

Analysis of Airbnb Prices in Amsterdam - R Appendix

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R Code Appendix

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr 0.3.4
## v tibble 3.1.3       v dplyr 1.0.7
## v tidyr 1.1.3        v stringr 1.4.0
## v readr 2.0.1        v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(GGally)

## Registered S3 method overwritten by 'GGally':
##   method from
##   +.gg      ggplot2

library(ggResidpanel)
library(broom)
library(car)

## Loading required package: carData

##
## Attaching package: 'car'

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

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

library(knitr)

# import dataset
airbnb <- read.csv("amsterdam_weekends.csv")
airbnb_mutate <- airbnb %>% mutate(case = row_number())

# EDA
summary(airbnb)

##           X           realSum      room_type      room_shared
## Min.      : 0   Min.      : 165.9   Length:977      Length:977
```

```
## 1st Qu.:244 1st Qu.: 362.8 Class :character Class :character
## Median :488 Median : 491.6 Mode :character Mode :character
## Mean :488 Mean : 604.8
## 3rd Qu.:732 3rd Qu.: 716.6
## Max. :976 Max. :8130.7
## room_private person_capacity host_is_superhost multi
## Length:977 Min. :2.00 Length:977 Min. :0.0000
## Class :character 1st Qu.:2.00 Class :character 1st Qu.:0.0000
## Mode :character Median :2.00 Mode :character Median :0.0000
## Mean :2.77 Mean :0.2549
## 3rd Qu.:4.00 3rd Qu.:1.0000
## Max. :6.00 Max. :1.0000
## biz cleanliness_rating guest_satisfaction_overall
## Min. :0.00000 Min. : 2.000 Min. : 20.00
## 1st Qu.:0.00000 1st Qu.: 9.000 1st Qu.: 93.00
## Median :0.00000 Median :10.000 Median : 96.00
## Mean :0.09417 Mean : 9.471 Mean : 94.69
## 3rd Qu.:0.00000 3rd Qu.:10.000 3rd Qu.: 99.00
## Max. :1.00000 Max. :10.000 Max. :100.00
## bedrooms dist metro_dist attr_index
## Min. :0.000 Min. : 0.01504 Min. :0.03652 Min. : 40.89
## 1st Qu.:1.000 1st Qu.: 1.41004 1st Qu.:0.46714 1st Qu.: 132.72
## Median :1.000 Median : 2.31455 Median :0.87617 Median : 208.90
## Mean :1.303 Mean : 2.80634 Mean :1.08929 Mean : 266.87
## 3rd Qu.:2.000 3rd Qu.: 3.61017 3rd Qu.:1.50086 3rd Qu.: 366.69
## Max. :5.000 Max. :11.19593 Max. :4.41190 Max. :1888.74
## attr_index_norm rest_index rest_index_norm lng
## Min. : 2.165 Min. : 50.82 Min. : 4.431 Min. :4.776
## 1st Qu.: 7.027 1st Qu.: 170.68 1st Qu.: 14.879 1st Qu.:4.869
## Median : 11.061 Median : 266.96 Median : 23.272 Median :4.888
## Mean : 14.129 Mean : 329.13 Mean : 28.692 Mean :4.890
## 3rd Qu.: 19.414 3rd Qu.: 438.11 3rd Qu.: 38.193 3rd Qu.:4.906
## Max. :100.000 Max. :1147.10 Max. :100.000 Max. :5.011
## lat
## Min. :52.29
## 1st Qu.:52.35
## Median :52.37
## Mean :52.36
## 3rd Qu.:52.38
## Max. :52.42
```

```
# fit a rich model
```

```
airbnb_lm <- lm(realSum ~ room_type+room_shared+room_private+person_capacity+host_is_superhost+multi+biz+lat)
summary(airbnb_lm)
```

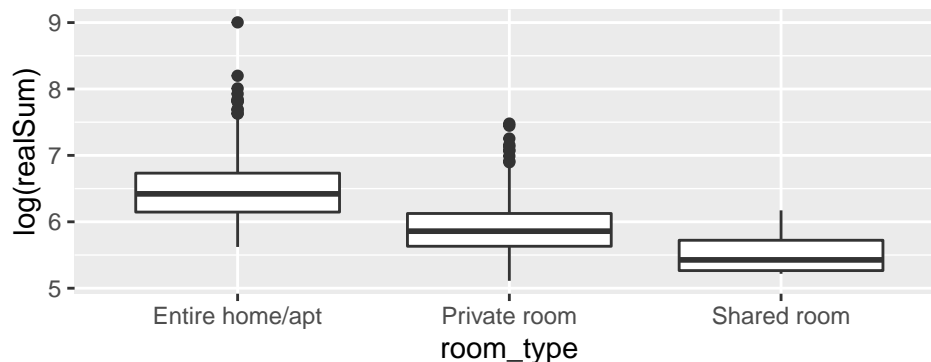
```
##
## Call:
## lm(formula = realSum ~ room_type + room_shared + room_private +
##     person_capacity + host_is_superhost + multi + biz + cleanliness_rating +
##     guest_satisfaction_overall + bedrooms + dist + metro_dist +
##     attr_index + attr_index_norm + rest_index + rest_index_norm +
##     lng + lat, data = airbnb)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -745.6 -138.3 -35.5 76.2 6772.0
##
## Coefficients: (4 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -6.109e+04  4.857e+04  -1.258  0.2088
## room_typePrivate room    -1.626e+02  2.561e+01  -6.349 3.35e-10 ***
## room_typeShared room    -3.139e+02  1.707e+02  -1.839  0.0663 .
## room_sharedTrue          NA         NA      NA      NA
## room_privateTrue         NA         NA      NA      NA
## person_capacity      1.275e+02  1.449e+01   8.799 < 2e-16 ***
## host_is_superhostTrue  -9.423e-01  2.559e+01  -0.037  0.9706
## multi                2.155e+01  2.653e+01   0.812  0.4167
## biz                 -3.615e+01  3.920e+01  -0.922  0.3567
## cleanliness_rating    1.263e+01  1.780e+01   0.710  0.4780
## guest_satisfaction_overall 1.942e+00  2.259e+00   0.860  0.3903
## bedrooms            1.589e+02  2.013e+01   7.893 8.03e-15 ***
## dist              -1.654e+01  1.133e+01  -1.459  0.1448
## metro_dist         -8.253e+00  1.797e+01  -0.459  0.6461
## attr_index          1.806e-01  1.072e-01   1.684  0.0925 .
## attr_index_norm      NA         NA      NA      NA
## rest_index           2.279e-01  1.111e-01   2.052  0.0404 *
## rest_index_norm      NA         NA      NA      NA
## lng                4.348e+02  2.928e+02   1.485  0.1379
## lat                1.121e+03  9.230e+02   1.214  0.2249
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 335.2 on 961 degrees of freedom
## Multiple R-squared:  0.438, Adjusted R-squared:  0.4292
## F-statistic: 49.93 on 15 and 961 DF, p-value: < 2.2e-16
```

```
# check model transformation
```

```
# room_type vs. realSum
```

```
ggplot(airbnb, aes(x=room_type, y=log(realSum))) + geom_boxplot()
```

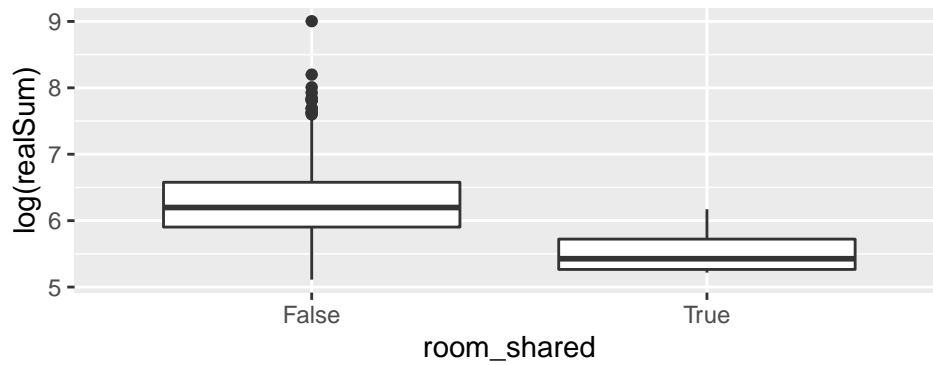


```
table(airbnb$room_type)
```

```
##
## Entire home/apt    Private room    Shared room
##           588           385           4
```

```
# room_shared vs. realSum
```

```
ggplot(airbnb, aes(x=room_shared, y=log(realSum))) + geom_boxplot()
```

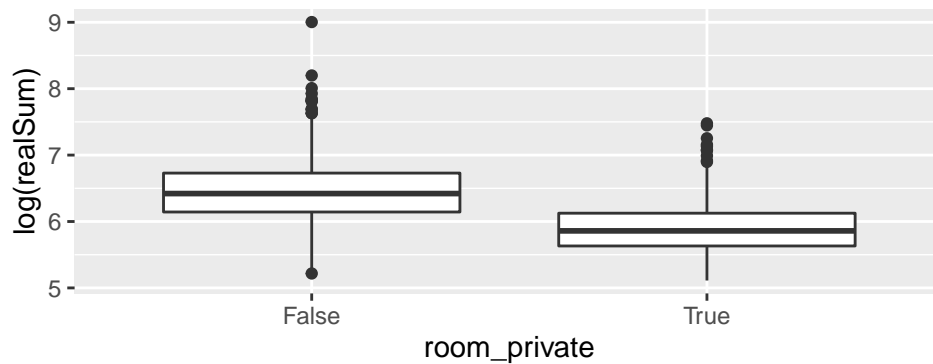


```
table(airbnb$room_shared)
```

```
##
## False  True
##   973    4
```

room_private vs. realSum

```
ggplot(airbnb, aes(x=room_private, y=log(realSum))) + geom_boxplot()
```



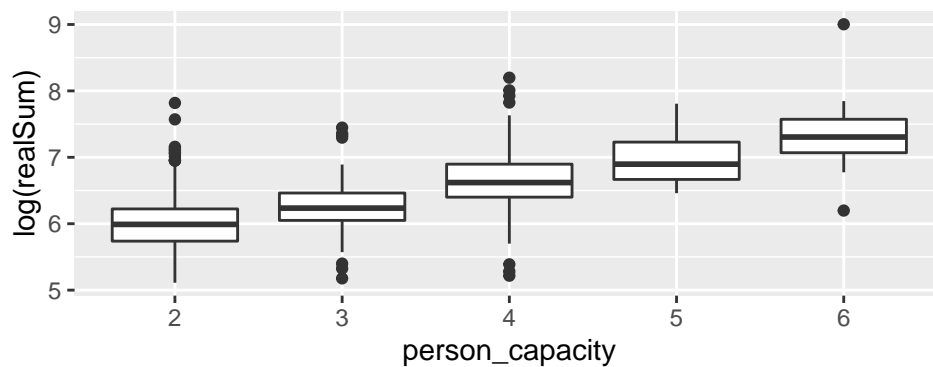
```
table(airbnb$room_private)
```

```
##
## False  True
##   592   385
```

person_capacity vs. realSum

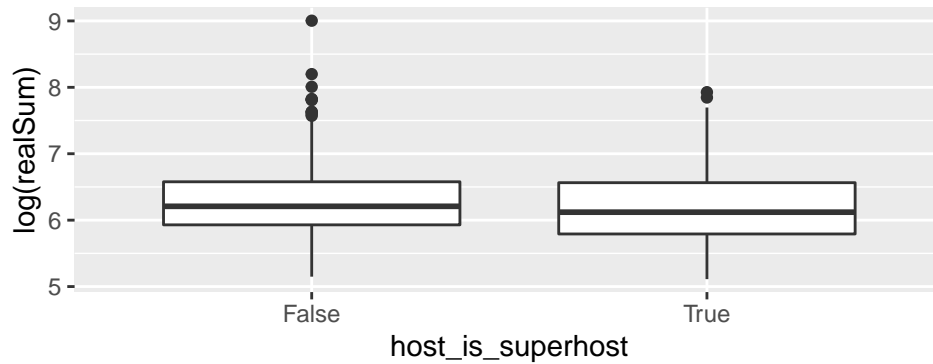
```
airbnb$person_capacity <- as.factor(airbnb$person_capacity)
```

```
ggplot(airbnb, aes(x=person_capacity, y=log(realSum))) + geom_boxplot()
```



```
table(airbnb$person_capacity)
```

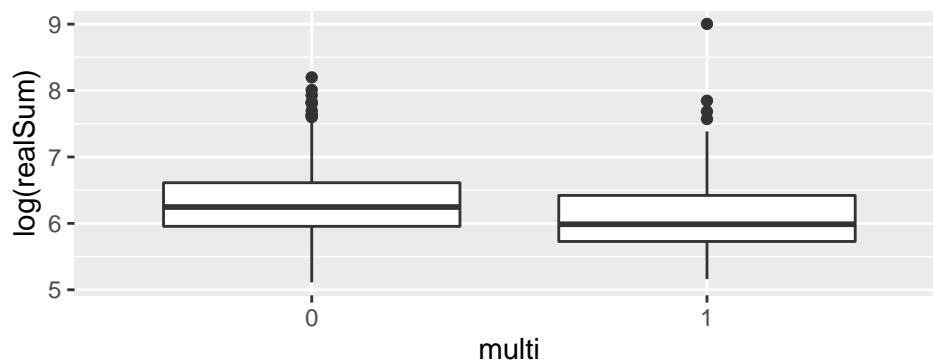
```
##
##      2      3      4      5      6
## 591    60   301    10    15
# host_is_superhost vs. realSum
ggplot(airbnb, aes(x=host_is_superhost, y=log(realSum))) + geom_boxplot()
```



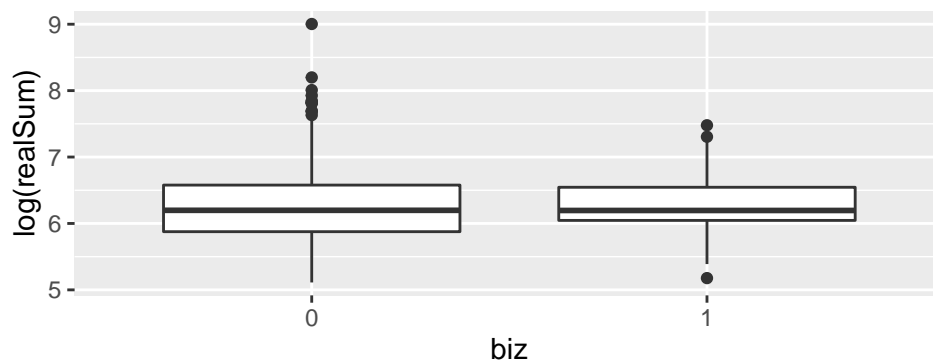
```
table(airbnb$host_is_superhost)
```

