Load Libraries

```
library(tidyverse)# data manipulation, create plots
library(DataExplorer) # correlation matrix
```

Load Data

df<-read.csv("houses_to_rent.csv")</pre> head(df)

furniture city area rooms bathroom parking.spaces floor animal <int> <chr> <chr> <chr> <int> <int> <int> <int> <int> 240 3 3 4 furnished 1 0 acept 1 2 1 64 2 1 1 10 not furnished 0 acept 5 3 2 443 5 4 3 furnished 1 acept 2 2 4 3 73 1 12 acept not furnished 1 5 4 1 19 1 0 not acept not furnished 6 5 1 13 1 1 0 2 not furnished acept 6 rows | 1-10 of 15 columns

df<- df %>% select(-c(floor,animal,hoa,total,X,city,property.tax))

Type of Data

df %>% select(fire.insurance, rent.amount) %>% summary() ## fire.insurance rent.amount ## Length:6080 Length:6080

Class :character Class :character ## Mode :character Mode :character

We note that the price of fire insurance, floor and the price of rent. They are character type values, due to the fact that you have special characters, R standardizes and converts them to character type. **Convert variables to numeric**

conv_to_numeric<-function(x){</pre> x<-gsub("[//R\$,]","",x) # eliminate special charterers</pre> x<-as.numeric(x) # transform to numeric data</pre> return(x) # return data attach(df)

fire_insurence<-sapply(fire.insurance,conv_to_numeric)</pre> rent_amount<-sapply(rent.amount,conv_to_numeric)</pre> df<-df %>%

mutate(fire.insurance=fire_insurence, rent.amount=rent_amount) # With the mutate function we make modifications to the data frame. Do furnished houses have a higher rental price compared to those that are not?

df %>% group_by(furniture) %>% summarise(rent_amount_mean=mean(rent.amount))

furniture rent_amount_mean <chr> <dpl> furnished 5387.092 not furnished 4047.211 2 rows

The average price of furnished houses is higher than those that are not.

histogram<-function(x,...){ df %>% ggplot(aes(x=x,y=..density..)) + geom_histogram(color="black", fill="#FFF0C9") + geom_density(color="black",lwd=1) + geom_vline(aes(xintercept=mean(x),color="mean")) + geom_vline(aes(xintercept=median(x),color="median")) + labs(col="Stadistcs") + theme(legend.position = "top") +

rent_amount_histogram=histogram(df\$rent.amount,labs(x="Rent Amount",title = "Rent Amount")) area_hsitogram=histogram(df\$area,labs(x="Area",title = "Area")) fire_insurence_histogram=histogram(df\$fire.insurance, labs(x="Fire Insurence", title = "Fire Insurence")) rooms_histogram=histogram(df\$rooms, labs(x="Rooms", title = "Rooms")) bathroom_histogram=histogram(df\$bathroom, labs(x="Bathroom", title = "Bathroom")) parking_spces_histogram<-histogram(df\$parking.spaces, labs(x="Parking Spaces", title="Parking Spaces"))</pre> scattter_plot<-function(x_feature,...){</pre> ggplot(data=df,aes(x=x_feature,y=rent.amount)) + geom_point(color="#77dd77",alpha=0.5) +

 $theme_light() +$ geom_smooth(method = "lm",color="red") + area_scatter<-scattter_plot(df\$area,labs(x="Area",y="Rent Amount"))</pre> fire_scatter<-scattter_plot(df\$fire.insurance, labs(x="Fire Insurence", y="Rent Amount"))</pre> rooms_scatter<-scattter_plot(df\$rooms,labs(x="Rooms",y="Rent Amount"))</pre> bathroom_scatter<-scattter_plot(df\$bathroom,labs(x="Bathroom",y="Rent Amount"))</pre> parking_scatter<-scattter_plot(df\$parking.spaces,labs(x="Parking Spaces",y="Charges"))</pre>

