.000000000		
## 100	0.08333300	0.000000000		
##	CASH_ADVANCE_FREQUENCY	CASH_ADVANCE_TRX	PURCHASES_TRX	CREDIT_LIMIT
## 1	0.27259049	6.2964775	0.001956947	4029.727
## 2	0.10984850	4.2500000	1.000000000	6375.000
## 3	0.33333300	7.0000000	1.000000000	1000.000
## 4	0.00000000	0.0000000	0.000000000	3000.000



## 5	0.16666700	3.0000000	1.000000000	3500.000
## 6	0.12500000	2.0000000	1.000000000	4100.000
## 7	0.08333300	1.0000000	2.000000000	4500.000
## 8	0.25000000	6.0000000	1.000000000	2500.000
## 9	0.41666700	11.0000000	3.000000000	16500.000
## 10	0.00000000	0.0000000	1.000000000	2200.000
## 11	0.00000000	0.0000000	1.000000000	2500.000
## 12	0.00000000	0.0000000	1.000000000	8500.000
## 13	0.16666700	1.0000000	1.000000000	1500.000
## 14	0.02777767	0.6666667	12.000000000	6400.000
## 15	0.00000000	0.0000000	3.000000000	5000.000
## 16	0.00000000	0.0000000	1.000000000	8000.000
## 17	0.00000000	0.0000000	1.000000000	5000.000
## 18	0.00000000	0.0000000	1.000000000	1900.000
## 19	0.00000000	0.0000000	1.000000000	1200.000
## 20	0.00000000	0.0000000	1.000000000	4000.000
## 21	0.08333300	1.0000000	1.000000000	4000.000
## 22	0.00000000	0.0000000	1.000000000	1200.000
## 23	0.00000000	0.0000000	2.000000000	10000.000
## 24	0.04166650	1.5000000	6.500000000	1400.000
## 25	0.16666700	2.0000000	1.000000000	3000.000
## 26	0.16666700	6.0000000	1.000000000	2500.000
## 27	0.41666700	13.0000000	1.000000000	10000.000
## 28	0.00000000	0.0000000	1.000000000	5000.000
## 29	0.25000000	5.0000000	1.000000000	1500.000
## 30	0.00000000	0.0000000	1.000000000	7000.000
## 31	0.00000000	0.0000000	1.000000000	2900.000
## 32	0.00000000	0.0000000	5.000000000	10500.000
## 33	0.25000000	3.0000000	1.000000000	1200.000
## 34	0.41666700	11.0000000	1.000000000	5700.000
## 35	0.00000000	0.0000000	1.000000000	1700.000
## 36	0.04545450	0.5000000	1.000000000	4500.000
## 37	0.00000000	0.0000000	1.000000000	5000.000
## 38	0.33333300	6.0000000	1.000000000	200.000
## 39	0.00000000	0.0000000	3.000000000	7000.000
## 40	0.33333300	9.0000000	1.000000000	5500.000
## 41	0.00000000	0.0000000	1.000000000	3350.000
## 42	0.00000000	0.0000000	1.000000000	6000.000
## 43	0.33333300	4.0000000	1.000000000	600.000
## 44	0.25000000	6.0000000	1.000000000	8954.545
## 45	0.00000000	0.0000000	1.000000000	4000.000
## 46	0.00000000	0.0000000	1.000000000	1500.000
## 47	0.12500000	1.5000000	1.000000000	2500.000
## 48	0.41666700	23.0000000	1.000000000	6000.000
## 49	0.08333300	1.0000000	1.000000000	1000.000
## 50	0.08333300	1.0000000	5.000000000	18000.000
## 51	0.00000000	0.0000000	1.000000000	1000.000
## 52	0.00000000	0.0000000	1.000000000	5000.000
## 53	0.00000000	0.0000000	2.000000000	7000.000
## 54	0.00000000	0.0000000	1.000000000	4000.000
## 55	0.00000000	0.0000000	1.000000000	1500.000
## 56	0.00000000	0.0000000	1.000000000	2000.000
## 57	0.33333300	20.0000000	2.000000000	900.000
## 58	0.08333350	1.0000000	1.500000000	3750.000

