

Principal Component Analysis on Istanbul Airbnb Dataset

```
library(data.table)#Data. table is an extension of data. frame package in R.  
It is widely used for fast aggregation of large datasets,  
## Warning: package 'data.table' was built under R version 3.6.2  
library(Hmisc)#data analysis funs  
## Warning: package 'Hmisc' was built under R version 3.6.2  
## Loading required package: lattice  
## Warning: package 'lattice' was built under R version 3.6.2  
## Loading required package: survival  
## Warning: package 'survival' was built under R version 3.6.2  
## Loading required package: Formula  
## Loading required package: ggplot2  
##  
## Attaching package: 'Hmisc'  
## The following objects are masked from 'package:base':  
##  
##   format.pval, units  
library(dplyr)  
## Warning: package 'dplyr' was built under R version 3.6.2  
##  
## Attaching package: 'dplyr'  
## The following objects are masked from 'package:Hmisc':  
##  
##   src, summarize  
## The following objects are masked from 'package:data.table':  
##  
##   between, first, last  
## The following objects are masked from 'package:stats':  
##  
##   filter, lag
```

```

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

library(tidyverse)

## Warning: package 'tidyverse' was built under R version 3.6.2

## -- Attaching packages ----- tidyverse 1.3.0 --

## v tibble 2.1.3      v purrr 0.3.3
## v tidyr 1.0.2       v stringr 1.4.0
## v readr 1.3.1       v forcats 0.4.0

## Warning: package 'tidyr' was built under R version 3.6.2
## Warning: package 'purrr' was built under R version 3.6.2
## Warning: package 'stringr' was built under R version 3.6.2

## -- Conflicts ----- tidyverse
e_conflicts() --
## x dplyr::between() masks data.table::between()
## x dplyr::filter() masks stats::filter()
## x dplyr::first() masks data.table::first()
## x dplyr::lag() masks stats::lag()
## x dplyr::last() masks data.table::last()
## x dplyr::src() masks Hmisc::src()
## x dplyr::summarize() masks Hmisc::summarize()
## x purrr::transpose() masks data.table::transpose()

library(ggplot2)
library(plotly)

## Warning: package 'plotly' was built under R version 3.6.2

##
## Attaching package: 'plotly'

## The following object is masked from 'package:Hmisc':
##
## subplot

## The following object is masked from 'package:ggplot2':
##
## last_plot

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

```

```
## The following object is masked from 'package:graphics':
##
## layout
library(GGally)
## Warning: package 'GGally' was built under R version 3.6.2
## Registered S3 method overwritten by 'GGally':
## method from
## +.gg ggplot2
##
## Attaching package: 'GGally'
## The following object is masked from 'package:dplyr':
##
## nasa
library(ggthemes)
## Warning: package 'ggthemes' was built under R version 3.6.2
library(psych)
## Warning: package 'psych' was built under R version 3.6.2
##
## Attaching package: 'psych'
## The following object is masked from 'package:Hmisc':
##
## describe
## The following objects are masked from 'package:ggplot2':
##
## %+%, alpha
library(relaimpo)
## Warning: package 'relaimpo' was built under R version 3.6.2
## Loading required package: MASS
##
## Attaching package: 'MASS'
## The following object is masked from 'package:plotly':
##
## select
## The following object is masked from 'package:dplyr':
##
## select
```

```
## Loading required package: boot
## Warning: package 'boot' was built under R version 3.6.2
##
## Attaching package: 'boot'
## The following object is masked from 'package:psych':
##
##     logit
## The following object is masked from 'package:survival':
##
##     aml
## The following object is masked from 'package:lattice':
##
##     melanoma
## Loading required package: survey
## Warning: package 'survey' was built under R version 3.6.2
## Loading required package: grid
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
## The following objects are masked from 'package:tidyr':
##
##     expand, pack, unpack
##
## Attaching package: 'survey'
## The following object is masked from 'package:Hmisc':
##
##     deff
## The following object is masked from 'package:graphics':
##
##     dotchart
## Loading required package: mitools
## Warning: package 'mitools' was built under R version 3.6.2
## This is the global version of package relaimpo.
## If you are a non-US user, a version with the interesting additional metric
pmvd is available
```

```
## from Ulrike Groempings web site at prof.beuth-hochschule.de/groemping.
```

```
library(e1071)
```

```
## Warning: package 'e1071' was built under R version 3.6.2
```

```
##
```

```
## Attaching package: 'e1071'
```

```
## The following object is masked from 'package:Hmisc':
```

```
##
```

```
##      impute
```

Loading Dataset and getting overall feel of data

```
AirbnbIstanbul<-read.csv("C:/Alok/OneDrive/Rutgers_MITA/Semester2/MVA/R/Airbn  
bIstanbul.csv",stringsAsFactors = FALSE)
```

```
Istanbul <- copy(AirbnbIstanbul)
```

```
View(Istanbul)
```

```
str(Istanbul)
```

```
## 'data.frame': 16251 obs. of 16 variables:
```

```
## $ id : int 4826 20815 25436 27271 28277 28308  
28318 29241 30697 33368 ...
```

```
## $ name : chr "The Place" "The Bosphorus from Th  
e Comfy Hill" "House for vacation rental furnutare" "LOVELY APT. IN PERFECT L  
OCATION" ...
```

```
## $ host_id : int 6603 78838 105823 117026 121607 12  
1695 121721 125742 132137 135136 ...
```

```
## $ host_name : chr "Kaan" "GÃ¼lder" "Yesim" "Mutlu" .  
..
```

```
## $ neighbourhood_group : logi NA NA NA NA NA NA ...
```

```
## $ neighbourhood : chr "Uskudar" "Besiktas" "Besiktas" "B  
eyoglu" ...
```

```
## $ latitude : num 41.1 41.1 41.1 41 41 ...
```

```
## $ longitude : num 29.1 29 29 29 29 ...
```

```
## $ room_type : chr "Entire home/apt" "Entire home/apt  
" "Entire home/apt" "Entire home/apt" ...
```

```
## $ price : int 554 100 211 237 591 237 633 264 59  
6 295 ...
```

```
## $ minimum_nights : int 1 30 21 5 3 1 3 3 1 2 ...
```

```
## $ number_of_reviews : int 1 41 0 2 0 0 0 0 1 1 ...
```

```
## $ last_review : chr "6/1/2009" "11/7/2018" "" "5/4/201  
8" ...
```

```
## $ reviews_per_month : num 0.01 0.38 NA 0.04 NA NA NA NA 0.01  
0.02 ...
```

```
## $ calculated_host_listings_count: int 1 2 1 1 13 1 1 1 1 2 ...
```

```
## $ availability_365 : int 365 49 83 228 356 365 365 365 365  
232 ...
```

```
#Checking number f rows and columns
```

```
dim(Istanbul)
```



```
## price                                0.030099631      0.04701
5234
## minimum_nights                       -0.020916303      0.01529
7298
## number_of_reviews                    0.174662879      0.04323
0154
## calculated_host_listings_count       1.000000000      0.17306
8073
## availability_365                     0.173068073      1.00000
0000
```

#Very little correlation between 'Number of reviews and calculated host listing' & 'calcHostlisting and availability365'

