

# Feature engineering

## Load Libraries

```
library(tidyverse) # common libraries
```

## Load Data

```
df<-read_csv("rent-amount.csv")
```

head(df)									
city	area	rooms	bathroom	parking_spaces	floor	animal	furniture	rent.amount	city
1 yes	240	3	3	4	0	accept	furnished	8000	
2 no	64	2	1	1	10	accept	not furnished	820	
3 yes	443	5	5	4	3	accept	furnished	7000	
4 yes	73	2	2	1	12	accept	not furnished	1250	
5 yes	19	1	1	0	0	not accept	not furnished	1200	
6 yes	13	1	1	0	2	accept	not furnished	2200	
6 rows x 10 of 10 columns									

In EDA we conclude that the data set has too many outliers, which is why they require a type of processing.

I can think of 3 ways to deal with the situation.

- Remove outliers.
- Replace abnormal values by statistical measures.
- Perform a logarithmic transformation
- Calculate an upper range and based on the range replace outliers with normal values.

The first and second options are ruled out, since the data set has very few observations. And if we replace the outliers with statistical measures such as the mean or the median, we would be changing the data set, we would be affecting the distribution of the data.

We can combine the 3 and 4 option, we can establish an **upper range**. Based on the interval we create a criterion, we make a **sample** of random values that oscillate in those ranges. To be able to establish them and then perform a **logarithmic transformation** to smooth the data, improving their distribution.

```
attach(df)
```

```
upper_limit<-function(x,limit){
  mean<-mean(x) # calculate mean
  sd<-sd(x) # calculate standard deviation
  mean+limit*sd # calculate upper range
}
```

```
calculate_upper_limit<-function(feature){
  limits<-c(2,2.5,3,3.5,4)
  for(index in 1:5){
    print(paste("Upper Limit",limits[index]))
    print(upper_limit(feature,limits[index]))
    print("=====")
  }
}
```

### Upper Limits

#### Rent Amount

```
calculate_upper_limit(feature = rent.amount)
```

```
[1] "Upper Limit 2"
[1] 11549.18
[1] "=====
[1] "Upper Limit 2.5"
[1] 13357.52
[1] "=====
[1] "Upper Limit 3"
[1] 15225.85
[1] "=====
[1] "Upper Limit 3.5"
[1] 16954.19
[1] "=====
[1] "Upper Limit 4"
[1] 18762.62
[1] "=====
```

The best upper limit is 3 which we can round up to R \$15,000. Since there are more rental houses that are around those prices as we can see in the histogram.

#### Fire Insurance

```
calculate_upper_limit(feature = fire.insurance)
```

```
[1] "Upper Limit 2"
[1] 156.8332
[1] "=====
[1] "Upper Limit 2.5"
[1] 188.1448
[1] "=====
[1] "Upper Limit 3"
[1] 206.1448
[1] "=====
[1] "Upper Limit 3.5"
[1] 238.8913
[1] "=====
[1] "Upper Limit 4"
[1] 255.458
[1] "=====
```

With an upper limit of 4 it is good. With an upper limit of 4 it is good. We can round it up to R \$250 since there are very few cases where the price of the fire insurance exceeds that amount.

#### Area

```
calculate_upper_limit(feature = Area)
```

```
[1] "Upper Limit 2"
[1] 962.2626
[1] "=====
[1] "Upper Limit 2.5"
[1] 1099.842
[1] "=====
[1] "Upper Limit 3"
[1] 1273.822
[1] "=====
[1] "Upper Limit 3.5"
[1] 1465.692
[1] "=====
[1] "Upper Limit 4"
[1] 1653.382
[1] "=====
```

With an upper limit of 2.5 it is a good point. Since most of the departments do not exceed 1000 square meters.

```
replace_outliers<-function(data,normal_sample,upper_limit){
  return(ifelse(data>upper_limit,normal_sample,data))
}
```

### Normal Values Sample for Upper Limit

```
set.seed(2008) # we define random seed, so that the numbers generated by the sample do not vary.
sample_rent<-sample(4000:15000,1000,size=25,replace = T)
sample_fire_insurance<-sample(240:250,1000,size=25,replace = T)
sample_area<-sample(500:1000,size=25,replace = T)
```

```
normal_values_rent<-replace_outliers(data=df$rent.amount,sample_rent,upper_limit = 15000)
normal_fire_insurance<-replace_outliers(data=df$fire.insurance,sample_fire_insurance,upper_limit = 250)
normal_area<-replace_outliers(data=area,sample_area,upper_limit = 1000)
```

## Replace Outliers Values

```
df<- df %>%
  mutate(rent.amount=normal_values_rent) %>%
  mutate(fire.insurance=normal_fire_insurance) %>%
  mutate(area=normal_area)
```

With the mutate function we make modifications. We replace outliers with normal values, using a random sample.

#### We check the changes

```
df %>% select(rent.amount,area,fire.insurance) %>% summary()
```

```
rent.amount      area      fire.insurance
Min.   : 420   Min.   : 30.8   Min.   : 3.80
1st Qu.: 1500   1st Qu.: 50.8   1st Qu.: 23.00
Median : 3333   Median : 100.8  Median : 41.80
Mean   : 4350   Mean   : 145.2   Mean   : 56.11
3rd Qu.: 9952   3rd Qu.: 200.8  3rd Qu.: 77.80
Max.   : 15000   Max.   : 1000.0   Max.   : 250.80
```

```
df<- df %>% mutate(fire.insurance=ifelse(fire.insurance>3.5,fire.insurance))
```

We replaced the minimum price value of fire insurance by five real brackets.

```
df %>% select(fire.insurance) %>% summary()
```

```
fire.insurance
Min.   : 4.00
1st Qu.: 23.00
Median : 41.80
Mean   : 56.11
3rd Qu.: 77.80
Max.   : 250.80
```

