

Airbnb Dataset Analysis

Abdul

2025-03-31

```
# Load the dataset
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)
```

Task 1: Structure of the Dataset

```
# Load the dataset
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)

# Print structure of the dataset
str(airbnb_data)
```

```
## 'data.frame': 102599 obs. of 26 variables:
## $ id : int 1001254 1002102 1002403 1002755 1003689 1004098 1004650 1005...
## $ NAME : chr "Clean & quiet apt home by the park" "Skylit Midtown Castle"
## $ host.id : num 8.00e+10 5.23e+10 7.88e+10 8.51e+10 9.20e+10 ...
## $ host_identity_verified : chr "unconfirmed" "verified" "" "unconfirmed" ...
## $ host.name : chr "Madaline" "Jenna" "Elise" "Garry" ...
## $ neighbourhood.group : chr "Brooklyn" "Manhattan" "Manhattan" "Brooklyn" ...
## $ neighbourhood : chr "Kensington" "Midtown" "Harlem" "Clinton Hill" ...
## $ lat : num 40.6 40.8 40.8 40.7 40.8 ...
## $ long : num -74 -74 -73.9 -74 -73.9 ...
## $ country : chr "United States" "United States" "United States" "United States" ...
## $ country.code : chr "US" "US" "US" "US" ...
## $ instant_bookable : logi FALSE FALSE TRUE TRUE FALSE TRUE ...
## $ cancellation_policy : chr "strict" "moderate" "flexible" "moderate" ...
## $ room.type : chr "Private room" "Entire home/apt" "Private room" "Entire home" ...
## $ Construction.year : int 2020 2007 2005 2005 2009 2013 2015 2009 2005 2015 ...
## $ price : chr "$966 " "$142 " "$620 " "$368 " ...
## $ service.fee : chr "$193 " "$28 " "$124 " "$74 " ...
## $ minimum.nights : int 10 30 3 30 10 3 45 45 2 2 ...
## $ number.of.reviews : int 9 45 0 270 9 74 49 49 430 118 ...
## $ last.review : chr "10/19/2021" "5/21/2022" "" "7/5/2019" ...
## $ reviews.per.month : num 0.21 0.38 NA 4.64 0.1 0.59 0.4 0.4 3.47 0.99 ...
## $ review.rate.number : int 4 4 5 4 3 3 5 5 3 5 ...
## $ calculated.host.listings.count : int 6 2 1 1 1 1 1 1 1 1 ...
## $ availability.365 : int 286 228 352 322 289 374 224 219 180 375 ...
## $ house_rules : chr "Clean up and treat the home the way you'd like your home to be" ...
## $ license : chr "" "" "" "" ...
```