#PCA

#Applying PCA on numeric data as it's not much recommended for categorical data

```
Istanbul_ip_pca <- prcomp(Istanbul_num2,scale=TRUE)
```

```
Istanbul_ip_pca
```

```
## Standard deviations (1, .., p=7):
```

```
## [1] 1.1329266 1.0916193 1.0156849 0.9917434 0.9620390 0.9009618 0.8788757
```

```
##
```

```
## Rotation (n x k) = (7 x 7):
```

```
##                                PC1          PC2          PC3
PC4
## latitude                      -0.13671624  0.67300964 -0.17720115 -0.0385
79150
## longitude                     0.24384922 -0.62742158  0.19983543  0.0991
97960
## price                         -0.15664516  0.18106565  0.61612379  0.5492
57096
## minimum_nights                0.02663696  0.09422947  0.58495893 -0.7914
22636
## number_of_reviews             -0.45778510 -0.27942467 -0.32052131 -0.2365
68651
## calculated_host_listings_count -0.64683557 -0.17371645 -0.04188819 -0.0071
09716
## availability_365              -0.51830063 -0.05899714  0.31998679  0.0679
92594
```

```
##                                PC5          PC6          PC7
## latitude                      0.068157824  0.69490244 -0.08912466
## longitude                     0.047268165  0.70233520 -0.03995339
## price                         0.504699877 -0.07281020 -0.03691007
## minimum_nights                0.129066213  0.01560196  0.07059059
## number_of_reviews             0.567198941 -0.03157838 -0.48056653
## calculated_host_listings_count -0.008167183  0.12443764  0.73080076
## availability_365              -0.632422989  0.04253481 -0.46808248
```

#PC1--> Dominated by negative effect of calculated_host_listings_count and availability_365 and no of reviews

*#PC2--> major +ve effect of Latitude and negative effect of Longitude
#PC3 --> Major +ve effect of minimum_nights and Price
#PC4 --> Major negative effect of minimum_nights and +ve effect of price
#PC5 --> Major negative effect of availability_365 and +ve effect of number_of_reviews*

#Summary pf PCAs
summary(Istanbul_ip_pca)

Importance of components:

##	PC1	PC2	PC3	PC4	PC5	PC6	PC7
## Standard deviation	1.1329	1.0916	1.0157	0.9917	0.9620	0.9010	0.8789
## Proportion of Variance	0.1834	0.1702	0.1474	0.1405	0.1322	0.1160	0.1104
## Cumulative Proportion	0.1834	0.3536	0.5010	0.6415	0.7737	0.8897	1.0000

#As per Summary output, 'Cumulative Proportion' field, 88.97% of Cumulative variance is explained by PC1, PC2,----PC6

#So we will have to include PC1 till PC6 to prevent Loss of Information.

Insights from Above PCA Output

Contents of Principal Components:

PC1→ Dominated by negative effect of `calculated_host_listings_count` and `availability_365` and no of reviews

PC2→ major +ve effect of latitude and negative effect of longitude

PC3 → Major +ve effect of `minimum_nights` and Price

PC4 → Major negative effect of `minimum_nights` and +ve effect of price

PC5 → Major negative effect of `availability_365` and +ve effect of `number_of_reviews`

From Summary of Principal components,

Proportion of Variance, does not explain much of variance individually.

‘Cumulative Proportion’ field, 88.97% of Cumulative variance is explained by PC1, PC2,—PC6

So we will have to include PC1 till PC6 to prevent loss of Information.

Plotting PCs

```
# A table containing eigenvalues and %'s accounted, follows
# Eigenvalues are sdev^2

(eigen_Istanbul <- Istanbul_ip_pca$sdev^2)

## [1] 1.2835228 1.1916328 1.0316158 0.9835550 0.9255190 0.8117322 0.7724225

names(eigen_Istanbul) <- paste("PC",1:7,sep="")
eigen_Istanbul

##          PC1          PC2          PC3          PC4          PC5          PC6          PC7
## 1.2835228 1.1916328 1.0316158 0.9835550 0.9255190 0.8117322 0.7724225

names(eigen_Istanbul)

## [1] "PC1" "PC2" "PC3" "PC4" "PC5" "PC6" "PC7"

#Taking Sum of all Eigen values
sumlambdas1 <- sum(eigen_Istanbul)
sumlambdas1 #sum of Eigenvalues is total var of ur dataset

## [1] 7
```

```

propvar1 <- eigen_Istanbul/sumlambdas1
propvar1 #Propvar1 gives the percentage of variance for each PC component

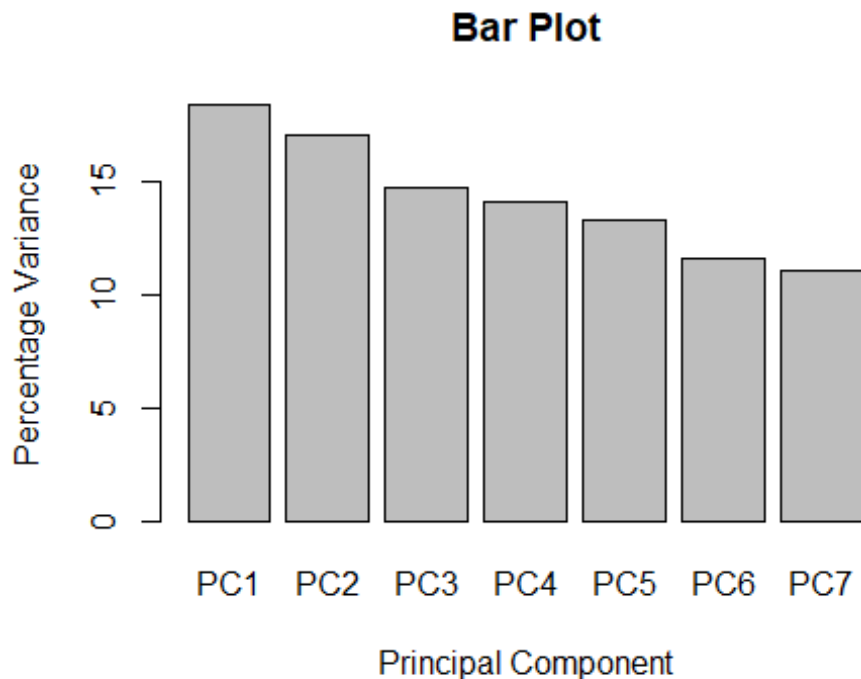
##      PC1      PC2      PC3      PC4      PC5      PC6      PC7
## 0.1833604 0.1702333 0.1473737 0.1405079 0.1322170 0.1159617 0.1103461

#Percentage of total variance
percentvar <- (eigen_Istanbul/sumlambdas1) *100
percentvar

##      PC1      PC2      PC3      PC4      PC5      PC6      PC7
## 18.33604 17.02333 14.73737 14.05079 13.22170 11.59617 11.03461

#Bar plot of Percentage variance
barplot(percentvar, main = "Bar Plot", xlab = "Principal Component", ylab = "
Percentage Variance")