### logarithmic transformation

```
df<- df %>%
  mutate(area_log=log(area)) %>%
  mutate(fire.insurance_log=log(fire.insurance)) %>%
  mutate(rent.amount_log=log(rent.amount))
```

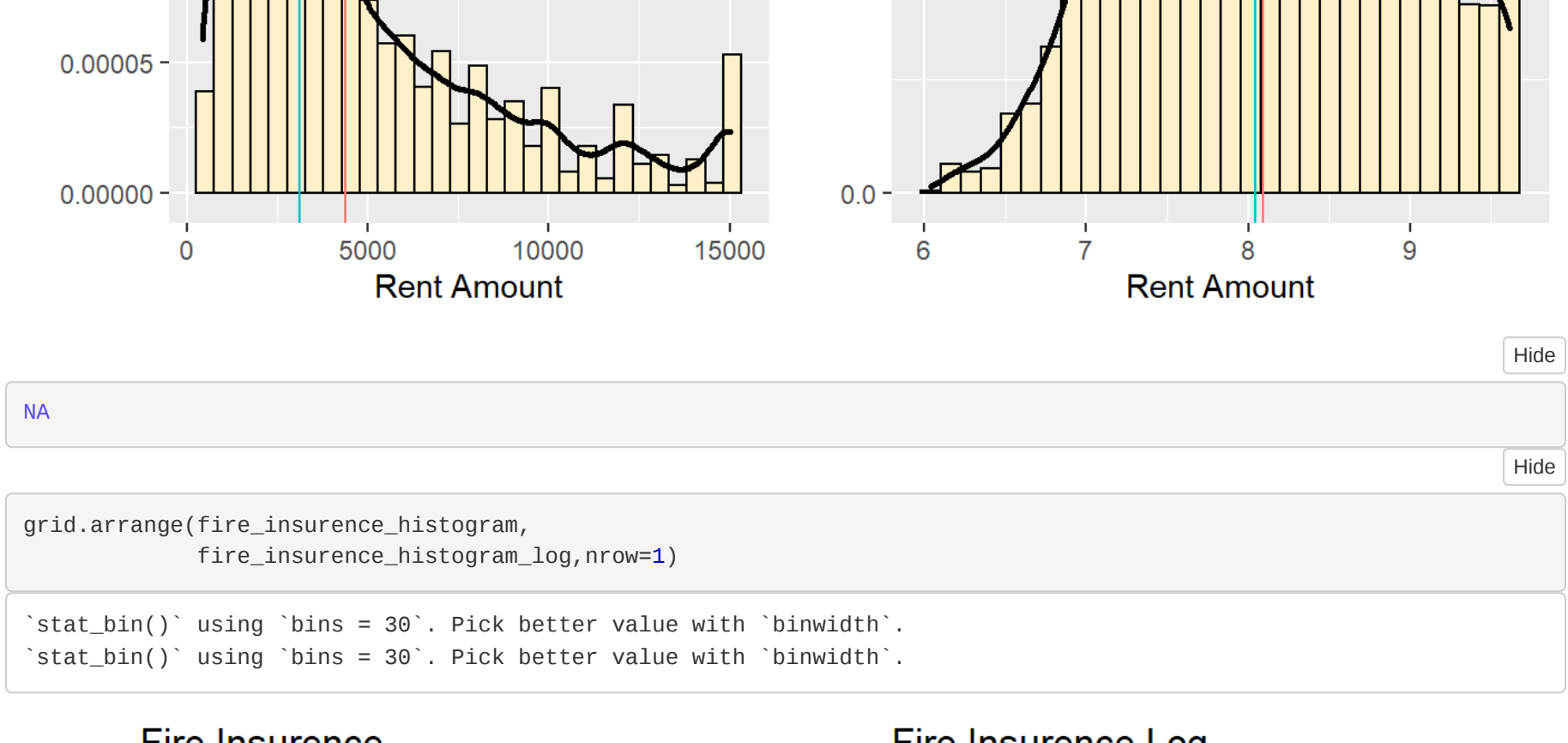
We perform logarithmic transformation to improve the distribution of continuous data.

