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

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#### 1 Download and Save Data

```
# exploratory: create a parquet file with selected variables
airbnb_select <-
  airbnb_raw |>
  select(
    host_id,
    host_response_time,
    host_is_superhost,
    host_total_listings_count,
    neighbourhood_cleansed,
    bathrooms,
    bedrooms,
    price,
    number_of_reviews,
    review_scores_rating,
    review_scores_accuracy,
    review_scores_value
  )
rm(airbnb_raw)
# save the parquet file
write_parquet(
```

<sup>\*</sup>Code and data are available at: https://github.com/YimiaoYuan09/Airbnb\_EDA\_Paris

```
x = airbnb_select,
sink =
   "../../inputs/data/2024-03-04-paris-airbnblistings-select_variables.parquet"
)
```

## 2 Distribution and properties of Price

```
airbnb_select$price |>
    head()
[1] "$150.00" "$146.00" "$110.00" "$140.00" "$180.00" "$71.00"
  # get character from price
  airbnb_select$price |>
   str_split("") |>
   unlist() |>
    unique()
 [1] "$" "1" "5" "0" "." "4" "6" "8" "7" "3" "2" "9" NA ","
  # , value
  airbnb_select |>
   select(price) |>
    filter(str_detect(price, ","))
# A tibble: 1,550 x 1
  price
  <chr>
1 $1,200.00
2 $8,000.00
3 $7,000.00
4 $1,997.00
5 $1,000.00
6 $1,286.00
7 $2,300.00
8 $1,500.00
9 $1,200.00
```

```
10 $1,357.00
# i 1,540 more rows
  # remove $
  airbnb_select <-
    airbnb_select |>
    mutate(
      price = str_remove_all(price, "[\\$,]"),
      price = as.integer(price)
  # distribution of price
  airbnb_select |>
    ggplot(aes(x = price)) +
    geom_histogram(binwidth = 10) +
    theme_classic() +
    labs(
      x = "Price per night",
      y = "Number of properties"
    )
  # distribution of price on log scale
  airbnb_select |>
    filter(price > 1000) |>
    ggplot(aes(x = price)) +
    geom_histogram(binwidth = 10) +
    theme_classic() +
    labs(
      x = "Price per night",
      y = "Number of properties"
    scale_y_log10()
  # focus on price < 1000
  airbnb_select |>
    filter(price < 1000) |>
    ggplot(aes(x = price)) +
    geom_histogram(binwidth = 10) +
    theme_classic() +
    labs(
      x = "Price per night",
```

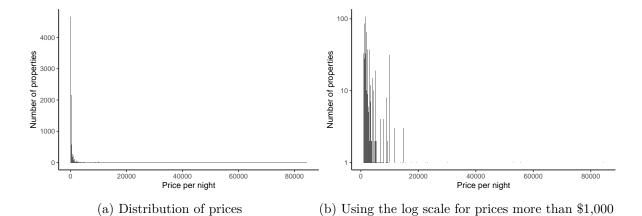
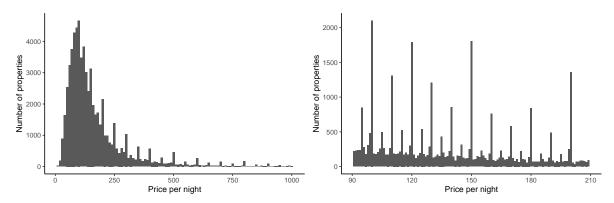


Figure 1: Distribution of prices of Paris Airbnb rentals in March 2024

```
y = "Number of properties"
)

airbnb_select |>
  filter(price > 90) |>
  filter(price < 210) |>
  ggplot(aes(x = price)) +
  geom_histogram(binwidth = 1) +
  theme_classic() +
  labs(
    x = "Price per night",
    y = "Number of properties"
)
```

### 3 Distribution and properties of Superhost



(a) Prices less than \$1,000 suggest some bunching (b) Prices between \$90 and \$210 illustrate the bunching more clearly

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

```
2 5869840 within a few hours NA
                                                                          7
3 35125972 within an hour
                                                                           3
4 13827149 within a few hours NA
                                                                           3
5 62919059 within a few hours NA
                                                                           3
                                                                           2
6 22167607 N/A
                                                                           2
7 10259782 N/A
                               NA
                                                                           3
8 62919059 within a few hours NA
                                                                           4
9 20056470 N/A
                               NA
10 20056470 N/A
                               NA
# i 73 more rows
# i 8 more variables: neighbourhood_cleansed <chr>, bathrooms <lgl>,
    bedrooms <dbl>, price <int>, number_of_reviews <dbl>,
#
   review_scores_rating <dbl>, review_scores_accuracy <dbl>,
    review_scores_value <dbl>
  # NA in review scores rating
  airbnb_data_no_superhost_na |>
    filter(is.na(review_scores_rating)) |>
    nrow()
```