```



```

#[1] 0.1833604 0.1702333 0.1473737 0.1405079 0.1322170 0.1159617 0.1103461
#OP says none of the component explains much variance so we will have to include all
#Can do plotting at this stage
cumvar_Istanbul <- cumsum(propvar1)
cumvar_Istanbul #This variable has cumulative sum of variance

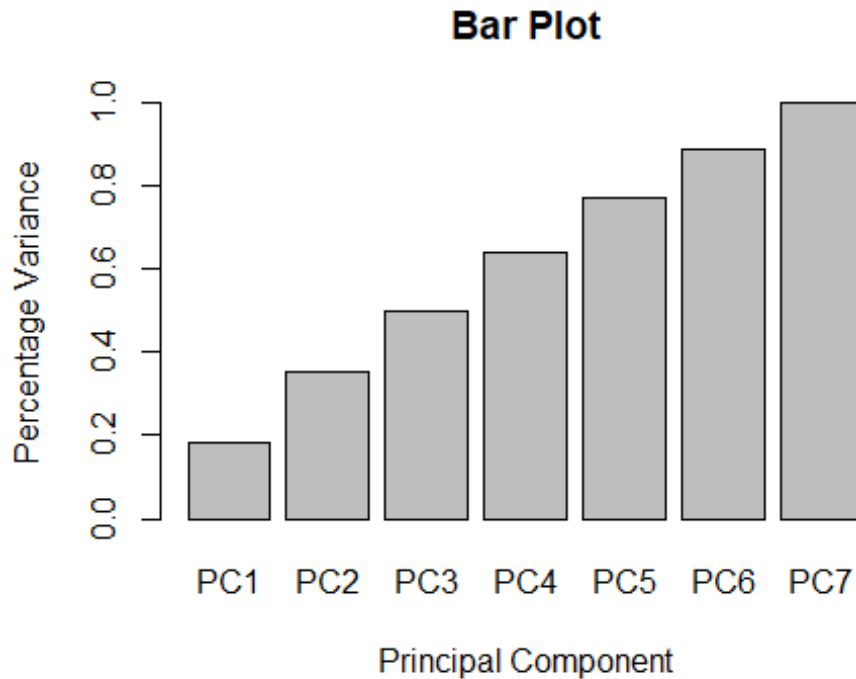
##      PC1      PC2      PC3      PC4      PC5      PC6      PC7
## 0.1833604 0.3535937 0.5009673 0.6414752 0.7736922 0.8896539 1.0000000

```

```
#PC1 to PC6 explains 88.96% of variance
```

```
#Bar plot of Cummulative Percentage variance
```

```
barplot(cumvar_Istanbul, main = "Bar Plot", xlab = "Principal Component", ylab = "Percentage Variance")
```



```
matlambdas <- rbind(eigen_Istanbul,propvar1,cumvar_Istanbul)
rownames(matlambdas) <- c("Eigenvalues","Prop. variance","Cum. prop. variance")
round(matlambdas,4)
```

```
##              PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Eigenvalues    1.2835  1.1916  1.0316  0.9836  0.9255  0.8117  0.7724
## Prop. variance  0.1834  0.1702  0.1474  0.1405  0.1322  0.1160  0.1103
## Cum. prop. variance 0.1834 0.3536 0.5010 0.6415 0.7737 0.8897 1.0000
```

```
summary(Istanbul_ip_pca)
```

```
## Importance of components:
```

```
##              PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation  1.1329  1.0916  1.0157  0.9917  0.9620  0.9010  0.8789
## Proportion of Variance 0.1834 0.1702 0.1474 0.1405 0.1322 0.1160 0.1104
## Cumulative Proportion 0.1834 0.3536 0.5010 0.6415 0.7737 0.8897 1.0000
```

```
# PCA Rotation
```

```
Istanbul_ip_pca$rotation #= print(Istanbul_ip_pca)
```

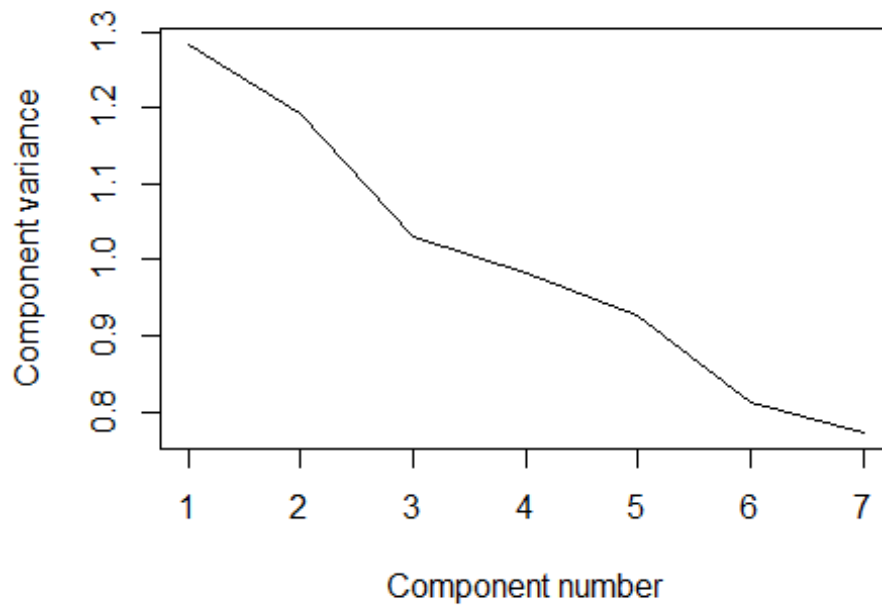
```

##                                PC1          PC2          PC3
PC4
## latitude                      -0.13671624  0.67300964 -0.17720115 -0.0385
79150
## longitude                      0.24384922 -0.62742158  0.19983543  0.0991
97960
## price                         -0.15664516  0.18106565  0.61612379  0.5492
57096
## minimum_nights                 0.02663696  0.09422947  0.58495893 -0.7914
22636
## number_of_reviews             -0.45778510 -0.27942467 -0.32052131 -0.2365
68651
## calculated_host_listings_count -0.64683557 -0.17371645 -0.04188819 -0.0071
09716
## availability_365              -0.51830063 -0.05899714  0.31998679  0.0679
92594
##                                PC5          PC6          PC7
## latitude                      0.068157824  0.69490244 -0.08912466
## longitude                     0.047268165  0.70233520 -0.03995339
## price                         0.504699877 -0.07281020 -0.03691007
## minimum_nights                0.129066213  0.01560196  0.07059059
## number_of_reviews             0.567198941 -0.03157838 -0.48056653
## calculated_host_listings_count -0.008167183  0.12443764  0.73080076
## availability_365              -0.632422989  0.04253481 -0.46808248

#Scree Plot
plot(eigen_Istanbul, xlab = "Component number", ylab = "Component variance",
type = "l", main = "Scree diagram")

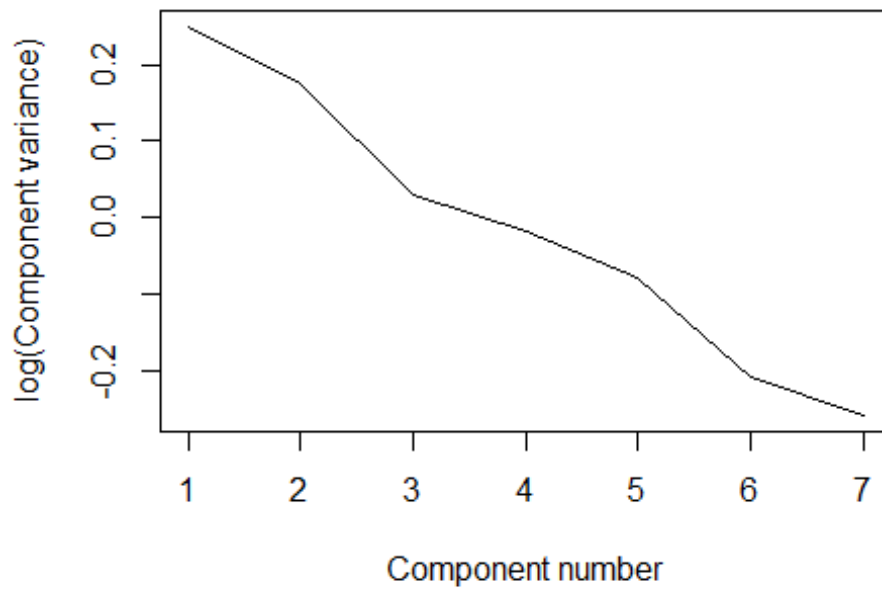
```