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

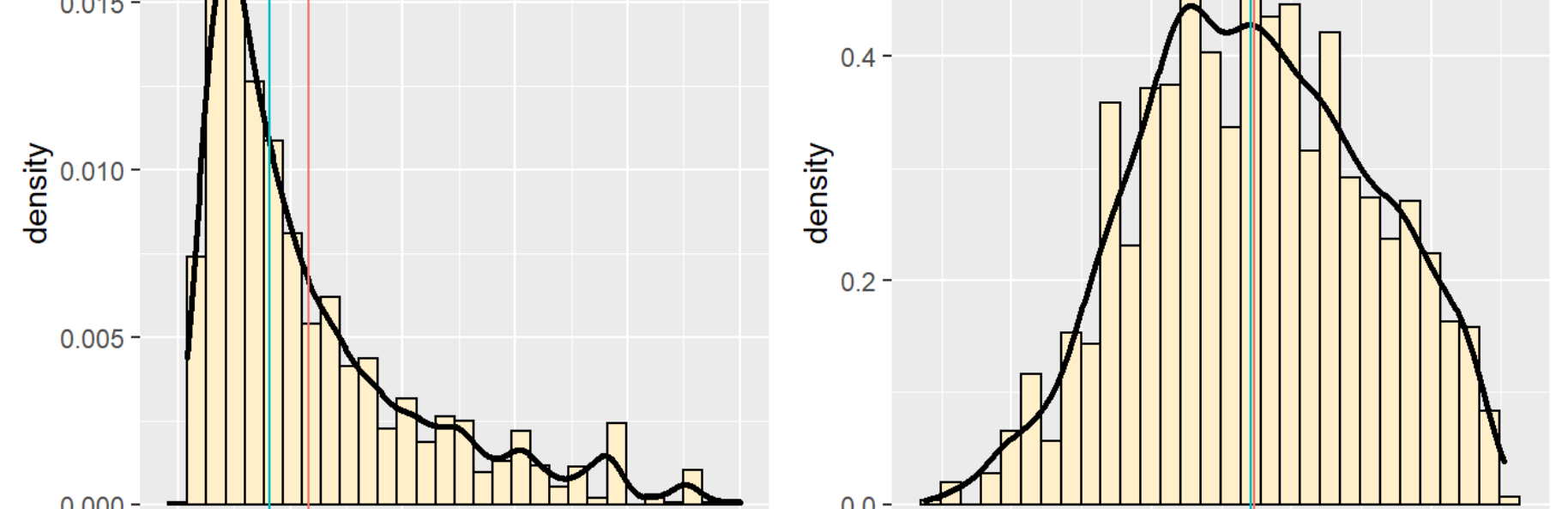
```
library(gridExtra)
```

```
rent.amount_histogram<-histogram(df$rent.amount,labs(x="Rent Amount",title = "Rent Amount"))
area_histogram<-histogram(df$area,labs(x="Area",title = "Area"))
fire.insurance_histogram<-histogram(df$fire.insurance,labs(x="Fire Insurance",title = "Fire Insurance"))
rent.amount_log_histogram<-histogram(df$rent.amount_log,labs(x="Rent Amount",title = "Rent Amount Log"))
area_log_histogram<-histogram(df$area_log,labs(x="Area Log",title = "Area Log"))
fire.insurance_log_histogram<-histogram(df$fire.insurance_log,labs(x="Fire Insurance Log",title = "Fire Insurance Log"))
```

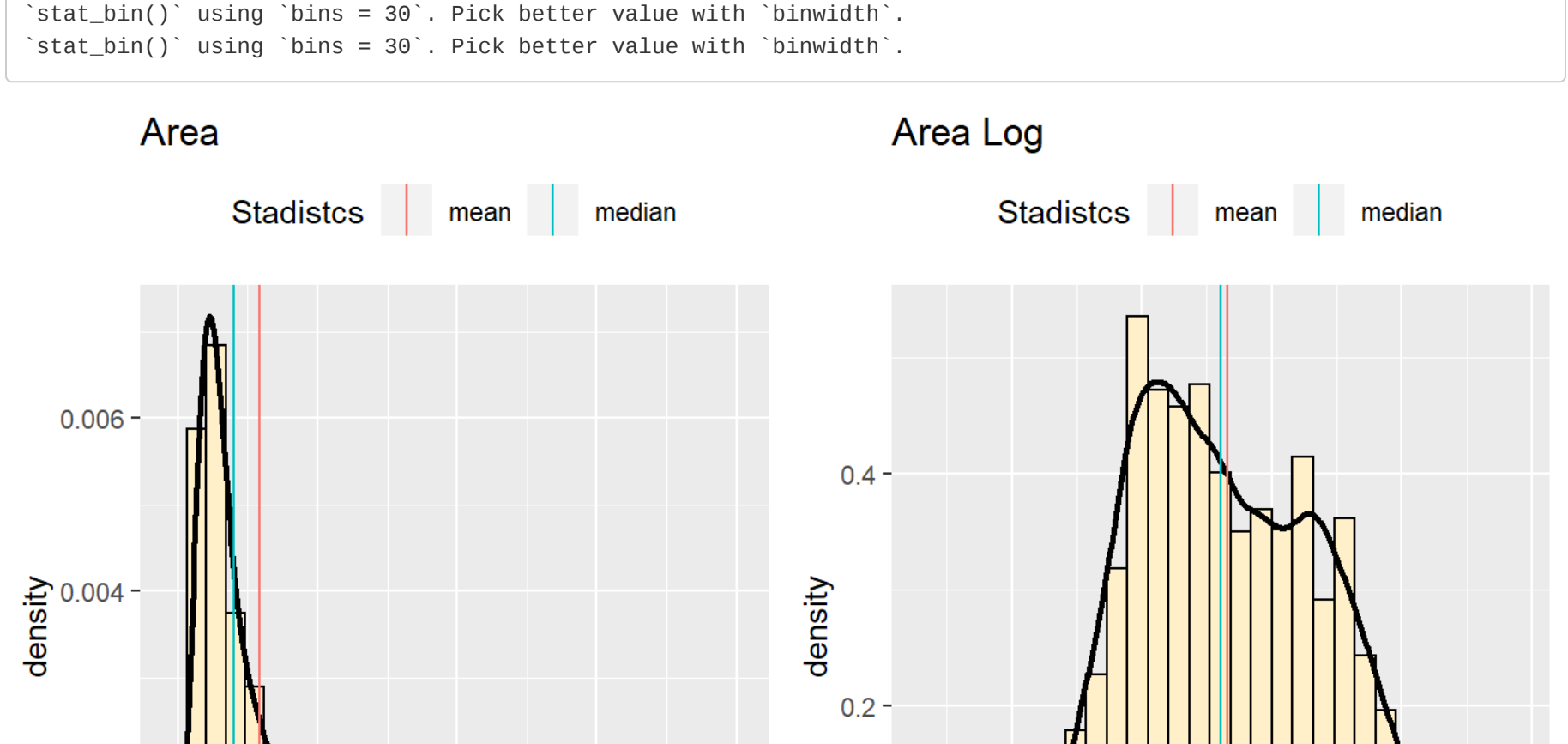
```
grid.arrange(rent.amount_histogram,
  area_histogram_log,rows=1)
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
```



```
grid.arrange(fire.insurance_histogram,
  fire.insurance_log,rows=1)
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
```



```
grid.arrange(area_histogram,
  area_log_histogram,rows=1)
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
stat_bin() using bins = 30 . Pick better value with 'binwidth'.
```

