

Explore Airbnb listings in Paris, France, as at 04 March 2024*

Yimiao Yuan

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1 Introduction

This report analyzes the Airbnb listings in Paris, France, as at 04 March 2024 using R (R Core Team (2022)). The dataset is read from the Inside Airbnb (Cox (2021)) website, and cleaned and explored using `tidyverse` (Wickham et al. (2019)), `janitor` (Firke (2023)), `knitr` (Xie (2023)), `lubridate` (Grolemund and Wickham (2011)), `mice` (van Buuren and Groothuis-Oudshoorn (2011)), `modelsummary` (Arel-Bundock (2022)), `ggplot2` (Wickham (2016)) and `naniar` (Tierney and Cook (2023)). This report will explore distribution and properties of different variables and analyze the relationship between them.

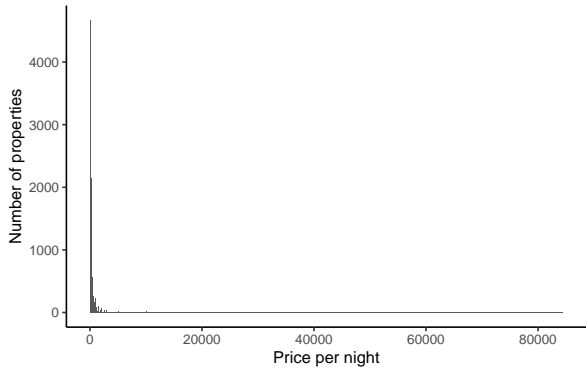
2 Distribution and Properties

The raw dataset is cleaned and used to create a parquet file with selected variables for exploratory purposes. I first look at the price of the Airbnb and plot the distribution in Figure 1. There are outliers on regular scale, so I also use a log scale to plot the distribution. From the graph, we can see that most properties have low prices, number of properties decreases as the price increases.

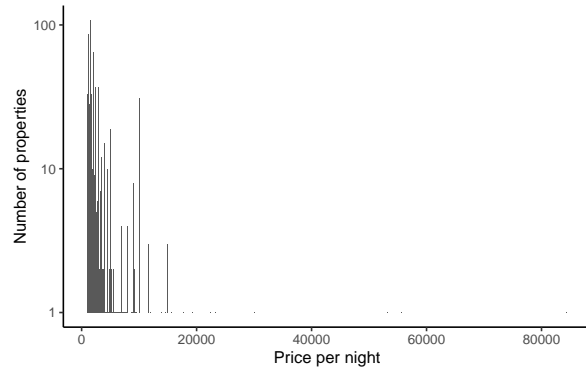
Then I turned my focus on price that less than \$1000, and the distribution contains more detailed. Figure 2a shows that there are some bunching in the price, so I choose to take a further look at prices between \$90 and \$250. Figure 2b also shows some bunching in the graph, which might due to the reason that prices around numbers ending in zero or nine.

Next I look at the superhosts in the Airbnb. After removing the NA value in superhost, I plot the distribution of review scores for all superhosts. Figure 3 shows that most of the superhosts' properties have 5 stars, they rarely get 1 or 2 stars.

*Code and data are available at: https://github.com/YimiaoYuan09/Airbnb_EDA_Paris