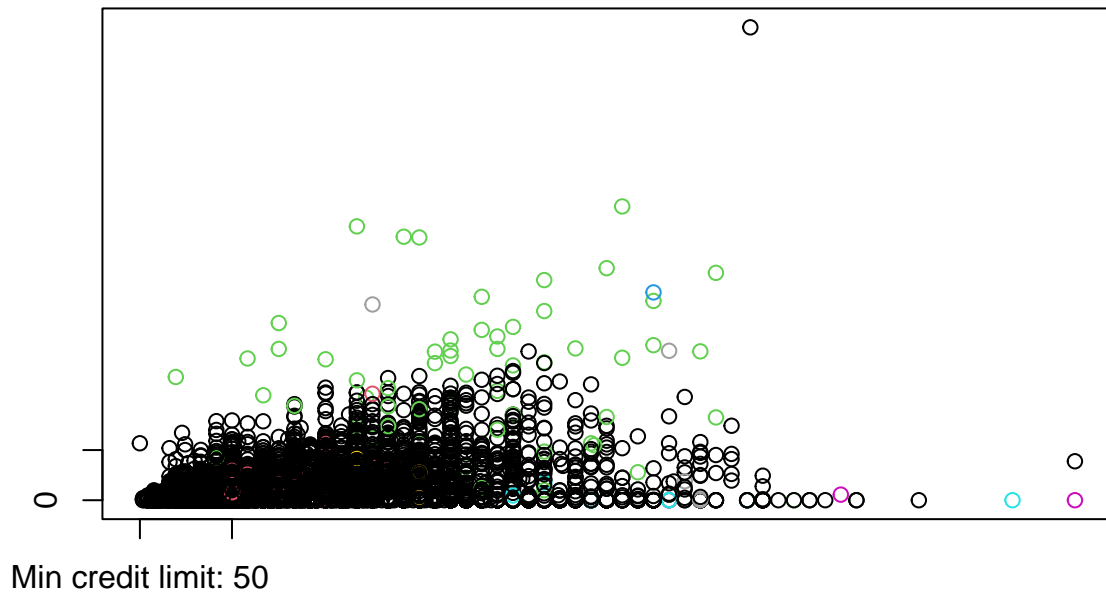
## 59	0.00000000	0.00000000	1.000000000	6000.000
## 60	0.25000000	3.00000000	1.000000000	3500.000
## 61	0.00000000	0.00000000	2.000000000	5500.000
## 62	0.15277767	2.1666667	1.500000000	3000.000
## 63	0.00000000	0.00000000	5.000000000	3200.000
## 64	0.00000000	0.00000000	2.000000000	1500.000
## 65	0.00000000	0.00000000	1.000000000	1000.000
## 66	0.00000000	0.00000000	1.000000000	11500.000
## 67	0.00000000	0.00000000	1.000000000	2900.000
## 68	0.33333300	5.00000000	1.000000000	3000.000
## 69	0.00000000	0.00000000	2.000000000	600.000
## 70	0.00000000	0.00000000	1.000000000	4000.000
## 71	0.00000000	0.00000000	1.000000000	9000.000
## 72	0.00000000	0.00000000	1.000000000	7500.000
## 73	0.00000000	0.00000000	3.000000000	6500.000
## 74	0.00000000	0.00000000	1.000000000	11500.000
## 75	0.00000000	0.00000000	1.000000000	4000.000
## 76	0.00000000	0.00000000	4.000000000	2000.000
## 77	0.08333300	2.00000000	1.000000000	500.000
## 78	0.00000000	0.00000000	4.000000000	15500.000
## 79	0.00000000	0.00000000	2.000000000	1500.000
## 80	0.00000000	0.00000000	5.000000000	4500.000
## 81	0.00000000	0.00000000	1.000000000	1800.000
## 82	0.33333300	6.00000000	1.000000000	2500.000
## 83	0.58333300	19.00000000	1.000000000	2500.000
## 84	0.12500000	1.50000000	1.500000000	3050.000
## 85	0.00000000	0.00000000	1.000000000	5500.000
## 86	0.00000000	0.00000000	1.000000000	4500.000
## 87	0.50000000	16.00000000	4.000000000	1200.000
## 88	0.08333300	1.00000000	1.000000000	1950.000
## 89	0.00000000	0.00000000	1.000000000	7500.000
## 90	0.00000000	0.00000000	1.000000000	2500.000
## 91	0.00000000	0.00000000	4.000000000	4500.000
## 92	0.42857100	11.00000000	1.000000000	500.000
## 93	0.00000000	0.00000000	1.000000000	1200.000
## 94	0.00000000	0.00000000	1.000000000	1200.000
## 95	0.41666700	15.00000000	1.000000000	5000.000
## 96	0.00000000	0.00000000	1.000000000	500.000
## 97	0.00000000	0.00000000	1.000000000	1800.000
## 98	0.42857100	3.00000000	1.000000000	500.000
## 99	0.08333300	1.00000000	1.000000000	4500.000
## 100	0.08333300	1.00000000	2.000000000	2000.000
##	PAYMENTS	MINIMUM_PAYMENTS	PRC_FULL_PAYMENT	TENURE
## 1	1653.130103	968.694784	0.04432393	11.31849
## 2	1273.540476	629.643771	0.03125000	11.75000
## 3	1084.281127	367.409536	0.00000000	12.00000
## 4	150.381107	53.294711	0.00000000	12.00000
## 5	0.000000	0.000000	0.00000000	12.00000
## 6	848.476405	604.791575	0.00000000	12.00000
## 7	0.000000	0.000000	0.00000000	12.00000
## 8	4678.254753	464.860442	0.08333300	12.00000
## 9	21440.298660	1350.823356	0.10000000	12.00000
## 10	70.517745	246.213204	0.00000000	12.00000
## 11	1127.028432	223.627537	0.08333300	12.00000

