

Proyecto Final. Precios Airbnb

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Introducción

Como es bien sabido, en los últimos años, los costos de alquiler y arrendamiento de viviendas en la Ciudad de México se han visto afectados de manera considerable, presentando una tendencia al alza que es mayormente visible en plataformas como Airbnb. Gracias a su gran volumen de propiedades en arrendamiento, Airbnb podría llegar a influir en los costos de viviendas circundantes a las registradas en la plataforma, lo que resultaría de gran interés tanto para arrendatarios como arrendadores.

Objetivo

Este trabajo tiene por objetivo ajustar un modelo de regresión lineal a una base de datos de Airbnb's de la Ciudad de México para mostrarle a la población general como es que el precio de uno de estos arrendamientos podrá variar en un futuro. Sin embargo, ya que muchas personas no están familiarizadas con lenguajes de programación, se decidió crear una calculadora interactiva que le permita a un usuario estimar el precio de un Airbnb en el futuro. Así pues, procedamos con el desarrollo del trabajo.

1. Las variables

Analizando la base de datos original en Excel es posible notar que existen múltiples variables que resultan irrelevantes para nuestro modelo, pues varias de ellas no nos aportan valores medibles que podamos incluir en el mismo. Debido a esto, se descartaron un total de 44 variables, cuyos motivos fueron los siguientes:

```
#Librerías utilizadas  
library(corrplot)
```

Warning: package 'corrplot' was built under R version 4.3.3

corrplot 0.92 loaded

```
library(MASS)  
library(dplyr)
```

Attaching package: 'dplyr'

The following object is masked from 'package:MASS':

select

The following objects are masked from 'package:stats':

filter, lag

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

```
library(readxl)  
library(nortest)  
library(ggplot2)
```

Warning: package 'ggplot2' was built under R version 4.3.2

```
library(car)
```

Loading required package: carData

Attaching package: 'car'

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

recode

```
library(DescTools)
```

Warning: package 'DescTools' was built under R version 4.3.3

Attaching package: 'DescTools'

The following object is masked from 'package:car':

Recode

```
#Usa el comando file.choose() para abrir la ubicación del excel  
#Lectura del excel  
excel <- "C:\\Users\\andre\\OneDrive\\Documents\\2024-2\\Datos_modelos.xlsx"  
datos <- read_excel(excel, sheet = "Datos")  
attach(datos);
```

Las variables que en ella aparecen fueron elegidas por su relación con el precio, descartando así un total de 44 variables que resultaban irrelevantes para el modelo. Los motivos por los que se descartaron el resto de variables se muestran a continuación:

- Letras: Debido a que muchas de estas variables presentaban únicamente palabras, links y caracteres que no eran cuantificables ni ajustables al modelo.
- Porcentajes: Al igual que el punto anterior los porcentajes al tener caracteres fueron ignorados para el modelo por contener datos que no eran medibles en R.
- Repetición de datos: Debido a que las columnas del máximo y mínimo de noches aparecían 6 veces con los mismos datos

Aunado a esto y, dado que el porcentaje con respecto al total no era muy elevado, se eliminaron los valores NA que contenía la base de datos.

```
#Primera limpieza
datos <- na.omit(datos)
datos <- datos %>%
  dplyr::select(-id)
```

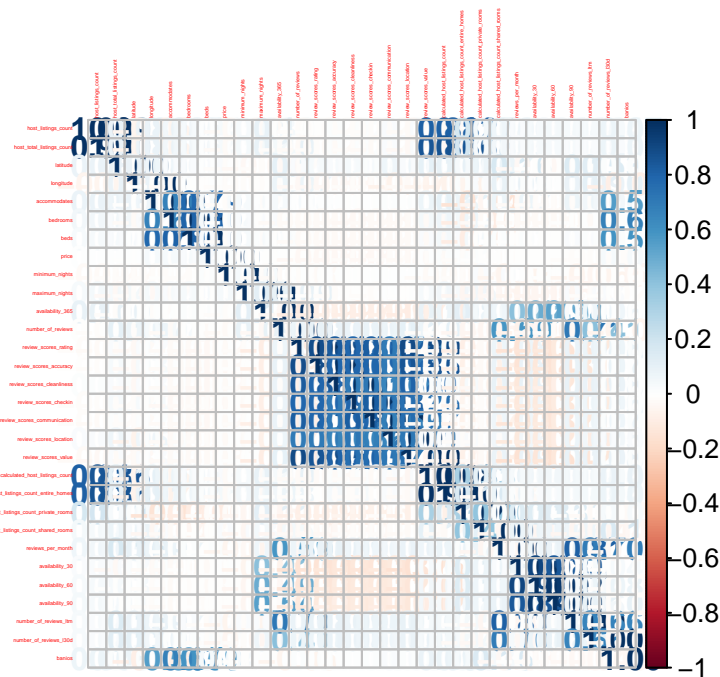
2. Análisis exploratorio de las variables

Para ser más precisos con los datos que estábamos trabajando se discutió sobre que variables podían llegar a estar más correlacionadas entre sí, siendo estas las siguientes:

- 1 y 2.- Ambas variables parametrizan datos muy similares
- 3 y 4.- Ambas tienen datos relativos a la ubicación del Airbnb
- 5, 6 y 7.- Hacen alusión a la cantidad de huéspedes que puede albergar el Airbnb
- 6 y 30.- En una mala tipificación de las habitaciones los baños podrían haber sido tomados como tales, o tal vez porque las habitaciones tengan un baño integrado
- 5, 6, 7 y 30.- Igual que el motivo anterior, una mala tipificación de los datos podría hacer que estas variables estén relacionadas
- 9, 10, 11, 25, 26 y 27.- Mientras mayor o menor sea el número de noches que alguien se hospede en el Airbnb menor o mayor será la cantidad de días que estuvo disponible
- 12 - 19.- Todas estas variables hacen alusión a las reviews del Airbnb
- 3, 4 y 18.- Las variables están relacionadas con la ubicación del Airbnb
- 6, 7, 15 y 30.- La limpieza del Airbnb es notoria en las recámaras, las camas y los baños
- 1, 2, 20, 21, 22, 23.- Son estadísticas del hospedador de cada Airbnb
- 11, 25, 26, 27.- Son las disponibilidades acumuladas en ciertos periodos de tiempo, siendo el más alto de un año, por lo que si descubrimos correlación podríamos quedarnos con el de periodicidad más grande
- 12, 24, 28, 29.- Hacen alusión al número de Reviews que el Airbnb ha tenido a lo largo de diferentes periodos de tiempo.

Para comprobar si nuestras suposiciones eran correctas se realizaron gráficos de correlación entre cada una de las variables anteriormente expuestas y, tras realizar 11 gráficos pudimos ver que nuestras suposiciones eran correctas, por lo que las variables reiterativas fueron eliminadas buscando cumplir con los supuestos del modelo.

```
#Gráfico de correlaciones
corrplot(cor(datos), method = "number", tl.cex = .2)
```



```
#Segunda limpieza después de ver correlaciones
attach(datos)
```

The following objects are masked from datos (pos = 3):

```
accommodates, availability_30, availability_365, availability_60,
availability_90, banios, bedrooms, beds,
calculated_host_listings_count,
calculated_host_listings_count_entire_homes,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms, host_listings_count,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d,
number_of_reviews_ltm, price, review_scores_accuracy,
review_scores_checkin, review_scores_cleanliness,
review_scores_communication, review_scores_location,
review_scores_rating, review_scores_value, reviews_per_month
```

```

datos <- datos %>%
  dplyr::select(-host_listings_count,
                -beds,
                -availability_30,
                -availability_60,
                -availability_90,
                -review_scores_rating,
                -review_scores_accuracy,
                -review_scores_communication,
                -review_scores_value,
                -calculated_host_listings_count,
                -calculated_host_listings_count_entire_homes,
                -number_of_reviews_ltm)

```

Nota: Para ver los 12 gráficos consultar el “Documento expandido”

3. Datos atípicos y el primer modelo

Teniendo ahora una base de datos con menos variables se procedió a analizar las escalas de cada una de ellas.

Habiendo eliminado las variables reiterativas se realizó el siguiente modelo:

```

#Modelo original
modelo1 <- lm(price~host_total_listings_count
               +latitude
               +longitude
               +accommodates
               +bedrooms
               +minimum_nights
               +maximum_nights
               +availability_365
               +number_of_reviews
               +review_scores_cleanliness
               +review_scores_checkin
               +review_scores_location
               +calculated_host_listings_count_private_rooms
               +calculated_host_listings_count_shared_rooms
               +reviews_per_month
               +number_of_reviews_l30d
               +banios)

```

```
summary(modelo1)
```

Call:

```
lm(formula = price ~ host_total_listings_count + latitude + longitude +  
    accommodates + bedrooms + minimum_nights + maximum_nights +  
    availability_365 + number_of_reviews + review_scores_cleanliness +  
    review_scores_checkin + review_scores_location + calculated_host_listings_count_private_rooms +  
    calculated_host_listings_count_shared_rooms + reviews_per_month +  
    number_of_reviews_l30d + banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-24848	-1079	-520	39	1830428

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-3.948e+05	4.161e+05	-0.949
host_total_listings_count	1.829e+00	1.583e+00	1.156
latitude	4.138e+03	3.291e+03	1.257
longitude	-3.238e+03	4.103e+03	-0.789
accommodates	1.412e+02	8.496e+01	1.662
bedrooms	1.116e+02	1.853e+02	0.602
minimum_nights	1.359e+01	5.796e+00	2.344
maximum_nights	-2.866e-01	2.879e-01	-0.995
availability_365	-5.307e-01	1.101e+00	-0.482
number_of_reviews	1.780e+00	2.504e+00	0.711
review_scores_cleanliness	-2.030e+03	4.412e+02	-4.601
review_scores_checkin	8.127e+02	5.332e+02	1.524
review_scores_location	-2.161e+01	5.402e+02	-0.040
calculated_host_listings_count_private_rooms	-4.979e+01	3.066e+01	-1.624
calculated_host_listings_count_shared_rooms	-5.730e+01	6.692e+01	-0.856
reviews_per_month	2.947e+01	1.112e+02	0.265
number_of_reviews_l30d	-1.548e+02	9.121e+01	-1.698
banios	5.143e+02	2.115e+02	2.431
Pr(> t)			
(Intercept)	0.3427		
host_total_listings_count	0.2479		
latitude	0.2087		
longitude	0.4299		
accommodates	0.0966		
bedrooms	0.5472		