```
##
## False  True
##   709    268
```

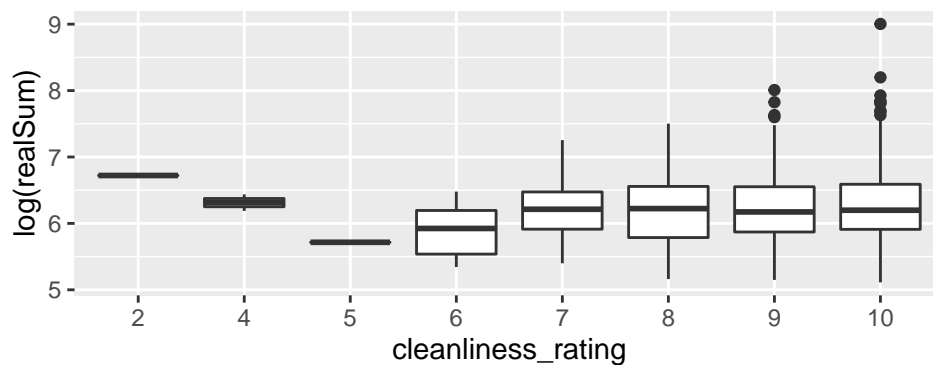
```
# multi vs. realSum
airbnb$multi <- as.factor(airbnb$multi)
ggplot(airbnb, aes(x=multi, y=log(realSum))) + geom_boxplot()
```



```
# biz vs. realSum
airbnb$biz <- as.factor(airbnb$biz)
ggplot(airbnb, aes(x=biz, y=log(realSum))) + geom_boxplot()
```



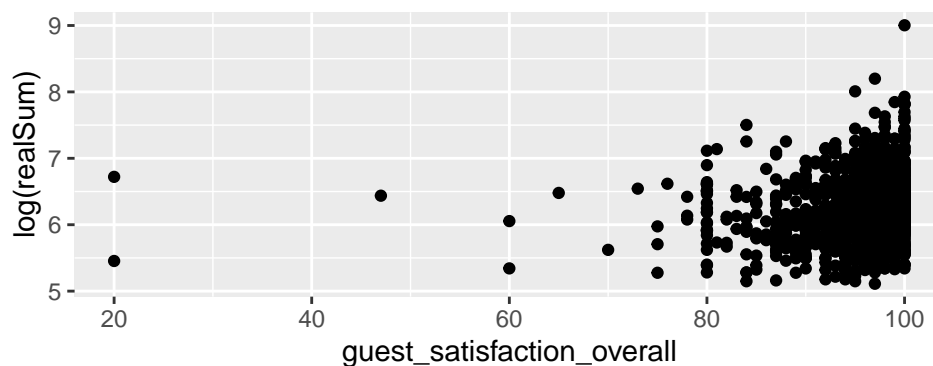
```
# cleanliness_rating vs. realSum
airbnb$cleanliness_rating <- as.factor(airbnb$cleanliness_rating)
ggplot(airbnb, aes(x=cleanliness_rating, y=log(realSum))) + geom_boxplot()
```



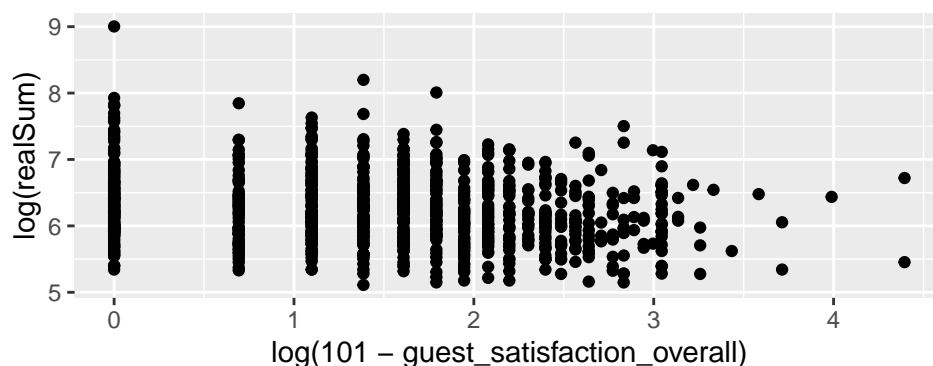
```
table(airbnb$cleanliness_rating)
```

```
##
##  2  4  5  6  7  8  9 10
##  1  2  1  7  8 81 278 599
# guest_satisfaction_overall vs. realSum
```

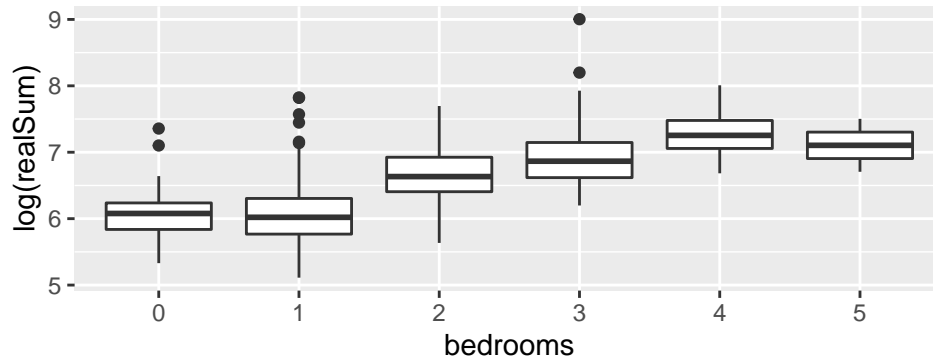
```
ggplot(airbnb, aes(x=guest_satisfaction_overall, y=log(realSum))) + geom_point()
```



```
ggplot(airbnb, aes(x=log(101-guest_satisfaction_overall), y=log(realSum))) + geom_point()
```



```
# bedrooms vs. realSum
airbnb$bedrooms <- as.factor(airbnb$bedrooms)
ggplot(airbnb, aes(x=bedrooms, y=log(realSum))) + geom_boxplot()
```

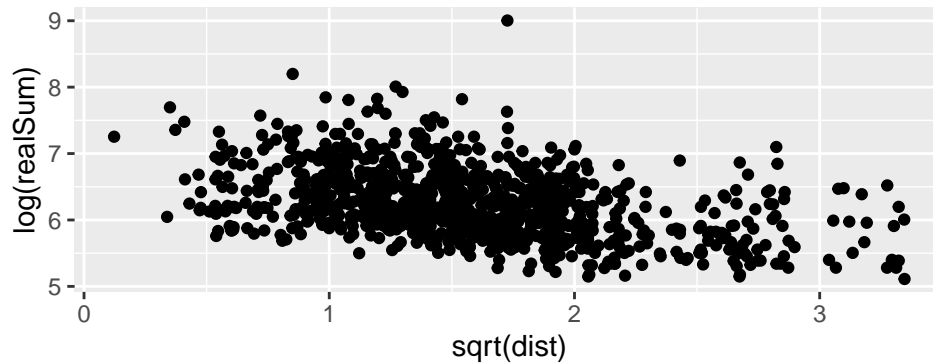


```
table(airbnb$bedrooms)
```

```
##
##    0    1    2    3    4    5
## 63 636 210  57   9   2
```

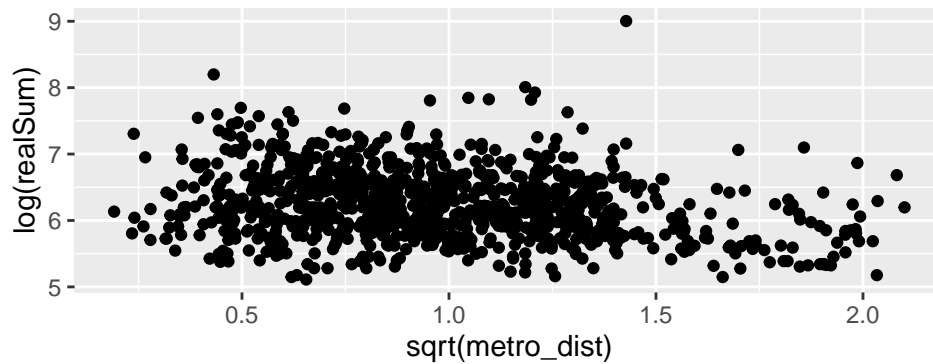
```
# dist vs. realSum
```

```
ggplot(airbnb, aes(x=sqrt(dist), y=log(realSum))) + geom_point()
```



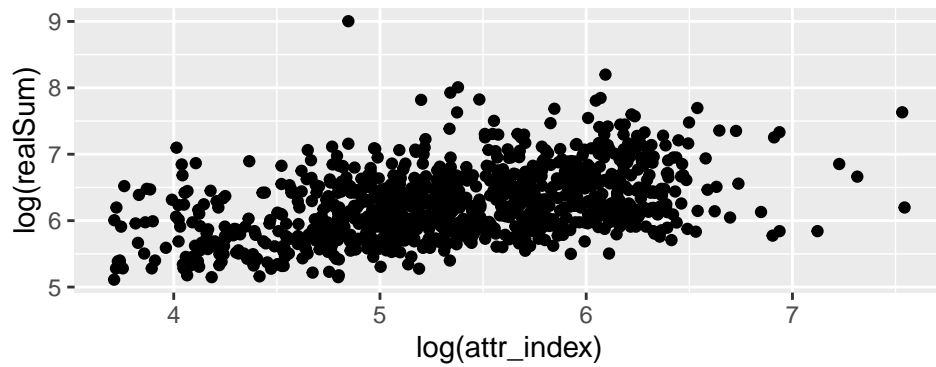
```
# metro_dist vs. realSum
```

```
ggplot(airbnb, aes(x=sqrt(metro_dist), y=log(realSum))) + geom_point()
```

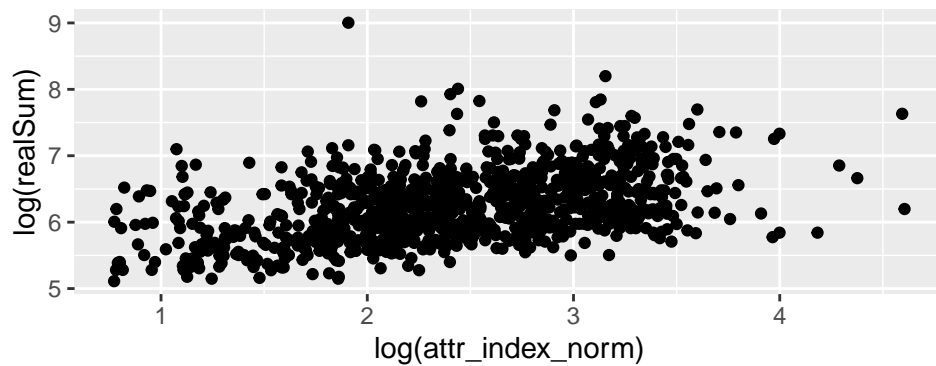