AirbnbData_Analysis

Trupti

2025-03-31

```
# Set global options for code chunks
```

```
knitr::opts_chunk$set(echo = TRUE)
```

```
library(readr)
```

```
#Load the dataset using read_csv from readr
```

```
data <- read_csv("C:/Users/dtrup/OneDrive/Desktop/Airbnb_Open_Data.csv")
```

```
## Warning: One or more parsing issues, call 'problems()' on your data frame for details,
```

```
## e.g.:
```

```
##   dat <- vroom(...)
```

```
##   problems(dat)
```

```
## Rows: 102599 Columns: 26
```

```
## -- Column specification -----
```

```
## Delimiter: ","
```

```
## chr (13): NAME, host_identity_verified, host name, neighbourhood group, neig...
```

```
## dbl (11): id, host id, lat, long, Construction year, minimum nights, number ...
```

```
## lgl (2): instant_bookable, license
```

```
##
```

```
## i Use 'spec()' to retrieve the full column specification for this data.
```

```
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
# List the variables (columns) in the dataset
```

```
print(colnames(data))
```

```
## [1] "id" "NAME"
## [3] "host id" "host_identity_verified"
## [5] "host name" "neighbourhood group"
## [7] "neighbourhood" "lat"
## [9] "long" "country"
## [11] "country code" "instant_bookable"
## [13] "cancellation_policy" "room type"
## [15] "Construction year" "price"
## [17] "service fee" "minimum nights"
## [19] "number of reviews" "last review"
## [21] "reviews per month" "review rate number"
## [23] "calculated host listings count" "availability 365"
## [25] "house_rules" "license"
```

```
# Print the top 15 rows of the dataset
top_15_rows <- as.data.frame(head(data, 15))
print(top_15_rows)
```

```
##      id                                     NAME      host id
## 1  1001254      Clean & quiet apt home by the park 80014485718
## 2  1002102      Skylit Midtown Castle 52335172823
## 3  1002403      THE VILLAGE OF HARLEM....NEW YORK ! 78829239556
## 4  1002755      <NA> 85098326012
## 5  1003689 Entire Apt: Spacious Studio/Loft by central park 92037596077
## 6  1004098      Large Cozy 1 BR Apartment In Midtown East 45498551794
## 7  1004650      BlissArtsSpace! 61300605564
## 8  1005202      BlissArtsSpace! 90821839709
## 9  1005754      Large Furnished Room Near B'way 79384379533
## 10 1006307      Cozy Clean Guest Room - Family Apt 75527839483
## 11 1006859      Cute & Cozy Lower East Side 1 bdrm 1280143094
## 12 1007411      Beautiful 1br on Upper West Side 18824631834
## 13 1007964      Central Manhattan/near Broadway 88136055909
## 14 1008516      Lovely Room 1, Garden, Best Area, Legal rental 26802410424
## 15 1009068 Wonderful Guest Bedroom in Manhattan for SINGLES 88920244552
##      host_identity_verified host name neighbourhood group      neighbourhood
## 1      unconfirmed      Madaline      Brooklyn      Kensington
## 2      verified      Jenna      Manhattan      Midtown
## 3      <NA>      Elise      Manhattan      Harlem
## 4      unconfirmed      Garry      Brooklyn      Clinton Hill
## 5      verified      Lyndon      Manhattan      East Harlem
## 6      verified      Michelle      Manhattan      Murray Hill
## 7      <NA>      Alberta      Brooklyn Bedford-Stuyvesant
## 8      unconfirmed      Emma      Brooklyn Bedford-Stuyvesant
## 9      verified      Evelyn      Manhattan      Hell's Kitchen
## 10     unconfirmed      Carl      Manhattan      Upper West Side
## 11     verified      Miranda      Manhattan      Chinatown
## 12     verified      Alan      Manhattan      Upper West Side
## 13     verified      <NA>      Manhattan      Hell's Kitchen
## 14     verified      Darcy      brookln      South Slope
## 15     verified      Leonardo      Manhattan      Upper West Side
##      lat      long      country country code instant_bookable
## 1  40.64749 -73.97237 United States      US      FALSE
## 2  40.75362 -73.98377 United States      US      FALSE
## 3  40.80902 -73.94190 United States      US      TRUE
## 4  40.68514 -73.95976 United States      US      TRUE
## 5  40.79851 -73.94399 United States      US      FALSE
## 6  40.74767 -73.97500 United States      US      TRUE
## 7  40.68688 -73.95596 United States      US      FALSE
## 8  40.68688 -73.95596 United States      US      FALSE
## 9  40.76489 -73.98493 United States      US      TRUE
## 10 40.80178 -73.96723 United States      US      FALSE
## 11 40.71344 -73.99037 United States      US      FALSE
## 12 40.80316 -73.96545 United States      US      TRUE
## 13 40.76076 -73.98867 United States      US      FALSE
## 14 40.66829 -73.98779 United States      US      TRUE
## 15 40.79826 -73.96113 United States      US      FALSE
##      cancellation_policy      room type Construction year      price service fee
```