```

minimum_nights          0.0191 *
maximum_nights          0.3195
availability_365        0.6297
number_of_reviews       0.4773
review_scores_cleanliness 4.23e-06 ***
review_scores_checkin    0.1275
review_scores_location   0.9681
calculated_host_listings_count_private_rooms 0.1044
calculated_host_listings_count_shared_rooms 0.3919
reviews_per_month        0.7910
number_of_reviews_l30d   0.0896 .
banios                  0.0151 *

```

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

Residual standard error: 18610 on 19957 degrees of freedom

Multiple R-squared: 0.004108, Adjusted R-squared: 0.003259

F-statistic: 4.842 on 17 and 19957 DF, p-value: 1.57e-10

```
seleccion <- stepAIC(modelo1, direction = c("both"));
```

Start: AIC=392780.2

```

price ~ host_total_listings_count + latitude + longitude + accommodates +
  bedrooms + minimum_nights + maximum_nights + availability_365 +
  number_of_reviews + review_scores_cleanliness + review_scores_checkin +
  review_scores_location + calculated_host_listings_count_private_rooms +
  calculated_host_listings_count_shared_rooms + reviews_per_month +
  number_of_reviews_l30d + banios

```

	Df	Sum of Sq	RSS	AIC
- review_scores_location	1	554359	6.9102e+12	392778
- reviews_per_month	1	24312502	6.9103e+12	392778
- availability_365	1	80513330	6.9103e+12	392778
- bedrooms	1	125483056	6.9104e+12	392779
- number_of_reviews	1	174837817	6.9104e+12	392779
- longitude	1	215749397	6.9104e+12	392779
- calculated_host_listings_count_shared_rooms	1	253824793	6.9105e+12	392779
- maximum_nights	1	343137576	6.9106e+12	392779
- host_total_listings_count	1	462404486	6.9107e+12	392780
- latitude	1	547216556	6.9108e+12	392780
<none>			6.9102e+12	392780

- review_scores_checkin	1	804517570	6.9110e+12	392781
- calculated_host_listings_count_private_rooms	1	913064952	6.9111e+12	392781
- accommodates	1	956094276	6.9112e+12	392781
- number_of_reviews_l30d	1	997833706	6.9112e+12	392781
- minimum_nights	1	1902107935	6.9121e+12	392784
- banios	1	2046714743	6.9123e+12	392784
- review_scores_cleanliness	1	7330167939	6.9176e+12	392799

Step: AIC=392778.2

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + minimum_nights + maximum_nights + availability_365 +
 number_of_reviews + review_scores_cleanliness + review_scores_checkin +
 calculated_host_listings_count_private_rooms + calculated_host_listings_count_shared_rooms +
 reviews_per_month + number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
- reviews_per_month	1	24411188	6.9103e+12	392776
- availability_365	1	80267601	6.9103e+12	392776
- bedrooms	1	126341845	6.9104e+12	392777
- number_of_reviews	1	174605356	6.9104e+12	392777
- longitude	1	215314274	6.9104e+12	392777
- calculated_host_listings_count_shared_rooms	1	253629258	6.9105e+12	392777
- maximum_nights	1	343687211	6.9106e+12	392777
- host_total_listings_count	1	461898349	6.9107e+12	392778
- latitude	1	546904937	6.9108e+12	392778
<none>			6.9102e+12	392778
- calculated_host_listings_count_private_rooms	1	916637488	6.9111e+12	392779
- accommodates	1	955563948	6.9112e+12	392779
- number_of_reviews_l30d	1	997840130	6.9112e+12	392779
- review_scores_checkin	1	1021629096	6.9113e+12	392779
+ review_scores_location	1	554359	6.9102e+12	392780
- minimum_nights	1	1902052074	6.9121e+12	392782
- banios	1	2046488717	6.9123e+12	392782
- review_scores_cleanliness	1	8089570356	6.9183e+12	392800

Step: AIC=392776.3

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + minimum_nights + maximum_nights + availability_365 +
 number_of_reviews + review_scores_cleanliness + review_scores_checkin +
 calculated_host_listings_count_private_rooms + calculated_host_listings_count_shared_rooms +
 number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
--	----	-----------	-----	-----

- availability_365	1	74636365	6.9103e+12	392774
- bedrooms	1	124377034	6.9104e+12	392775
- longitude	1	213920223	6.9105e+12	392775
- calculated_host_listings_count_shared_rooms	1	260654110	6.9105e+12	392775
- number_of_reviews	1	309502558	6.9106e+12	392775
- maximum_nights	1	366141039	6.9106e+12	392775
- host_total_listings_count	1	470117738	6.9107e+12	392776
- latitude	1	569012696	6.9108e+12	392776
<none>			6.9103e+12	392776
- calculated_host_listings_count_private_rooms	1	901134162	6.9112e+12	392777
- accommodates	1	976502697	6.9112e+12	392777
- review_scores_checkin	1	1014462013	6.9113e+12	392777
- number_of_reviews_l30d	1	1288579692	6.9115e+12	392778
+ reviews_per_month	1	24411188	6.9102e+12	392778
+ review_scores_location	1	653045	6.9103e+12	392778
- minimum_nights	1	1892821305	6.9121e+12	392780
- banios	1	2039276398	6.9123e+12	392780
- review_scores_cleanliness	1	8067414789	6.9183e+12	392798

Step: AIC=392774.5

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + minimum_nights + maximum_nights + number_of_reviews +
 review_scores_cleanliness + review_scores_checkin + calculated_host_listings_count_private_rooms +
 calculated_host_listings_count_shared_rooms + number_of_reviews_l30d +
 banios

	Df	Sum of Sq	RSS	AIC
- bedrooms	1	126134867	6.9105e+12	392773
- longitude	1	212674304	6.9105e+12	392773
- calculated_host_listings_count_shared_rooms	1	264471579	6.9106e+12	392773
- number_of_reviews	1	331889537	6.9107e+12	392773
- maximum_nights	1	415978787	6.9107e+12	392774
- host_total_listings_count	1	446598432	6.9108e+12	392774
- latitude	1	581407402	6.9109e+12	392774
<none>			6.9103e+12	392774
- calculated_host_listings_count_private_rooms	1	952872607	6.9113e+12	392775
- accommodates	1	969723147	6.9113e+12	392775
- review_scores_checkin	1	1039385517	6.9114e+12	392775
+ availability_365	1	74636365	6.9103e+12	392776
- number_of_reviews_l30d	1	1361460566	6.9117e+12	392776
+ reviews_per_month	1	18779953	6.9103e+12	392776
+ review_scores_location	1	381577	6.9103e+12	392776
- minimum_nights	1	1904831793	6.9122e+12	392778

- banios	1	2041877773	6.9124e+12	392778
- review_scores_cleanliness	1	8042925784	6.9184e+12	392796

Step: AIC=392772.9

price ~ host_total_listings_count + latitude + longitude + accommodates +
 minimum_nights + maximum_nights + number_of_reviews + review_scores_cleanliness +
 review_scores_checkin + calculated_host_listings_count_private_rooms +
 calculated_host_listings_count_shared_rooms + number_of_reviews_l30d +
 banios

	Df	Sum of Sq	RSS	AIC
- longitude	1	202514625	6.9107e+12	392771
- calculated_host_listings_count_shared_rooms	1	305926411	6.9108e+12	392772
- number_of_reviews	1	320768395	6.9108e+12	392772
- maximum_nights	1	407134715	6.9109e+12	392772
- host_total_listings_count	1	443176761	6.9109e+12	392772
- latitude	1	579976098	6.9110e+12	392773
<none>			6.9105e+12	392773
- calculated_host_listings_count_private_rooms	1	978310409	6.9114e+12	392774
- review_scores_checkin	1	1022978418	6.9115e+12	392774
+ bedrooms	1	126134867	6.9103e+12	392774
+ availability_365	1	76394198	6.9104e+12	392775
- number_of_reviews_l30d	1	1364921604	6.9118e+12	392775
+ reviews_per_month	1	16996576	6.9104e+12	392775
+ review_scores_location	1	1116768	6.9105e+12	392775
- accommodates	1	1673088884	6.9121e+12	392776
- minimum_nights	1	1907723288	6.9124e+12	392776
- banios	1	3411676750	6.9139e+12	392781
- review_scores_cleanliness	1	8079482545	6.9185e+12	392794

Step: AIC=392771.4

price ~ host_total_listings_count + latitude + accommodates +
 minimum_nights + maximum_nights + number_of_reviews + review_scores_cleanliness +
 review_scores_checkin + calculated_host_listings_count_private_rooms +
 calculated_host_listings_count_shared_rooms + number_of_reviews_l30d +
 banios

	Df	Sum of Sq	RSS	AIC
- number_of_reviews	1	306400645	6.9110e+12	392770
- calculated_host_listings_count_shared_rooms	1	326101523	6.9110e+12	392770
- maximum_nights	1	397595990	6.9111e+12	392771
- host_total_listings_count	1	477963842	6.9111e+12	392771
- latitude	1	537241689	6.9112e+12	392771

<none>			6.9107e+12	392771