```
# attr_index vs. realSum
```

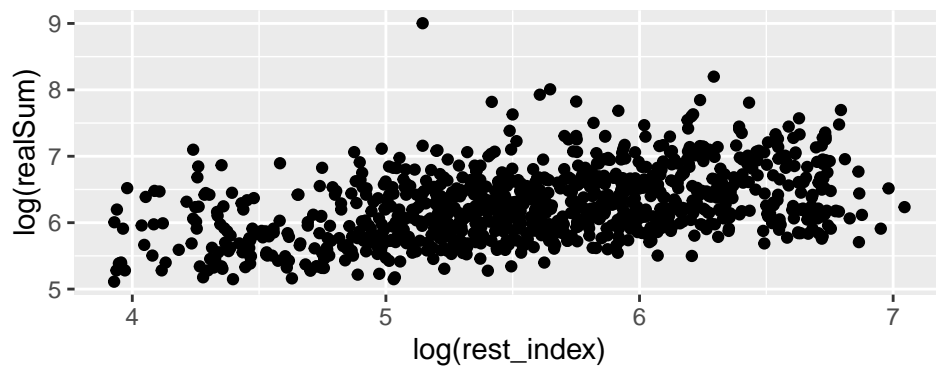
```
ggplot(airbnb, aes(x=log(attr_index), y=log(realSum))) + geom_point()
```



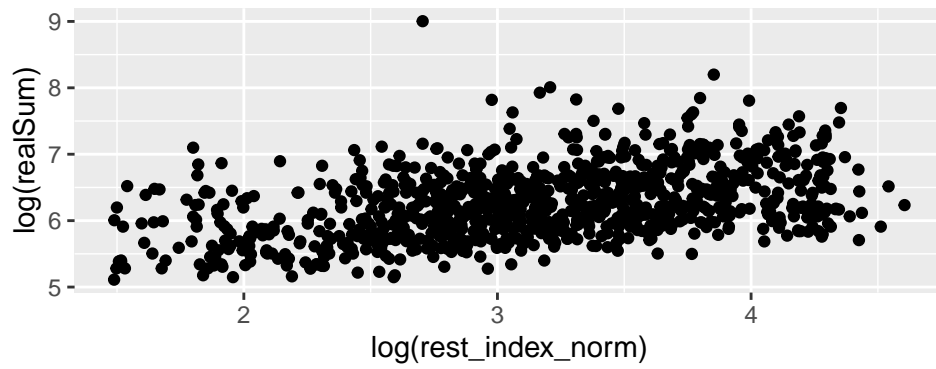
```
# attr_index_norm vs. realSum
ggplot(airbnb, aes(x=log(attr_index_norm), y=log(realSum))) + geom_point()
```



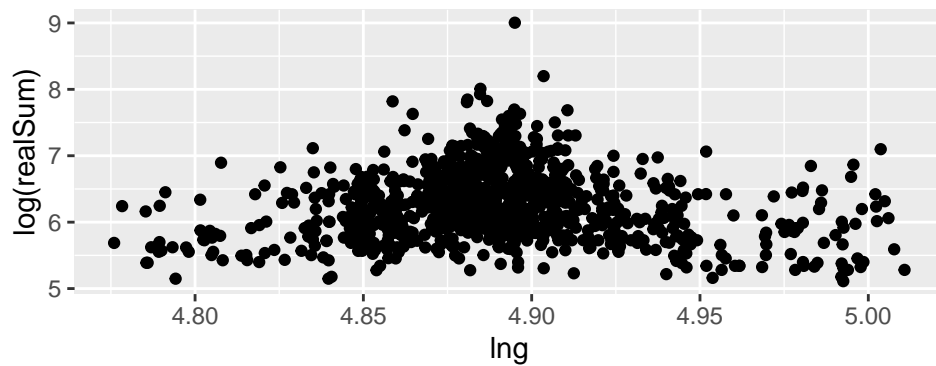
```
# rest_index vs. realSum
ggplot(airbnb, aes(x=log(rest_index), y=log(realSum))) + geom_point()
```



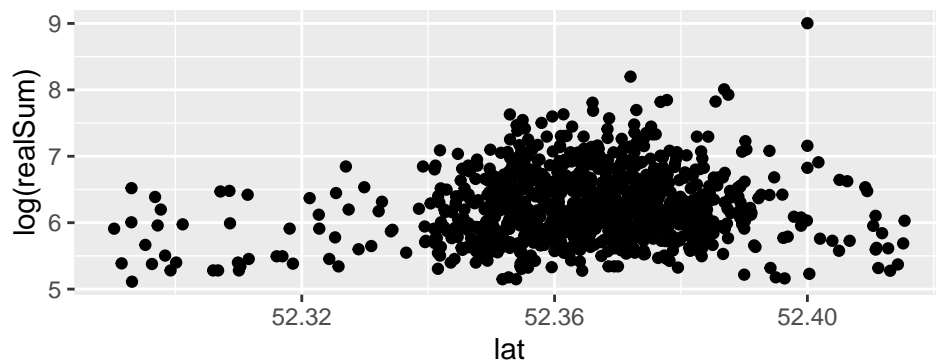
```
# rest_index_norm vs. realSum
ggplot(airbnb, aes(x=log(rest_index_norm), y=log(realSum))) + geom_point()
```

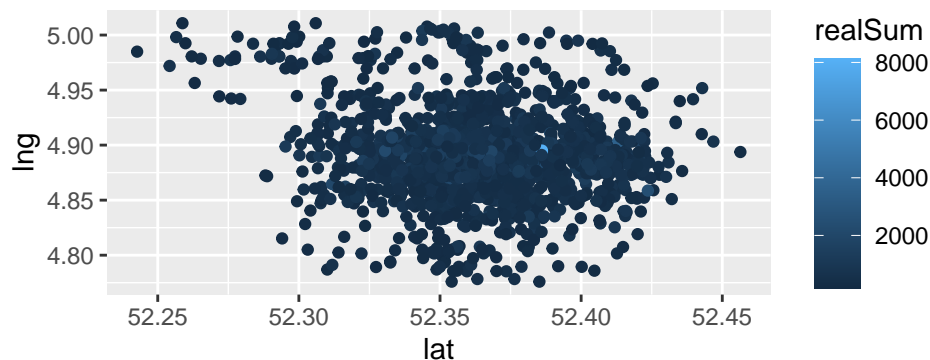
```
# lng vs. realSum
ggplot(airbnb, aes(x=lng, y=log(realSum))) + geom_point()
```



```
# lat vs. realSum
ggplot(airbnb, aes(x=lat, y=log(realSum))) + geom_point()
```



```
# lng vs. lat
ggplot(airbnb, aes(x=lat, y=lng, color = realSum)) + geom_point() + geom_jitter(width = 0.05)
```



```
# transformed model
airbnb_lm_transformed <- lm(log(realSum) ~ room_type+room_shared+room_private+person_capacity+host_is_s
summary(airbnb_lm_transformed)
```

```
##
## Call:
## lm(formula = log(realSum) ~ room_type + room_shared + room_private +
##     person_capacity + host_is_superhost + multi + biz + cleanliness_rating +
##     log(101 - guest_satisfaction_overall) + bedrooms + sqrt(dist) +
##     sqrt(metro_dist) + log(attr_index) + log(attr_index_norm) +
##     log(rest_index) + log(rest_index_norm), data = airbnb)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96796 -0.18472 -0.02441  0.14165  1.72381
##
## Coefficients: (4 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      4.8137741   0.4835551    9.955 < 2e-16
## room_typePrivate room -0.3250904   0.0222121  -14.636 < 2e-16
## room_typeShared room -0.6211706   0.1483561   -4.187 3.09e-05
## room_sharedTrue      NA          NA      NA      NA
## room_privateTrue     NA          NA      NA      NA
## person_capacity3      0.1933472   0.0411843    4.695 3.06e-06
## person_capacity4      0.3266046   0.0311079   10.499 < 2e-16
## person_capacity5      0.5413083   0.0972736    5.565 3.41e-08
## person_capacity6      0.7711199   0.0852378    9.047 < 2e-16
## host_is_superhostTrue -0.0006424   0.0222932   -0.029 0.977016
## multi1              -0.0170297   0.0230957   -0.737 0.461088
## biz1                 0.0061857   0.0340945    0.181 0.856071
## cleanliness_rating4    0.2543658   0.3582216    0.710 0.477829
## cleanliness_rating5   -0.4621271   0.4160099   -1.111 0.266911
## cleanliness_rating6   -0.4156111   0.3128272   -1.329 0.184311
## cleanliness_rating7   -0.3738294   0.3100169   -1.206 0.228181
## cleanliness_rating8   -0.3189049   0.2954928   -1.079 0.280759
## cleanliness_rating9   -0.3080339   0.2944443   -1.046 0.295756
## cleanliness_rating10  -0.2933455   0.2956053   -0.992 0.321277
## log(101 - guest_satisfaction_overall) -0.0394297   0.0122789   -3.211 0.001366
## bedrooms1             0.0454986   0.0393075    1.158 0.247358
## bedrooms2             0.2215028   0.0493335    4.490 8.00e-06
## bedrooms3             0.4509334   0.0623491    7.232 9.76e-13
## bedrooms4             0.7568799   0.1079190    7.013 4.41e-12
```

```

## bedrooms5                0.7232943  0.2102449  3.440 0.000606
## sqrt(dist)                0.0152180  0.0570692  0.267 0.789789
## sqrt(metro_dist)          0.0059329  0.0297082  0.200 0.841752
## log(attr_index)            0.0913190  0.0656785  1.390 0.164735
## log(attr_index_norm)      NA          NA          NA    NA
## log(rest_index)           0.2092932  0.0780708  2.681 0.007472
## log(rest_index_norm)      NA          NA          NA    NA
##
## (Intercept)                ***
## room_typePrivate room      ***
## room_typeShared room      ***
## room_sharedTrue
## room_privateTrue
## person_capacity3           ***
## person_capacity4           ***
## person_capacity5           ***
## person_capacity6           ***
## host_is_superhostTrue
## multi1
## biz1
## cleanliness_rating4
## cleanliness_rating5
## cleanliness_rating6
## cleanliness_rating7
## cleanliness_rating8
## cleanliness_rating9
## cleanliness_rating10
## log(101 - guest_satisfaction_overall) **
## bedrooms1
## bedrooms2                  ***
## bedrooms3                  ***
## bedrooms4                  ***
## bedrooms5                  ***
## sqrt(dist)
## sqrt(metro_dist)
## log(attr_index)
## log(attr_index_norm)
## log(rest_index)            **
## log(rest_index_norm)
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.289 on 950 degrees of freedom
## Multiple R-squared:  0.6997, Adjusted R-squared:  0.6915
## F-statistic: 85.14 on 26 and 950 DF, p-value: < 2.2e-16

airbnb_lm_transformed <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+cleanl.
summary(airbnb_lm_transformed)

##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(dist) + sqrt(metro_dist) + log(attr_index) +
##     log(rest_index), data = airbnb)

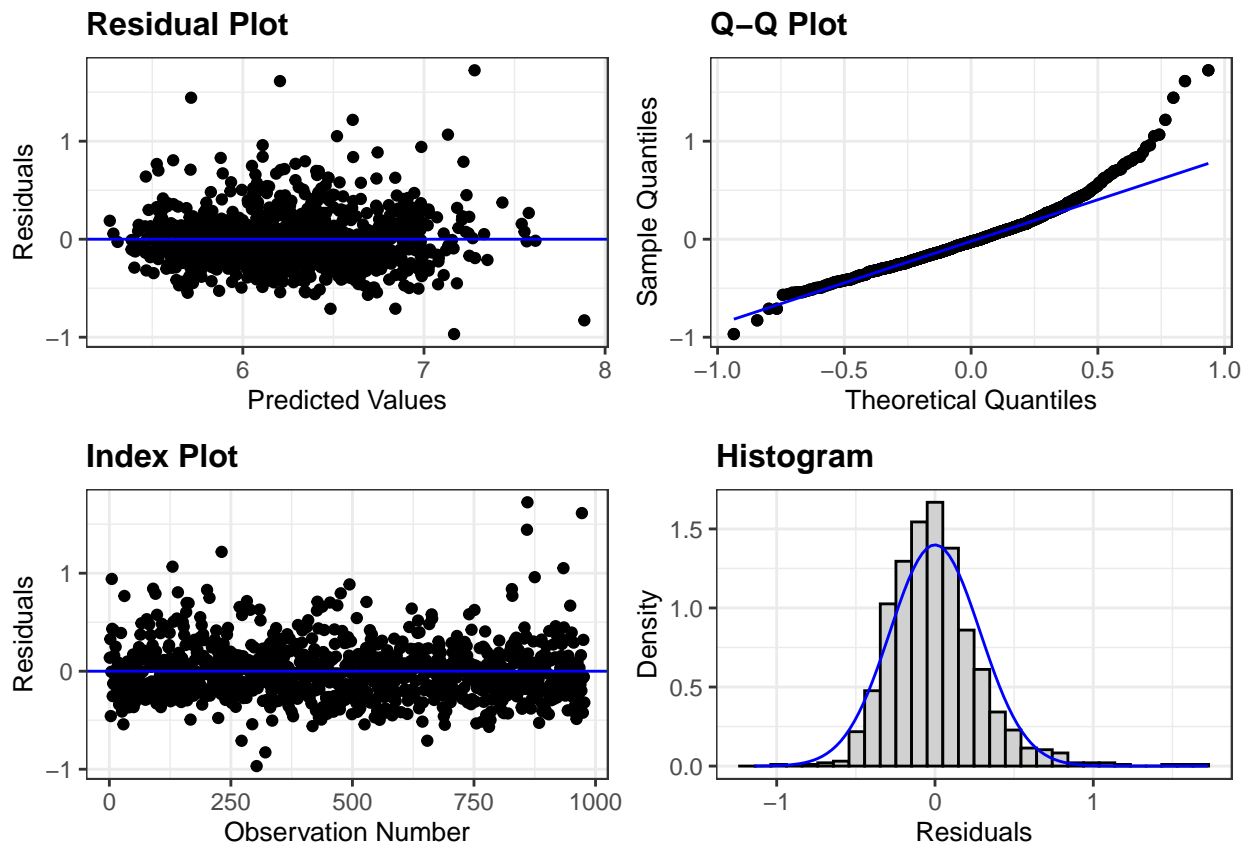
```

```

##
## Residuals:
##      Min        1Q      Median        3Q        Max
## -0.96796 -0.18472 -0.02441  0.14165  1.72381
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      4.8137741   0.4835551    9.955 < 2e-16
## room_typePrivate room -0.3250904   0.0222121  -14.636 < 2e-16
## room_typeShared room -0.6211706   0.1483561   -4.187 3.09e-05
## person_capacity3      0.1933472   0.0411843    4.695 3.06e-06
## person_capacity4      0.3266046   0.0311079   10.499 < 2e-16
## person_capacity5      0.5413083   0.0972736    5.565 3.41e-08
## person_capacity6      0.7711199   0.0852378    9.047 < 2e-16
## host_is_superhostTrue -0.0006424   0.0222932   -0.029 0.977016
## multi1              -0.0170297   0.0230957   -0.737 0.461088
## biz1                0.0061857   0.0340945    0.181 0.856071
## cleanliness_rating4    0.2543658   0.3582216    0.710 0.477829
## cleanliness_rating5   -0.4621271   0.4160099   -1.111 0.266911
## cleanliness_rating6   -0.4156111   0.3128272   -1.329 0.184311
## cleanliness_rating7   -0.3738294   0.3100169   -1.206 0.228181
## cleanliness_rating8   -0.3189049   0.2954928   -1.079 0.280759
## cleanliness_rating9   -0.3080339   0.2944443   -1.046 0.295756
## cleanliness_rating10  -0.2933455   0.2956053   -0.992 0.321277
## log(101 - guest_satisfaction_overall) -0.0394297   0.0122789   -3.211 0.001366
## bedrooms1            0.0454986   0.0393075    1.158 0.247358
## bedrooms2            0.2215028   0.0493335    4.490 8.00e-06
## bedrooms3            0.4509334   0.0623491    7.232 9.76e-13
## bedrooms4            0.7568799   0.1079190    7.013 4.41e-12
## bedrooms5            0.7232943   0.2102449    3.440 0.000606
## sqrt(dist)           0.0152180   0.0570692    0.267 0.789789
## sqrt(metro_dist)      0.0059329   0.0297082    0.200 0.841752
## log(attr_index)       0.0913190   0.0656785    1.390 0.164735
## log(rest_index)       0.2092932   0.0780708    2.681 0.007472
##
## (Intercept) ***
## room_typePrivate room ***
## room_typeShared room ***
## person_capacity3 ***
## person_capacity4 ***
## person_capacity5 ***
## person_capacity6 ***
## host_is_superhostTrue
## multi1
## biz1
## cleanliness_rating4
## cleanliness_rating5
## cleanliness_rating6
## cleanliness_rating7
## cleanliness_rating8
## cleanliness_rating9
## cleanliness_rating10
## log(101 - guest_satisfaction_overall) **
## bedrooms1