Scree diagram



```
plot(log(eigen_Istanbul), xlab = "Component number", ylab = "log(Component variance)", type="l", main = "Log(eigenvalue) diagram")
```

Log(eigenvalue) diagram



*#Conclusion : As per Summary output, 'Cumulative Proportion' field, 88.97% of Cumulative variance is explained by PC1, PC2,----PC6
#So we will have to include PC1 till PC6 to prevent loss of Information.*

From Summary of Pincipal components,

Proportion of Variance, does not explain much of variance individually.

‘Cumulative Proportion’ field, 88.97% of Cumulative variance is explained by PC1, PC2,—PC6

So we will have to include PC1 till PC6 to prevent loss of Information.

Binding Principal Components and categorical columns together

Printing our new Dataset after PCA

```
Istanbultyp_pca <- cbind(data.frame(neighbourhood,room_type),Istanbul_ip_pca$x)
names(Istanbultyp_pca)

## [1] "neighbourhood" "room_type"      "PC1"           "PC2"
## [5] "PC3"           "PC4"           "PC5"           "PC6"
## [9] "PC7"

#Istanbultyp_pca This is our new dataset
head(Istanbultyp_pca,5)

##  neighbourhood      room_type      PC1      PC2      PC3      P
C4
## 1      Uskudar Entire home/apt -0.0198208 0.20770197 0.3882550 0.338905
61
## 2      Besiktas Entire home/apt 0.2611837 0.09049684 -0.6360656 -1.248468
31
## 3      Besiktas Entire home/apt 1.0337257 0.76983155 -0.1140098 -0.501509
68
## 4      Beyoglu Entire home/apt 0.4432095 0.24065558 -0.0297182 -0.012200
13
## 5      Sisli Entire home/apt -1.0889860 0.15267916 0.3022794 0.245316
84
##          PC5      PC6      PC7
## 1 -0.56734954 0.88516538 -0.6549925
## 2 1.94434798 0.94668791 -0.3117832
## 3 0.71092336 1.07091534 0.3568210
## 4 -0.06654501 0.02912375 -0.1133420
## 5 -0.58952855 0.47097856 0.5969920

#Renaming Principal components
names(Istanbultyp_pca) <- c("Neighbourhood", "Room_Type", "calc_Review_365_Ne
gative", "Lattitude_Positive_Longi_Negate",
                           "MinNight_Price_Positive", "MinNightNegative_Price
```

```

Pos", "availabilityNegate_Reviews_Pos",
      "Positive_Lat_Long", "CalcHostListing_Pos")

#This is Our new dataset
names(Istanbultyp_pca)

## [1] "Neighbourhood"          "Room_Type"
## [3] "calc_Review_365_Negative" "Latitude_Positive_Longi_Negate"
## [5] "MinNight_Price_Positive" "MinNightNegative_PricePos"
## [7] "availabilityNegate_Reviews_Pos" "Positive_Lat_Long"
## [9] "CalcHostListing_Pos"

#View(Istanbultyp_pca)
dim(Istanbultyp_pca)

## [1] 16251      9

head(Istanbultyp_pca, 5)

##   Neighbourhood      Room_Type calc_Review_365_Negative
## 1      Uskudar Entire home/apt          -0.0198208
## 2      Besiktas Entire home/apt           0.2611837
## 3      Besiktas Entire home/apt           1.0337257
## 4      Beyoglu Entire home/apt           0.4432095
## 5      Sisli Entire home/apt            -1.0889860
##   Latitude_Positive_Longi_Negate MinNight_Price_Positive
## 1                   0.20770197              0.3882550
## 2                   0.09049684             -0.6360656
## 3                   0.76983155             -0.1140098
## 4                   0.24065558             -0.0297182
## 5                   0.15267916              0.3022794
##   MinNightNegative_PricePos availabilityNegate_Reviews_Pos Positive_Lat_Long
## 1                   0.33890561                -0.56734954      0.885165
## 2                   -1.24846831                1.94434798      0.946687
## 3                   -0.50150968                0.71092336      1.070915
## 4                   -0.01220013               -0.06654501      0.029123
## 5                   0.24531684               -0.58952855      0.470978
##   CalcHostListing_Pos
## 1          -0.6549925
## 2          -0.3117832
## 3           0.3568210
## 4          -0.1133420
## 5           0.5969920

```

Finding Means and standard deviations by Room Types

Means of scores for all the PC's classified by Survival status so da u can perform ttest on that

```
tabmeansPC1 <- aggregate(Istanbultyp_pca[,c(3,4,5,6,7,8,9)],by=list(room_type=  
=Istanbul$room_type),mean)
```

tabmeansPC1 #Means of all the columns per Room Type

```
##      room_type calc_Review_365_Negative Lattitude_Positive_Longi_Negate  
## 1 Entire home/apt          -0.1756988          0.02117150  
## 2 Private room            0.1297107          -0.02394794  
## 3 Shared room             0.3080373          0.10680772  
##  MinNight_Price_Positive MinNightNegative_PricePos  
## 1          -0.01821017          -0.072516130  
## 2          0.01739272          0.060570262  
## 3          -0.03640266          0.005412508  
##  availabilityNegate_Reviews_Pos Positive_Lat_Long CalcHostListing_Pos  
## 1          0.1342928          -0.01136218          -0.02408167  
## 2          -0.1050311          0.01863751          0.01675934  
## 3          -0.1335527          -0.15742400          0.05985366
```