- calculated_host_listings_count_private_rooms	1	989880866	6.9116e+12	392772
- review_scores_checkin	1	1012277787	6.9117e+12	392772
+ longitude	1	202514625	6.9105e+12	392773
+ bedrooms	1	115975188	6.9105e+12	392773
+ availability_365	1	75091637	6.9106e+12	392773
+ reviews_per_month	1	15977289	6.9106e+12	392773
+ review_scores_location	1	346	6.9107e+12	392773
- number_of_reviews_l30d	1	1451715034	6.9121e+12	392774
- accommodates	1	1626611459	6.9123e+12	392774
- minimum_nights	1	1920448935	6.9126e+12	392775
- banios	1	3591836302	6.9143e+12	392780
- review_scores_cleanliness	1	7991262869	6.9187e+12	392793

Step: AIC=392770.3

price ~ host_total_listings_count + latitude + accommodates +
 minimum_nights + maximum_nights + review_scores_cleanliness +
 review_scores_checkin + calculated_host_listings_count_private_rooms +
 calculated_host_listings_count_shared_rooms + number_of_reviews_l30d +
 banios

	Df	Sum of Sq	RSS	AIC
- calculated_host_listings_count_shared_rooms	1	328463902	6.9113e+12	392769
- maximum_nights	1	329194234	6.9113e+12	392769
- host_total_listings_count	1	458083266	6.9114e+12	392770
- latitude	1	596562471	6.9116e+12	392770
<none>			6.9110e+12	392770
- calculated_host_listings_count_private_rooms	1	1013689334	6.9120e+12	392771
- review_scores_checkin	1	1050404184	6.9120e+12	392771
+ number_of_reviews	1	306400645	6.9107e+12	392771
- number_of_reviews_l30d	1	1152251188	6.9121e+12	392772
+ longitude	1	188146874	6.9108e+12	392772
+ reviews_per_month	1	137093255	6.9108e+12	392772
+ bedrooms	1	105900428	6.9109e+12	392772
+ availability_365	1	96575074	6.9109e+12	392772
+ review_scores_location	1	29334	6.9110e+12	392772
- accommodates	1	1678765135	6.9126e+12	392773
- minimum_nights	1	1900597837	6.9129e+12	392774
- banios	1	3493069785	6.9145e+12	392778
- review_scores_cleanliness	1	7934576527	6.9189e+12	392791

Step: AIC=392769.3

price ~ host_total_listings_count + latitude + accommodates +

minimum_nights + maximum_nights + review_scores_cleanliness +
review_scores_checkin + calculated_host_listings_count_private_rooms +
number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
- maximum_nights	1	318368749	6.9116e+12	392768
- host_total_listings_count	1	499841711	6.9118e+12	392769
- latitude	1	537031988	6.9118e+12	392769
<none>			6.9113e+12	392769
- review_scores_checkin	1	1042923276	6.9123e+12	392770
+ calculated_host_listings_count_shared_rooms	1	328463902	6.9110e+12	392770
+ number_of_reviews	1	308763023	6.9110e+12	392770
- number_of_reviews_l30d	1	1120810345	6.9124e+12	392771
+ longitude	1	207632077	6.9111e+12	392771
+ reviews_per_month	1	152509315	6.9111e+12	392771
+ bedrooms	1	145239654	6.9111e+12	392771
+ availability_365	1	101780683	6.9112e+12	392771
+ review_scores_location	1	99026	6.9113e+12	392771
- calculated_host_listings_count_private_rooms	1	1740258091	6.9130e+12	392772
- accommodates	1	1824876886	6.9131e+12	392773
- minimum_nights	1	1852913798	6.9131e+12	392773
- banios	1	3225026657	6.9145e+12	392777
- review_scores_cleanliness	1	7898342177	6.9192e+12	392790

Step: AIC=392768.2

price ~ host_total_listings_count + latitude + accommodates +
minimum_nights + review_scores_cleanliness + review_scores_checkin +
calculated_host_listings_count_private_rooms + number_of_reviews_l30d +
banios

	Df	Sum of Sq	RSS	AIC
- host_total_listings_count	1	459301557	6.9121e+12	392768
- latitude	1	569399112	6.9122e+12	392768
<none>			6.9116e+12	392768
- review_scores_checkin	1	1031012810	6.9126e+12	392769
+ maximum_nights	1	318368749	6.9113e+12	392769
+ calculated_host_listings_count_shared_rooms	1	317638417	6.9113e+12	392769
- number_of_reviews_l30d	1	1104304349	6.9127e+12	392769
+ number_of_reviews	1	241037644	6.9114e+12	392770
+ longitude	1	200216272	6.9114e+12	392770
+ reviews_per_month	1	166603960	6.9114e+12	392770
+ availability_365	1	147715630	6.9115e+12	392770
+ bedrooms	1	137577988	6.9115e+12	392770

+ review_scores_location	1	186	6.9116e+12	392770
- calculated_host_listings_count_private_rooms	1	1671651621	6.9133e+12	392771
- accommodates	1	1771327382	6.9134e+12	392771
- minimum_nights	1	1807626416	6.9134e+12	392771
- banios	1	3288028434	6.9149e+12	392776
- review_scores_cleanliness	1	7847049869	6.9195e+12	392789

Step: AIC=392767.5

price ~ latitude + accommodates + minimum_nights + review_scores_cleanliness +
 review_scores_checkin + calculated_host_listings_count_private_rooms +
 number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
- latitude	1	647298458	6.9127e+12	392767
<none>			6.9121e+12	392768
+ host_total_listings_count	1	459301557	6.9116e+12	392768
- review_scores_checkin	1	967646743	6.9130e+12	392768
+ calculated_host_listings_count_shared_rooms	1	357696029	6.9117e+12	392768
- number_of_reviews_l30d	1	1083105824	6.9132e+12	392769
+ maximum_nights	1	277828595	6.9118e+12	392769
+ longitude	1	236386945	6.9118e+12	392769
+ number_of_reviews	1	227442540	6.9118e+12	392769
+ reviews_per_month	1	179657202	6.9119e+12	392769
+ bedrooms	1	136469344	6.9119e+12	392769
+ availability_365	1	109601920	6.9120e+12	392769
+ review_scores_location	1	376682	6.9121e+12	392770
- calculated_host_listings_count_private_rooms	1	1402762384	6.9135e+12	392770
- minimum_nights	1	1776421786	6.9138e+12	392771
- accommodates	1	1922717394	6.9140e+12	392771
- banios	1	3331524076	6.9154e+12	392775
- review_scores_cleanliness	1	7661143327	6.9197e+12	392788

Step: AIC=392767.4

price ~ accommodates + minimum_nights + review_scores_cleanliness +
 review_scores_checkin + calculated_host_listings_count_private_rooms +
 number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
<none>			6.9127e+12	392767
+ latitude	1	647298458	6.9121e+12	392768
+ host_total_listings_count	1	537200902	6.9122e+12	392768
- number_of_reviews_l30d	1	905021911	6.9136e+12	392768
- review_scores_checkin	1	1013490743	6.9137e+12	392768

+ maximum_nights	1	306957867	6.9124e+12	392769
+ calculated_host_listings_count_shared_rooms	1	292418356	6.9124e+12	392769
+ number_of_reviews	1	276660876	6.9124e+12	392769
+ reviews_per_month	1	253299212	6.9125e+12	392769
+ longitude	1	184651623	6.9125e+12	392769
+ bedrooms	1	130309733	6.9126e+12	392769
+ availability_365	1	126966152	6.9126e+12	392769
- calculated_host_listings_count_private_rooms	1	1311302950	6.9140e+12	392769
+ review_scores_location	1	3490765	6.9127e+12	392769
- minimum_nights	1	1763907876	6.9145e+12	392770
- accommodates	1	2073656799	6.9148e+12	392771
- banios	1	3278290753	6.9160e+12	392775
- review_scores_cleanliness	1	7682627305	6.9204e+12	392788

```
summary(seleccion)
```

Call:

```
lm(formula = price ~ accommodates + minimum_nights + review_scores_cleanliness +
    review_scores_checkin + calculated_host_listings_count_private_rooms +
    number_of_reviews_l30d + banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-21998	-1061	-552	22	1830686

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	6154.460	1669.668	3.686
accommodates	182.156	74.429	2.447
minimum_nights	13.060	5.786	2.257
review_scores_cleanliness	-1976.152	419.502	-4.711
review_scores_checkin	796.235	465.371	1.711
calculated_host_listings_count_private_rooms	-53.706	27.596	-1.946
number_of_reviews_l30d	-106.279	65.733	-1.617
banios	552.808	179.647	3.077

	Pr(> t)
(Intercept)	0.000228 ***
accommodates	0.014399 *
minimum_nights	0.024006 *
review_scores_cleanliness	2.49e-06 ***

```

review_scores_checkin          0.087103 .
calculated_host_listings_count_private_rooms 0.051647 .
number_of_reviews_l30d        0.105933
banios                        0.002092 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 18610 on 19967 degrees of freedom
Multiple R-squared:  0.003749, Adjusted R-squared:  0.0034
F-statistic: 10.73 on 7 and 19967 DF, p-value: 1.428e-13

```

Dado que el anterior modelo contempla la base datos, hay que tener en cuenta que dichos datos tienen valores atípicos, por lo que será indispensable quitar para obtener un mejor modelo.