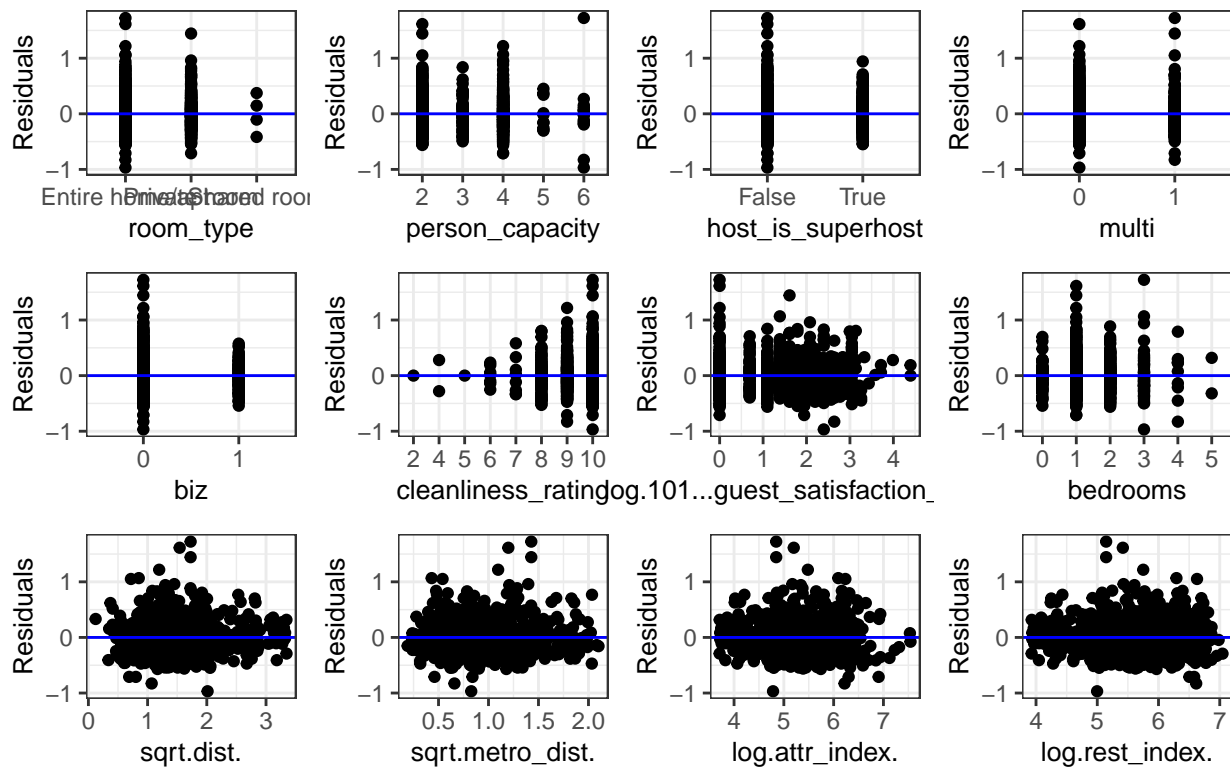
```

```
## bedrooms2 ***
## bedrooms3 ***
## bedrooms4 ***
## bedrooms5 ***
## sqrt(dist)
## sqrt(metro_dist)
## log(attr_index)
## log(rest_index) **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.289 on 950 degrees of freedom
## Multiple R-squared:  0.6997, Adjusted R-squared:  0.6915
## F-statistic: 85.14 on 26 and 950 DF,  p-value: < 2.2e-16
# check assumptions
resid_panel(airbnb_lm_transformed)
```

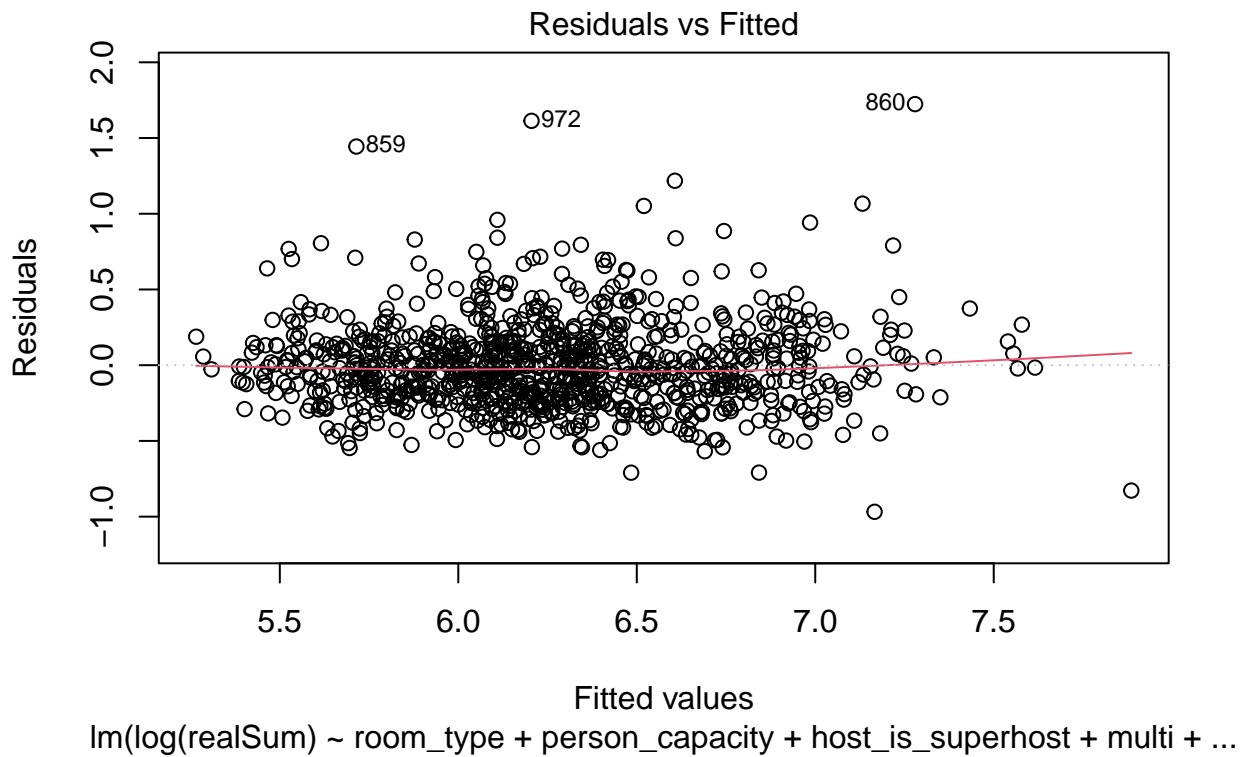


```
resid_xpanel(airbnb_lm_transformed)
```

Plots of Residuals vs Predictor Variables

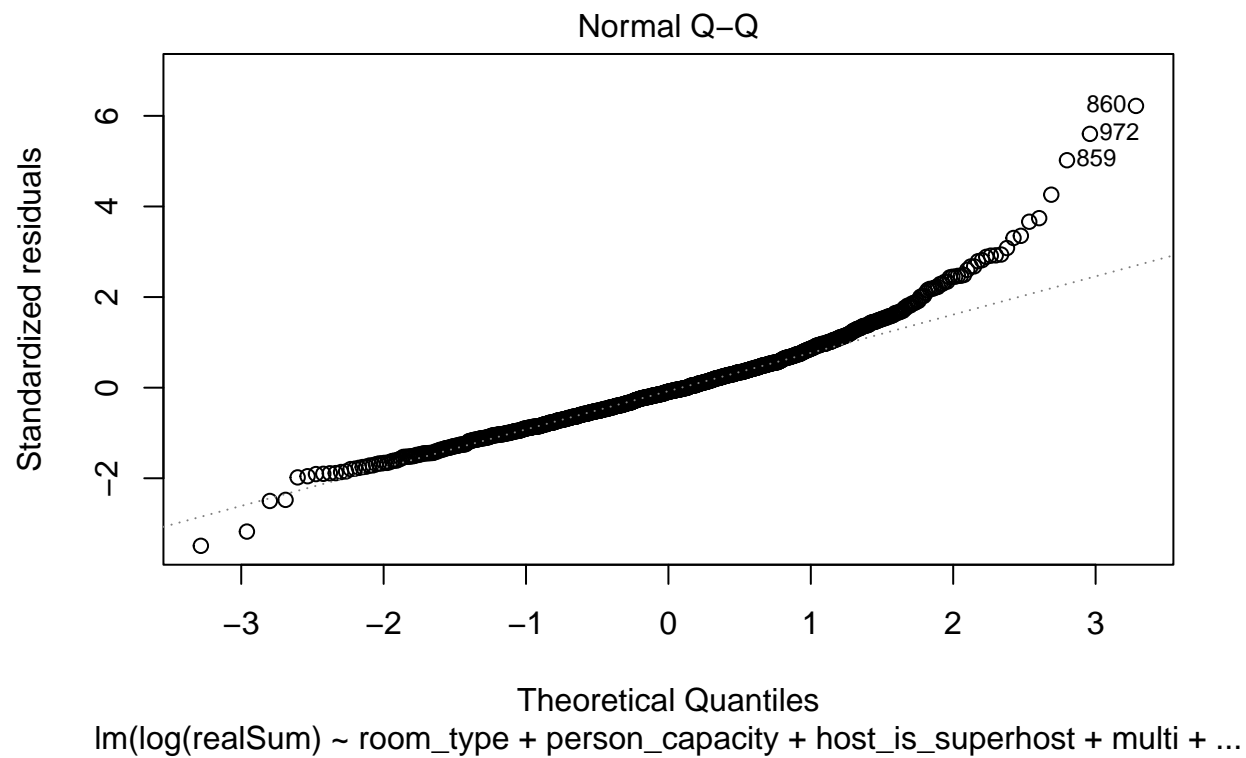


```
plot(airbnb_lm_transformed, which = 1)
```

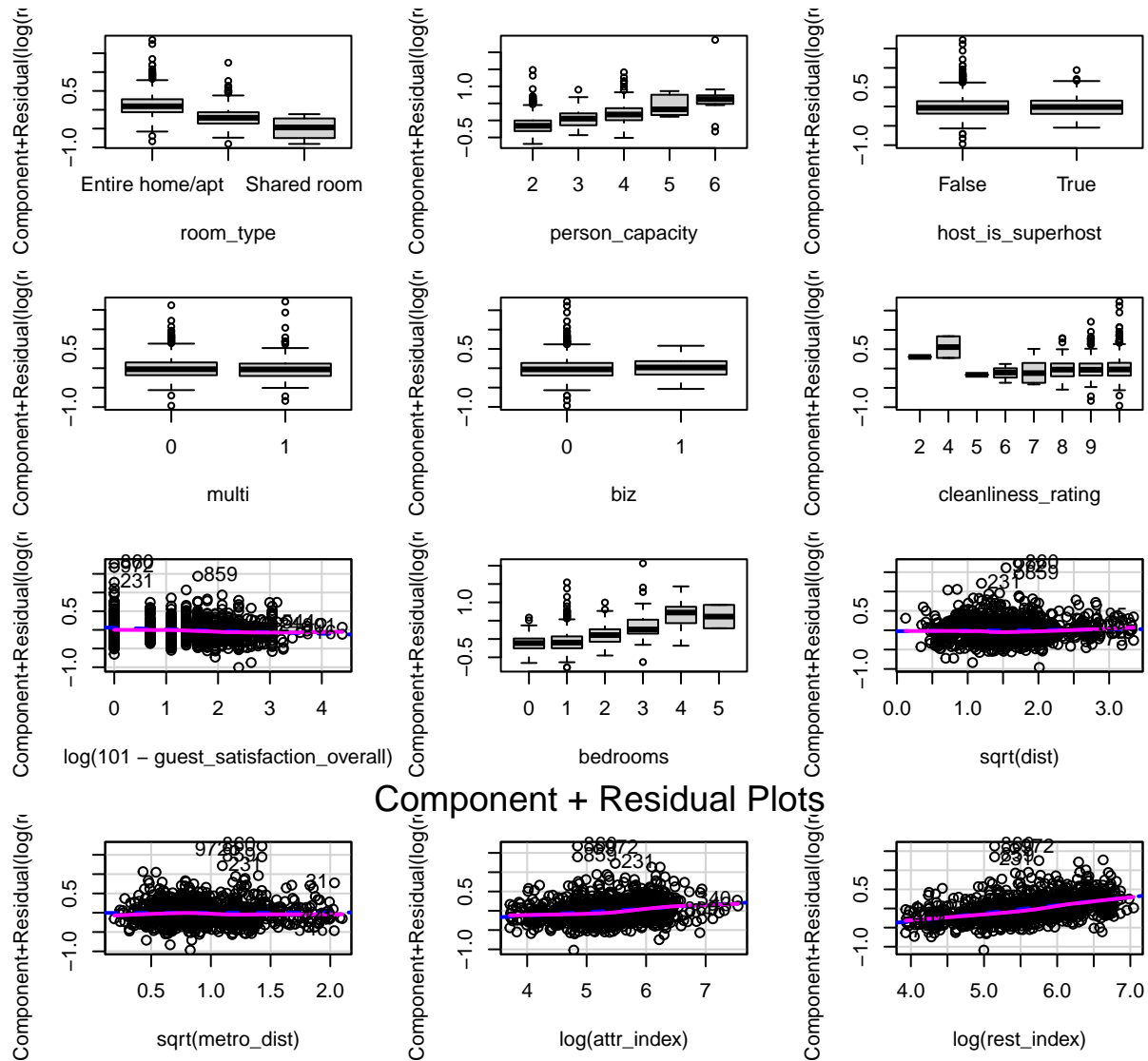