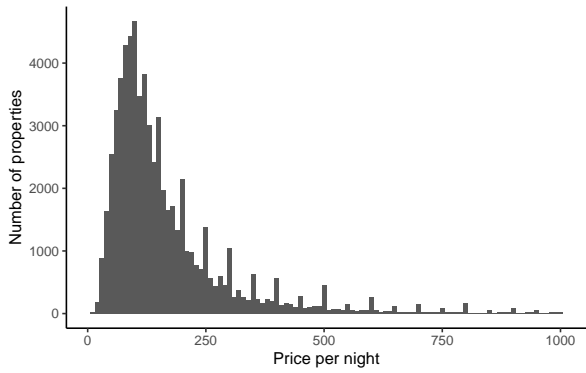


(a) Distribution of prices

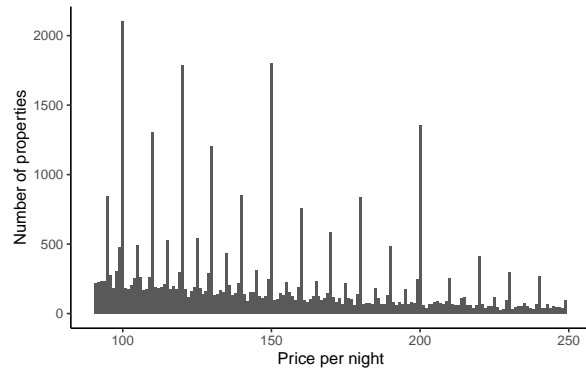


(b) Using the log scale for prices more than \$1,000

Figure 1: Distribution of prices of Paris Airbnb rentals in March 2024



(a) Prices less than \$1,000 suggest some bunching



(b) Prices between \$90 and \$250 illustrate the bunching more clearly

Figure 2: Distribution of prices of Paris Airbnb rentals in March 2024

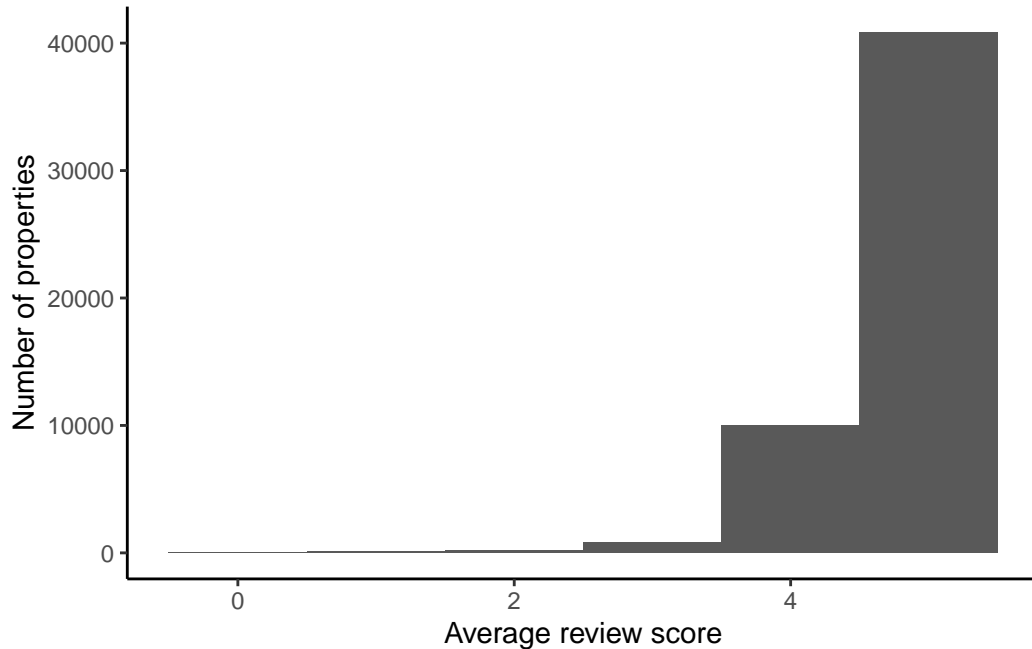


Figure 3: Distribution of review scores for Paris Airbnb rentals in March 2024

There are a lot of NA values in the review scores rating since they do not have enough reviews. For this report, I just remove NA values in review scores rating.

I also interested in the host response time, which is how quickly a host responds to an inquiry. There are also a lot of NAs in this variable, so I want to see that if there is a relationship with the review score. Figure 4 shows that most of properties get a rate over 4 stars. However, `ggplot2` drops a lot of missing values, so I use `geom_miss_point()` from `naniar` to include them in the graph (Figure 5).

I also interested in how many properties a host has on Airbnb. From Figure 6, we can see that most hosts have 1 properties, some hosts have less than 10 properties. Few of them have more than 100 properties, which is a little bit strange.