Creemos el siguiente código para eliminar datos atípicos por columna:

```

#Variables sin datos atípicos

# Crear una función para eliminar valores atípicos basados en el rango intercuartílico en

remove_outliers_iqr <- function(data, multiplier) {
  for (columna in names(data)) {
    q1 <- quantile(data[[columna]], 0.25)
    q3 <- quantile(data[[columna]], 0.75)
    iqr <- q3 - q1
    lower_limit <- q1 - multiplier * iqr
    upper_limit <- q3 + multiplier * iqr
    data <- subset(data, data[[columna]] >= lower_limit & data[[columna]] <= upper_limit)
  }
  return(data)
}

# Llamamos a la función para eliminar valores atípicos en todas las columnas del dataframe

data <- remove_outliers_iqr(datos, 1.5) # Elimina los valores que están fuera de 1.5 veces

```

Así probamos nuestro modelo 1, pero ahora utilizando una base de datos sin valores atípicos, y seleccionamos un modelo sin considerar variables que no son significantes.

```
attach(data)
```

The following objects are masked from `datos` (pos = 3):


```
accommodates, availability_365, banios, bedrooms,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d, price,
review_scores_checkin, review_scores_cleanliness,
review_scores_location, reviews_per_month
```

The following objects are masked from `datos` (`pos = 4`):

```
accommodates, availability_365, banios, bedrooms,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d, price,
review_scores_checkin, review_scores_cleanliness,
review_scores_location, reviews_per_month
```

```
#Modelo sin atípicos
modelo1 <- lm(price~host_total_listings_count
              +latitude
              +longitude
              +accommodates
              +bedrooms
              +minimum_nights
              +maximum_nights
              +availability_365
              +number_of_reviews
              +review_scores_cleanliness
              +review_scores_checkin
              +review_scores_location
              +calculated_host_listings_count_private_rooms
              +calculated_host_listings_count_shared_rooms
              +reviews_per_month
              +number_of_reviews_l30d
              +banios)
summary(modelo1)
```

Call:

```
lm(formula = price ~ host_total_listings_count + latitude + longitude +
```

accommodates + bedrooms + minimum_nights + maximum_nights +
availability_365 + number_of_reviews + review_scores_cleanliness +
review_scores_checkin + review_scores_location + calculated_host_listings_count_private_1
calculated_host_listings_count_shared_rooms + reviews_per_month +
number_of_reviews_l30d + banios)

Residuals:

Min	1Q	Median	3Q	Max
-1223.32	-290.61	-84.62	200.74	2326.81

Coefficients: (1 not defined because of singularities)

	Estimate	Std. Error	t value
(Intercept)	-5.459e+05	2.775e+04	-19.674
host_total_listings_count	1.423e+01	7.847e-01	18.132
latitude	1.571e+03	1.622e+02	9.691
longitude	-5.160e+03	2.813e+02	-18.346
accommodates	1.225e+02	5.786e+00	21.172
bedrooms	3.214e+01	1.449e+01	2.218
minimum_nights	-7.371e+00	4.870e+00	-1.514
maximum_nights	4.259e-03	1.053e-02	0.404
availability_365	3.485e-01	4.096e-02	8.508
number_of_reviews	-6.820e-01	1.787e-01	-3.816
review_scores_cleanliness	3.189e+02	3.385e+01	9.419
review_scores_checkin	-1.521e+02	6.294e+01	-2.416
review_scores_location	6.474e+02	6.008e+01	10.776
calculated_host_listings_count_private_rooms	-9.331e+01	4.297e+00	-21.718
calculated_host_listings_count_shared_rooms	NA	NA	NA
reviews_per_month	-5.054e+00	6.402e+00	-0.789
number_of_reviews_l30d	-7.070e+00	4.729e+00	-1.495
banios	2.115e+02	1.456e+01	14.525

Pr(>|t|)

(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
bedrooms	0.026605 *
minimum_nights	0.130148
maximum_nights	0.686019
availability_365	< 2e-16 ***
number_of_reviews	0.000137 ***
review_scores_cleanliness	< 2e-16 ***
review_scores_checkin	0.015722 *

```

review_scores_location          < 2e-16 ***
calculated_host_listings_count_private_rooms < 2e-16 ***
calculated_host_listings_count_shared_rooms      NA
reviews_per_month               0.429878
number_of_reviews_l30d         0.134925
banios                          < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 433.9 on 7684 degrees of freedom
Multiple R-squared:  0.3946,    Adjusted R-squared:  0.3933
F-statistic:   313 on 16 and 7684 DF,  p-value: < 2.2e-16

```

```

seleccion <- stepAIC(modelo1, direction = c("both"));

```

Start: AIC=93549.56

```

price ~ host_total_listings_count + latitude + longitude + accommodates +
  bedrooms + minimum_nights + maximum_nights + availability_365 +
  number_of_reviews + review_scores_cleanliness + review_scores_checkin +
  review_scores_location + calculated_host_listings_count_private_rooms +
  calculated_host_listings_count_shared_rooms + reviews_per_month +
  number_of_reviews_l30d + banios

```

Step: AIC=93549.56

```

price ~ host_total_listings_count + latitude + longitude + accommodates +
  bedrooms + minimum_nights + maximum_nights + availability_365 +
  number_of_reviews + review_scores_cleanliness + review_scores_checkin +
  review_scores_location + calculated_host_listings_count_private_rooms +
  reviews_per_month + number_of_reviews_l30d + banios

```

	Df	Sum of Sq	RSS	AIC
- maximum_nights	1	30767	1446520887	93548
- reviews_per_month	1	117319	1446607439	93548
<none>			1446490120	93550
- number_of_reviews_l30d	1	420802	1446910921	93550
- minimum_nights	1	431318	1446921437	93550
- bedrooms	1	925826	1447415946	93552
- review_scores_checkin	1	1098658	1447588778	93553
- number_of_reviews	1	2741239	1449231359	93562
- availability_365	1	13627115	1460117234	93620

- review_scores_cleanliness	1	16702585	1463192705	93636
- latitude	1	17678074	1464168194	93641
- review_scores_location	1	21858850	1468348969	93663
- banios	1	39716258	1486206377	93756
- host_total_listings_count	1	61889682	1508379802	93870
- longitude	1	63362412	1509852532	93878
- accommodates	1	84383882	1530874001	93984
- calculated_host_listings_count_private_rooms	1	88787633	1535277753	94006

Step: AIC=93547.72

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + minimum_nights + availability_365 + number_of_reviews +
 review_scores_cleanliness + review_scores_checkin + review_scores_location +
 calculated_host_listings_count_private_rooms + reviews_per_month +
 number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
- reviews_per_month	1	132193	1446653079	93546
<none>			1446520887	93548
- number_of_reviews_l30d	1	420887	1446941773	93548
- minimum_nights	1	434881	1446955768	93548
+ maximum_nights	1	30767	1446490120	93550
- bedrooms	1	926548	1447447435	93551
- review_scores_checkin	1	1103685	1447624572	93552
- number_of_reviews	1	2716146	1449237033	93560
- availability_365	1	14081628	1460602515	93620
- review_scores_cleanliness	1	16673092	1463193978	93634
- latitude	1	17666620	1464187507	93639
- review_scores_location	1	21830396	1468351283	93661
- banios	1	39702573	1486223460	93754
- host_total_listings_count	1	62003073	1508523960	93869
- longitude	1	63431078	1509951964	93876
- accommodates	1	84577833	1531098720	93983
- calculated_host_listings_count_private_rooms	1	88857001	1535377888	94005

Step: AIC=93546.42

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + minimum_nights + availability_365 + number_of_reviews +
 review_scores_cleanliness + review_scores_checkin + review_scores_location +
 calculated_host_listings_count_private_rooms + number_of_reviews_l30d +
 banios

	Df	Sum of Sq	RSS	AIC
--	----	-----------	-----	-----

- minimum_nights	1	370566	1447023646	93546
<none>			1446653079	93546
+ reviews_per_month	1	132193	1446520887	93548
+ maximum_nights	1	45641	1446607439	93548
- bedrooms	1	962735	1447615814	93550
- number_of_reviews_l30d	1	1021929	1447675008	93550
- review_scores_checkin	1	1060000	1447713079	93550
- number_of_reviews	1	3198838	1449851917	93561
- availability_365	1	13975687	1460628766	93618
- review_scores_cleanliness	1	16636163	1463289242	93632
- latitude	1	17548216	1464201295	93637
- review_scores_location	1	21913560	1468566639	93660
- banios	1	39636627	1486289706	93753
- host_total_listings_count	1	61889420	1508542500	93867
- longitude	1	63515062	1510168141	93875
- accommodates	1	84774450	1531427529	93983
- calculated_host_listings_count_private_rooms	1	89256077	1535909156	94005

Step: AIC=93546.4

price ~ host_total_listings_count + latitude + longitude + accommodates +
 bedrooms + availability_365 + number_of_reviews + review_scores_cleanliness +
 review_scores_checkin + review_scores_location + calculated_host_listings_count_private_rooms +
 number_of_reviews_l30d + banios

	Df	Sum of Sq	RSS	AIC
<none>			1447023646	93546
+ minimum_nights	1	370566	1446653079	93546
+ reviews_per_month	1	67878	1446955768	93548
+ maximum_nights	1	44982	1446978664	93548
- bedrooms	1	907465	1447931111	93549
- number_of_reviews_l30d	1	941796	1447965442	93549
- review_scores_checkin	1	1117206	1448140852	93550
- number_of_reviews	1	3283850	1450307496	93562
- availability_365	1	14729023	1461752669	93622
- review_scores_cleanliness	1	16752559	1463776205	93633
- latitude	1	17596789	1464620434	93637
- review_scores_location	1	21978521	1469002167	93660
- banios	1	39379111	1486402757	93751
- longitude	1	63398815	1510422461	93875
- host_total_listings_count	1	63672663	1510696308	93876
- accommodates	1	85745632	1532769278	93988
- calculated_host_listings_count_private_rooms	1	89001653	1536025299	94004

```
summary(seleccion)
```