## 12	4.523555	4.763689	0.00000000	12.00000
## 13	431.732367	92.246251	0.00000000	6.00000
## 14	769.002348	1076.353714	0.00000000	12.00000
## 15	0.000000	0.000000	0.00000000	12.00000
## 16	72.282497	17.530337	0.00000000	12.00000
## 17	151.732627	98.617471	0.50000000	12.00000
## 18	186.923852	7243.733403	0.00000000	7.00000
## 19	918.160924	166.281511	0.00000000	12.00000
## 20	0.000000	0.000000	0.00000000	12.00000
## 21	745.984976	1243.564770	0.00000000	12.00000
## 22	421.348577	206.637191	0.00000000	12.00000
## 23	7280.584479	1013.780486	0.08333300	12.00000
## 24	79.733163	65.870176	0.00000000	12.00000
## 25	628.314620	1288.695715	0.00000000	12.00000
## 26	481.589626	238.636185	0.08333300	12.00000
## 27	10813.829330	1661.518089	0.08333300	12.00000
## 28	0.000000	0.000000	0.00000000	12.00000
## 29	396.056561	308.356999	0.00000000	12.00000
## 30	1703.339459	208.367436	0.00000000	11.00000
## 31	2165.403988	219.961065	0.00000000	12.00000
## 32	1263.292815	1486.059666	0.08333300	12.00000
## 33	347.568201	463.992512	0.00000000	12.00000
## 34	1289.284068	1172.289863	0.00000000	12.00000
## 35	607.535166	43.582049	0.00000000	12.00000
## 36	352.719365	261.538020	0.00000000	11.50000
## 37	190.254352	117.892476	0.00000000	12.00000
## 38	255.103798	198.933162	0.00000000	12.00000
## 39	1542.583506	375.127536	0.00000000	12.00000
## 40	1032.183632	1129.747227	0.00000000	12.00000
## 41	667.190047	182.614350	0.50000000	12.00000
## 42	57.595083	67.192500	0.00000000	12.00000
## 43	243.905711	291.892373	0.00000000	12.00000
## 44	10226.601760	1024.126428	0.10000000	12.00000
## 45	589.674674	472.763536	0.00000000	12.00000
## 46	18.825023	10.074393	0.00000000	12.00000
## 47	5253.799504	411.220052	0.50000000	12.00000
## 48	483.729934	720.971058	0.00000000	12.00000
## 49	233.886707	230.806229	0.00000000	12.00000
## 50	22281.700460	1592.560164	0.09090900	12.00000
## 51	1407.291593	424.623595	0.00000000	12.00000
## 52	0.000000	0.000000	0.00000000	12.00000
## 53	170.180973	161.234787	0.00000000	12.00000
## 54	0.000000	0.000000	0.00000000	12.00000
## 55	95.789433	119.880198	0.00000000	12.00000
## 56	402.155365	322.394919	0.00000000	12.00000
## 57	1072.638864	373.912056	0.08333300	12.00000
## 58	132.130910	111.680274	0.25000000	11.00000
## 59	610.491342	246.804668	0.00000000	12.00000
## 60	218.851473	332.899226	0.00000000	12.00000
## 61	593.220384	1159.310924	0.00000000	12.00000
## 62	711.253366	388.624279	0.00000000	11.50000
## 63	0.000000	0.000000	0.00000000	12.00000
## 64	168.489938	98.859855	0.00000000	12.00000
## 65	173.167499	263.360908	0.00000000	9.00000