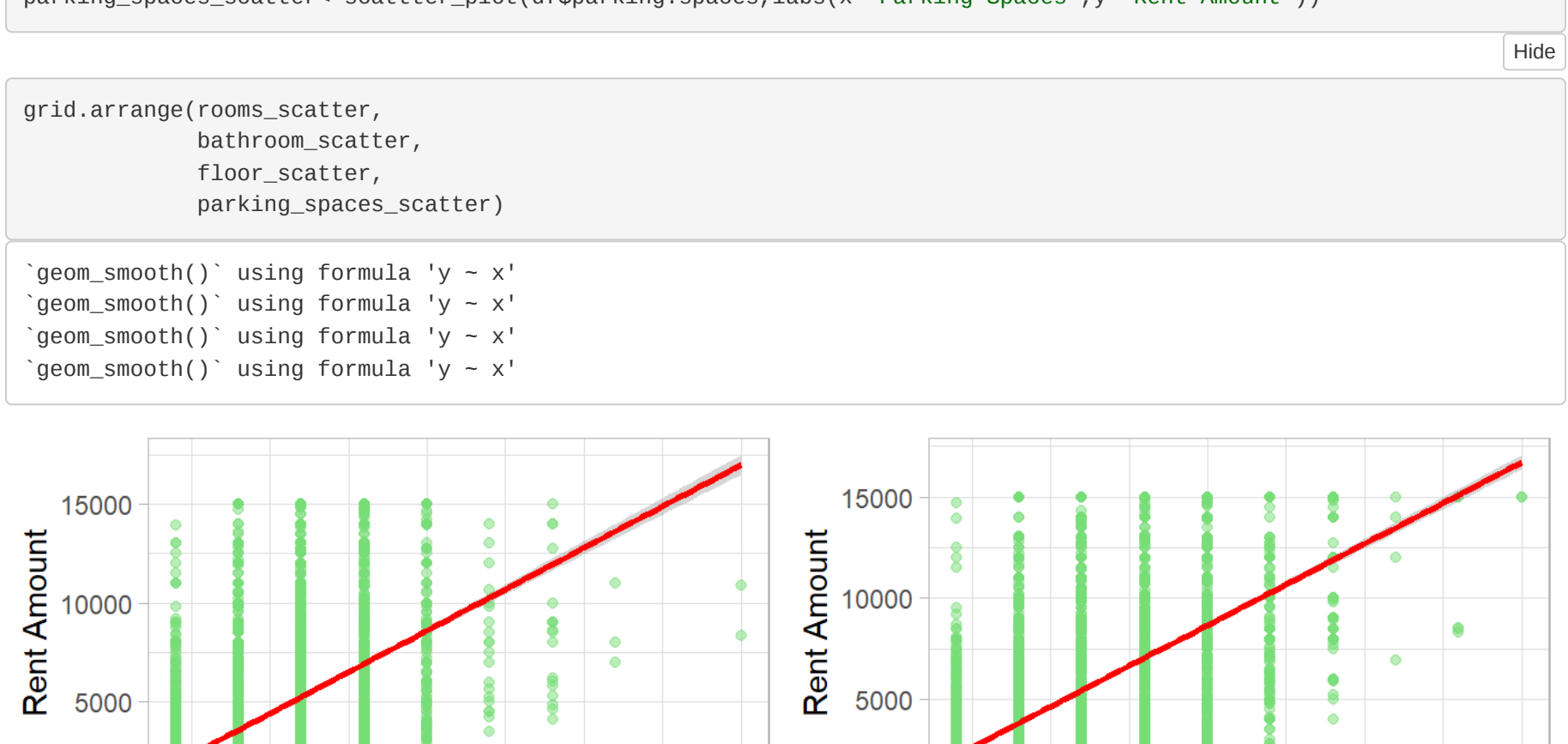


We observed a great improvement in the distribution of the data.

```
scatter_plot<-function(x,feature,...){
  ggplot(data=df,aes(x=x,feature,y=rent.amount)) +
  geom_point(color="gray60",alpha=0.5) +
  theme_lignt() +
  geom_smooth(method = "lm",color="red") +
  ...
}
```

```
rooms_scatter<-scatter_plot(df$rooms,labs(x="Rooms",y="Rent Amount"))
bathroom_scatter<-scatter_plot(df$bathroom,labs(x="Bathroom",y="Rent Amount"))
floor_scatter<-scatter_plot(df$floor,labs(x="Floor",y="Rent Amount"))
parking_spaces_scatter<-scatter_plot(df$parking_spaces,labs(x="Parking Spaces",y="Rent Amount"))
```

```
grid.arrange(rooms_scatter,
  bathroom_scatter,
  floor_scatter,
  parking_spaces_scatter)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



We can replace values greater than 7, since there are few departments that exceed that number of rooms. I will replace it with a value of 7, since there is little data and it will not affect the distribution of the data.