Call:

```
lm(formula = price ~ host_total_listings_count + latitude + longitude +  
    accommodates + bedrooms + availability_365 + number_of_reviews +  
    review_scores_cleanliness + review_scores_checkin + review_scores_location +  
    calculated_host_listings_count_private_rooms + number_of_reviews_l30d +  
    banios)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-1215.86	-291.61	-84.42	200.10	2329.41

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-5.457e+05	2.774e+04	-19.674
host_total_listings_count	1.431e+01	7.782e-01	18.391
latitude	1.565e+03	1.619e+02	9.668
longitude	-5.160e+03	2.812e+02	-18.352
accommodates	1.226e+02	5.743e+00	21.343
bedrooms	3.174e+01	1.446e+01	2.196
availability_365	3.557e-01	4.021e-02	8.846
number_of_reviews	-7.156e-01	1.713e-01	-4.177
review_scores_cleanliness	3.191e+02	3.382e+01	9.434
review_scores_checkin	-1.529e+02	6.275e+01	-2.436
review_scores_location	6.488e+02	6.004e+01	10.805
calculated_host_listings_count_private_rooms	-9.232e+01	4.246e+00	-21.744
number_of_reviews_l30d	-8.842e+00	3.953e+00	-2.237
banios	2.104e+02	1.455e+01	14.464

Pr(>|t|)

(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
bedrooms	0.0281 *
availability_365	< 2e-16 ***
number_of_reviews	2.99e-05 ***
review_scores_cleanliness	< 2e-16 ***
review_scores_checkin	0.0149 *
review_scores_location	< 2e-16 ***

```

calculated_host_listings_count_private_rooms < 2e-16 ***
number_of_reviews_l30d                      0.0253 *
banios                                         < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 433.9 on 7687 degrees of freedom
Multiple R-squared:  0.3944,    Adjusted R-squared:  0.3934
F-statistic: 385.1 on 13 and 7687 DF,  p-value: < 2.2e-16

```

Teniendo ahora la base de datos definitiva, se ajustó un primer modelo para la misma, sin embargo, tras esto nos pudimos percatar de que la base de datos contenía valores atípicos, por lo que resultó indispensable quitarlos para obtener un mejor modelo y, tras eliminarlos, pudimos ver un aumento en la precisión del modelo tanto en su R^2 ajustada como en los p-values de cada una de sus variables.

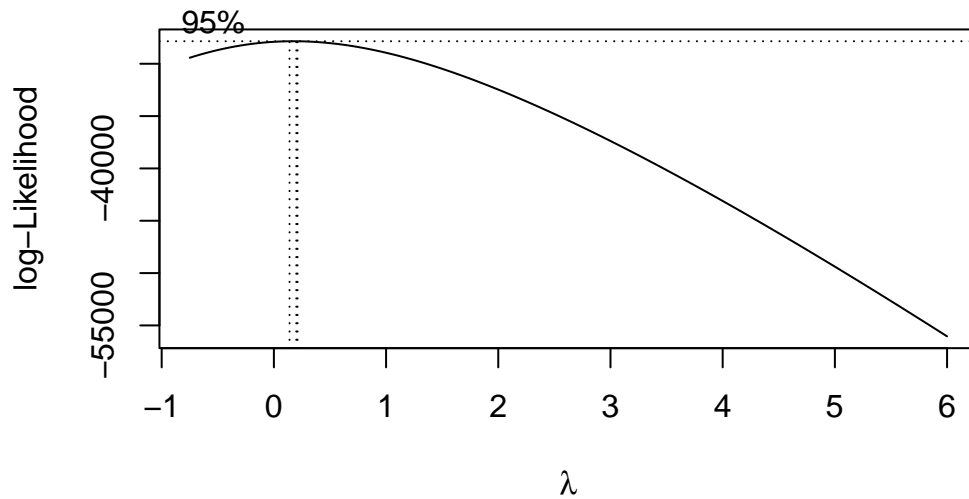
Nota: Para más información de cada transformación consultar el archivo “Documento expandido”

4. Los datos atípicos y los supuestos

Finalmente procedimos a acotar nuestro modelo, haciéndolo efectivo únicamente para Airbnb's que cumplen ciertas características.

Para tratar de subir un poco nuestra R^2 ajustada, utilizamos la función de Box-Cox, para obtener una potencia (que llamaremos lambda) que nos ayudara a reescalar nuestros datos de la variable precios para así poder trabajar con ellos de manera más fácil. Dicha lambda es la siguiente:

```
BC <-boxcox(seleccion, lambda = seq(-0.75,6, by = 0.05), plotit = TRUE)
```



```
lamda <- BC$x[which.max(BC$y)]
```

Una vez encontrada la lambda adecuada, transformamos nuestra variable de precios. Así, obtenemos el siguiente modelo, el cual es el mejor según el análisis previo.

Para lograrlo, aplicaremos la siguiente transformación:

$$y = \frac{price^\lambda - 1}{\lambda}$$

```
modelo2 <- lm((((price^lamda)-1)/lamda) ~ host_total_listings_count + latitude + longitude +
  accommodates + bedrooms + availability_365 + number_of_reviews +
  review_scores_cleanliness + review_scores_checkin + review_scores_location +
  calculated_host_listings_count_private_rooms + number_of_reviews_l30d +
  banios)
summary(modelo2)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +
  latitude + longitude + accommodates + bedrooms + availability_365 +
```


number_of_reviews + review_scores_cleanliness + review_scores_checkin +
review_scores_location + calculated_host_listings_count_private_rooms +
number_of_reviews_l30d + banios)

Residuals:

Min	1Q	Median	3Q	Max
-5.7521	-1.1399	-0.1166	1.0424	7.6814

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.259e+03	1.054e+02	-21.437
host_total_listings_count	5.677e-02	2.957e-03	19.199
latitude	5.802e+00	6.151e-01	9.433
longitude	-2.161e+01	1.068e+00	-20.228
accommodates	5.696e-01	2.182e-02	26.101
bedrooms	1.263e-02	5.493e-02	0.230
availability_365	1.507e-03	1.528e-04	9.862
number_of_reviews	-2.145e-03	6.510e-04	-3.295
review_scores_cleanliness	1.277e+00	1.285e-01	9.937
review_scores_checkin	-6.342e-01	2.384e-01	-2.660
review_scores_location	2.537e+00	2.281e-01	11.121
calculated_host_listings_count_private_rooms	-4.686e-01	1.613e-02	-29.046
number_of_reviews_l30d	-2.827e-02	1.502e-02	-1.882
banios	6.706e-01	5.527e-02	12.133

Pr(>|t|)

(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
bedrooms	0.818192
availability_365	< 2e-16 ***
number_of_reviews	0.000988 ***
review_scores_cleanliness	< 2e-16 ***
review_scores_checkin	0.007833 **
review_scores_location	< 2e-16 ***
calculated_host_listings_count_private_rooms	< 2e-16 ***
number_of_reviews_l30d	0.059881 .
banios	< 2e-16 ***

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

Residual standard error: 1.649 on 7687 degrees of freedom

Multiple R-squared: 0.4471, Adjusted R-squared: 0.4462
 F-statistic: 478.2 on 13 and 7687 DF, p-value: < 2.2e-16

```
seleccion <- stepAIC(modelo2, direction = c("both"));
```

Start: AIC=7713.96

```
((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude +
  longitude + accommodates + bedrooms + availability_365 +
  number_of_reviews + review_scores_cleanliness + review_scores_checkin +
  review_scores_location + calculated_host_listings_count_private_rooms +
  number_of_reviews_l30d + banios
```

	Df	Sum of Sq	RSS	AIC
- bedrooms	1	0.14	20893	7712.0
<none>			20893	7714.0
- number_of_reviews_l30d	1	9.63	20902	7715.5
- review_scores_checkin	1	19.23	20912	7719.0
- number_of_reviews	1	29.51	20922	7722.8
- latitude	1	241.85	21135	7800.6
- availability_365	1	264.32	21157	7808.8
- review_scores_cleanliness	1	268.36	21161	7810.2
- review_scores_location	1	336.13	21229	7834.9
- banios	1	400.10	21293	7858.0
- host_total_listings_count	1	1001.87	21895	8072.7
- longitude	1	1112.13	22005	8111.4
- accommodates	1	1851.59	22744	8365.9
- calculated_host_listings_count_private_rooms	1	2293.00	23186	8513.9

Step: AIC=7712.01

```
((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude +
  longitude + accommodates + availability_365 + number_of_reviews +
  review_scores_cleanliness + review_scores_checkin + review_scores_location +
  calculated_host_listings_count_private_rooms + number_of_reviews_l30d +
  banios
```

	Df	Sum of Sq	RSS	AIC
<none>			20893	7712.0
- number_of_reviews_l30d	1	9.7	20903	7713.6
+ bedrooms	1	0.1	20893	7714.0
- review_scores_checkin	1	19.2	20912	7717.1
- number_of_reviews	1	29.7	20923	7721.0

- latitude	1	241.7	21135	7798.6
- availability_365	1	264.7	21158	7807.0
- review_scores_cleanliness	1	268.2	21161	7808.2
- review_scores_location	1	336.1	21229	7832.9
- banios	1	451.2	21344	7874.6
- host_total_listings_count	1	1008.8	21902	8073.2
- longitude	1	1112.4	22005	8109.5
- calculated_host_listings_count_private_rooms	1	2301.0	23194	8514.6
- accommodates	1	3367.4	24260	8860.8