```
plot(airbnb_lm_transformed, which = 2)
```

```
## Warning: not plotting observations with leverage one:  
##      816
```

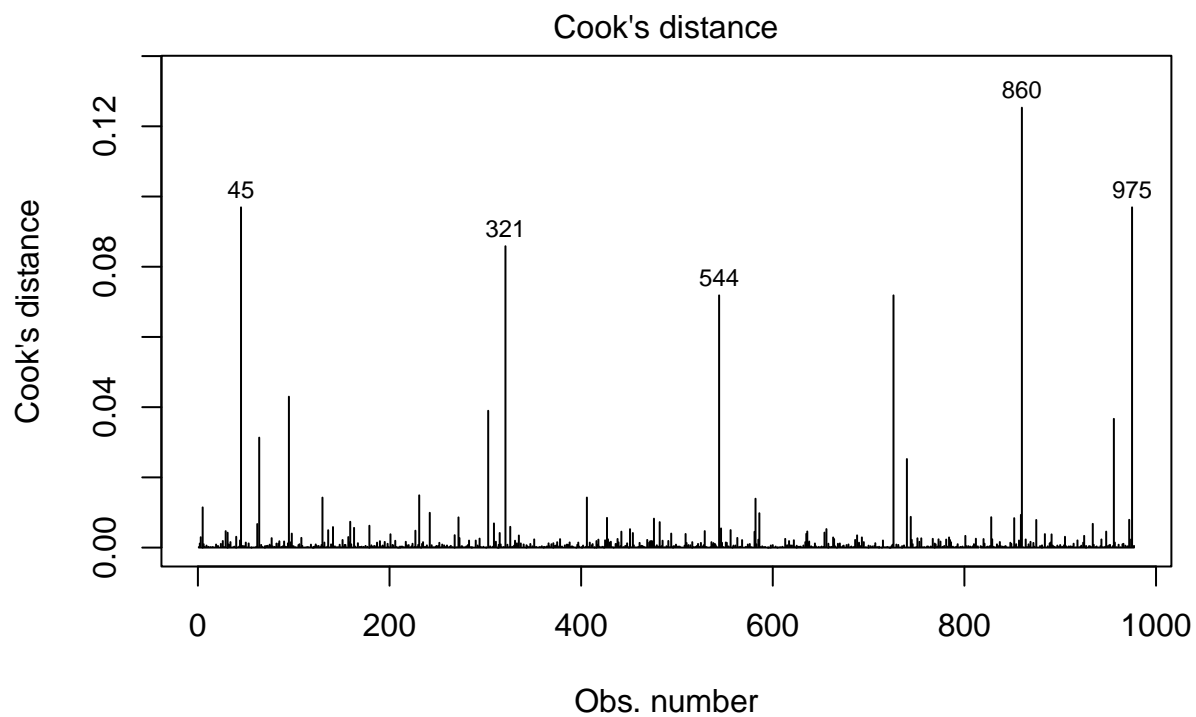


```
# partial residuals  
crp(airbnb_lm_transformed, id = list(n = 4))
```



Component + Residual Plots

```
# check outliers
plot(airbnb_lm_transformed, which = 4, id.n = 5)
```

$\text{lm}(\log(\text{realSum}) \sim \text{room_type} + \text{person_capacity} + \text{host_is_superhost} + \text{multi} + \dots)$

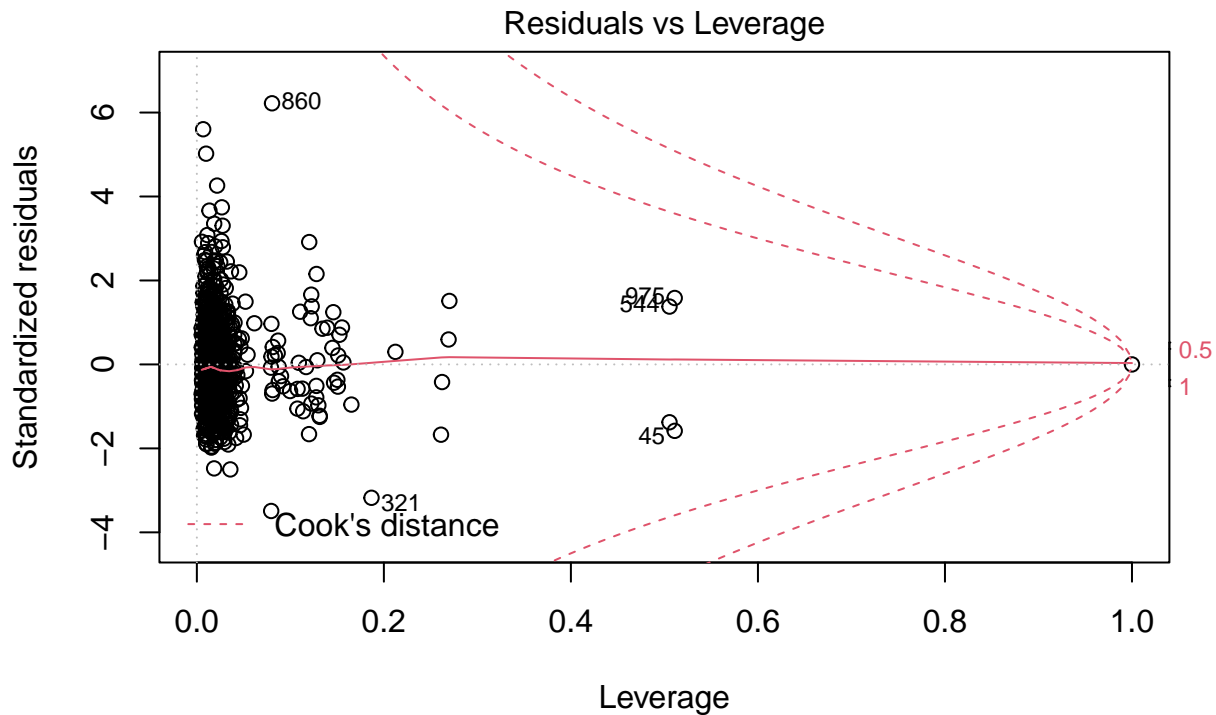
```
plot(airbnb_lm_transformed, which = 5, id.n = 5)
```

```
## Warning: not plotting observations with leverage one:
```

```
## 816
```

```
## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced
```

```
## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced
```



$\text{lm}(\log(\text{realSum}) \sim \text{room_type} + \text{person_capacity} + \text{host_is_superhost} + \text{multi} + \dots$

```
# refit model without potential influential cases
airbnb_lm_transformed_no_inf <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+
# bedrooms5 is not significant anymore, only has 2 samples
summary(airbnb_lm_transformed_no_inf)
```

```
##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(dist) + sqrt(metro_dist) + log(attr_index) +
##     log(rest_index), data = airbnb, subset = -c(975))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96477 -0.18614 -0.02309  0.14060  1.72310
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.825770   0.483231   9.986  < 2e-16 ***
## room_typePrivate room    -0.323184   0.022227  -14.540  < 2e-16 ***
## room_typeShared room    -0.617570   0.148256   -4.166 3.39e-05 ***
## person_capacity3      0.193258   0.041152   4.696 3.04e-06 ***
## person_capacity4      0.326863   0.031084  10.516  < 2e-16 ***
## person_capacity5      0.541535   0.097197   5.572 3.29e-08 ***
## person_capacity6      0.770055   0.085173   9.041  < 2e-16 ***
## host_is_superhostTrue  -0.001519   0.022282  -0.068 0.945661
## multi1             -0.016696   0.023078  -0.723 0.469575
## biz1               0.010443   0.034173   0.306 0.759988
## cleanliness_rating4    0.249953   0.357949   0.698 0.485166
## cleanliness_rating5   -0.465473   0.415685  -1.120 0.263094
```

```

## cleanliness_rating6          -0.418250    0.312584   -1.338  0.181203
## cleanliness_rating7          -0.375817    0.309774   -1.213  0.225356
## cleanliness_rating8          -0.326964    0.295302   -1.107  0.268480
## cleanliness_rating9          -0.312138    0.294222   -1.061  0.289009
## cleanliness_rating10         -0.297743    0.295384   -1.008  0.313718
## log(101 - guest_satisfaction_overall) -0.040749    0.012297   -3.314  0.000956 ***
## bedrooms1                    0.045765    0.039277    1.165  0.244235
## bedrooms2                    0.222333    0.049297    4.510  7.29e-06 ***
## bedrooms3                    0.451472    0.062301    7.247  8.84e-13 ***
## bedrooms4                    0.757066    0.107834    7.021  4.20e-12 ***
## bedrooms5                    0.395990    0.294705    1.344  0.179372
## sqrt(dist)                   0.013939    0.057030    0.244  0.806961
## sqrt(metro_dist)             0.006623    0.029688    0.223  0.823514
## log(attr_index)              0.094123    0.065650    1.434  0.151985
## log(rest_index)              0.205588    0.078044    2.634  0.008569 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2888 on 949 degrees of freedom
## Multiple R-squared:  0.6987, Adjusted R-squared:  0.6905
## F-statistic: 84.65 on 26 and 949 DF,  p-value: < 2.2e-16

airbnb_lm_transformed_no_inf2 <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+
# bedrooms5 is gone
summary(airbnb_lm_transformed_no_inf2)

##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(dist) + sqrt(metro_dist) + log(attr_index) +
##     log(rest_index), data = airbnb, subset = -c(975, 45))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96477 -0.18625 -0.02339  0.14064  1.72310
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.825770   0.483231   9.986 < 2e-16 ***
## room_typePrivate room    -0.323184   0.022227  -14.540 < 2e-16 ***
## room_typeShared room    -0.617570   0.148256   -4.166 3.39e-05 ***
## person_capacity3         0.193258   0.041152    4.696 3.04e-06 ***
## person_capacity4         0.326863   0.031084   10.516 < 2e-16 ***
## person_capacity5         0.541535   0.097197    5.572 3.29e-08 ***
## person_capacity6         0.770055   0.085173    9.041 < 2e-16 ***
## host_is_superhostTrue   -0.001519   0.022282   -0.068 0.945661
## multi1             -0.016696   0.023078   -0.723 0.469575
## biz1                0.010443   0.034173    0.306 0.759988
## cleanliness_rating4      0.249953   0.357949    0.698 0.485166
## cleanliness_rating5     -0.465473   0.415685   -1.120 0.263094
## cleanliness_rating6     -0.418250   0.312584   -1.338 0.181203
## cleanliness_rating7     -0.375817   0.309774   -1.213 0.225356
## cleanliness_rating8     -0.326964   0.295302   -1.107 0.268480
## cleanliness_rating9     -0.312138   0.294222   -1.061 0.289009

```

```

## cleanliness_rating10          -0.297743    0.295384   -1.008  0.313718
## log(101 - guest_satisfaction_overall) -0.040749    0.012297   -3.314  0.000956 ***
## bedrooms1                     0.045765    0.039277    1.165  0.244235
## bedrooms2                     0.222333    0.049297    4.510  7.29e-06 ***
## bedrooms3                     0.451472    0.062301    7.247  8.84e-13 ***
## bedrooms4                     0.757066    0.107834    7.021  4.20e-12 ***
## sqrt(dist)                    0.013939    0.057030    0.244  0.806961
## sqrt(metro_dist)              0.006623    0.029688    0.223  0.823514
## log(attr_index)               0.094123    0.065650    1.434  0.151985
## log(rest_index)               0.205588    0.078044    2.634  0.008569 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2888 on 949 degrees of freedom
## Multiple R-squared:  0.6985, Adjusted R-squared:  0.6905
## F-statistic: 87.94 on 25 and 949 DF,  p-value: < 2.2e-16

airbnb_lm_transformed_no_inf3 <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz1+
# bedrooms5 is gone, no significant change
summary(airbnb_lm_transformed_no_inf3)

##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz1 + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(dist) + sqrt(metro_dist) + log(attr_index) +
##     log(rest_index), data = airbnb, subset = -c(975, 45, 544,
##     726))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96162 -0.18657 -0.02314  0.14045  1.72403
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.801261   0.483313   9.934 < 2e-16 ***
## room_typePrivate room -0.323445   0.022217 -14.558 < 2e-16 ***
## room_typeShared room -0.617442   0.148183  -4.167 3.37e-05 ***
## person_capacity3    0.193065   0.041132   4.694 3.08e-06 ***
## person_capacity4    0.326619   0.031069  10.513 < 2e-16 ***
## person_capacity5    0.540814   0.097150   5.567 3.38e-08 ***
## person_capacity6    0.768837   0.085135   9.031 < 2e-16 ***
## host_is_superhostTrue -0.001383   0.022272  -0.062 0.950500
## multi1           -0.016080   0.023071  -0.697 0.485986
## biz1             0.014562   0.034284   0.425 0.671114
## cleanliness_rating5 -0.464509   0.415480  -1.118 0.263849
## cleanliness_rating6 -0.419596   0.312431  -1.343 0.179591
## cleanliness_rating7 -0.376327   0.309621  -1.215 0.224499
## cleanliness_rating8 -0.328626   0.295159  -1.113 0.265825
## cleanliness_rating9 -0.314079   0.294080  -1.068 0.285790
## cleanliness_rating10 -0.300169   0.295243  -1.017 0.309563
## log(101 - guest_satisfaction_overall) -0.041387   0.012300  -3.365 0.000797 ***
## bedrooms1         0.046476   0.039261   1.184 0.236801
## bedrooms2         0.223182   0.049277   4.529 6.68e-06 ***
## bedrooms3         0.452142   0.062272   7.261 8.01e-13 ***

```