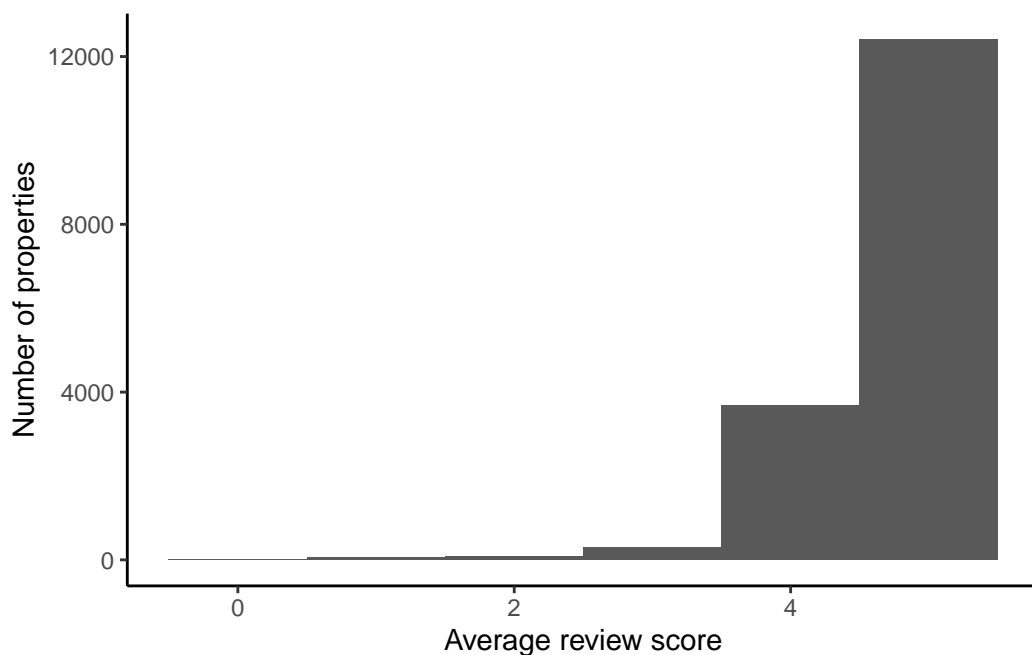


Figure 4: Distribution of review scores for properties with NA response time, for Paris Airbnb rentals in March 2024

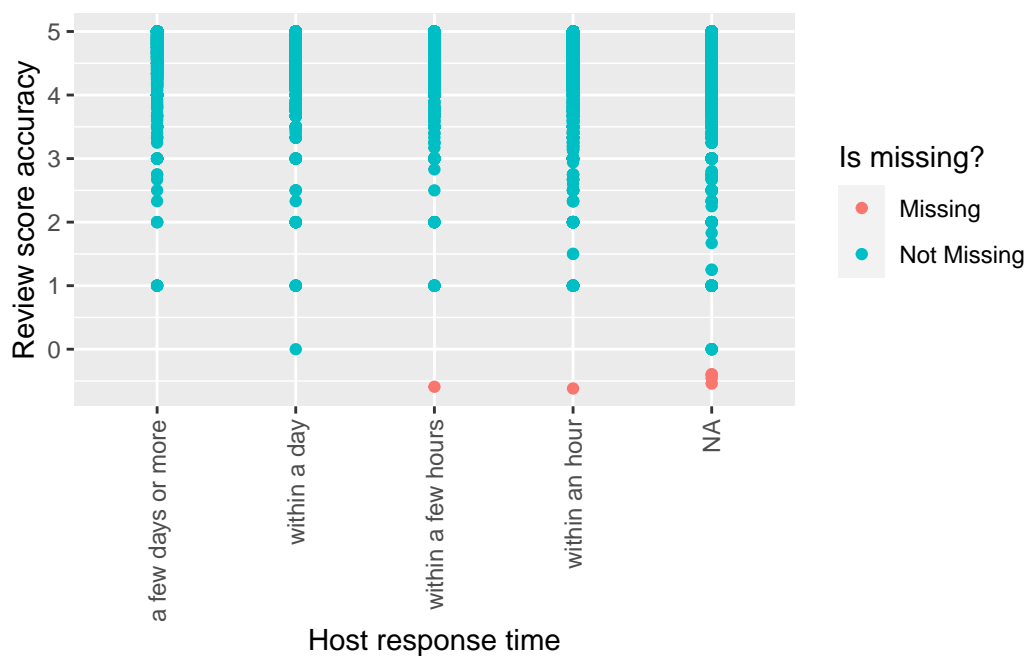
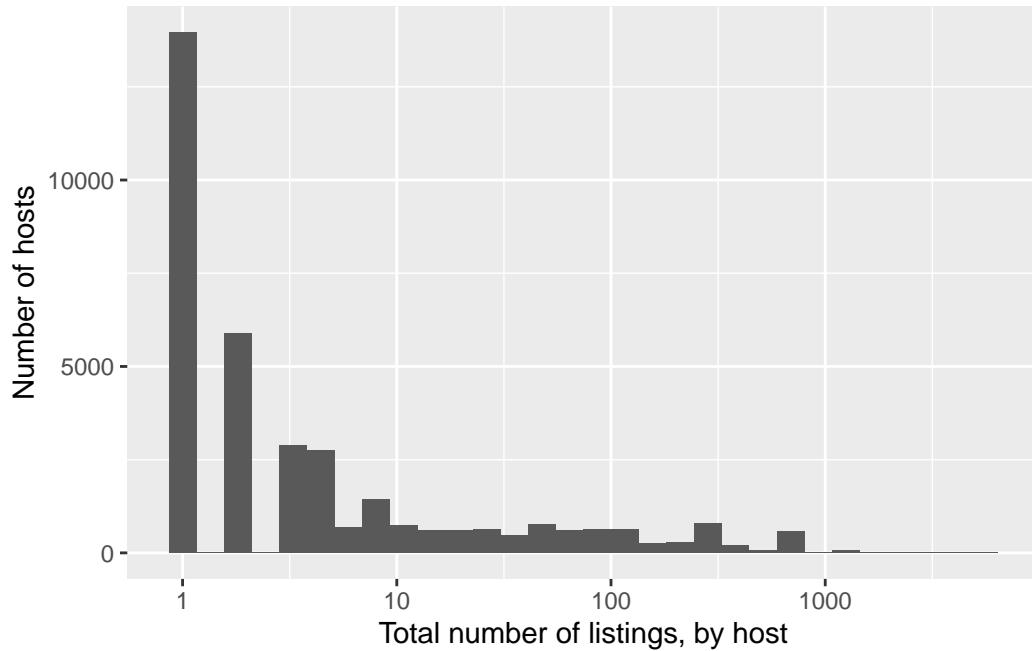