```
summary(seleccion)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +
  latitude + longitude + accommodates + availability_365 +
  number_of_reviews + review_scores_cleanliness + review_scores_checkin +
  review_scores_location + calculated_host_listings_count_private_rooms +
  number_of_reviews_l30d + banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-5.7439	-1.1389	-0.1156	1.0403	7.6815

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.260e+03	1.054e+02	-21.440
host_total_listings_count	5.671e-02	2.943e-03	19.267
latitude	5.799e+00	6.149e-01	9.431
longitude	-2.161e+01	1.068e+00	-20.232
accommodates	5.729e-01	1.628e-02	35.201
availability_365	1.508e-03	1.528e-04	9.869
number_of_reviews	-2.151e-03	6.504e-04	-3.308
review_scores_cleanliness	1.276e+00	1.285e-01	9.935
review_scores_checkin	-6.341e-01	2.384e-01	-2.660
review_scores_location	2.537e+00	2.281e-01	11.121
calculated_host_listings_count_private_rooms	-4.688e-01	1.611e-02	-29.098
number_of_reviews_l30d	-2.832e-02	1.502e-02	-1.886
banios	6.746e-01	5.236e-02	12.885
	Pr(> t)		
(Intercept)	< 2e-16	***	

```

host_total_listings_count      < 2e-16 ***
latitude                      < 2e-16 ***
longitude                     < 2e-16 ***
accommodates                   < 2e-16 ***
availability_365               < 2e-16 ***
number_of_reviews              0.000945 ***
review_scores_cleanliness      < 2e-16 ***
review_scores_checkin          0.007840 **
review_scores_location         < 2e-16 ***
calculated_host_listings_count_private_rooms < 2e-16 ***
number_of_reviews_l30d         0.059334 .
banios                         < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.649 on 7688 degrees of freedom
Multiple R-squared:  0.4471,    Adjusted R-squared:  0.4463
F-statistic: 518.1 on 12 and 7688 DF,  p-value: < 2.2e-16

```

Tras este reescalamiento, se eliminaron nuevamente los valores atípicos mediante dos métodos siendo el primero el uso de la distancia de Cook para su eliminación y el segundo acotando el costo de arrendamiento de un Airbnb, acotando a su vez a nuestro modelo para Airbnb's cuyo costo anteriormente mencionado no supere los \$2385 MXN (en escala transformada).

```

# Calcular la distancia de Cook del mejor modelo
distancia_cook <- cooks.distance(seleccion)

# Definir el umbral para la distancia de Cook
umbral <- 4/length(distancia_cook) # Utilizamos el umbral 4/n, donde n es el número de ob

# Identificar las observaciones influyentes
observaciones_influyentes <- which(distancia_cook > umbral)

# Paso 4: Eliminar las observaciones influyentes del conjunto de datos
datos_filtrados <- data[-observaciones_influyentes, ]

summary(datos_filtrados$price)

```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
180	600	932	1015	1314	2750

Finalmente con el resumen anterior, observemos que con la siguiente fórmula

$$1314 + (1.5(1314 - 600)) = 2385$$

Así eliminamos precios que esten por encima de dicho valor, con el siguiente código:

```
#Quitamos los últimos valore atípicos en los precios
data_clean <- subset(datos_filtrados, datos_filtrados$price <= 2385)
#Volvemos a eliminar datos atípicos
data_2 <- remove_outliers_iqr(data_clean, 1.5) # Elimina los valores que están fuera de 1
```

Finalmente nuestro mejor modelo queda de la siguiente manera:

```
attach(data_2)
```

The following objects are masked from data:

```
accommodates, availability_365, banios, bedrooms,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d, price,
review_scores_checkin, review_scores_cleanliness,
review_scores_location, reviews_per_month
```

The following objects are masked from datos (pos = 4):

```
accommodates, availability_365, banios, bedrooms,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d, price,
review_scores_checkin, review_scores_cleanliness,
review_scores_location, reviews_per_month
```

The following objects are masked from datos (pos = 5):

```
accommodates, availability_365, banios, bedrooms,
calculated_host_listings_count_private_rooms,
calculated_host_listings_count_shared_rooms,
host_total_listings_count, latitude, longitude, maximum_nights,
minimum_nights, number_of_reviews, number_of_reviews_l30d, price,
```

```
review_scores_checkin, review_scores_cleanliness,
review_scores_location, reviews_per_month
```

```
mejor_modelo_ajustado <- lm((((price^lamda) - 1)/lamda) ~ host_total_listings_count +
                             latitude + longitude + accommodates + availability_365 +
                             number_of_reviews + review_scores_cleanliness + review_scores_checkin +
                             review_scores_location + calculated_host_listings_count_private_rooms +
                             number_of_reviews_l30d + banios)
summary(mejor_modelo_ajustado)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +
    latitude + longitude + accommodates + availability_365 +
    number_of_reviews + review_scores_cleanliness + review_scores_checkin +
    review_scores_location + calculated_host_listings_count_private_rooms +
    number_of_reviews_l30d + banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-4.1785	-0.9435	-0.0778	0.9031	5.3215

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.250e+03	1.237e+02	-18.183
host_total_listings_count	9.689e-02	7.029e-03	13.783
latitude	8.512e+00	6.684e-01	12.735
longitude	-2.095e+01	1.254e+00	-16.712
accommodates	4.496e-01	1.804e-02	24.923
availability_365	9.828e-04	1.609e-04	6.110
number_of_reviews	-1.507e-03	8.487e-04	-1.776
review_scores_cleanliness	1.431e+00	1.612e-01	8.880
review_scores_checkin	-8.546e-02	3.241e-01	-0.264
review_scores_location	2.615e+00	2.999e-01	8.719
calculated_host_listings_count_private_rooms	-1.118e+00	3.427e-02	-32.634
number_of_reviews_l30d	-1.353e-02	1.660e-02	-0.815
banios	6.992e-01	5.463e-02	12.797

	Pr(> t)
(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***

```

longitude                < 2e-16 ***
accommodates              < 2e-16 ***
availability_365          1.08e-09 ***
number_of_reviews         0.0758 .
review_scores_cleanliness < 2e-16 ***
review_scores_checkin     0.7920
review_scores_location    < 2e-16 ***
calculated_host_listings_count_private_rooms < 2e-16 ***
number_of_reviews_l30d    0.4153
banios                   < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Residual standard error: 1.372 on 4733 degrees of freedom
Multiple R-squared: 0.539, Adjusted R-squared: 0.5378
F-statistic: 461.2 on 12 and 4733 DF, p-value: < 2.2e-16

Nuevamente tomamos el mejor modelo con la variable seleccion:

```
seleccion <- stepAIC(mejor_modelo_ajustado, direction = c("both"));
```

Start: AIC=3012.33

```
((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude +
  longitude + accommodates + availability_365 + number_of_reviews +
  review_scores_cleanliness + review_scores_checkin + review_scores_location +
  calculated_host_listings_count_private_rooms + number_of_reviews_l30d +
  banios
```

	Df	Sum of Sq	RSS	AIC
- review_scores_checkin	1	0.13	8904.4	3010.4
- number_of_reviews_l30d	1	1.25	8905.5	3011.0
<none>			8904.3	3012.3
- number_of_reviews	1	5.93	8910.2	3013.5
- availability_365	1	70.23	8974.5	3047.6
- review_scores_location	1	143.02	9047.3	3086.0
- review_scores_cleanliness	1	148.36	9052.6	3088.8
- latitude	1	305.11	9209.4	3170.2
- banios	1	308.11	9212.4	3171.8
- host_total_listings_count	1	357.42	9261.7	3197.1
- longitude	1	525.44	9429.7	3282.4
- accommodates	1	1168.58	10072.9	3595.6
- calculated_host_listings_count_private_rooms	1	2003.59	10907.9	3973.5

Step: AIC=3010.4

```
((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude +
  longitude + accommodates + availability_365 + number_of_reviews +
  review_scores_cleanliness + review_scores_location + calculated_host_listings_count_private_rooms +
  number_of_reviews_l30d + banios
```

	Df	Sum of Sq	RSS	AIC
- number_of_reviews_l30d	1	1.24	8905.7	3009.1
<none>			8904.4	3010.4
- number_of_reviews	1	5.81	8910.2	3011.5
+ review_scores_checkin	1	0.13	8904.3	3012.3
- availability_365	1	70.76	8975.2	3046.0
- review_scores_location	1	149.68	9054.1	3087.5
- review_scores_cleanliness	1	158.41	9062.8	3092.1
- latitude	1	306.52	9210.9	3169.0
- banios	1	308.67	9213.1	3170.1
- host_total_listings_count	1	358.88	9263.3	3195.9
- longitude	1	526.32	9430.7	3280.9
- accommodates	1	1170.44	10074.9	3594.5
- calculated_host_listings_count_private_rooms	1	2003.60	10908.0	3971.6

Step: AIC=3009.06

```
((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude +
  longitude + accommodates + availability_365 + number_of_reviews +
  review_scores_cleanliness + review_scores_location + calculated_host_listings_count_private_rooms +
  banios
```