## 66	23150.571840	1863.225391	0.33333300	12.00000
## 67	243.137299	112.019269	0.00000000	12.00000
## 68	255.779858	489.808320	0.00000000	12.00000
## 69	162.272808	271.112945	0.00000000	12.00000
## 70	0.000000	0.000000	0.00000000	11.00000
## 71	568.781100	30528.432400	0.00000000	12.00000
## 72	68.205028	200.034390	0.00000000	12.00000
## 73	1441.188109	248.194890	0.00000000	12.00000
## 74	906.092735	224.709183	0.00000000	12.00000
## 75	0.000000	0.000000	0.00000000	12.00000
## 76	500.515114	475.627205	0.00000000	12.00000
## 77	584.679478	139.651483	0.08333300	12.00000
## 78	31.845894	22.081988	0.11111100	12.00000
## 79	9.533313	8.842600	0.00000000	12.00000
## 80	1349.938041	203.987924	0.00000000	12.00000
## 81	45.640106	26.849523	0.00000000	12.00000
## 82	540.687217	570.576789	0.00000000	9.00000
## 83	410.617369	503.627101	0.00000000	12.00000
## 84	819.279904	526.338710	0.00000000	12.00000
## 85	563.748018	83.648417	0.50000000	12.00000
## 86	163.352538	128.336305	0.00000000	12.00000
## 87	1218.866837	200.752923	0.08333300	12.00000
## 88	525.947589	627.660656	0.00000000	12.00000
## 89	78.573329	42.095280	0.00000000	12.00000
## 90	155.972776	82.345774	0.00000000	12.00000
## 91	18.336805	8.745383	0.09090900	12.00000
## 92	28.654864	256.522546	0.00000000	7.00000
## 93	219.692369	310.328990	0.00000000	12.00000
## 94	104.687781	300.921255	0.00000000	12.00000
## 95	507.240956	1102.034141	0.00000000	12.00000
## 96	160.536841	220.943631	0.00000000	12.00000
## 97	228.417814	175.836378	1.00000000	12.00000
## 98	733.901545	122.835190	0.25000000	7.00000
## 99	1215.965938	99.950697	1.00000000	12.00000
## 100	440.707307	517.115912	0.00000000	12.00000

*# emerging group 5 & 6*

```
plot(jitter(as.numeric(boa.df.new$CASH_ADVANCE)) ~
      jitter(as.numeric(boa.df.new$CREDIT_LIMIT)),
     col=boa.hc.purchases,
     yaxt="n",
     xaxt="n",
     ylab="",
     xlab="")
axis(1, at=c(50, 3000), labels=c("Min credit limit: 50", "Max credit limit: 3000"))
axis(2, at=c(0, 5000), labels=levels(boa.df.new$CASH_ADVANCE))
```



## K Means

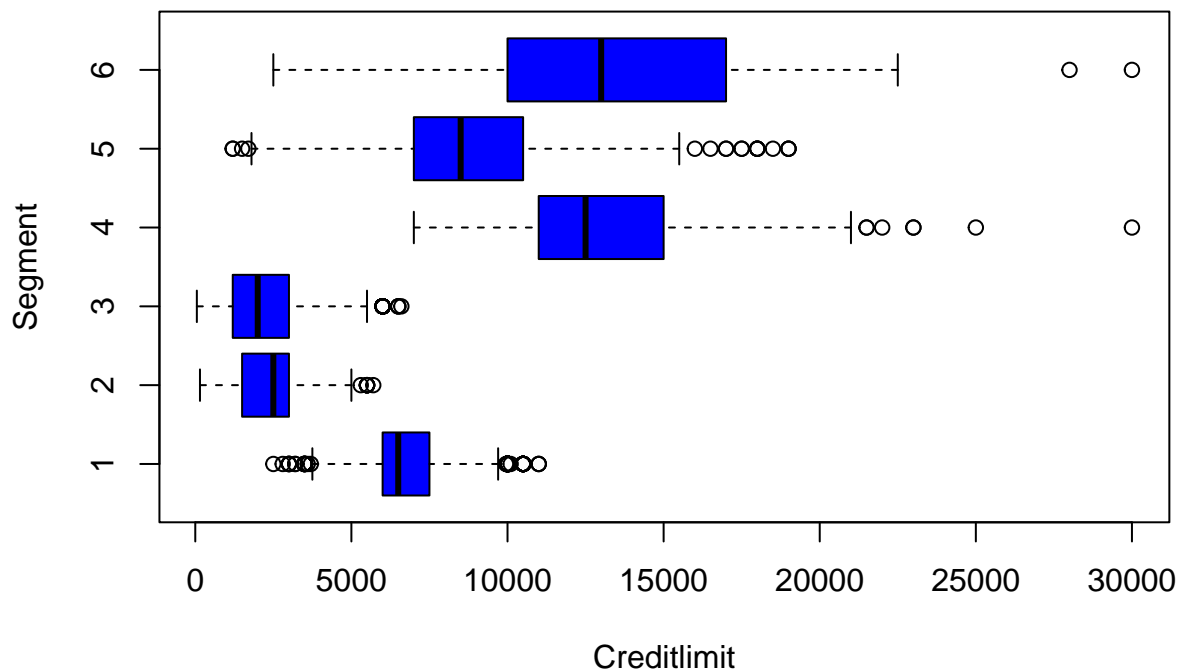
```
boa.df.num <- boa.df.new
set.seed(12345)
```

```
boa.k<-kmeans(boa.df.num,centers = 6)
boa.summ(boa.df.new,boa.k$cluster)
```