## 1	strict	Private room	2020	\$966	\$193
## 2	moderate	Entire home/apt	2007	\$142	\$28
## 3	flexible	Private room	2005	\$620	\$124
## 4	moderate	Entire home/apt	2005	\$368	\$74
## 5	moderate	Entire home/apt	2009	\$204	\$41
## 6	flexible	Entire home/apt	2013	\$577	\$115
## 7	moderate	Private room	2015	\$71	\$14
## 8	moderate	Private room	2009	\$1,060	\$212
## 9	strict	Private room	2005	\$1,018	\$204
## 10	strict	Private room	2015	\$291	\$58
## 11	flexible	Entire home/apt	2004	\$319	\$64
## 12	flexible	Entire home/apt	2008	\$606	\$121
## 13	strict	Private room	2008	\$714	\$143
## 14	moderate	Private room	2010	\$580	\$116
## 15	flexible	Private room	2019	\$149	\$30

##	minimum nights	number of reviews	last review	reviews per month
----	----------------	-------------------	-------------	-------------------

## 1	10	9	10/19/2021	0.21
## 2	30	45	5/21/2022	0.38
## 3	3	0	<NA>	NA
## 4	30	270	7/5/2019	4.64
## 5	10	9	11/19/2018	0.10
## 6	3	74	6/22/2019	0.59
## 7	45	49	10/5/2017	0.40
## 8	45	49	10/5/2017	0.40
## 9	2	430	6/24/2019	3.47
## 10	2	118	7/21/2017	0.99
## 11	1	160	6/9/2019	1.33
## 12	5	53	6/22/2019	0.43
## 13	2	188	6/23/2019	1.50
## 14	4	167	6/24/2019	1.34
## 15	2	113	7/5/2019	0.91

##	review rate	number calculated	host listings	count	availability	365
----	-------------	-------------------	---------------	-------	--------------	-----

## 1	4			6		286
## 2	4			2		228
## 3	5			1		352
## 4	4			1		322
## 5	3			1		289
## 6	3			1		374
## 7	5			1		224
## 8	5			1		219
## 9	3			1		180
## 10	5			1		375
## 11	3			4		1
## 12	4			1		163
## 13	4			1		258
## 14	4			3		47
## 15	3			1		68

##

1
2
3
4
5
6

```

## 7
## 8 House Guidelines for our BnB We are delighted to welcome you. Check in Sun - Thurs by 8PM and Fr
## 9
## 10
## 11
## 12
## 13
## 14
## 15
##      license
## 1      NA
## 2      NA
## 3      NA
## 4      NA
## 5      NA
## 6      NA
## 7      NA
## 8      NA
## 9      NA
## 10     NA
## 11     NA
## 12     NA
## 13     NA
## 14     NA
## 15     NA

```

```

---
Title: "Task 4 - AirBnb Data Analysis"
Author: Akshat Singh
Date: "2025-03-31"
Output:
  pdf_document: default
  df_print: paged
output: pdf_document
---

```

```

```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```

```

R Markdown

This is Task 4 of the AirBnb Data Analysis our group has been conducting. In this task we had to choose a variable of our choice from the dataset and define a user defined function for the same variable.

We have used the Construction Year as our variable to print information of all the properties listed and constructed after 2020.

Below is the code for the same Task.

```

```{r cars}
getwd()
Airbnb_Open_Data <- read.csv("/cloud/project/Airbnb_Open_Data.csv", header=FALSE)
recent_properties <- as.numeric(Airbnb_Open_Data$V15)
recent_properties <- Airbnb_Open_Data[Airbnb_Open_Data$V15 > 2020,]
print(recent_properties)
```

```

****Title: "Task 5 - AirBnb Data Analysis"****

In this task we were supposed to filter data regarding cancellation policy based on the location of the property.

For this output we had set a condition for Brooklyn and hence it showed us the date of the property as well as the cancellation policy of the particular BnB.

The code for the function is given below.