#In this op coz of +ve -VE signs u can see the means are different

```
tabmeansPC1 <- tabmeansPC1[rev(order(tabmeansPC1$room_type)),]
```

```
tabmeansPC1
```

```
##      room_type calc_Review_365_Negative Lattitude_Positive_Longi_Negate  
## 3 Shared room            0.3080373          0.10680772  
## 2 Private room            0.1297107          -0.02394794  
## 1 Entire home/apt          -0.1756988          0.02117150  
##  MinNight_Price_Positive MinNightNegative_PricePos  
## 3          -0.03640266          0.005412508  
## 2          0.01739272          0.060570262  
## 1          -0.01821017          -0.072516130  
##  availabilityNegate_Reviews_Pos Positive_Lat_Long CalcHostListing_Pos  
## 3          -0.1335527          -0.15742400          0.05985366  
## 2          -0.1050311          0.01863751          0.01675934  
## 1          0.1342928          -0.01136218          -0.02408167
```

```
tabfmeans1 <- t(tabmeansPC1[, -1]) #transpose
```

```
tabfmeans1
```

```
##              3              2              1  
## calc_Review_365_Negative      0.308037254  0.12971065 -0.17569882  
## Lattitude_Positive_Longi_Negate 0.106807721 -0.02394794  0.02117150  
## MinNight_Price_Positive      -0.036402663  0.01739272 -0.01821017  
## MinNightNegative_PricePos      0.005412508  0.06057026 -0.07251613  
## availabilityNegate_Reviews_Pos -0.133552683 -0.10503106  0.13429281  
## Positive_Lat_Long            -0.157424001  0.01863751 -0.01136218  
## CalcHostListing_Pos          0.059853663  0.01675934 -0.02408167
```

```
colnames(tabfmeans1) <- t(as.vector(tabmeansPC1[1]))
```

tabfmeans1 #This is means for all PCs per room Type


```
##                               Shared room Private room Entire home/apt
## calc_Review_365_Negative      0.308037254   0.12971065   -0.17569882
## Lattitude_Positive_Longi_Negate 0.106807721  -0.02394794    0.02117150
## MinNight_Price_Positive      -0.036402663   0.01739272   -0.01821017
## MinNightNegative_PricePos     0.005412508   0.06057026   -0.07251613
## availabilityNegate_Reviews_Pos -0.133552683  -0.10503106   0.13429281
## Positive_Lat_Long            -0.157424001   0.01863751   -0.01136218
## CalcHostListing_Pos          0.059853663   0.01675934   -0.02408167

# Standard deviations of scores for all the PC's classified by Room Type
tabsdsPC1 <- aggregate(Istanbultyp_pca[,c(3,4,5,6,7,8,9)],by=list(room_type=I
stanbul$room_type),sd)
tabfsds1 <- t(tabsdsPC1[,-1])
colnames(tabfsds1) <- t(as.vector(tabsdsPC1[1]))
tabfsds1 #This is Std Deviation for all PCs per room Type

##                               Entire home/apt Private room Shared room
## calc_Review_365_Negative      1.1362586    1.1226645    0.8624386
## Lattitude_Positive_Longi_Negate 1.1418609    1.0397350    1.2051611
## MinNight_Price_Positive      0.9987636    1.0233101    1.1184727
## MinNightNegative_PricePos     1.0239683    0.9748299    0.6917807
## availabilityNegate_Reviews_Pos 0.9581406    0.9562878    0.8705255
## Positive_Lat_Long            0.9391580    0.8631261    0.9572905
## CalcHostListing_Pos          0.9043482    0.8717852    0.5592181

class(tabfsds1) #changed to matrix
## [1] "matrix"
```

Levens Tests

```
# Levene's tests (one-sided)
library(car)

## Warning: package 'car' was built under R version 3.6.2
## Loading required package: carData

##
## Attaching package: 'car'

## The following object is masked from 'package:boot':
##
##     logit

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

```

## The following object is masked from 'package:purrr':
##
##      some

## The following object is masked from 'package:dplyr':
##
##      recode

library(carData)
names(Istanbultyp_pca)

## [1] "Neighbourhood"          "Room_Type"
## [3] "calc_Review_365_Negative" "Latitude_Positive_Longi_Negate"
## [5] "MinNight_Price_Positive"  "MinNightNegative_PricePos"
## [7] "availabilityNegate_Reviews_Pos" "Positive_Lat_Long"
## [9] "CalcHostListing_Pos"

(LTPC1 <- leveneTest(calc_Review_365_Negative~Istanbul$room_type,data=Istanbultyp_pca))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2  32.39 9.141e-15 ***
##           16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC1_1sided1 <- LTPC1[[3]][1]/2)

## [1] 4.570698e-15

(LTPC2 <- leveneTest(Latitude_Positive_Longi_Negate~Istanbul$room_type,data=Istanbultyp_pca))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2  15.797 1.4e-07 ***
##           16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC2_1sided=LTPC2[[3]][1]/2)

## [1] 6.999728e-08

(LTPC3 <- leveneTest(MinNight_Price_Positive~Istanbul$room_type,data=Istanbultyp_pca))

```

```

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2  15.166 2.628e-07 ***
##      16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC3_1sided <- LTPC3[[3]][1]/2)

## [1] 1.313987e-07

(LTPC4 <- leveneTest(MinNightNegative_PricePos~Istanbul$room_type,data=Istanb
ultyp_pca))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2   26.52 3.17e-12 ***
##      16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC4_1sided <- LTPC4[[3]][1]/2)

## [1] 1.585111e-12

(LTPC5 <- leveneTest(availabilityNegate_Reviews_Pos~Istanbul$room_type,data=I
stanbultyp_pca))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2  18.916 6.229e-09 ***
##      16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC5_1sided <- LTPC5[[3]][1]/2)

## [1] 3.114674e-09

(LTPC6 <- leveneTest(Positive_Lat_Long~Istanbul$room_type,data=Istanbultyp_pc
a))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

```

```
## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2  9.7007 6.16e-05 ***
##       16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC6_1sided <- LTPC6[[3]][1]/2)

## [1] 3.079758e-05

(LTPC7 <- leveneTest(CalcHostListing_Pos~Istanbul$room_type,data=Istanbultyp_pca))

## Warning in leveneTest.default(y = y, group = group, ...): group coerced to
## factor.

## Levene's Test for Homogeneity of Variance (center = median)
##           Df F value    Pr(>F)
## group      2 19.239 4.515e-09 ***
##       16248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(p_PC7_1sided <- LTPC7[[3]][1]/2)

## [1] 2.257301e-09
```

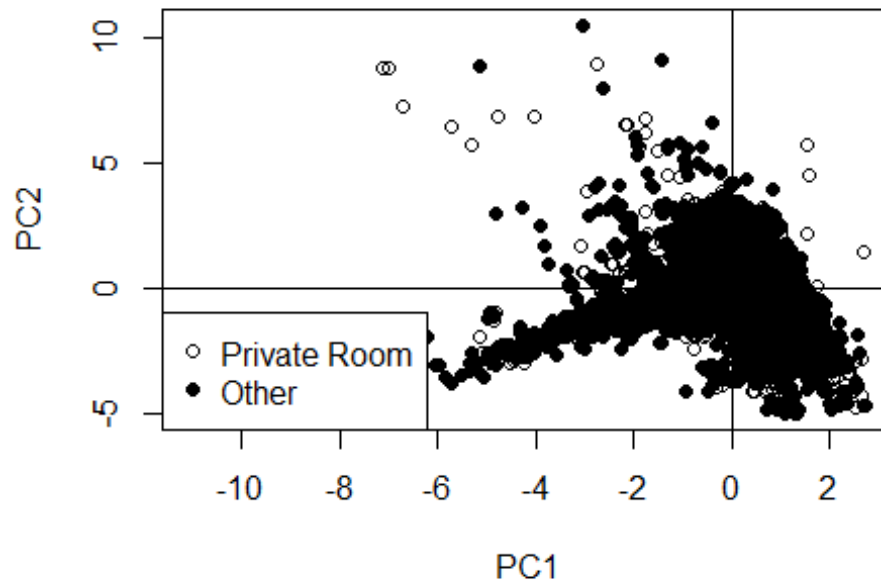
Plotting the scores for the first and second components for Private Rooms