For the bathrooms we can change replace those values that are greater than 8, since there are very few bathrooms greater than the mentioned amount.

There are very few departments where the buildings of the rental houses are greater than 30.

There are very few apartments, where parking spaces exceed 8 units.

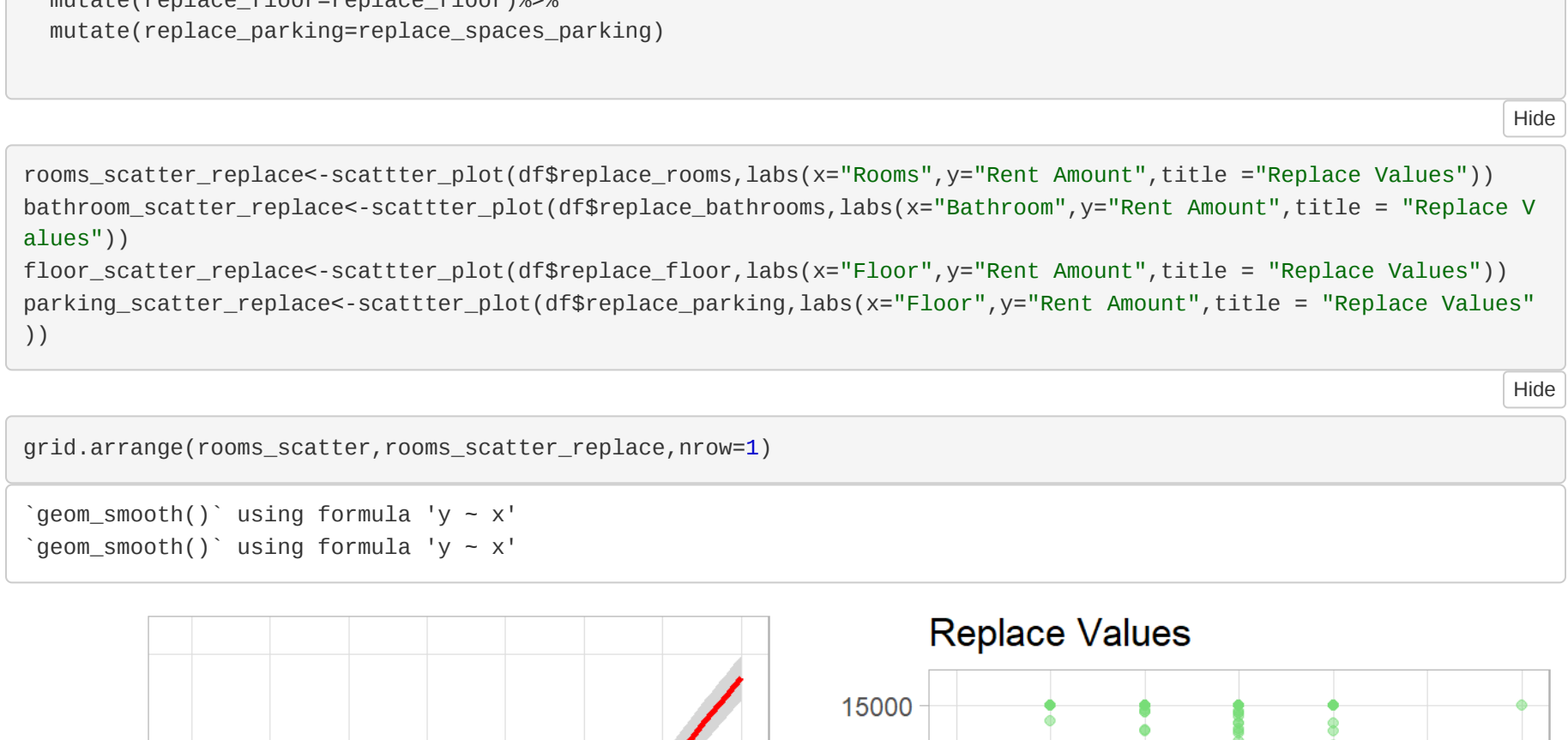
```
replace_values<-function(x,limit){
  ifelse(x>limit,limit,x)
}
```

```
replace_rooms<-apply(replace_values,rooms,7)
replace_bathrooms<-apply(replace_values,bathroom,8)
replace_floor<-apply(replace_values,floor,30)
replace_spaces_parking<-apply(replace_values,parking_spaces,8)
```

```
df<- df %>%
  mutate(replace_rooms=replace_rooms) %>%
  mutate(replace_bathrooms=replace_bathrooms) %>%
  mutate(replace_floor=replace_floor) %>%
  mutate(replace_parking=replace_spaces_parking)
```