```

```{r cars2}
install.packages("dplyr")
library(dplyr)
Airbnb_Open_Data$V15 <- as.numeric(Airbnb_Open_Data$V15)
brooklyn_new_properties <- Airbnb_Open_Data %>%
 filter(Airbnb_Open_Data$V15 > 2010, Airbnb_Open_Data$V6 == "Brooklyn") %>%
 select(V15,V13)
print(brooklyn_new_properties)
```

```

Including Plots

```
` `{r pressure, echo=FALSE}  
plot(pressure)  
` }
```

Note that the ``echo = FALSE`` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

Task 6

Abdul Ali

2025-04-01

```
## Task 2: Identify Variables and Reshape the Data
```

```
# Load the necessary libraries
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
# Set working directory and load the dataset
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)

# Convert price to numeric
airbnb_data$price <- gsub("\\$", "", airbnb_data$price) # Remove $
airbnb_data$price <- gsub(",", "", airbnb_data$price)  # Remove commas
airbnb_data$price <- as.numeric(airbnb_data$price)     # Convert to number

# Select variables and remove NA
selected_data <- airbnb_data %>%
  select(price, Construction.year, room.type, number.of.reviews, country) %>%
  na.omit()

# Reshape data - average price by room type and country
reshaped_data <- selected_data %>%
  group_by(room.type, country) %>%
  summarise(Average_Price = mean(price, na.rm = TRUE), .groups = "drop")

# Display the first few rows
head(reshaped_data)
```

```
## # A tibble: 6 x 3
##   room.type      country Average_Price
```


| ## | <chr> | <chr> | <dbl> |
|------|-----------------|-----------------|-------|
| ## 1 | Entire home/apt | " " | 604. |
| ## 2 | Entire home/apt | "United States" | 625. |
| ## 3 | Hotel room | " " | 859 |
| ## 4 | Hotel room | "United States" | 663. |
| ## 5 | Private room | " " | 621. |
| ## 6 | Private room | "United States" | 625. |

```
---
title: "AirbnbDataset_Analysis"
author: "Sohini Biswas"
date: "2025-03-30"
output: html_document
---
```

```
```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```
```

```
## R Markdown
```

When you click the ****Knit**** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
```{r cars}
summary(cars)
```
```

```
## Including Plots
```

You can also embed plots, for example:

```
```{r pressure, echo=FALSE}
plot(pressure)
```
```

```
# AUTHOR - SOHINI BISWAS
```

```
setwd("C:\\Users\\sohin\\Downloads")
```

```
getwd()
```

```
airbnb_data <- read.csv("airbnb_Data.csv", check.names = FALSE)
```

```
# TASK 1 : Remove Missing Values in the Dataset
# Check for missing values in each column
missing_values_per_column <- colSums(is.na(airbnb_data))
print("Missing values per column:")
print(missing_values_per_column)
```

```
# Total missing values in the entire dataset
total_missing_values <- sum(is.na(airbnb_data))
print(paste("Total missing values:", total_missing_values))
```

```
# Remove rows with any missing values
airbnb_cleaned_rows <- na.omit(airbnb_data)
```

```
# Remove columns with any missing values
airbnb_cleaned_cols <- airbnb_data[, colSums(is.na(airbnb_data)) == 0]
```

```
# TASK 2 : Identify and remove Duplicate Values in the Dataset
# Identify duplicate rows
duplicate_rows <- duplicated(airbnb_data)
```

```
# Count the number of duplicate rows
number_of_duplicate_rows <- sum(duplicate_rows)
print(paste("Number of duplicate rows:", number_of_duplicate_rows))
```

```
# View the duplicate rows
```

```
print("Duplicate rows:")
print(airbnb_data[duplicate_rows, ])

# Remove duplicate rows
airbnb_unique <- airbnb_data[!duplicate_rows, ]

# Verify removal by counting rows before and after
print(paste("Number of rows before removing duplicates:", nrow(airbnb_data)))
print(paste("Number of rows after removing duplicates:", nrow(airbnb_unique)))
```

Airbnb Dataset Analysis

Sonali

2025-03-30

##Introduction

In this assignment, we are performing a data analysis on an Airbnb dataset using R Programming. The objective of this assignment is to apply basic data manipulation techniques on the dataset. We will perform two tasks: 1. Reorder the dataset rows in descending order based on a numeric column. 2. Rename selected column names for better clarity.

##Load Libraries and Dataset

In this section, we will load the required library and the Airbnb dataset. We will use the `dplyr` package for data manipulation and load the dataset from the local folder.