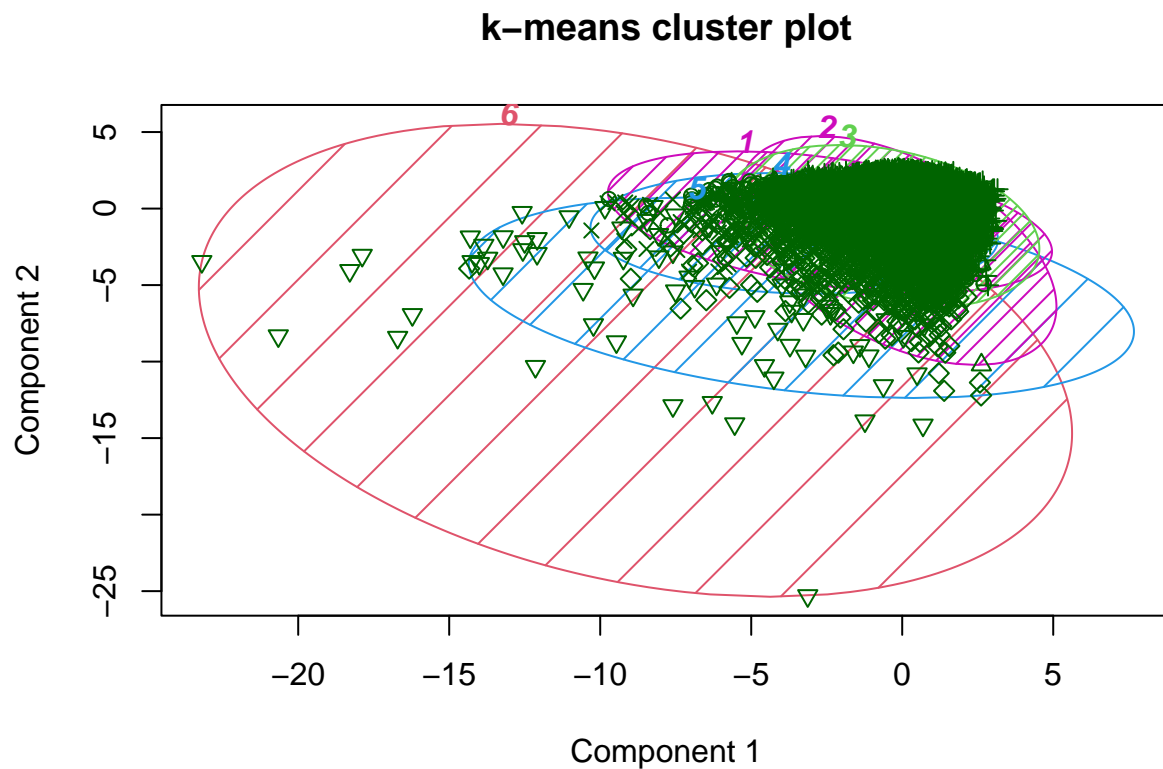
```
##   Group.1      X  BALANCE BALANCE_FREQUENCY ONEOFF_PURCHASES
## 1      1 3473.192 1666.250          0.9066200          876.0499
## 2      2 2383.478 1033.415          0.8789913          318.9521
## 3      3 7032.866  699.745          0.8353757          242.3384
## 4      4 2528.185 2890.655          0.9091980         2042.4956
## 5      5 3414.552 6100.887          0.9765424          476.2747
## 6      6 2505.634 5043.583          0.8593260         7289.1705
##  INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1          527.9718          786.3986          0.5690613
## 2          257.2039          509.2856          0.4292796
## 3          284.8894          514.0729          0.4751120
## 4         1031.9138          564.0622          0.7049679
## 5          453.9301         4984.7457          0.3654152
## 6         2768.6097         7546.1825          0.6296676
##  ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 1          0.3235480          0.3949300
## 2          0.1546873          0.3115449
## 3          0.1246642          0.3757905
```

```
## 4          0.4798681          0.5005175
## 5          0.1574601          0.2772320
## 6          0.4841991          0.5139087
##  CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS
## 1          0.11583796          2.555057      20.183857      6801.164 1991.591
## 2          0.10876211          2.191609      10.001298      2454.562 1009.112
## 3          0.11333867          2.503684          9.976717      2219.845  886.118
## 4          0.07900764          1.497942      36.674897     13110.905 3347.170
## 5          0.41935725         12.966616      15.556904      8741.130 3863.819
## 6          0.28261880         11.978495      65.526882     13292.473 21525.445
##  MINIMUM_PAYMENTS PRC_FULL_PAYMENT   TENURE
## 1          600.2507          0.19050166 11.80219
## 2          612.3474          0.11460082 11.59299
## 3          476.9178          0.16859332 11.23224
## 4          855.5973          0.23079449 11.84156
## 5         3916.8317          0.02294102 11.57511
## 6         2457.0669          0.31323406 11.78495
```

```
boxplot(boa.df.num$CREDIT_LIMIT~boa.k$cluster,
        col="blue",
        xlab="Creditlimit",
        ylab="Segment",
        horizontal = TRUE)
```



```
# visualizing the overall clusters
library(cluster)
clusplot(boa.df.new,boa.k$cluster,
          color="TRUE",
          shade="TRUE",
          labels= 4,
          main="k-means cluster plot")
```