	Df	Sum of Sq	RSS	AIC
<none>			8905.7	3009.1
+ number_of_reviews_l30d	1	1.24	8904.4	3010.4
+ review_scores_checkin	1	0.12	8905.5	3011.0
- number_of_reviews	1	7.58	8913.2	3011.1
- availability_365	1	69.54	8975.2	3044.0
- review_scores_location	1	149.91	9055.6	3086.3
- review_scores_cleanliness	1	157.88	9063.5	3090.5
- latitude	1	305.35	9211.0	3167.1
- banios	1	309.40	9215.1	3169.1
- host_total_listings_count	1	358.13	9263.8	3194.2
- longitude	1	530.65	9436.3	3281.7
- accommodates	1	1169.84	10075.5	3592.8
- calculated_host_listings_count_private_rooms	1	2006.32	10912.0	3971.3


```
summary(seleccion)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +  
  latitude + longitude + accommodates + availability_365 +  
  number_of_reviews + review_scores_cleanliness + review_scores_location +  
  calculated_host_listings_count_private_rooms + banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-4.1836	-0.9444	-0.0743	0.9033	5.3290

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.256e+03	1.235e+02	-18.275
host_total_listings_count	9.685e-02	7.019e-03	13.799
latitude	8.494e+00	6.666e-01	12.742
longitude	-2.101e+01	1.251e+00	-16.797
accommodates	4.489e-01	1.800e-02	24.940
availability_365	9.697e-04	1.595e-04	6.080
number_of_reviews	-1.630e-03	8.120e-04	-2.007
review_scores_cleanliness	1.417e+00	1.546e-01	9.162
review_scores_location	2.598e+00	2.910e-01	8.928
calculated_host_listings_count_private_rooms	-1.116e+00	3.418e-02	-32.661
banios	7.003e-01	5.460e-02	12.826

Pr(>|t|)

(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
availability_365	1.29e-09 ***
number_of_reviews	0.0448 *
review_scores_cleanliness	< 2e-16 ***
review_scores_location	< 2e-16 ***
calculated_host_listings_count_private_rooms	< 2e-16 ***
banios	< 2e-16 ***

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

Residual standard error: 1.371 on 4735 degrees of freedom

Multiple R-squared: 0.5389, Adjusted R-squared: 0.538
 F-statistic: 553.5 on 10 and 4735 DF, p-value: < 2.2e-16

```
modelo_final<- lm((((price^lamda) - 1)/lamda) ~ host_total_listings_count +
  latitude + longitude + accommodates + availability_365 +
  + review_scores_cleanliness + review_scores_location +
  calculated_host_listings_count_private_rooms + banios)
summary(modelo_final)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +
  latitude + longitude + accommodates + availability_365 +
  +review_scores_cleanliness + review_scores_location + calculated_host_listings_count_pri
  banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-4.1586	-0.9596	-0.0736	0.9105	5.3492

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.258e+03	1.235e+02	-18.287
host_total_listings_count	9.588e-02	7.004e-03	13.689
latitude	8.513e+00	6.668e-01	12.767
longitude	-2.102e+01	1.252e+00	-16.799
accommodates	4.465e-01	1.796e-02	24.853
availability_365	9.943e-04	1.591e-04	6.251
review_scores_cleanliness	1.467e+00	1.526e-01	9.611
review_scores_location	2.709e+00	2.858e-01	9.479
calculated_host_listings_count_private_rooms	-1.114e+00	3.417e-02	-32.600
banios	7.028e-01	5.460e-02	12.872

	Pr(> t)
(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
availability_365	4.43e-10 ***
review_scores_cleanliness	< 2e-16 ***
review_scores_location	< 2e-16 ***

```

calculated_host_listings_count_private_rooms < 2e-16 ***
baños < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.372 on 4736 degrees of freedom
Multiple R-squared:  0.5385,    Adjusted R-squared:  0.5377 
F-statistic: 614.1 on 9 and 4736 DF,  p-value: < 2.2e-16

```

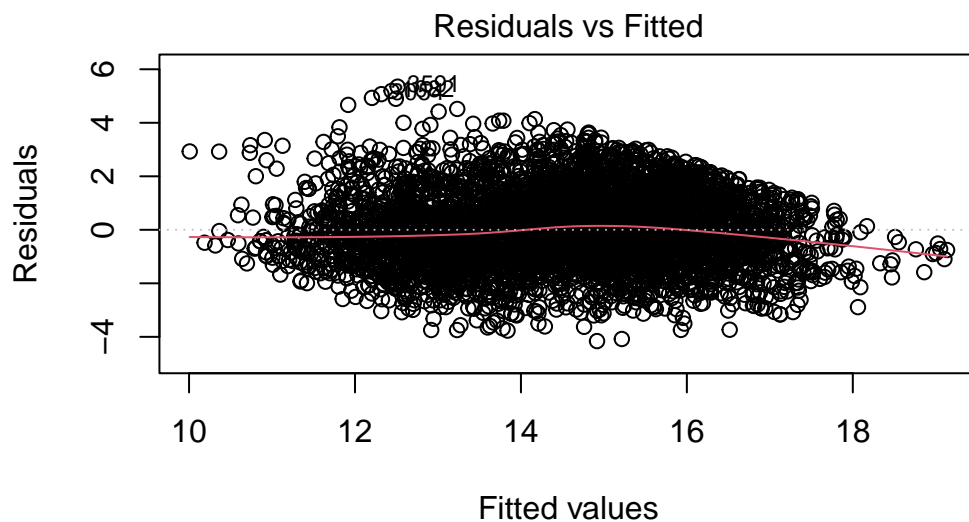
Finalmente, procederemos a la validación de nuestro modelo y, a continuación, realizaremos una interpretación detallada del mismo:

Linealidad:

```

#Hacemos la prueba
plot(modelo_final,1)

```



$\text{lm}(\text{(((price}^{\text{lamda}}) - 1)/\text{lamda}) \sim \text{host_total_listings_count} + \text{latitude} + \text{lor}$

```

cor.test((((price^lamda) - 1)/lamda), host_total_listings_count)

```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and host_total_listings_count
t = 13.486, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.1646058 0.2194092
sample estimates:
      cor
0.1921573
```

```
cor.test(((price^lamda) - 1)/lamda), latitude)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and latitude
t = 14.793, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.1826212 0.2370170
sample estimates:
      cor
0.2099816
```

```
cor.test(((price^lamda) - 1)/lamda), longitude)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and longitude
t = -16.794, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 -0.2635561 -0.2098437
sample estimates:
      cor
-0.2368809
```

```
cor.test(((price^lamda) - 1)/lamda), accommodates)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and accommodates
t = 45.585, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.5318038 0.5713839
sample estimates:
      cor
0.5519046
```

```
cor.test(((price^lamda) - 1)/lamda), availability_365)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and availability_365
t = 5.4775, df = 4744, p-value = 4.538e-08
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.05093873 0.10748430
sample estimates:
      cor
0.07927529
```

```
cor.test(((price^lamda) - 1)/lamda), review_scores_cleanliness)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and review_scores_cleanliness
t = 5.7414, df = 4744, p-value = 9.972e-09
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.05474806 0.11125860
sample estimates:
```

```
cor
0.08307011
```

```
cor.test(((price^lamda) - 1)/lamda), review_scores_location)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and review_scores_location
t = 4.9214, df = 4744, p-value = 8.883e-07
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 0.04290655 0.09952063
sample estimates:
      cor
0.07127099
```

```
cor.test(((price^lamda) - 1)/lamda), calculated_host_listings_count_private_rooms)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and calculated_host_listings_count_private_rooms
t = -45.505, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 -0.5707371 -0.5311151
sample estimates:
      cor
-0.5512368
```

```
cor.test(((price^lamda) - 1)/lamda), banios)
```

Pearson's product-moment correlation

```
data: (((price^lamda) - 1)/lamda) and banios
t = 27.648, df = 4744, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to 0
```

```

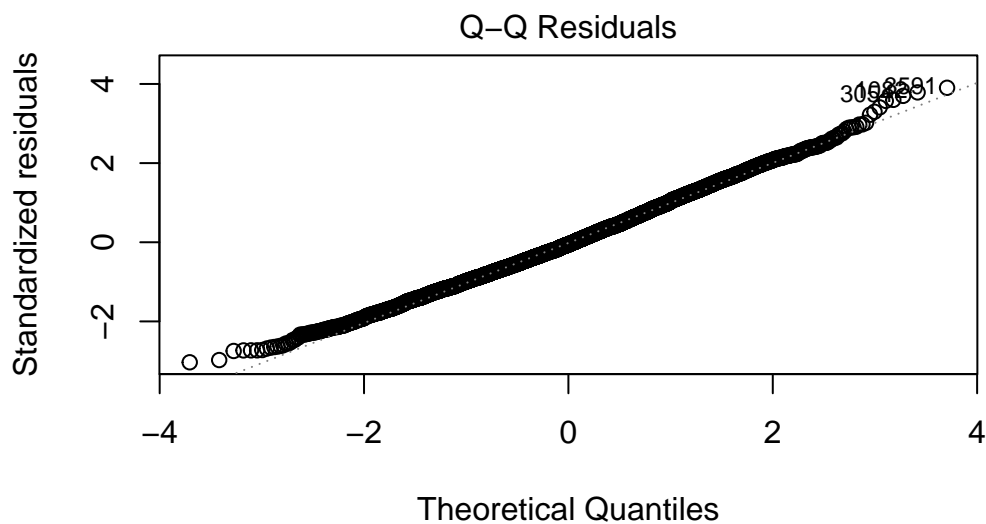
95 percent confidence interval:
 0.3477585 0.3967702
sample estimates:
      cor
0.3725241

```

Al analizar la correlación entre los precios y variables independientes, observamos que los p-value son menores a la significancia de 0.05, por lo que concluimos que hay linealidad en nuestro modelo.

Normalidad