```
rooms_scatter_replace<-scatter_plot(df$replace_rooms,labs(x="Rooms",y="Rent Amount",title = "Replace Values"))
bathroom_scatter_replace<-scatter_plot(df$replace_bathrooms,labs(x="Bathroom",y="Rent Amount",title = "Replace V
alue"))
floor_scatter_replace<-scatter_plot(df$replace_floor,labs(x="Floor",y="Rent Amount",title = "Replace Values"))
parking_scatter_replace<-scatter_plot(df$replace_parking,labs(x="Floor",y="Rent Amount",title = "Replace Values"))
```

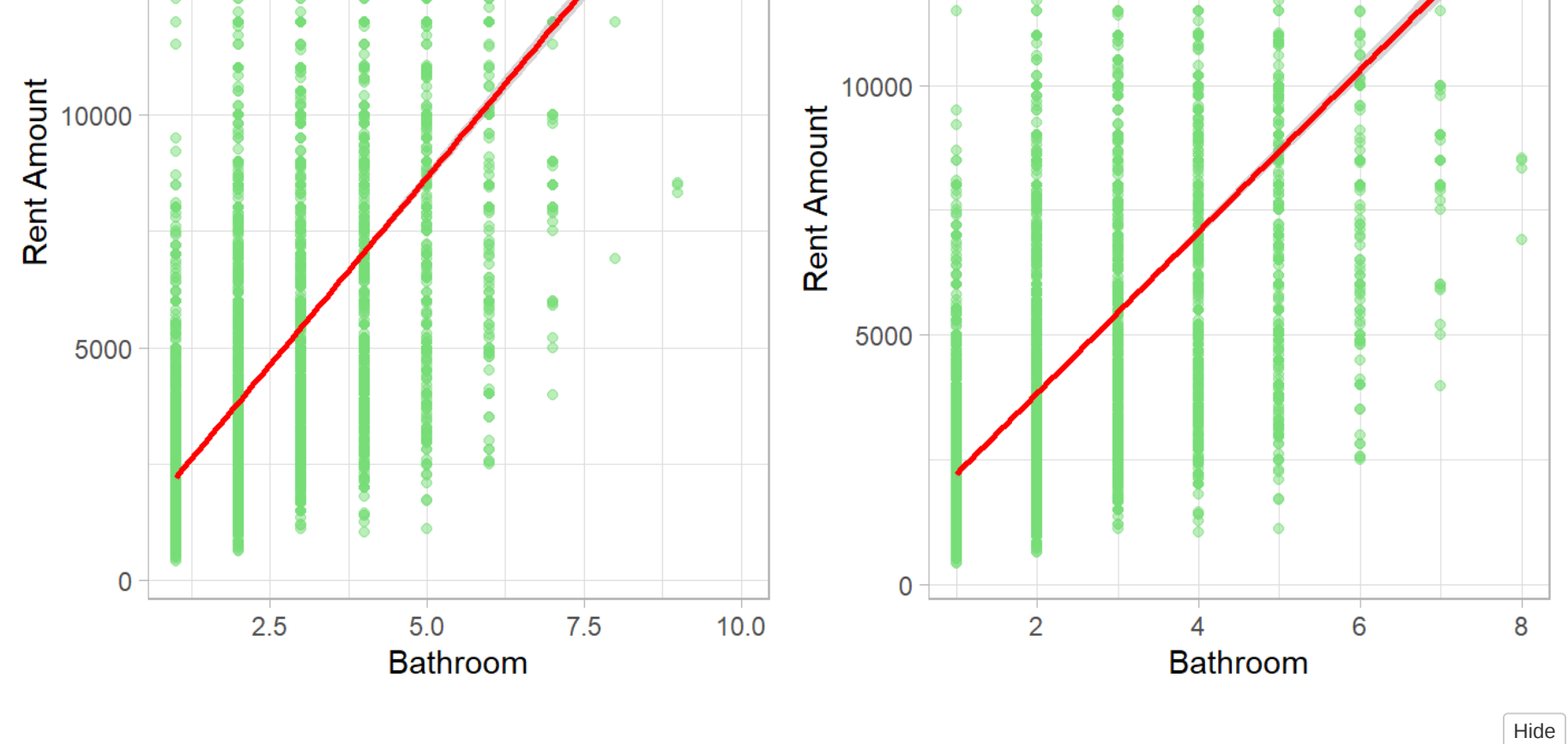
```
grid.arrange(rooms_scatter_replace,rooms_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