```
# Plotting the scores for the first and second components for Private Rooms

plot(Istanbultyp_pca$calc_Review_365_Negative, Istanbultyp_pca$Latitude_Positive_Longi_Negate,
     pch=ifelse(Istanbultyp_pca$Room_Type == "Private room",1,16),xlab="PC1",
     ylab="PC2", main="Private rooms against values for PC1 & PC2")
abline(h=0)
abline(v=0)

legend("bottomleft", legend=c("Private Room","Other"), pch=c(1,16))
```

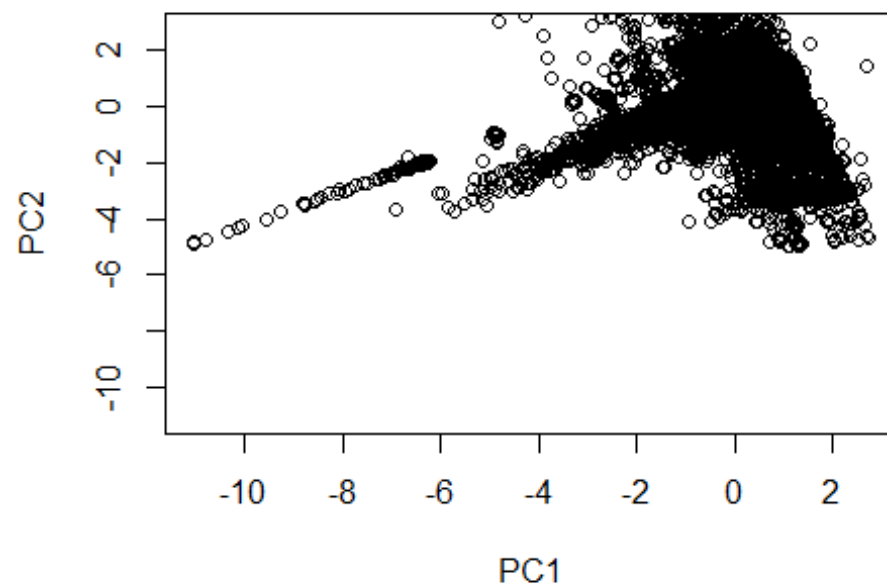
Private rooms against values for PC1 & PC2



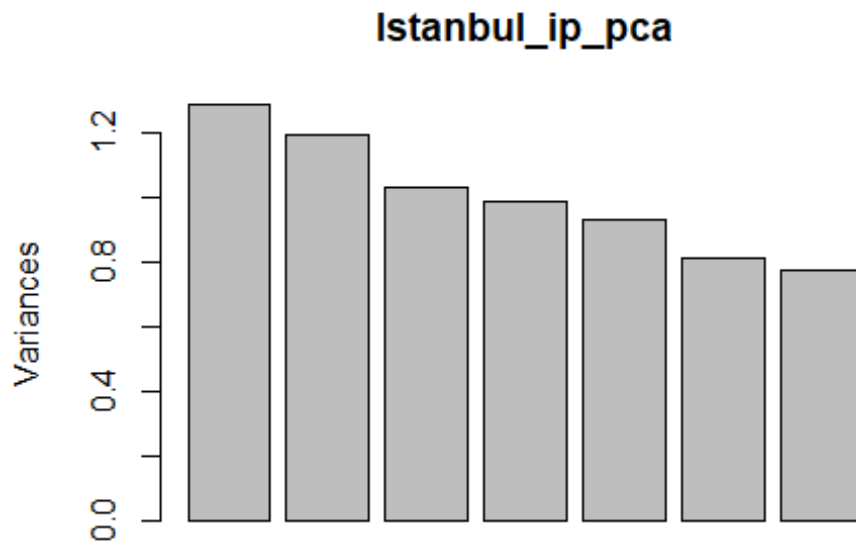
```
#View(Istanbul_ip_pca)
diag(cov(Istanbul_ip_pca$x))

##          PC1          PC2          PC3          PC4          PC5          PC6          PC7
## 1.2835228 1.1916328 1.0316158 0.9835550 0.9255190 0.8117322 0.7724225

xlim <- range(Istanbul_ip_pca$x[,1])
#xlim
#Istanbul_ip_pca$x[,1]
#Istanbul_ip_pca$x
plot(Istanbul_ip_pca$x,xlim=xlim,ylim=xlim)
```



```
#Istanbul_ip_pca$rotation[,1]  
#Istanbul_ip_pca$rotation  
#plot(Istanbul[, -1])  
#Plotting Variances  
plot(Istanbul_ip_pca)
```



original value of the data based on PCA

To get the

#get the original value of the data based on PCA

```
center <- Istanbul_ip_pca$center
```

```
scale <- Istanbul_ip_pca$scale
```

```
new_Istanbul <- as.matrix(Istanbul[, -1])
```

```
head(new_Istanbul, 5)
```

```
##      name                                     host_id  host_name
## [1,] "The Place"                             "    6603" "Kaan"
## [2,] "The Bosphorus from The Comfy Hill"      "   78838" "GÃ¼lder"
## [3,] "House for vacation rental furnutare"    "  105823" "Yesim"
## [4,] "LOVELY APT. IN PERFECT LOCATION"        "  117026" "Mutlu"
## [5,] "Duplex Apartment with Terrace "         "  121607" "Alen"
##      neighbourhood_group neighbourhood latitude longitude room_type
## [1,] NA                  "Uskudar"    "41.05650" "29.05367" "Entire home/
apt"
## [2,] NA                  "Besiktas"    "41.06984" "29.04545" "Entire home/
apt"
## [3,] NA                  "Besiktas"    "41.07731" "29.03891" "Entire home/
apt"
## [4,] NA                  "Beyoglu"     "41.03220" "28.98216" "Entire home/
apt"
## [5,] NA                  "Sisli"       "41.04471" "28.98567" "Entire home/
apt"
##      price  minimum_nights number_of_reviews last_review reviews_per_mont
h
```

```
## [1,] " 554" " 1" " 1" "6/1/2009" " 0.01"
## [2,] " 100" " 30" " 41" "11/7/2018" " 0.38"
## [3,] " 211" " 21" " 0" "" NA
## [4,] " 237" " 5" " 2" "5/4/2018" " 0.04"
## [5,] " 591" " 3" " 0" "" NA
##      calculated_host_listings_count availability_365
## [1,] " 1" "365"
## [2,] " 2" " 49"
## [3,] " 1" " 83"
## [4,] " 1" "228"
## [5,] "13" "356"
```

Predict