```
plot(modelo_final, 2)
```



lm((((price^lamda) - 1)/lamda) ~ host_total_listings_count + latitude + lor

```
shapiro.test(modelo_final$residuals)
```

Shapiro-Wilk normality test

data: modelo_final\$residuals

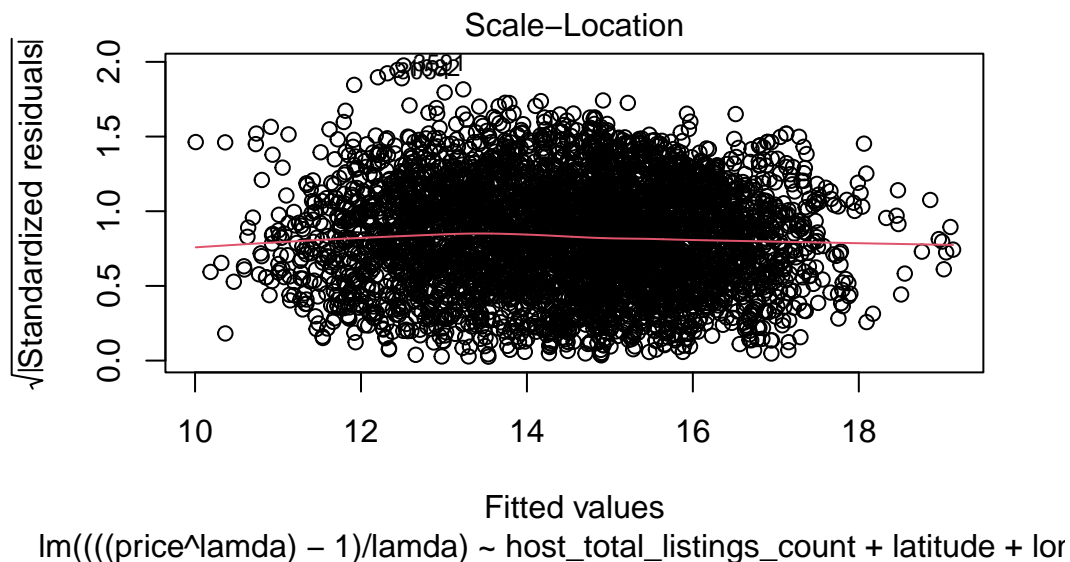
W = 0.9971, p-value = 6.541e-08

Para validar este supuesto, se realizó una prueba Shapiro-Wilk, notando que nuestro p-value < 0.05 , por lo que se rechaza la hipótesis nula: “ H_0 Los residuos siguen una distribución normal”, sin embargo, tras realizar un gráfico QQ podemos decir con seguridad que nuestro modelo si cumple con el supuesto de normalidad ya que nuestros datos se ajustan perfectamente a la gráfica.

Nota: Esta discrepancia suele darse cuando los datos se enfrentan a un reescalamiento severo, lo que hace que las pruebas de bondad de ajuste den falsos negativos o positivos.

Homocedasticidad

```
plot(modelo_final, 3)
```



Para validar este supuesto se realizó un gráfico de dispersión, observando que los puntos no siguen algún tipo de patrón, además de que la línea roja va de manera horizontal, por lo que nuestro modelo cumple con la Homocedasticidad.

Multicolinealidad

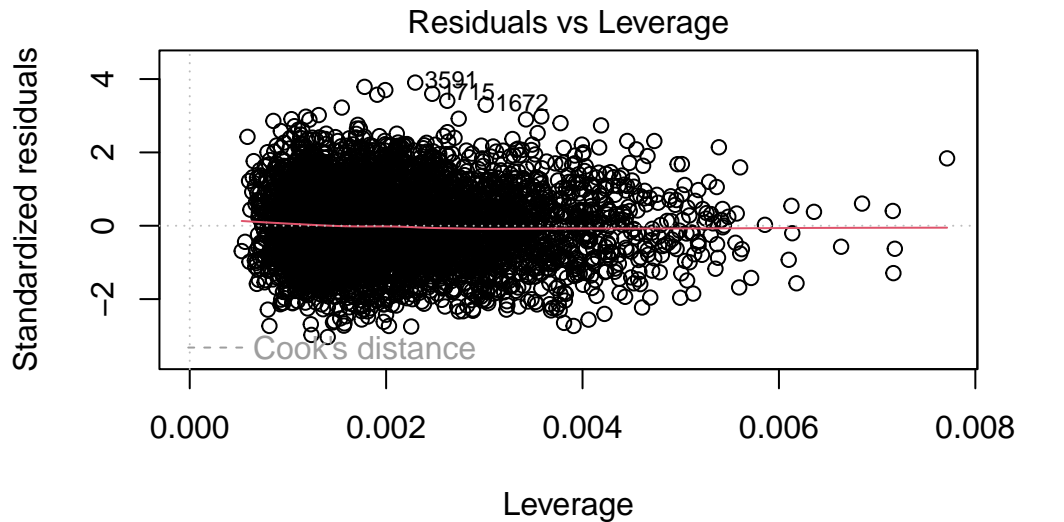
```
vif(modelo_final)
```

```
host_total_listings_count
1.026718
latitude
1.022833
longitude
1.041577
accommodates
1.573110
availability_365
1.014967
review_scores_cleanliness
1.074754
review_scores_location
1.098347
calculated_host_listings_count_private_rooms
1.291497
bancos
1.296549
```

Este supuesto fue validado prácticamente al inicio cuando las variables más fuertemente correlacionadas fueron eliminadas, corroborado por la función `vif` que nos ayuda a determinar si un modelo de regresión lineal presenta colinealidad o no, y, dado que los datos obtenidos para nuestras variables son menores a 5, podemos concluir que nuestro modelo cumple también con este supuesto.

Valores influyentes

```
plot(modelo_final, 5)
```



$\text{lm}(\text{(((price}^{\text{lamda}} - 1)/\text{lamda}) \sim \text{host_total_listings_count} + \text{latitude} + \text{lor}$

En un caso similar al anterior, al haber eliminado los valores atípicos del modelo y al haberlo acotado solamente a ciertos Airbnb's, nos podemos percatar de que este supuesto también se cumple.

Interpretación

```
summary(modelo_final)
```

Call:

```
lm(formula = (((price^lamda) - 1)/lamda) ~ host_total_listings_count +
  latitude + longitude + accommodates + availability_365 +
  +review_scores_cleanliness + review_scores_location + calculated_host_listings_count_pri
  banios)
```

Residuals:

Min	1Q	Median	3Q	Max
-4.1586	-0.9596	-0.0736	0.9105	5.3492

Coefficients:

	Estimate	Std. Error	t value
(Intercept)	-2.258e+03	1.235e+02	-18.287
host_total_listings_count	9.588e-02	7.004e-03	13.689
latitude	8.513e+00	6.668e-01	12.767
longitude	-2.102e+01	1.252e+00	-16.799
accommodates	4.465e-01	1.796e-02	24.853
availability_365	9.943e-04	1.591e-04	6.251
review_scores_cleanliness	1.467e+00	1.526e-01	9.611
review_scores_location	2.709e+00	2.858e-01	9.479
calculated_host_listings_count_private_rooms	-1.114e+00	3.417e-02	-32.600
banios	7.028e-01	5.460e-02	12.872

Pr(>|t|)

(Intercept)	< 2e-16 ***
host_total_listings_count	< 2e-16 ***
latitude	< 2e-16 ***
longitude	< 2e-16 ***
accommodates	< 2e-16 ***
availability_365	4.43e-10 ***
review_scores_cleanliness	< 2e-16 ***
review_scores_location	< 2e-16 ***
calculated_host_listings_count_private_rooms	< 2e-16 ***
banios	< 2e-16 ***

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

Residual standard error: 1.372 on 4736 degrees of freedom

Multiple R-squared: 0.5385, Adjusted R-squared: 0.5377

F-statistic: 614.1 on 9 and 4736 DF, p-value: < 2.2e-16

Tras haber realizado la validación de supuestos de nuestro modelo obtuvimos la siguiente ecuación:

$$y = \frac{price^\lambda - 1}{\lambda} = -225.8 + 0.09588 * (host\ total\ listings\ count) \\ + 8.513 * (latitude) - 21.02 * (longitude) + 0.4465 * (accommodates) + 0.0009943 * (availability\ 365) \\ + 1.467 * (review\ scores\ cleanliness) + 2.709 * (review\ scores\ location)$$

Viendo que nuestra R^2 ajustada es del 0.5377, implica que nuestro modelo nos explica en un 53.77% nuestro problema de predecir el precio de un Airbnb, y además es muy significativa, dado que nuestro p-value es menor que un nivel de significancia 0.05, así como también lo es en cada una de nuestras variables explicativas.

Finalmente, podemos concluir lo siguiente de nuestro modelo:

- Decimos que con cada 0.09588 de Airbnb's que posea el anfitrión el precio aumentará.
- Decimos por cada 8.513 que la latitud se mueva en la CDMX, el precio aumentará.
- Dado que la longitud en la CDMX es negativa, decimos que precio aumenta con 21.02 por cada vez la longitud se mueva.
- Decimos que aproximadamente por cada media persona que se hospede en el AIRBNB, el precio igual aumentará, por lo que una persona completa casi duplica el precio.
- Vemos que mientras casi no haya disponibilidad el precio aumentará en 0.0009943
- Vemos que por cada punto que se le da a la limpieza de un Airbnb, nuestro precio aumentará con una razón de 1.467
- Análogamente al punto anterior, mientras mayor sea la calificación que tenga por ubicación el Airbnb, mayor será el precio con una razón de 2.709