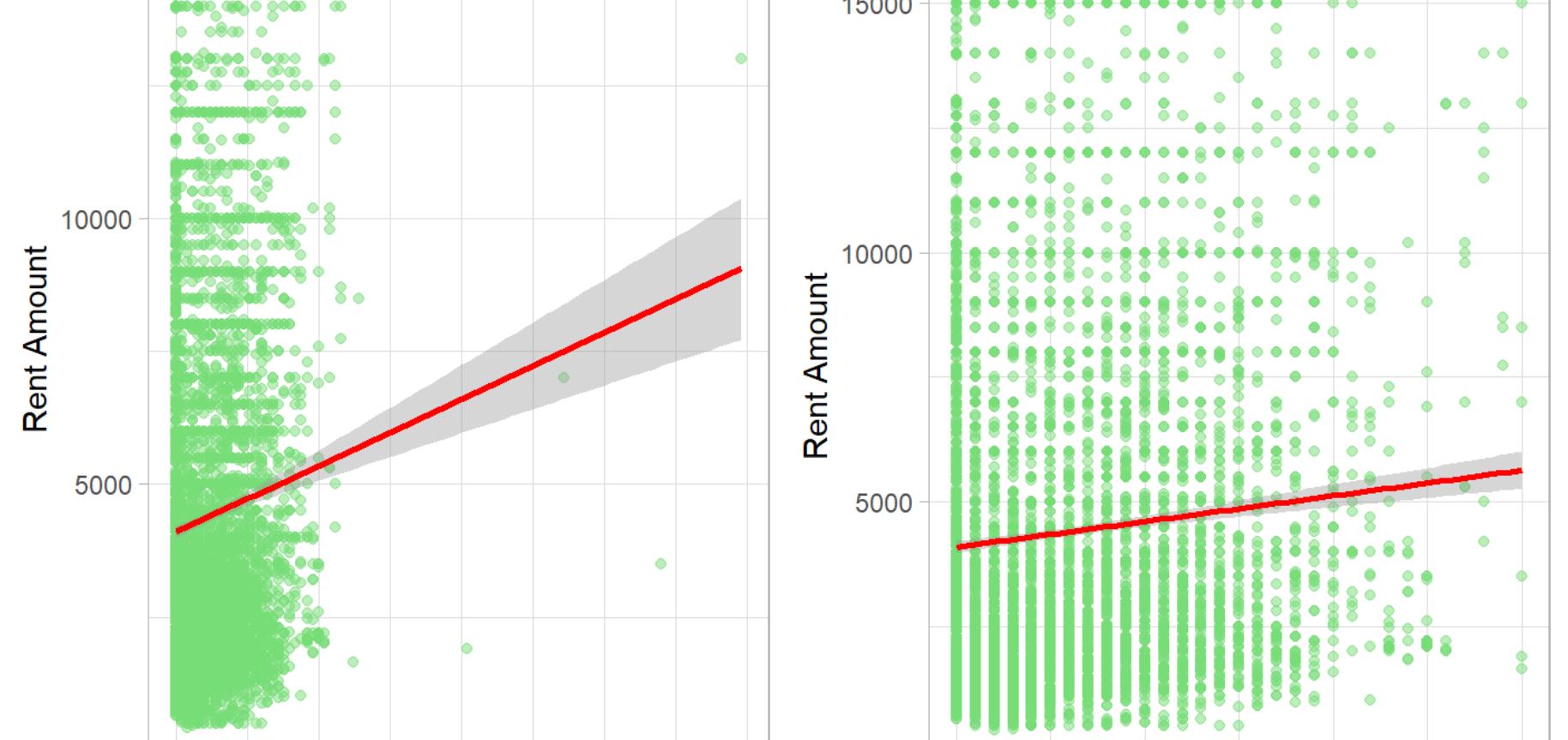
```
grid.arrange(floor_scatter_replace,floor_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



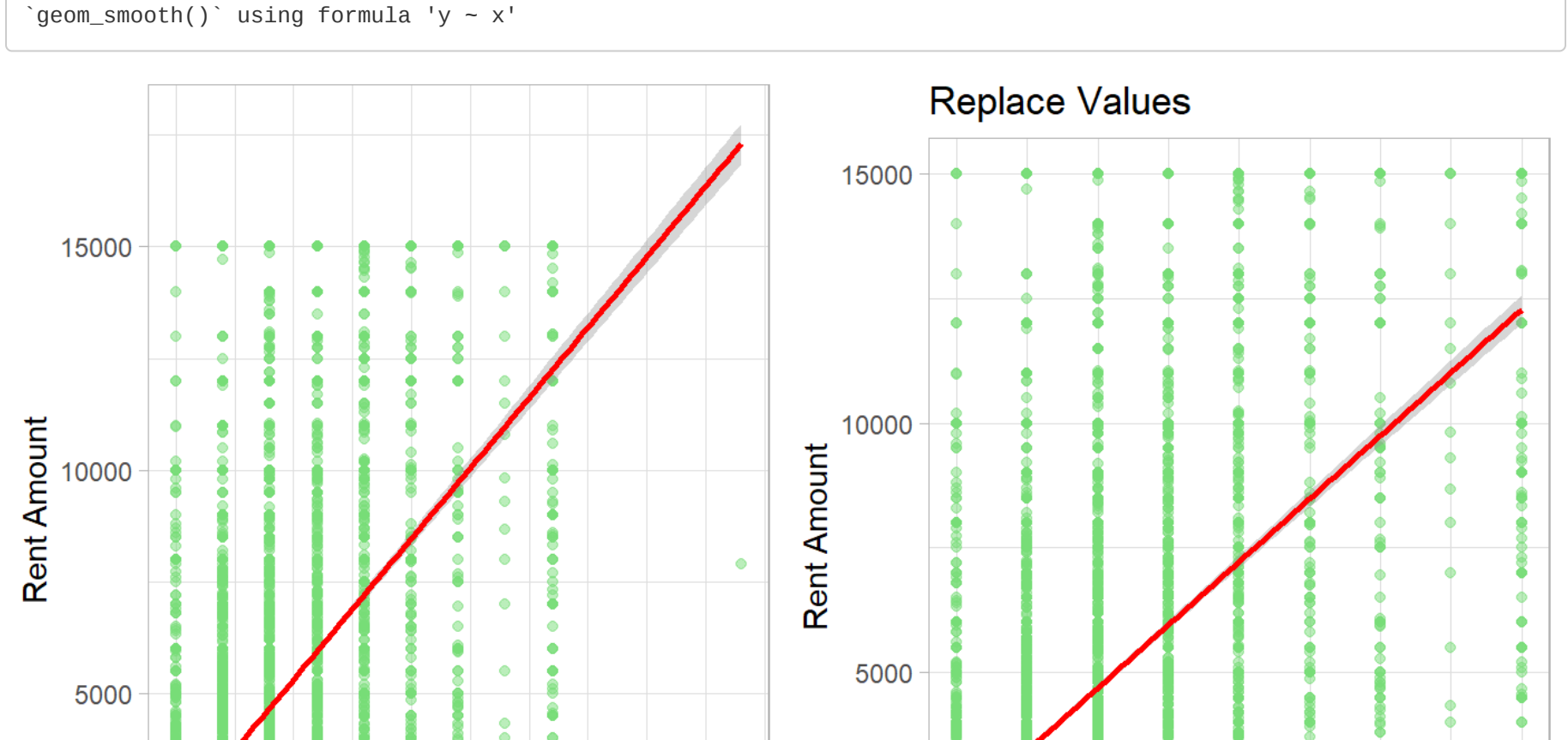
```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