```
## bedrooms4          0.757246    0.107780    7.026 4.06e-12 ***
## sqrt(dist)         0.017512    0.057059    0.307 0.758986
## sqrt(metro_dist)   0.007753    0.029684    0.261 0.794006
## log(attr_index)    0.095472    0.065625    1.455 0.146050
## log(rest_index)    0.207846    0.078022    2.664 0.007854 **
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 0.2886 on 948 degrees of freedom
```

```
## Multiple R-squared:  0.6991, Adjusted R-squared:  0.6914
```

```
## F-statistic: 91.75 on 24 and 948 DF,  p-value: < 2.2e-16
```

```
# check collinearity
```

```
vif(airbnb_lm_transformed_no_inf2)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## room_type          1.428492  2      1.093250
## person_capacity    2.825552  4      1.138644
## host_is_superhost  1.157108  1      1.075690
## multi              1.184242  1      1.088229
## biz                1.155490  1      1.074937
## cleanliness_rating 1.789273  7      1.042433
## log(101 - guest_satisfaction_overall) 1.576988  1      1.255782
## bedrooms           2.965420  4      1.145541
## sqrt(dist)         12.362496  1      3.516034
## sqrt(metro_dist)   1.449851  1      1.204098
## log(attr_index)    23.828914  1      4.881487
## log(rest_index)    32.714445  1      5.719654
```

```
# transformed model - removing dist
```

```
airbnb_lm_transformed_1 <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+cleanliness_rating+log(101 - guest_satisfaction_overall)+bedrooms+sqrt(metro_dist)+log(attr_index)+log(rest_index),
summary(airbnb_lm_transformed_1)
```

```
##
```

```
## Call:
```

```
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(attr_index) + log(rest_index),
##      data = airbnb, subset = -c(975, 45))
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -0.96597 -0.18627 -0.02322  0.13955  1.72091
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.913499   0.323386  15.194 < 2e-16 ***
## room_typePrivate room -0.322826   0.022168 -14.563 < 2e-16 ***
## room_typeShared room -0.615948   0.148034  -4.161 3.46e-05 ***
## person_capacity3    0.193467   0.041122   4.705 2.92e-06 ***
## person_capacity4    0.326752   0.031065  10.518 < 2e-16 ***
## person_capacity5    0.540862   0.097109   5.570 3.32e-08 ***
## person_capacity6    0.769682   0.085117   9.043 < 2e-16 ***
## host_is_superhostTrue -0.001447   0.022269  -0.065 0.948219
## multi1            -0.016646   0.023066  -0.722 0.470669
## biz1              0.009801   0.034056   0.288 0.773556
```

```
## cleanliness_rating4          0.251165    0.357737    0.702 0.482792
## cleanliness_rating5         -0.467376    0.415407   -1.125 0.260829
## cleanliness_rating6         -0.416314    0.312329   -1.333 0.182872
## cleanliness_rating7         -0.375736    0.309620   -1.214 0.225226
## cleanliness_rating8         -0.326106    0.295135   -1.105 0.269467
## cleanliness_rating9         -0.310796    0.294025   -1.057 0.290763
## cleanliness_rating10        -0.296547    0.295197   -1.005 0.315358
## log(101 - guest_satisfaction_overall) -0.040662    0.012286   -3.310 0.000969 ***
## bedrooms1                   0.045354    0.039221    1.156 0.247828
## bedrooms2                   0.222247    0.049272    4.511 7.27e-06 ***
## bedrooms3                   0.451765    0.062258    7.256 8.26e-13 ***
## bedrooms4                   0.757603    0.107758    7.031 3.93e-12 ***
## sqrt(metro_dist)            0.004537    0.028421    0.160 0.873199
## log(attr_index)              0.092578    0.065313    1.417 0.156678
## log(rest_index)              0.195448    0.066070    2.958 0.003171 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2886 on 950 degrees of freedom
## Multiple R-squared:  0.6985, Adjusted R-squared:  0.6908
## F-statistic: 91.69 on 24 and 950 DF,  p-value: < 2.2e-16
```

anova test

```
anova(airbnb_lm_transformed_1, airbnb_lm_transformed_no_inf2)
```

Analysis of Variance Table

```
##
## Model 1: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(attr_index) + log(rest_index)
## Model 2: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(dist) + sqrt(metro_dist) + log(attr_index) +
##      log(rest_index)
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      950 79.139
## 2      949 79.134  1 0.0049815 0.0597  0.807
```

collinearity between attr_index and rest_index is high after removing sqrt(dist)

```
vif(airbnb_lm_transformed_1)
```

```
##                               GVIF Df GVIF^(1/(2*Df))
## room_type                    1.420219  2      1.091664
## person_capacity               2.820576  4      1.138393
## host_is_superhost            1.156903  1      1.075594
## multi                        1.184149  1      1.088186
## biz                          1.148677  1      1.071764
## cleanliness_rating           1.775364  7      1.041853
## log(101 - guest_satisfaction_overall) 1.575671  1      1.255257
## bedrooms                    2.948508  4      1.144723
## sqrt(metro_dist)             1.330032  1      1.153270
## log(attr_index)              23.607847  1      4.858791
## log(rest_index)              23.469291  1      4.844511
```

transformed model - removing attr_index

```
airbnb_lm_transformed_2 <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+cleanliness_rating+log(101 - guest_satisfaction_overall)+sqrt(metro_dist)+log(rest_index))
```

```
summary(airbnb_lm_transformed_2)
```

```
##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(metro_dist) + log(rest_index), data = airbnb,
##     subset = -c(975, 45))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96708 -0.18494 -0.02415  0.14086  1.71727
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      4.917225   0.323547  15.198 < 2e-16 ***
## room_typePrivate room    -0.322895   0.022180 -14.558 < 2e-16 ***
## room_typeShared room    -0.611118   0.148073  -4.127 4.00e-05 ***
## person_capacity3         0.193625   0.041144   4.706 2.90e-06 ***
## person_capacity4         0.329172   0.031034  10.607 < 2e-16 ***
## person_capacity5         0.542202   0.097156   5.581 3.12e-08 ***
## person_capacity6         0.771936   0.085147   9.066 < 2e-16 ***
## host_is_superhostTrue    -0.000113   0.022261  -0.005 0.99595
## multi1                 -0.019445   0.022994  -0.846 0.39795
## biz1                   0.008167   0.034054   0.240 0.81051
## cleanliness_rating4      0.240282   0.357844   0.671 0.50208
## cleanliness_rating5     -0.481796   0.415502  -1.160 0.24652
## cleanliness_rating6     -0.427650   0.312392  -1.369 0.17134
## cleanliness_rating7     -0.387241   0.309678  -1.250 0.21144
## cleanliness_rating8     -0.338553   0.295161  -1.147 0.25167
## cleanliness_rating9     -0.321782   0.294079  -1.094 0.27414
## cleanliness_rating10    -0.309194   0.295219  -1.047 0.29521
## log(101 - guest_satisfaction_overall) -0.040611   0.012293  -3.304 0.00099 ***
## bedrooms1              0.045930   0.039240   1.170 0.24210
## bedrooms2              0.221624   0.049296   4.496 7.79e-06 ***
## bedrooms3              0.450972   0.062289   7.240 9.25e-13 ***
## bedrooms4              0.770713   0.107417   7.175 1.45e-12 ***
## sqrt(metro_dist)        0.001025   0.028327   0.036 0.97116
## log(rest_index)         0.286335   0.015941  17.962 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2888 on 951 degrees of freedom
## Multiple R-squared:  0.6978, Adjusted R-squared:  0.6905
## F-statistic: 95.49 on 23 and 951 DF,  p-value: < 2.2e-16
```

```
# anova test
```

```
anova(airbnb_lm_transformed_1, airbnb_lm_transformed_2)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Model 1: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##     multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##     bedrooms + sqrt(metro_dist) + log(attrib_index) + log(rest_index)
```

```

## Model 2: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(rest_index)
## Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      950 79.139
## 2      951 79.307 -1   -0.16737 2.0092 0.1567

# transformed model - removing rest_index
airbnb_lm_transformed_3 <- lm(log(realSum) ~ room_type+person_capacity+host_is_superhost+multi+biz+cleanliness_rating+log(101 - guest_satisfaction_overall)+sqrt(metro_dist)+log(attr_index), data = airbnb, subset = -c(975, 45))
summary(airbnb_lm_transformed_3)

##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(attr_index), data = airbnb,
##      subset = -c(975, 45))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96985 -0.18250 -0.02421  0.14363  1.72845
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      4.991055    0.323632   15.422 < 2e-16 ***
## room_typePrivate room      -0.324203    0.022253  -14.569 < 2e-16 ***
## room_typeShared room     -0.633116    0.148522   -4.263 2.22e-05 ***
## person_capacity3         0.191190    0.041282    4.631 4.14e-06 ***
## person_capacity4         0.322806    0.031163   10.359 < 2e-16 ***
## person_capacity5         0.543614    0.097500    5.576 3.21e-08 ***
## person_capacity6         0.765685    0.085452    8.960 < 2e-16 ***
## host_is_superhostTrue    -0.004363    0.022338   -0.195 0.845195
## multi1                -0.010307    0.023060   -0.447 0.654986
## biz1                   0.015493    0.034140    0.454 0.650062
## cleanliness_rating4      0.260581    0.359178    0.725 0.468329
## cleanliness_rating5     -0.459788    0.417088   -1.102 0.270577
## cleanliness_rating6     -0.404090    0.313571   -1.289 0.197826
## cleanliness_rating7     -0.361713    0.310843   -1.164 0.244856
## cleanliness_rating8     -0.310411    0.296288   -1.048 0.295058
## cleanliness_rating9     -0.297257    0.295185   -1.007 0.314182
## cleanliness_rating10    -0.279799    0.296343   -0.944 0.345322
## log(101 - guest_satisfaction_overall) -0.040923    0.012336  -3.317 0.000943 ***
## bedrooms1              0.041021    0.039353    1.042 0.297507
## bedrooms2              0.219498    0.049463    4.438 1.02e-05 ***
## bedrooms3              0.447713    0.062496    7.164 1.57e-12 ***
## bedrooms4              0.730524    0.107805    6.776 2.16e-11 ***
## sqrt(metro_dist)       0.002186    0.028525    0.077 0.938935
## log(attr_index)        0.280083    0.015814   17.711 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2898 on 951 degrees of freedom
## Multiple R-squared:  0.6957, Adjusted R-squared:  0.6883
## F-statistic: 94.53 on 23 and 951 DF,  p-value: < 2.2e-16

```



```
# anova test
anova(airbnb_lm_transformed_1, airbnb_lm_transformed_3)

## Analysis of Variance Table
##
## Model 1: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(attr_index) + log(rest_index)
## Model 2: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##      multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##      bedrooms + sqrt(metro_dist) + log(attr_index)
##   Res.Df    RSS Df Sum of Sq    F   Pr(>F)
## 1      950 79.139
## 2      951 79.868 -1   -0.72899 8.7509 0.003171 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# no significant collinearity
vif(airbnb_lm_transformed_2)
```

	GVIF	Df	GVIF ^{1/(2*Df)}
room_type	1.419437	2	1.091513
person_capacity	2.811435	4	1.137931
host_is_superhost	1.154838	1	1.074634
multi	1.175472	1	1.084192
biz	1.147361	1	1.071149
cleanliness_rating	1.769499	7	1.041606
log(101 - guest_satisfaction_overall)	1.575657	1	1.255252
bedrooms	2.919040	4	1.143286
sqrt(metro_dist)	1.319920	1	1.148878
log(rest_index)	1.364777	1	1.168236

```
# anova test
```

```
airbnb_lm_transformed_reduced <- lm(log(realSum) ~ room_type+person_capacity+log(101-guest_satisfaction_overall)+bedrooms+log(rest_index),
data = airbnb, subset = -c(975, 45))
summary(airbnb_lm_transformed_reduced)
```

```
##
## Call:
## lm(formula = log(realSum) ~ room_type + person_capacity + log(101 -
##      guest_satisfaction_overall) + bedrooms + log(rest_index),
##      data = airbnb, subset = -c(975, 45))
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.95869	-0.18670	-0.02772	0.13846	1.69852

```
##
## Coefficients:
```