#### [1] 13497

```
airbnb_data_no_superhost_na |>
  filter(is.na(review_scores_rating)) |>
  select(number_of_reviews) |>
```

```
table()
```

```
number_of_reviews
     0
13497
```

```
# no NA in review scores rating
airbnb_data_no_superhost_na |>
  filter(!is.na(review_scores_rating)) |>
  ggplot(aes(x = review_scores_rating)) +
  geom_histogram(binwidth = 1) +
  theme_classic() +
  labs(
    x = "Average review score",
    y = "Number of properties"
)
```

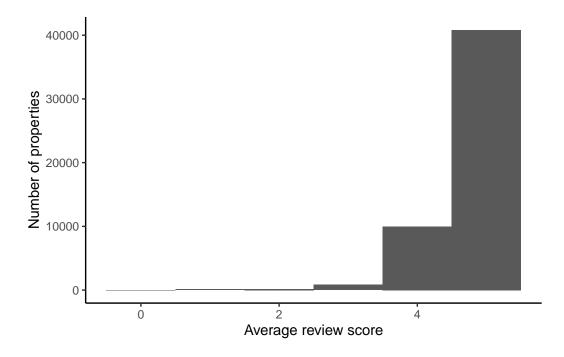


Figure 3: Distribution of review scores for Paris Airbnb rentals in March 2024

```
# host response time
  airbnb_data_has_reviews |>
    count(host_response_time)
# A tibble: 6 x 2
 host_response_time
  <chr>>
                     <int>
1 N/A
                     16531
2 a few days or more 1243
3 within a day
                      5297
4 within a few hours 6811
5 within an hour
                     22094
6 <NA>
  # host response time NA
  # relationship with review scores rating
  airbnb_data_has_reviews |>
    filter(is.na(host_response_time)) |>
    ggplot(aes(x = review_scores_rating)) +
    geom_histogram(binwidth = 1) +
    theme_classic() +
    labs(
      x = "Average review score",
      y = "Number of properties"
```

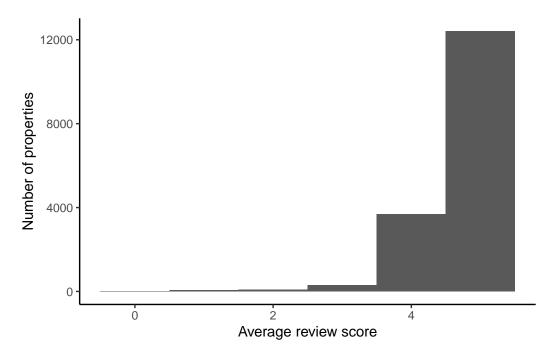


Figure 4: Distribution of review scores for properties with NA response time, for Paris Airbnb rentals in March 2024

```
# include missing data
airbnb_data_has_reviews |>
    ggplot(aes(
        x = host_response_time,
        y = review_scores_accuracy
)) +
    geom_miss_point() +
    labs(
        x = "Host response time",
        y = "Review score accuracy",
        color = "Is missing?"
    ) +
    theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```

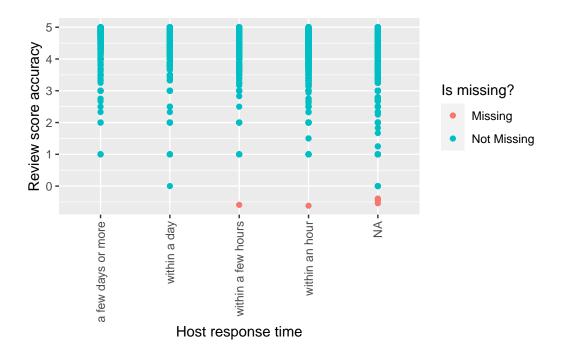


Figure 5: Missing values in Paris Airbnb data, by host response time

```
# remove NA in host_response_time
# superhost, has revies, has response time
airbnb_select <-
   airbnb_data_has_reviews |>
   filter(!is.na(host_response_time))
```

## 4 Distribution and properties of Host Properties

```
# how many properties a host has on Airbnb
airbnb_select |>
    ggplot(aes(x = host_total_listings_count)) +
    geom_histogram() +
    scale_x_log10() +
    labs(
        x = "Total number of listings, by host",
        y = "Number of hosts"
    )
```

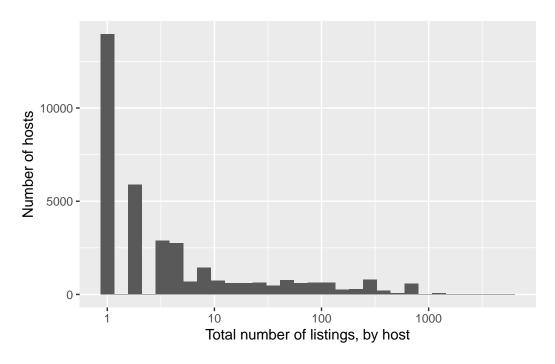


Figure 6: Distribution of the number of properties a host has on Airbnb, for Paris Airbnb rentals in March 2024