These two components explain 43.82 % of the point variability.

We can focus on Group 2,4,because 5,6 are overlapping

M Clust

```
library(mclust)
boa.mc<-Mclust(boa.df.num)
summary(boa.mc)
```

```
## -----
## Gaussian finite mixture model fitted by EM algorithm
## -----
##
## Mclust VEV (ellipsoidal, equal shape) model with 4 components:
##
## log-likelihood    n  df      BIC      ICL
##      -604665 8950 635 -1215108 -1215387
##
## Clustering table:
```

```
##      1      2      3      4
## 2292 1367 4002 1289
```

```
boa.mc4<-Mclust(boa.df.num, G=8)
summary(boa.mc4)
```

```
## -----
## Gaussian finite mixture model fitted by EM algorithm
## -----
##
## Mclust EEV (ellipsoidal, equal volume and shape) model with 8 components:
##
## log-likelihood    n    df      BIC      ICL
##      -605311.5 8950 1248 -1221979 -1223711
##
## Clustering table:
##      1      2      3      4      5      6      7      8
## 285 239 629 1665 2024 1382 2593 133
```

We will use the 1 model because it has a lower log likelihood.

Comparing the Different Models with BIC

```
BIC(boa.mc,boa.mc4)
```

```
##      df      BIC
## boa.mc  635 1215108
## boa.mc4 1248 1221979
```

The 1st model has a lower BIC meaning it is group better than the first model.