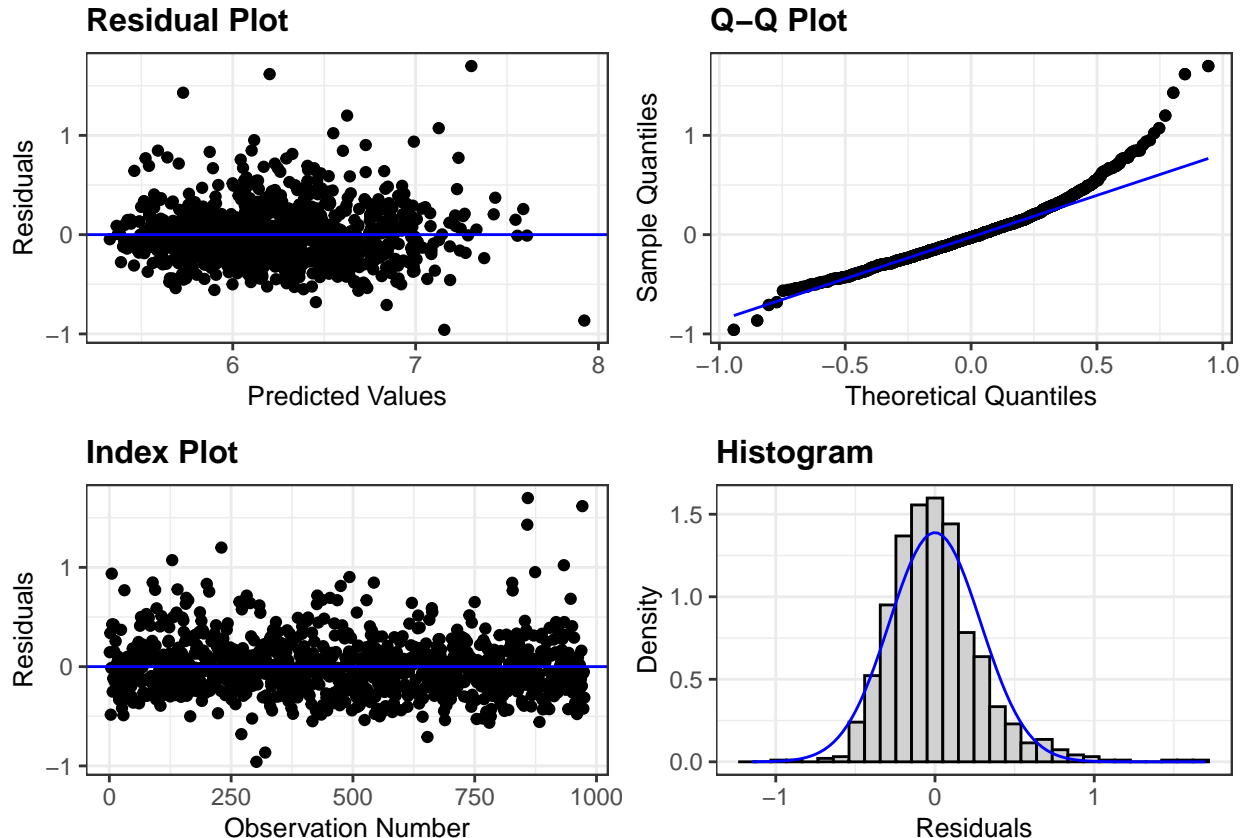
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	4.59366	0.09305	49.366	< 2e-16 ***
room_typePrivate room	-0.32412	0.02076	-15.610	< 2e-16 ***
room_typeShared room	-0.62969	0.14651	-4.298	1.90e-05 ***
person_capacity3	0.19913	0.04074	4.888	1.19e-06 ***
person_capacity4	0.32786	0.03061	10.709	< 2e-16 ***
person_capacity5	0.53585	0.09648	5.554	3.61e-08 ***
person_capacity6	0.77503	0.08428	9.196	< 2e-16 ***

```
## log(101 - guest_satisfaction_overall) -0.04402    0.01008   -4.366 1.40e-05 ***
## bedrooms1    0.05141    0.03868    1.329    0.184
## bedrooms2    0.22553    0.04832    4.667 3.49e-06 ***
## bedrooms3    0.45785    0.06163    7.429 2.41e-13 ***
## bedrooms4    0.76923    0.10681    7.202 1.20e-12 ***
## log(rest_index) 0.28734    0.01397   20.563 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.289 on 962 degrees of freedom
## Multiple R-squared:  0.6939, Adjusted R-squared:  0.69
## F-statistic: 181.7 on 12 and 962 DF,  p-value: < 2.2e-16

anova(airbnb_lm_transformed_reduced, airbnb_lm_transformed_2)

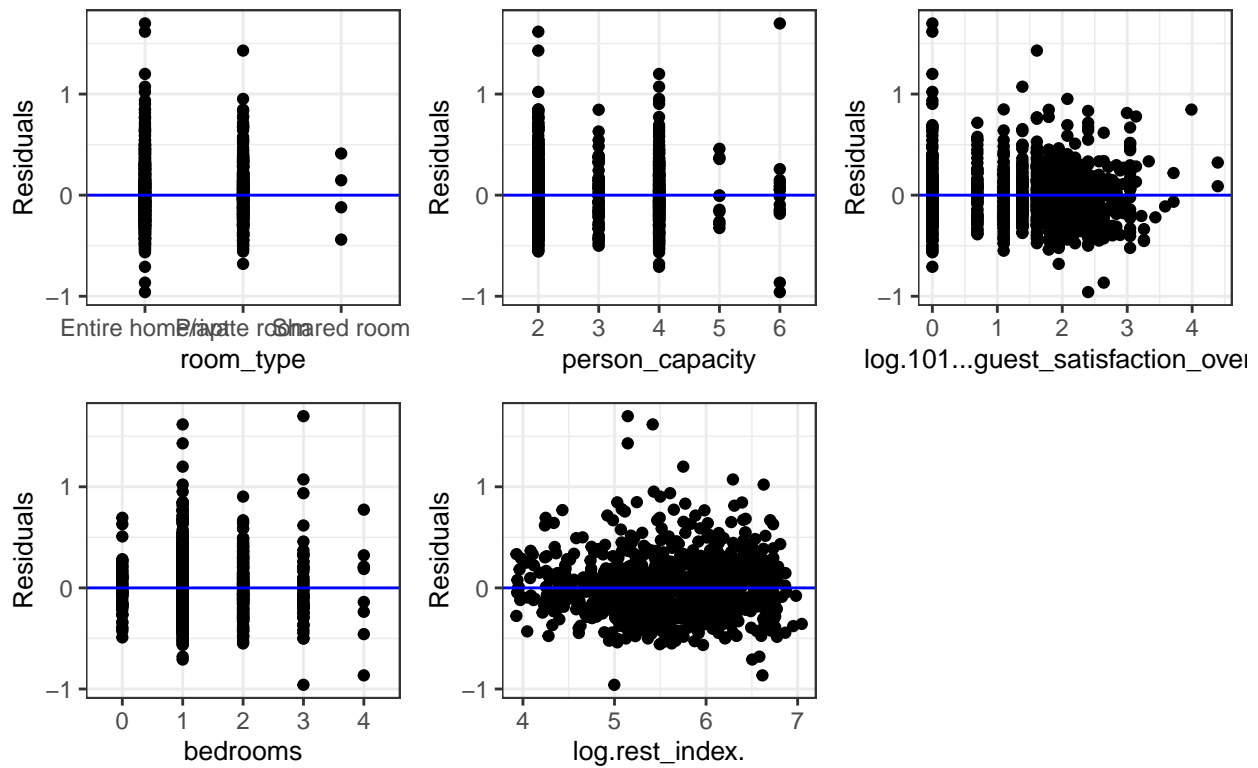
## Analysis of Variance Table
##
## Model 1: log(realSum) ~ room_type + person_capacity + log(101 - guest_satisfaction_overall) +
##   bedrooms + log(rest_index)
## Model 2: log(realSum) ~ room_type + person_capacity + host_is_superhost +
##   multi + biz + cleanliness_rating + log(101 - guest_satisfaction_overall) +
##   bedrooms + sqrt(metro_dist) + log(rest_index)
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      962 80.346
## 2      951 79.307 11     1.0392 1.1329 0.3315

# check assumptions
resid_panel(airbnb_lm_transformed_reduced)
```

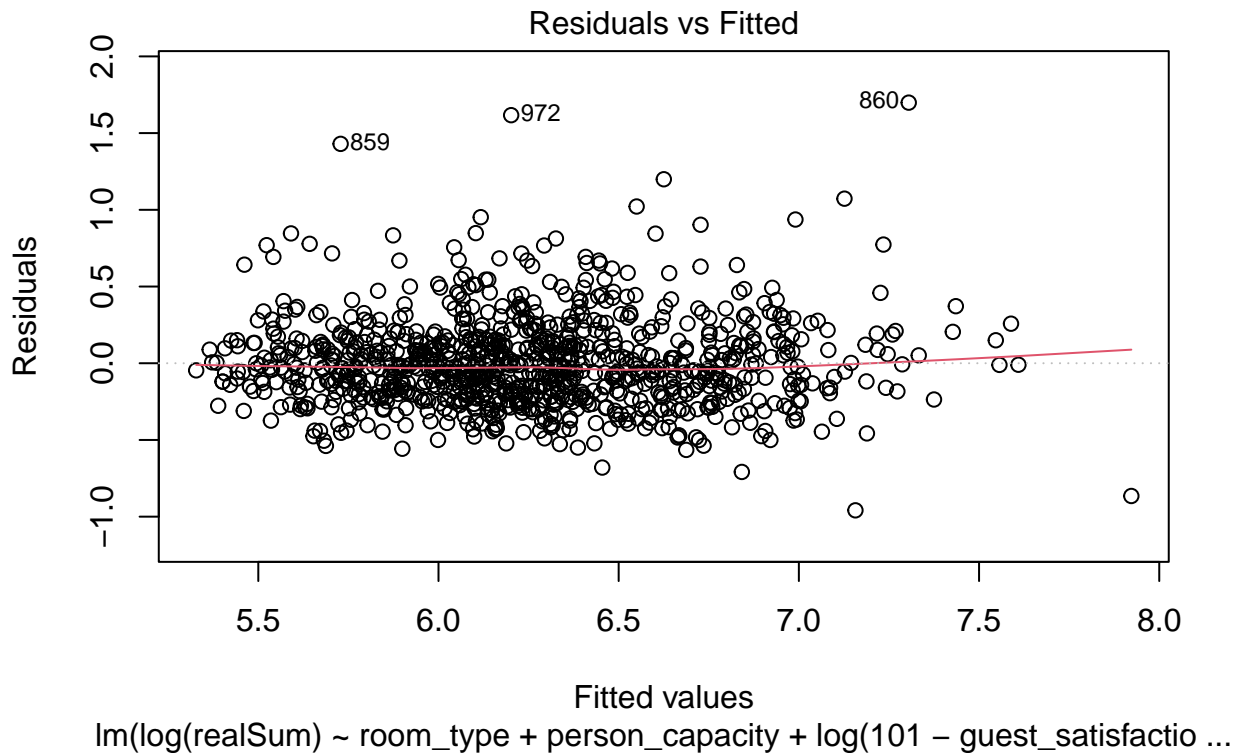


```
resid_xpanel(airbnb_lm_transformed_reduced)
```

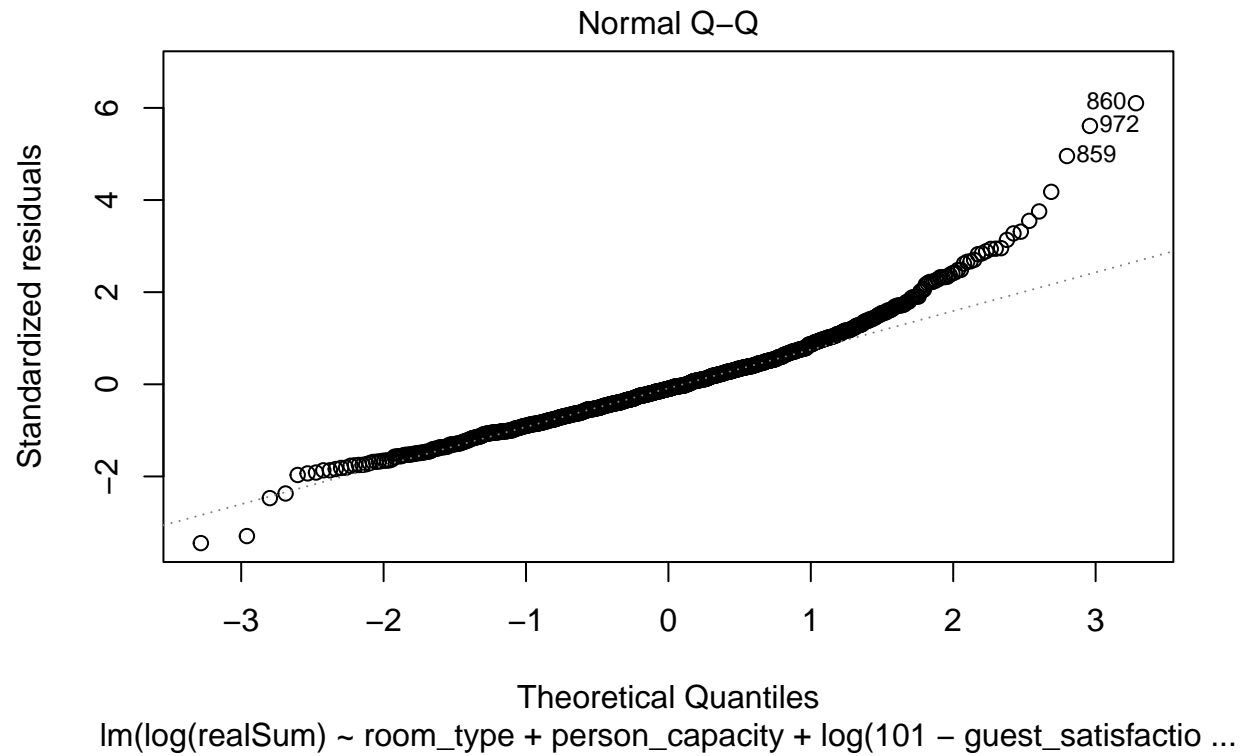
Plots of Residuals vs Predictor Variables



```
plot(airbnb_lm_transformed_reduced, which = 1)
```