```
# host with number of listings > 500
airbnb_select |>
  filter(host_total_listings_count >= 500) |>
  head()
```

#### # A tibble: 6 x 13

```
host_id host_response_time host_is_superhost host_total_listings_count
     <dbl> <fct>
                                                                      <dbl>
                               <lgl>
1 50502817 within an hour
                               FALSE
                                                                        778
2 50502817 within an hour
                               FALSE
                                                                        778
3 50502817 within an hour
                               FALSE
                                                                        778
4 50502817 within an hour
                               FALSE
                                                                        778
5 50502817 within an hour
                               FALSE
                                                                        778
6 50502817 within an hour
                               FALSE
                                                                        778
```

- # i 9 more variables: neighbourhood\_cleansed <chr>>, bathrooms <lgl>,
- # bedrooms <dbl>, price <int>, number\_of\_reviews <dbl>,
- # review\_scores\_rating <dbl>, review\_scores\_accuracy <dbl>,
- # review\_scores\_value <dbl>, host\_is\_superhost\_binary <dbl>

```
# focus on host with only 1 property
airbnb_select <-
   airbnb_select |>
   add_count(host_id) |>
   filter(n == 1) |>
   select(-n)
```

## 5 Relationship between prices and reviews, superhosts, number of properties, neighborhood

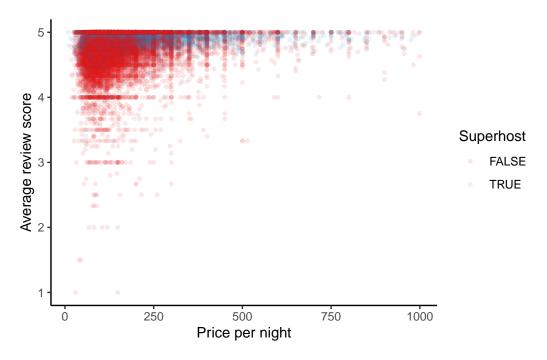


Figure 7: Relationship between price and review and whether a host is a superhost, for Paris Airbnb rentals in March 2024

```
# proportion of superhost
  airbnb_select |>
    count(host_is_superhost) |>
    mutate(
      proportion = n / sum(n),
      proportion = round(proportion, digits = 2)
    )
# A tibble: 2 x 3
 host_is_superhost
                        n proportion
  <lgl>
                                <dbl>
                    <int>
1 FALSE
                    15820
                                 0.72
2 TRUE
                     6227
                                0.28
  # host response time, by superhost
  airbnb_select |>
    tabyl(host_response_time, host_is_superhost) |>
    adorn_percentages("col") |>
```

```
adorn_pct_formatting(digits = 0) |>
   adorn_ns() |>
   adorn_title()
                   host_is_superhost
                               FALSE
                                            TRUE
host_response_time
                         6%
                               (953) 0%
a few days or more
                                             (24)
      within a day
                         22% (3,511) 12%
                                            (770)
within a few hours
                         24% (3,802) 26% (1,614)
    within an hour
                         48% (7,554) 61% (3,819)
 # neighbourhood
 airbnb_select |>
   tabyl(neighbourhood_cleansed) |>
   adorn_pct_formatting() |>
   arrange(-n) |>
   filter(n > 100) |>
   adorn totals("row") |>
   head()
neighbourhood_cleansed
                          n percent
     Buttes-Montmartre 2842
                              12.9%
            Popincourt 2202
                              10.0%
              Entrepôt 1713
                               7.8%
             Vaugirard 1681
                               7.6%
                             6.5%
          Ménilmontant 1438
       Buttes-Chaumont 1430
                               6.5%
```

#### 6 Model

```
# forecast whether someone is a superhost
# logistic regression
# affected by faster responses and better reviews
logistic_reg_superhost_response_review <-
    glm(
    host_is_superhost ~
    host_response_time +
    review_scores_rating,
    data = airbnb_select,</pre>
```

	(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.	22047
AIC	24165.0
BIC	24205.0
Log.Lik.	-12077.507
RMSE	0.43

```
family = binomial
)

modelsummary(logistic_reg_superhost_response_review)
```

## 7 Save Analysis Dataset

```
# save analysis data
write_parquet(
    x = airbnb_select,
    sink = "../../outputs/data/2024-03-04-paris-airbnblistings-analysis_dataset.parquet"
)
```