Figure 5: Missing values in Paris Airbnb data, by host response time



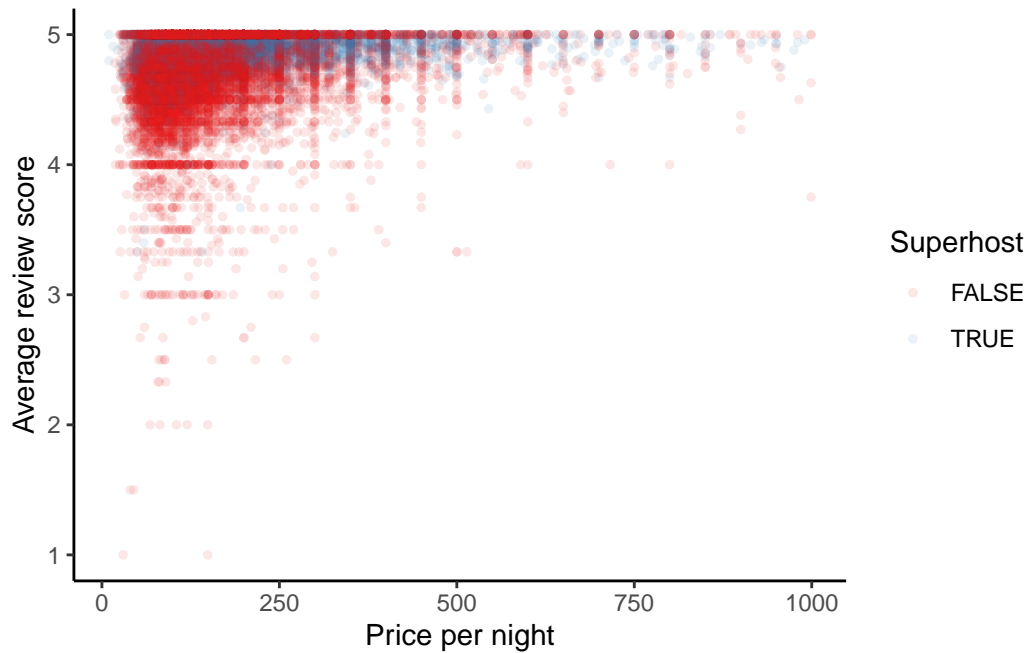


Figure 7: Relationship between price and review and whether a host is a superhost, for Paris Airbnb rentals in March 2024

4 Model

A model is run on the dataset to gain a better understanding of relationships between multiple variables. This model is going to predict whether someone is a superhost, using host response time and review scores rating. Since the output is a binary value, I use the logistic regression for this model. The result shows that each variable is positively associated with the probability of being a superhost.

	(1)
(Intercept)	−16.262 (0.481)
host_response_timewithin a day	2.019 (0.211)
host_response_timewithin a few hours	2.695 (0.210)
host_response_timewithin an hour	2.972 (0.209)
review_scores_rating	2.624 (0.089)
Num.Obs.	22 047
AIC	24 165.0
BIC	24 205.0
Log.Lik.	−12 077.507
RMSE	0.43

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