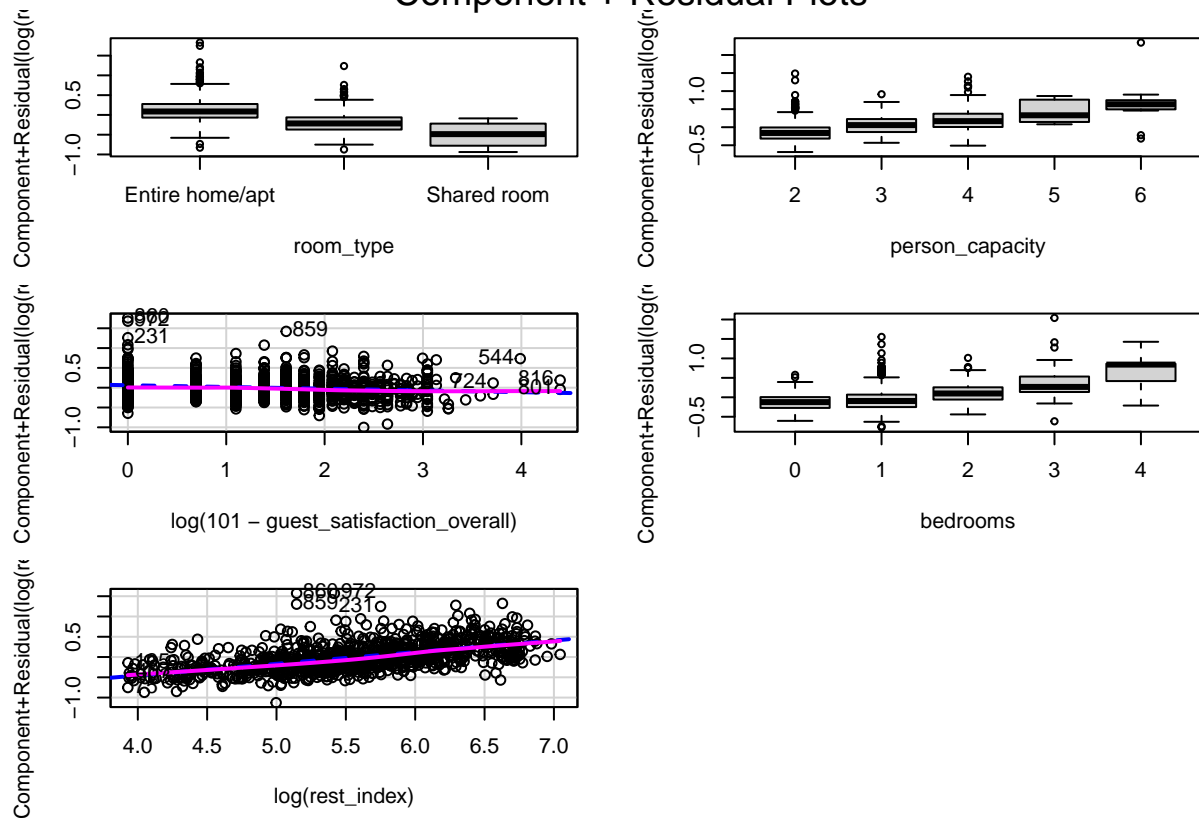


```
plot(airbnb_lm_transformed_reduced, which = 2)
```



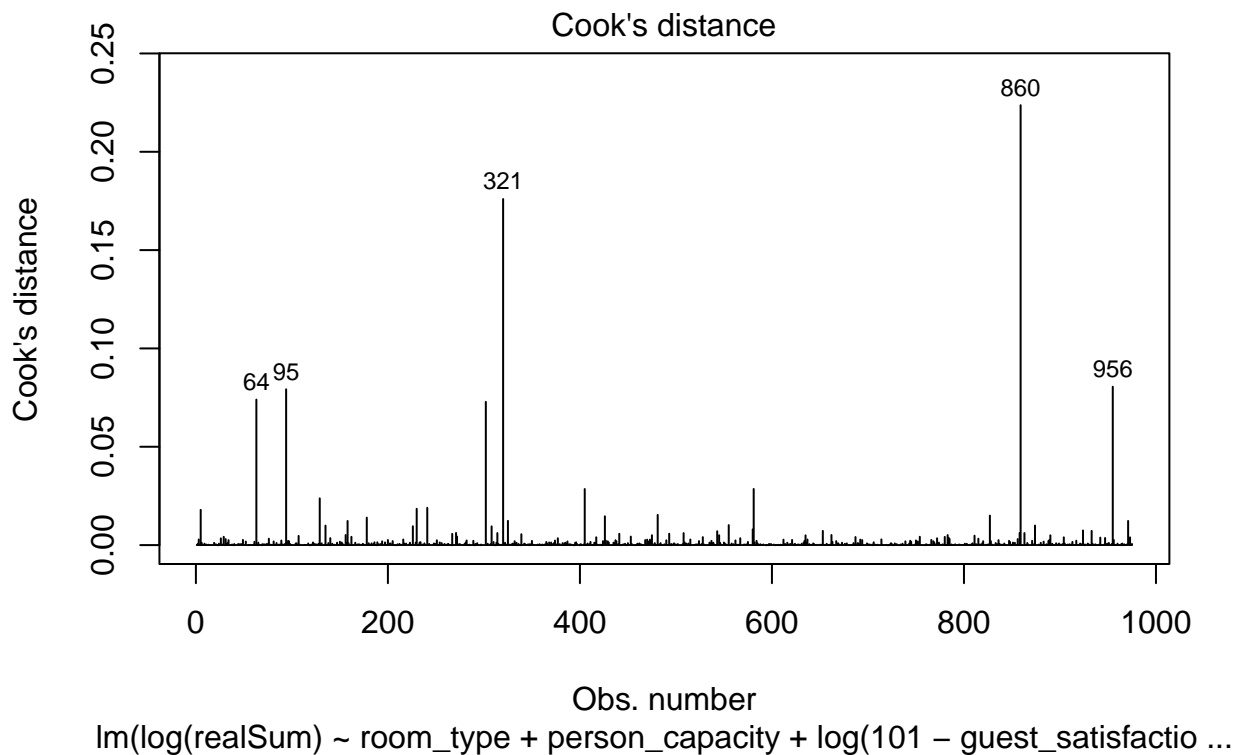
```
# partial residuals  
crp(airbnb_lm_transformed_reduced, id = list(n = 4))
```

Component + Residual Plots

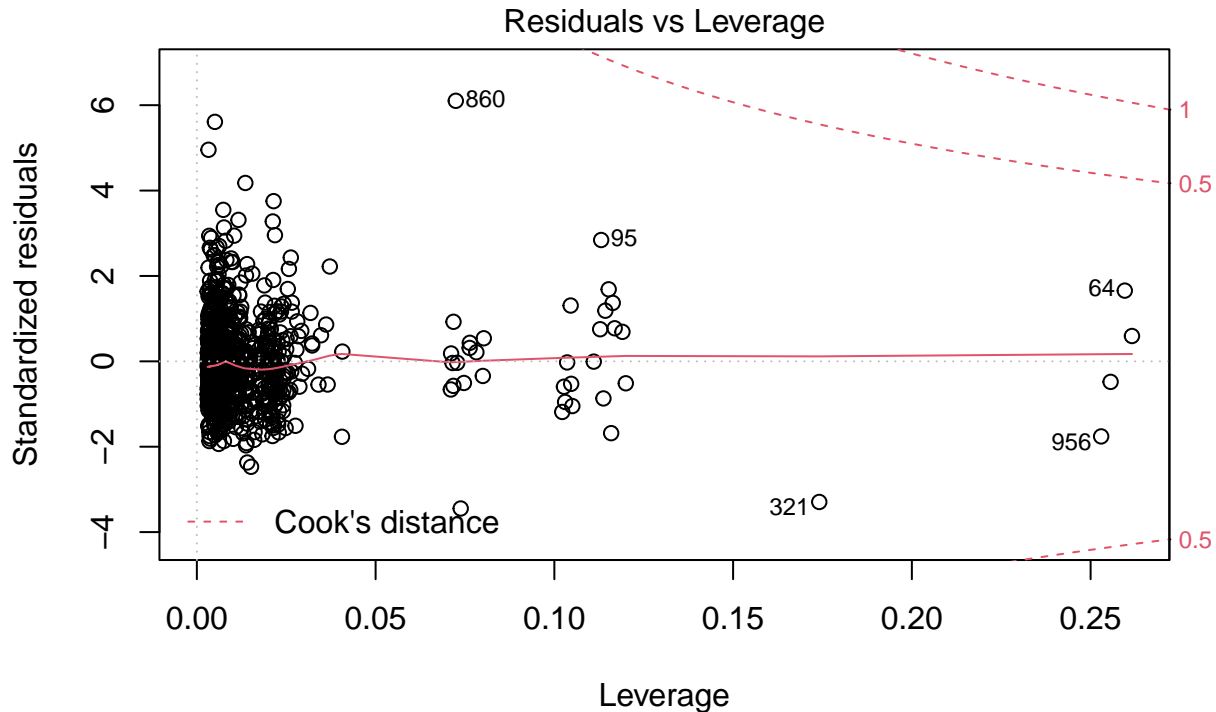


check outliers

```
plot(airbnb_lm_transformed_reduced, which = 4, id.n = 5)
```



```
plot(airbnb_lm_transformed_reduced, which = 5, id.n = 5)
```



lm(log(realSum) ~ room_type + person_capacity + log(101 - guest_satisfaction ...

```
# refit model without potential influential cases
```

```
airbnb_lm_transformed_reduced_no_inf <- lm(log(realSum) ~ room_type+person_capacity+log(101-guest_satisfaction_overall) + bedrooms + log(rest_index),
```

```
# no change in significance
```

```
summary(airbnb_lm_transformed_reduced_no_inf)
```

```
##
```

```
## Call:
```

```
## lm(formula = log(realSum) ~ room_type + person_capacity + log(101 -
```

```
##   guest_satisfaction_overall) + bedrooms + log(rest_index),
```

```
##   data = airbnb, subset = -c(975, 45, 321, 956, 860, 95, 64))
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -0.88852 -0.18304 -0.02872  0.13789  1.62212
```

```
##
```

```
## Coefficients:
```

```
##
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)      4.576224   0.090793  50.403 < 2e-16 ***
```

```
## room_typePrivate room      -0.324917   0.020222 -16.068 < 2e-16 ***
```

```
## room_typeShared room     -0.617941   0.201574  -3.066  0.00223 **
```

```
## person_capacity3         0.201259   0.039726   5.066 4.87e-07 ***
```

```
## person_capacity4         0.331446   0.029881  11.092 < 2e-16 ***
```

```
## person_capacity5         0.542840   0.093902   5.781 1.00e-08 ***
```

```
## person_capacity6         0.728302   0.086770   8.393 < 2e-16 ***
```

```
## log(101 - guest_satisfaction_overall) -0.040699   0.009832  -4.139 3.79e-05 ***
```

```
## bedrooms1              0.053011   0.037632   1.409  0.15926
```

```
## bedrooms2              0.225328   0.047058   4.788 1.95e-06 ***
```

```
## bedrooms3              0.434056   0.060092   7.223 1.04e-12 ***
```

```
## bedrooms4                0.781077    0.115223    6.779 2.12e-11 ***
## log(rest_index)          0.289307    0.013647   21.199 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2812 on 957 degrees of freedom
## Multiple R-squared:  0.6974, Adjusted R-squared:  0.6936
## F-statistic: 183.8 on 12 and 957 DF,  p-value: < 2.2e-16
```

```
# check collinearity
vif(airbnb_lm_transformed_reduced)
```

```
##                                GVIF Df GVIF^(1/(2*Df))
## room_type                     1.220335  2      1.051041
## person_capacity               2.584994  4      1.126049
## log(101 - guest_satisfaction_overall) 1.058207  1      1.028692
## bedrooms                     2.722158  4      1.133350
## log(rest_index)              1.047144  1      1.023300
```

```
# interpretation stats
(exp(-0.324) - 1)*100
```

```
## [1] -27.67498
```

```
(exp(-0.630) - 1)*100
```

```
## [1] -46.74082
```

```
(exp(0.199) - 1)*100
```

```
## [1] 22.0182
```

```
(exp(0.328) - 1)*100
```

```
## [1] 38.8189
```

```
(exp(0.536) - 1)*100
```

```
## [1] 70.91565
```

```
(exp(0.775) - 1)*100
```

```
## [1] 117.0592
```

```
(2^-0.044 - 1)*100
```

```
## [1] -3.003809
```

```
(exp(0.226) - 1)*100
```

```
## [1] 25.35757
```

```
(exp(0.458) - 1)*100
```

```
## [1] 58.0909
```

```
(exp(0.769) - 1)*100
```

```
## [1] 115.7608
```

```
(2^0.287 - 1)*100
```

```
## [1] 22.01005
```

```

# confidence interval
(exp(-0.365) - 1)*100

## [1] -30.58033
(exp(-0.283) - 1)*100

## [1] -24.64802
(exp(-0.917) - 1)*100

## [1] -60.02836
(exp(-0.342) - 1)*100

## [1] -28.96518
(exp(0.119) - 1)*100

## [1] 12.63699
(exp(0.279) - 1)*100

## [1] 32.18073
(exp(0.268) - 1)*100

## [1] 30.73471
(exp(0.388) - 1)*100

## [1] 47.40298
(exp(0.347) - 1)*100

## [1] 41.48167
(exp(0.725) - 1)*100

## [1] 106.4731
(exp(0.610) - 1)*100

## [1] 84.04314
(exp(0.940) - 1)*100

## [1] 155.9981
(2^-0.064 - 1)*100

## [1] -4.339184
(2^-0.024 - 1)*100

## [1] -1.649793
(exp(0.131) - 1)*100

## [1] 13.99678
(exp(0.320) - 1)*100

## [1] 37.71278

```



```
(exp(0.337) - 1)*100
```

```
## [1] 40.07391
```

```
(exp(0.579) - 1)*100
```

```
## [1] 78.42533
```

```
(exp(0.560) - 1)*100
```

```
## [1] 75.06725
```

```
(exp(0.979) - 1)*100
```

```
## [1] 166.1793
```

```
(2^0.260 - 1)*100
```

```
## [1] 19.74787
```

```
(2^0.315 - 1)*100
```

```
## [1] 24.40117
```