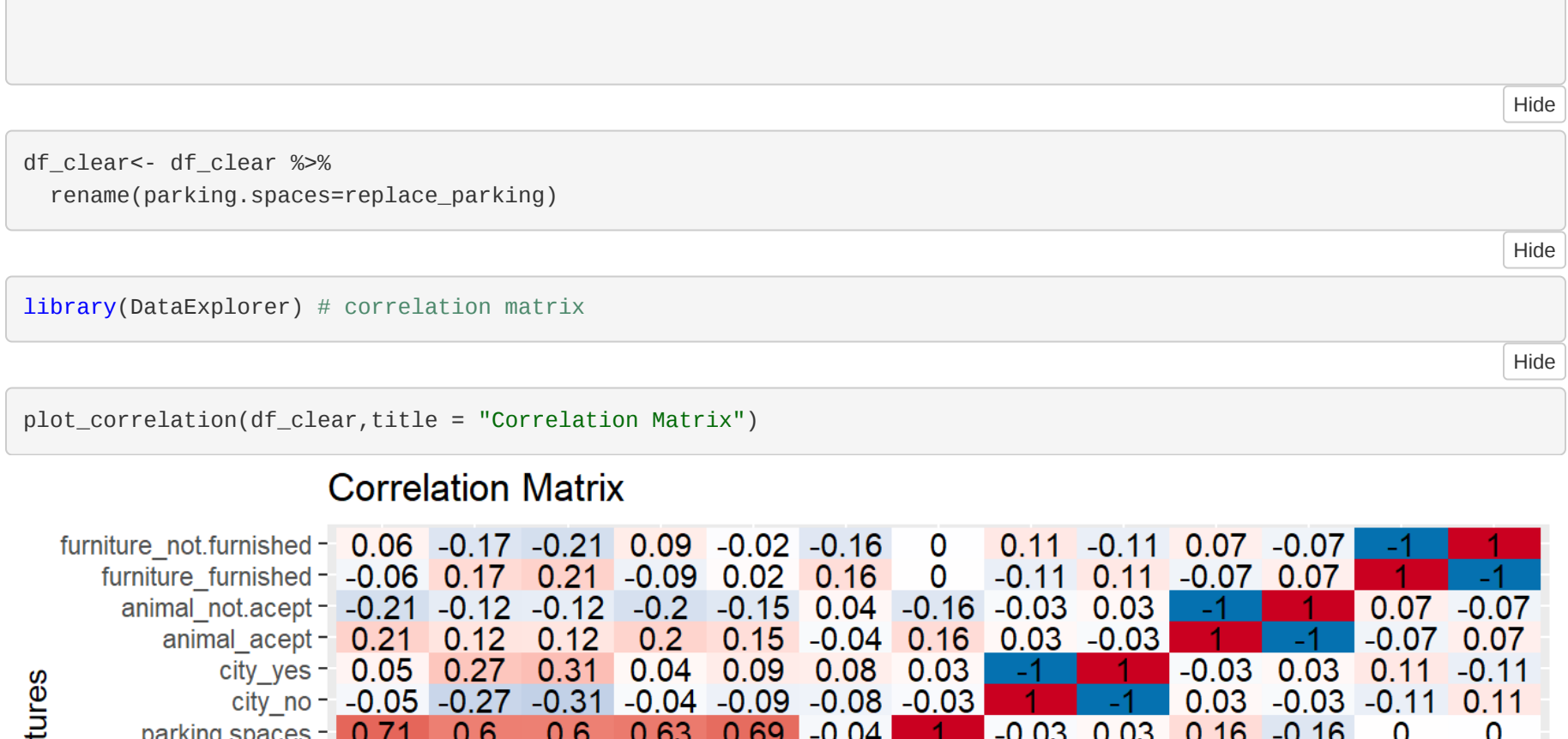
```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



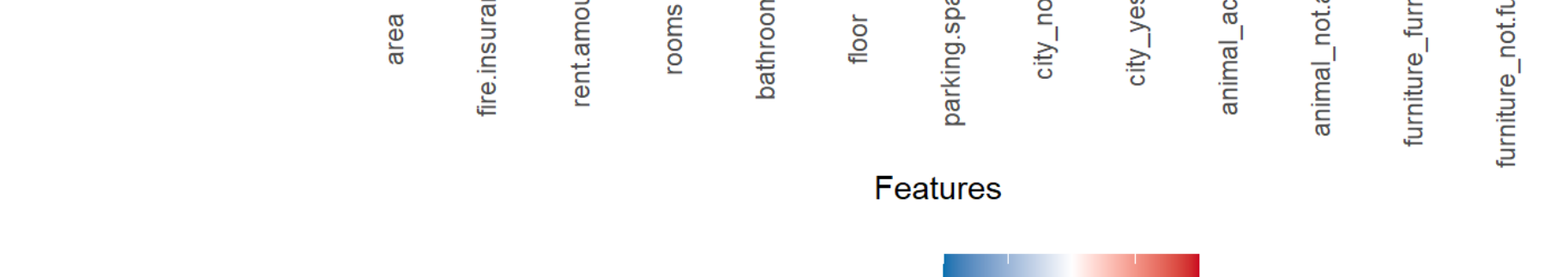
```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



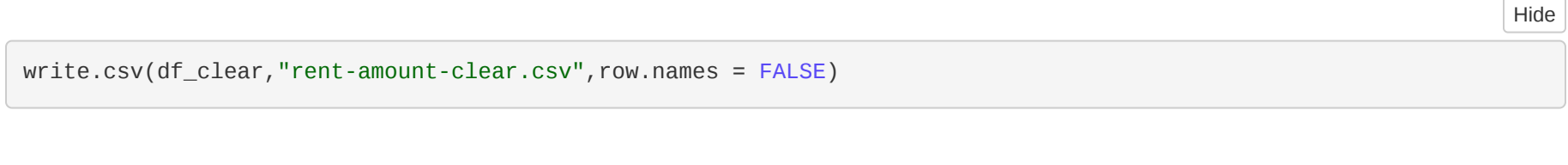
```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



```
grid.arrange(parking_spaces_scatter_replace,parking_spaces_scatter_replace,rows=1)
geom_smooth() using formula 'y ~ x'
geom_smooth() using formula 'y ~ x'
```



### We remove old variables

```
df<-df %>%
  select(-rent.amount,-area,-fire.insurance) %>%
  select(rooms,bathroom,floor,parking_spaces)
```

#### Rename Features

```
df<-clear<-df %>%
  rename(fire.insurance=fire.insurance_log,
  area=area_log,
  rent.amount=rent.amount_log) %>%
  rename(bathrooms=replace_bathrooms,
  floor=replace_floor,
  rooms=replace_rooms)
```

```
df<-clear<- df<-clear %>%
  rename(parking_spaces=replace_parking)
```

```
library(DataExplorer) # correlation matrix
```

```
plot_correlation(df_clear,title = "Correlation Matrix")
```