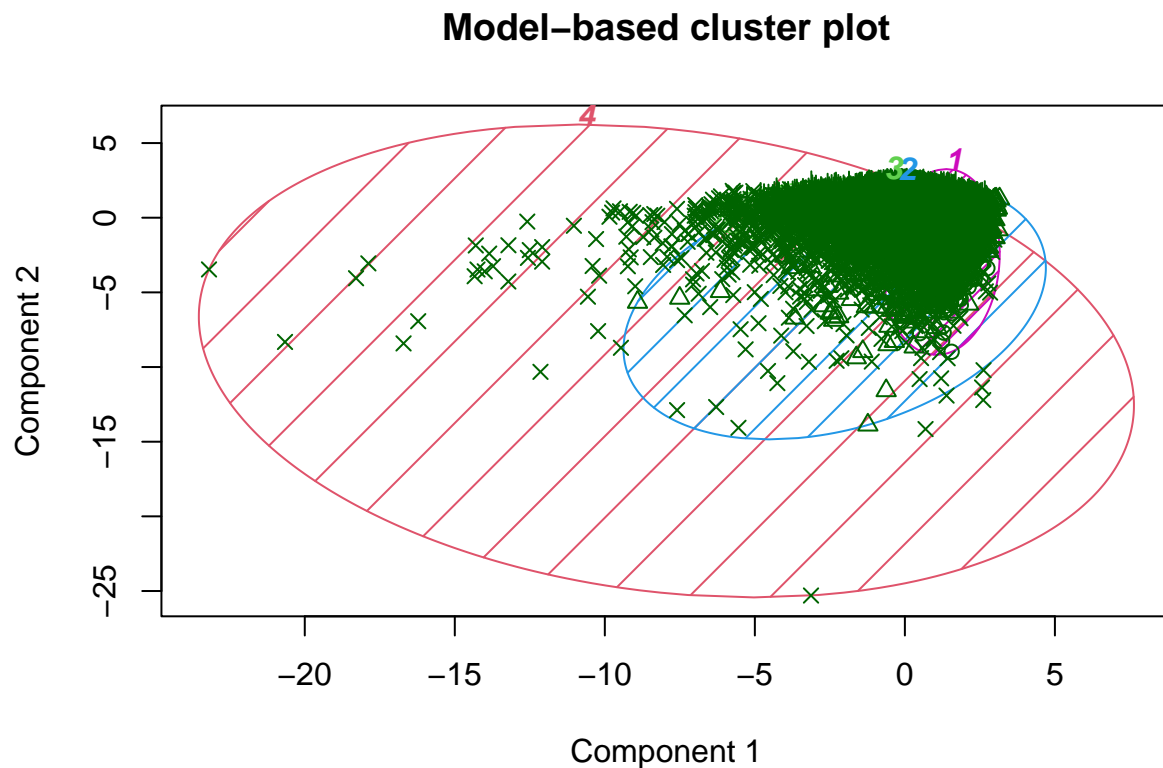
```
# check the summary
boa.summ(boa.df.new,boa.mc$class)
```

```
## Group.1      X  BALANCE BALANCE_FREQUENCY ONEOFF_PURCHASES
## 1      1 4418.907 2012.5306      0.9239582      63.03483
## 2      2 4521.628 2236.8750      0.8571661      464.04798
## 3      3 4653.367 561.6782      0.8217121      554.93689
## 4      4 3974.980 3168.1034      0.9880706      1786.36743
## INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1      2.405345 1433.551849      0.04999687
## 2      184.925099 2214.636661      0.41279185
## 3      457.894718 1.007874      0.67792812
## 4      1232.160667 1895.852435      0.77322715
## ONEOFF_PURCHASES_FREQUENCY PURCHASES_INSTALLMENTS_FREQUENCY
## 1      0.04559752      0.004399357
## 2      0.16354572      0.253374064
## 3      0.24510078      0.515989632
## 4      0.39024551      0.651883651
## CASH_ADVANCE_FREQUENCY CASH_ADVANCE_TRX PURCHASES_TRX CREDIT_LIMIT PAYMENTS
## 1      0.2417214939      5.452006981      0.7966841      3694.815 1125.984
## 2      0.2358160680      5.029992685      7.9378200      5039.927 2775.862
```



```
## 3      0.0005849343      0.006996502      15.9882559      4222.633 1064.733
## 4      0.2566434888      7.507370054      42.6617533      6181.730 3782.168
##  MINIMUM_PAYMENTS PRC_FULL_PAYMENT  TENURE
## 1      678.4855      0.01970663 11.87347
## 2      885.0488      0.09316473 10.65838
## 3      275.9230      0.26566551 11.58646
## 4      2788.9509      0.10863381 11.58029
```

```
library(cluster)
clusplot(boa.df.new, boa.mc$class, color=TRUE, shade=TRUE,
         labels=5, lines=0, main="Model-based cluster plot")
```



These two components explain 43.82 % of the point variability.

```
# summarizing clusters
boa.summ(boa.df.new, boa.k$cluster)
```

```
##  Group.1      X  BALANCE BALANCE_FREQUENCY ONEOFF_PURCHASES
## 1      1 3473.192 1666.250      0.9066200      876.0499
## 2      2 2383.478 1033.415      0.8789913      318.9521
## 3      3 7032.866  699.745      0.8353757      242.3384
## 4      4 2528.185 2890.655      0.9091980     2042.4956
## 5      5 3414.552 6100.887      0.9765424      476.2747
## 6      6 2505.634 5043.583      0.8593260     7289.1705
##  INSTALLMENTS_PURCHASES CASH_ADVANCE PURCHASES_FREQUENCY
## 1      527.9718      786.3986      0.5690613
## 2      257.2039      509.2856      0.4292796
```