```
predict(Istanbul_ip_pca)[,1]
```

```
##      [1] -1.982080e-02  2.611837e-01  1.033726e+00  4.432095e-01 -1.088986e
+00
##      [6] -3.392863e-02 -1.087338e-01 -1.549697e-01 -1.166976e-01  3.196913e
-01
##     [11] -2.810846e-01 -1.220210e-01 -2.331986e-01 -5.027803e-02 -8.092017e
-01
##     [16] -7.545404e-02 -7.402963e-01  5.570317e-01 -2.858194e-01  1.008379e
+00
##     [21] -3.839433e-01 -2.057558e-01 -3.360221e-02 -3.840837e-02 -3.324720e
-01
##     [26] -4.421108e-01 -5.355561e-01  1.134305e+00 -2.323594e-01 -1.611261e
+00
##     [31] -3.276429e-02  6.685052e-01 -6.898364e-02 -9.518836e-01 -2.161913e
-01
##     [36] -1.789718e+00  1.327175e+00 -1.113483e+00 -1.128657e-01 -1.648513e
-01
##     [41] -9.395780e-02  1.237452e-01 -1.092384e+00 -3.950938e-01  2.464463e
-01
##     [46]  1.155281e+00 -2.574804e-01 -1.205699e+00  2.545080e-01 -7.782369e
-02
##     [51]  8.137489e-01  7.795472e-02 -2.967713e+00 -3.875881e-01 -1.004389e
-01
##     [56] -1.972576e-01  7.501653e-01 -2.355558e-01  4.169608e-01 -1.075607e
-01
##     [61]  1.006752e+00 -9.757401e-01 -3.250128e+00  5.891643e-01 -1.195963e
+00
##     [66] -1.557194e-01 -6.603129e-01 -1.430680e-01 -5.384394e-01  1.348814e
+00
##     [71] -7.498563e-01 -8.312277e-01 -1.135657e+00 -9.759375e-01 -1.067088e
+00
##     [76] -1.715597e+00 -2.232542e+00 -8.404877e-01  9.172502e-02 -3.306277e
-01
##     [81] -6.902405e-01 -1.193715e+00 -1.462379e-01 -6.713765e-01  1.300603e
-01
##     [86] -3.557103e-01 -7.488786e-02  3.369594e-02 -1.126006e+00 -2.377932e
```



```
+00
## [91] -6.540497e-01 -4.368060e-01 3.464927e-01 -9.154547e-02 -8.320408e
-01
## [96] -6.584618e-01 8.499784e-01 -1.084718e+00 -2.310425e+00 -7.230624e
-01
## [101] -1.296381e-01 9.539511e-01 1.153902e-01 -6.239424e-02 1.088967e
+00
## [106] -8.386576e-02 -1.708867e+00 -7.740343e-01 -2.687296e-02 -1.417552e
+00
## [111] -7.784377e-02 -6.864227e-02 -4.141733e+00 -6.473901e-01 -2.282851e
+00
## [116] 9.311354e-01 -7.955900e-02 -2.428303e-01 -2.839345e+00 -9.086922e
-02
## [121] -4.219141e+00 -1.109395e-01 -2.062707e-01 8.589579e-01 -2.786242e
-01
## [126] 2.409104e-01 -3.965124e+00 -5.998445e-02 -1.069070e+00 -3.863835e
-01
## [131] -1.245134e+00 -1.112636e+00 -1.278711e+00 -4.173623e-01 -5.494119e
-01
## [136] -1.570588e+00 -2.572670e-01 -2.821681e-01 -3.825581e-01 -1.037351e
-01
## [141] -8.235494e-01 -1.497293e-02 -9.974701e-01 -1.011142e+00 -8.767852e
-02
## [146] -1.901515e-01 -9.717040e-02 -1.150255e-01 -1.600301e-02 -1.167109e
-01
## [151] -1.300358e-01 -1.352541e-01 -5.569849e-02 1.013302e+00 -2.116819e
-01
## [156] -4.404628e-01 -1.015299e-01 -3.565791e-02 -2.388967e-01 -3.039721e
-02
## [161] -2.442102e+00 -3.449361e+00 2.020249e-01 -7.870325e-01 -3.331865e
+00
## [166] -1.704644e-01 -1.711917e-01 1.055289e+00 3.258597e-01 -1.471583e
-01
## [171] -4.302418e+00 -1.067108e+00 -2.599282e+00 -8.883889e-02 1.078873e
-01
## [176] -4.005199e-01 -4.304152e+00 -1.464351e+00 -4.993317e-01 -1.098409e
+00
## [181] -1.480237e+00 -1.420427e+00 -5.425331e-01 -1.103854e-01 6.259847e
-01
## [186] 2.167565e-01 -5.842942e-01 -2.688149e+00 -4.174007e-01 -6.300420e
-02
## [191] -4.065615e-01 -3.576724e+00 -5.068692e-01 -1.127862e-01 -6.599756e
-01
## [196] 9.549614e-01 6.011233e-01 -4.556286e-01 -8.016814e-01 1.347489e
+00
## [201] -8.158979e-02 -1.241353e-01 -2.547744e-02 -3.650825e-01 3.639016e
-01
## [206] -1.739817e-01 -9.669459e-01 -1.433957e+00 -1.859376e+00 2.704572e
-01
## [211] -2.088320e+00 -3.377852e+00 -1.705615e+00 -6.423824e-01 -9.442331e
```

```
-01
## [216] -1.695883e-01  1.704450e+00  1.722005e-01 -1.209673e+00  1.235584e
+00
## [221] -3.744665e-01 -1.151915e-01 -1.567328e-01  1.250163e-01  1.321930e
+00
## [226] -6.529419e-01 -9.019988e-01 -4.988458e+00  2.844205e-01 -6.461650e
-01
## [231]  1.505858e-01  1.062841e+00 -1.273287e+00 -1.368076e-01 -6.926739e
-01
## [236] -3.078794e-02 -3.879273e-01  3.679952e-01  6.217408e-01 -3.428627e
-01
## [241] -2.479444e+00 -6.326378e-01 -5.851018e-01 -5.946826e-01 -6.816518e
-01
## [246] -3.392980e+00  5.329706e-01  7.083182e-01  1.232485e-01 -4.705729e
-02
## [251]  4.365217e-01  7.388403e-01  1.737495e-01 -9.778856e-01 -1.057725e
-01
## [256]  1.296131e-01 -2.230399e-02 -2.645520e+00 -2.555077e-01  1.079501e
-02
## [261]  8.932809e-02  1.252459e+00 -3.167480e-01 -3.252802e+00 -6.371606e
-01
## [266] -4.895416e-01 -1.102454e+00 -4.941450e-01 -9.253630e-02 -1.161590e
+00
## [271] -5.195123e-02 -4.753659e-02 -3.475402e+00 -1.693226e-01 -8.808210e
-02
## [276] -3.390209e+00 -1.864724e+00  3.035989e-01 -2.653189e+00 -5.548788e
-01
## [281] -2.570770e-01 -1.054588e-01 -3.676471e-01 -3.265438e-02 -5.033397e
-02
## [286] -2.068026e-01 -2.734400e+00  3.666016e-02  1.003833e+00  3.696921e
-01
## [291] -3.328837e-02  1.105063e+00 -1.774027e+00  8.655015e-01  1.237508e
+00
## [296] -6.484451e-01 -1.212418e+00 -9.981482e-01  1.321613e-01 -9.706043e
-01
## [301] -9.527078e-01 -4.711859e-01 -3.654901e+00 -1.417470e-01 -3.194388e
+00
## [306] -3.286248e+00  1.864465e-01 -9.818818e-01 -2.832022e-02 -2.302121e
-01
## [311] -3.197792e+00 -3.802189e+00 -2.982941e-01 -1.005432e-01 -6.894561e
-01
## [316]  9.268751e-01 -1.241733e+00  1.388262e+00 -1.180826e+00 -1.237381e
-01
## [321] -3.921369e-01 -1.077849e+00 -5.909146e-02 -2.619316e-01 -2.625444e
-01
## [326] -2.921921e-01 -5.081994e-01 -6.784224e-01 -2.223503e-01 -8.669702e
-01
## [331] -5.860475e-01 -4.250465e+00 -2.536849e+00 -1.965285e+00 -5.142430e
+00
## [336] -6.029848e-01 -1.838733e-01  1.959318e+00 -7.901505e-03 -7.072318e
```