library(gridExtra) **Histogram Plots**

fire_insurence_histogram ## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

area_hsitogram

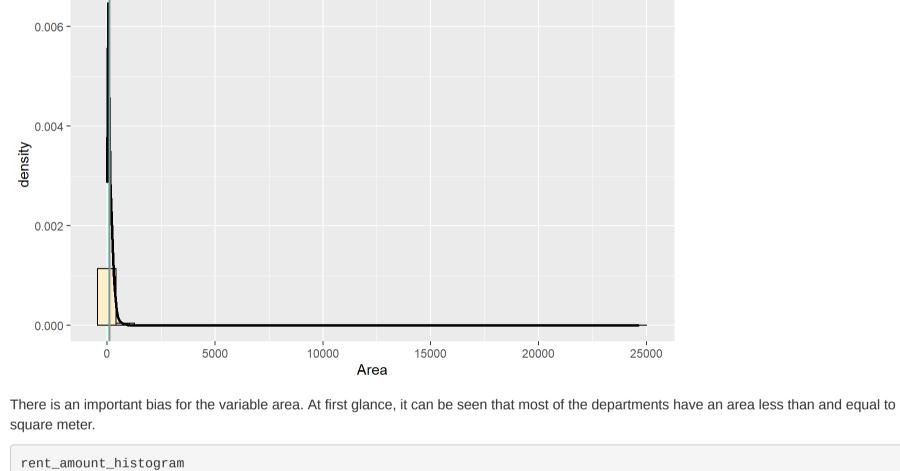
Fire Insurence

0.015 -0.010 density 0.005 -0.000 -400 600 200 Fire Insurence

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

For the variable fire insurence there is a high amount of abnormal values. There are very few cases where they exceed R \$250.

Area



`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

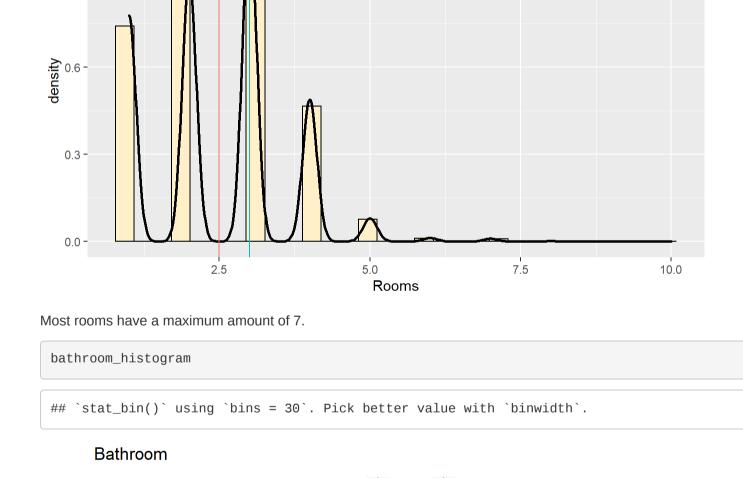
Rent Amount

0.00020 0.00015 density 0.00005 -0.00000 -10000 20000 30000 40000 Rent Amount It can be considered that most of the rental houses are around a price below R \$15,000. rooms_histogram ## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Rooms