## 3	284.8894	514.0729	0.4751120		
## 4	1031.9138	564.0622	0.7049679		
## 5	453.9301	4984.7457	0.3654152		
## 6	2768.6097	7546.1825	0.6296676		
##	ONEOFF_PURCHASES_FREQUENCY	PURCHASES_INSTALLMENTS_FREQUENCY			
## 1	0.3235480		0.3949300		
## 2	0.1546873		0.3115449		
## 3	0.1246642		0.3757905		
## 4	0.4798681		0.5005175		
## 5	0.1574601		0.2772320		
## 6	0.4841991		0.5139087		
##	CASH_ADVANCE_FREQUENCY	CASH_ADVANCE_TRX	PURCHASES_TRX	CREDIT_LIMIT	PAYMENTS
## 1	0.11583796	2.555057	20.183857	6801.164	1991.591
## 2	0.10876211	2.191609	10.001298	2454.562	1009.112
## 3	0.11333867	2.503684	9.976717	2219.845	886.118
## 4	0.07900764	1.497942	36.674897	13110.905	3347.170
## 5	0.41935725	12.966616	15.556904	8741.130	3863.819
## 6	0.28261880	11.978495	65.526882	13292.473	21525.445
##	MINIMUM_PAYMENTS	PRC_FULL_PAYMENT	TENURE		
## 1	600.2507	0.19050166	11.80219		
## 2	612.3474	0.11460082	11.59299		
## 3	476.9178	0.16859332	11.23224		
## 4	855.5973	0.23079449	11.84156		
## 5	3916.8317	0.02294102	11.57511		
## 6	2457.0669	0.31323406	11.78495		

**REPORT** Our business goal was to see which variables would emerge when we remove purchases from the variables so we can better see what to focus on in order to generate and maintain customers. We removed purchases because we believe this is a dominant variable in determining the clusters so we wanted to explore other variables that would drive insight. Purchases was believed to be dominant as this is the purpose of the credit card and is the outcome that Bank of America wants their customers to do. After exploring the emerging variables, we can develop a business strategy to leverage these factors. After modeling these variables in clusters, we found that balance (how much is left in their account to make purchases), cash advance (cash in advance given by the user), and credit limit (limit of credit card for user) were most significant. With these variables in mind, we created six customer segments. We identified the characteristics of these customers, and created a plan to appeal to them.

The first customer segment we identified are those that have low balance, low cash advance, and high full payment. We propose to increase their credit limit with low interest rate and discounts on purchases. This will appeal to them since they save money with low rates and discounts, but have the advantage of a higher credit limit which isn't a big risk for Bank of America as this segment pays in full.

Customer segment 2 are those that have low balance and low frequency of purchase. For these customers, we would provide incentives like reward programs with attractive benefits such as increase free air miles in order to encourage them to purchase more frequently.

Customer segment 3 are those that have low balance, low cash advance, low purchase frequency, low credit limit, low minimum payment, and a considerably high full payment. We would propose promotions for when they make a certain amount of payments in a month. This would encourage them to not only spend more but spend more often. Because they have a low credit limit and pay in full, it isn't a high risk for the customer to spend more than they have.

Customer segment 4 are those that have low balance, high one off purchases (maximum purchase amount done in one go), high cash advance frequency, high minimum, and full payment. Since these customers are typically bigger spenders, we would propose to increase their credit limit with cash backs for purchases. This

would benefit them a lot since they are spending large amounts, they'll get a significant amount with cash back rewards.

Customer segment 5 are those that have a high balance, high one off purchases, high cash advance, high minimum payments, but low full payments. Since these customers aren't able to pay in full we can offer educational programs and materials for them to plan better, but also increase credit limit with a certain percent since this helps with ROI. Since this customer has a higher balance, raising the credit limit would be appealing for them.

Lastly, customer segment 6 are those that have a high balance, high cash advance, and high frequency of purchase. We can offer them a higher credit limit, cash back benefits, and reward programs to show them that they are a valued customer and to maintain their purchasing habits.

After determining which variables were dominant when taking out purchases, we were able to identify core customer profiles and propose this marketing strategy to generate more purchase amounts and frequencies to get customers to use credit cards more.