```
-01
## [341] 6.204968e-01 3.638214e-01 -1.236216e-01 -3.464861e-01 -8.105086e
-03
## [346] -7.684888e-01 -1.467215e-01 5.105006e-01 -5.988965e+00 -9.026224e
-02
## [351] -4.104784e-02 -2.022596e-01 -1.830572e-01 -7.652918e-02 -4.949180e
-01
## [356] 1.323803e+00 -8.430715e-02 -4.168333e-01 -5.186920e-01 -3.985782e
-03
## [361] -3.244058e-01 2.008787e-01 -2.015891e-01 1.158165e+00 1.436465e
+00
## [366] -1.970279e-01 -5.286266e-01 -2.539343e-01 -1.324746e+00 -3.056241e
-01
## [371] -8.340517e-01 -3.065072e+00 -1.583484e+00 -6.829869e-02 -5.910895e
-01
## [376] -1.034744e+00 -1.569323e+00 2.408498e-01 -5.123845e-01 3.142843e
-02
## [381] -5.445187e-01 -5.486889e-01 -5.454307e-01 1.108570e-01 -1.719849e
+00
## [386] -1.311266e-01 -1.912987e-01 8.749633e-01 -4.131435e-01 -6.605668e
-02
## [391] -2.897227e+00 -6.278987e-01 -2.338910e+00 1.041685e-01 -1.033187e
-02
## [396] 1.320492e+00 -2.096394e-01 -2.803434e-01 1.486372e-01 -1.748606e
-01
## [401] -4.251138e+00 -2.556311e-01 -1.745456e-01 -1.693310e-01 -2.765124e
+00
## [406] -7.285400e-01 -1.436064e+00 1.686599e-01 1.069908e-02 -1.142294e
-01
## [411] -9.751302e-01 -4.209342e-01 -2.720712e-01 -5.879651e-02 -7.073787e
-02
## [416] -2.630785e-01 -1.646618e-01 -4.060008e-02 -5.097411e-02 -2.915763e
-01
## [421] -1.598913e+00 -1.111920e+00 -2.530662e-01 -3.873790e-02 -7.204242e
-01
## [426] -2.412394e+00 3.510422e-01 -1.587573e+00 5.638859e-01 -4.564205e
-01
## [431] -8.057868e-02 2.967512e-01 1.135373e+00 -8.670239e-01 -1.000450e
-01
## [436] -2.540858e-01 3.240985e-01 9.318218e-02 6.010804e-01 2.182560e
-01
## [441] -5.176646e-01 -1.002451e+00 -1.387413e-01 -1.228382e-01 -1.592737e
-01
## [446] -4.616297e-01 -5.226009e-01 -5.555407e-02 -6.964795e-02 -1.412537e
+00
## [451] -2.114266e+00 -1.580878e+00 -2.067595e+00 -3.106964e-02 -1.825734e
-01
## [456] 1.828780e+00 -8.708678e-01 -1.124852e-01 -2.573184e+00 -2.977100e
-01
## [461] -3.306565e+00 -1.183614e-01 4.225446e-02 -6.663781e-01 -5.965215e
```

```

-01
## [16186] -7.569162e-01  1.839086e+00  8.859075e-01  1.331249e+00 -4.882132e
-01
## [16191] -1.106459e-01  8.982957e-01 -5.219667e-01 -4.082493e-01 -2.031369e
-02
## [16196]  4.376895e-02  8.745035e-01  1.054471e+00  1.177472e+00 -1.463083e
-01
## [16201]  9.272812e-01 -1.600413e-01  9.126685e-01  8.721023e-01 -1.385621e
-01
## [16206]  7.008098e-02  1.052904e+00  8.685269e-02 -1.572937e+00 -7.178983e
-01
## [16211] -1.051766e+00 -7.875838e-02 -1.182372e-01 -1.644315e-01 -1.358004e
-01
## [16216] -2.710485e-01 -2.251084e-01 -2.388787e-01  1.031306e+00  7.217415e
-01
## [16221]  2.602975e-01 -4.444977e-01 -8.079784e-02  9.315375e-01 -6.898640e
-01
## [16226]  2.468887e-01  1.027389e+00  1.269236e+00  1.036332e+00  6.079825e
-01
## [16231] -2.324746e-01  1.203770e-01 -2.185624e-01  1.084634e+00  9.868980e
-01
## [16236]  1.195266e+00  1.100947e+00  6.673032e-01  1.079111e+00 -9.962855e
-01
## [16241] -6.105469e-01  6.412052e-01  4.413661e-01 -5.921832e-01 -4.883565e
-01
## [16246]  1.633144e+00 -3.206824e-01 -2.940352e-01  4.240007e-02 -3.057399e
-02
## [16251]  2.519838e-01

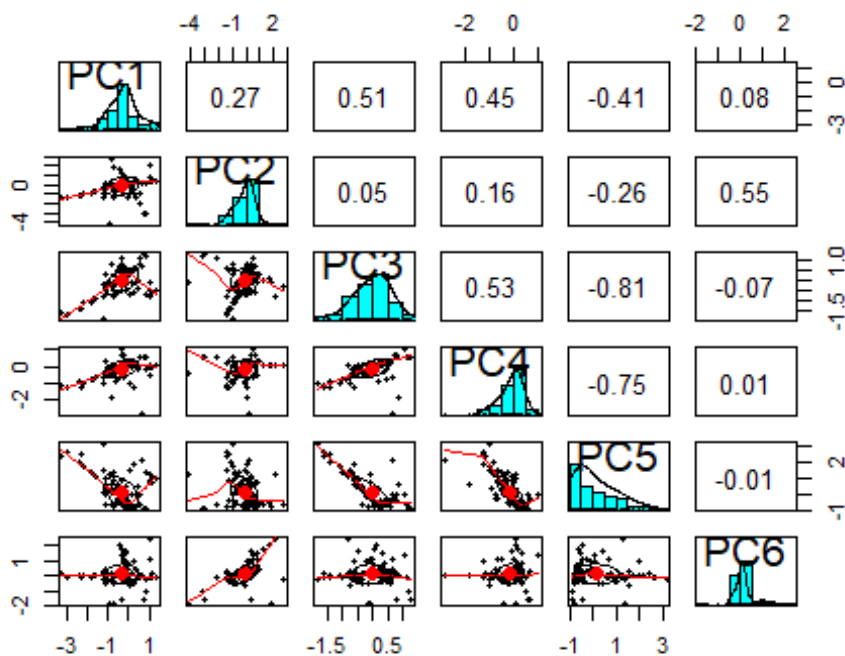
```

Pairs

```

pairs.panels(Istanbul_ip_pca$x[1:100,c(1,2,3,4,5,6)])

```



Summarizing our Principal Component Analysis for Istanbul Airbnb Data:

From Summary of Principal components:

Principal components do not explain much of variance individually.

As per, 'Cumulative Proportion' field, 88.97% of Cumulative variance is explained by PC1, PC2, PC3, PC4, PC5 and PC6.

So we will have to include PC1 till PC6 to prevent loss of Information.