0.9 -

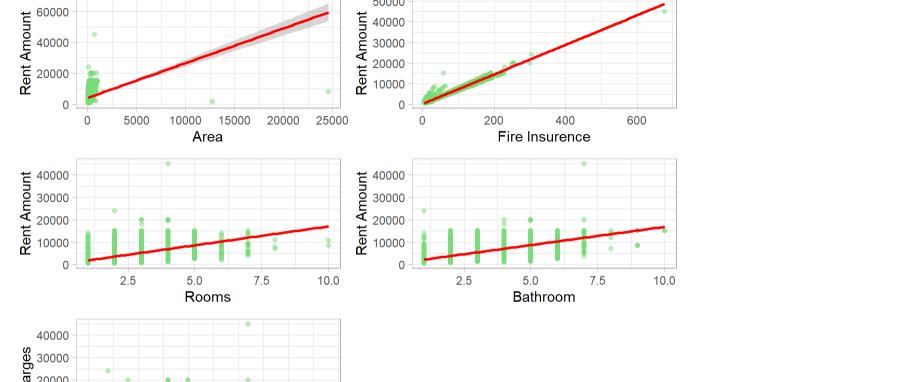
0.9 -



1.25 -1.00 -

0.75 density 0.50 -0.25 -0.00 -7.5 10.0 Bathroom In most rental houses records have a maximum of 6 bathrooms. parking_spces_histogram ## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`. Parking Spaces

0.3 -0.0 -Parking Spaces There are very few departments where it exceeds more than 6 places to park. **Scatter Plots** grid.arrange(area_scatter, fire_scatter, rooms_scatter, bathroom_scatter,parking_scatter) ## `geom_smooth()` using formula 'y ~ x' ## $geom_smooth()$ using formula 'y ~ x'



50000

Oparde 20000 -0.0 2.5 7.5 10.0 5.0 12.5 Parking Spaces The area variable apparently does not show any possible correlation with the rental price. Since it contains a good number of outliers, which causes the data to be skewed. There is a clear correlation between the price of fire insurance with respect to the price of the rental house, since the more expensive the price of the apartment, the more you will have to pay to protect it against fire. These variables have a linear relationship, that is, they increase

0.02

insurance, since you will have to cover more costs due to the proportion of the apartment.

0

plot_correlation(df,title = "Correelation Matrix") **Correelation Matrix** 0.09 -0.02 0 -0.16 -0.13

0.16

0.13

Shows the degree of relationship of the variables. They are measured from 0 to 1 if it is a positive correlation, otherwise it is measured from 1 to -1.

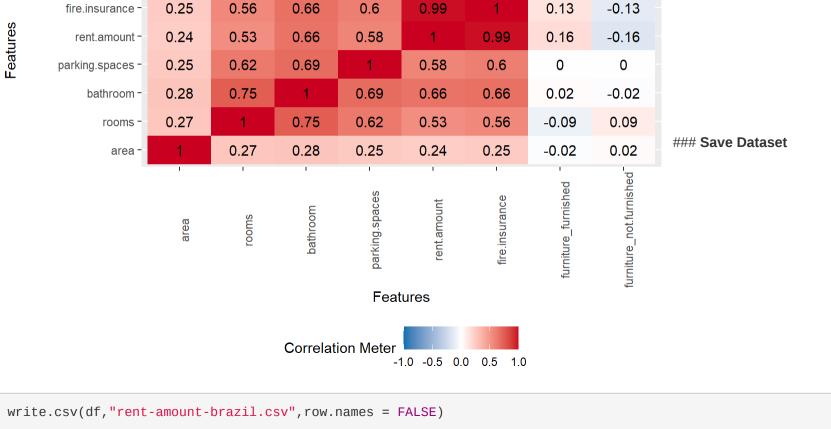
furniture_not.furnished -0.02 -0.02 -0.09 furniture_furnished -0.25 fire.insurance -

Conclusion

proportionally with another.

Correlation Matrix

0.53 0.66 0.58 0.99 0.24 0.16 -0.16 rent.amount -0.58 0 0.25 0.62 0.69 0.6 0 parking.spaces -0.75 0.69 0.28 0.66 0.66 0.02 -0.02 bathroom -



There is a strong presence of outliers in the data set. Especially for the area variable. For continuous variables, that is, those values with decimals, we can perform a logarithmic transformation to transform the outliers. Variables such as the size of the apartment area, the number of bathrooms, number of bedrooms and fire insurance. It makes all the sense in the world for it to increase prices, since these qualities increase the size of houses. The higher the cost of the apartment, the higher the price of fire