```
# Install dplyr package (only one time)
# install.packages("dplyr")
```

```
# Load Library
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
# Load dataset
airbnb_data <- read.csv("Airbnb_Sample.csv.csv", check.names = FALSE)

# View column names
colnames(airbnb_data)
```

```
## [1] "id" "NAME"
## [3] "host id" "host_identity_verified"
## [5] "host name" "neighbourhood group"
## [7] "neighbourhood" "lat"
## [9] "long" "country"
## [11] "country code" "instant_bookable"
## [13] "cancellation_policy" "room type"
## [15] "Construction year" "price"
## [17] "service fee" "minimum nights"
## [19] "number of reviews" "last review"
## [21] "reviews per month" "review rate number"
## [23] "calculated host listings count" "availability 365"
## [25] "house_rules" "license"
```

Task 1: Reorder Multiple Rows in Descending Order

In this task, we will reorder the rows of the Airbnb dataset in descending order based on the `price` column. This will help us to see the listings with the highest prices at the top of the dataset.

```
# Reorder rows in descending order by price
airbnb_sorted <- airbnb_data %>%
  arrange(desc(price))

# View top 10 rows after sorting
head(airbnb_sorted, 10)
```

| ## | id | NAME | host id |
|-------|---------|--|-------------|
| ## 1 | 1248766 | Hancock Town House!-Stuyvesant Mews | 17302896128 |
| ## 2 | 1068717 | 3 Story Town House in Park Slope | 9853816495 |
| ## 3 | 1536515 | Great room priv/bathrm Eastside location 70's ST | 33469116543 |
| ## 4 | 1445938 | Lovely Hell's Kitchen Studio... | 91331960557 |
| ## 5 | 1551427 | Stay with a Jazz Singer in Harlem! | 38298412868 |
| ## 6 | 1201821 | Colorful Artistic Williamsburg Apt | 59139947513 |
| ## 7 | 1173101 | 1BR: See Central Park from Terrace! | 948625792 |
| ## 8 | 1363645 | At a very nice area in the WestSide | 27086952528 |
| ## 9 | 1526021 | Eastern Parkway Brooklyn 1BR Flat | 55368535774 |
| ## 10 | 1511661 | Beautiful Room in a Beautiful New NYC Apartment | 58156347387 |

| ## | host_identity_verified | host name | neighbourhood | group | neighbourhood |
|-------|------------------------|-----------|---------------|-------|--------------------|
| ## 1 | unconfirmed | Lucia | | | Bedford-Stuyvesant |
| ## 2 | | Darcy | Brooklyn | | South Slope |
| ## 3 | verified | Perry | Manhattan | | Upper East Side |
| ## 4 | verified | Scott | Manhattan | | Hell's Kitchen |
| ## 5 | verified | Taylor | Manhattan | | Harlem |
| ## 6 | verified | Parker | Brooklyn | | Williamsburg |
| ## 7 | verified | Howard | Manhattan | | Upper West Side |
| ## 8 | unconfirmed | Bennett | Manhattan | | Washington Heights |
| ## 9 | verified | Holmes | Brooklyn | | Crown Heights |
| ## 10 | unconfirmed | Edwards | Manhattan | | Harlem |

| ## | lat | long | country | country code | instant_bookable |
|-------|----------|-----------|---------------|--------------|------------------|
| ## 1 | 40.68669 | -73.91989 | United States | US | False |
| ## 2 | 40.66499 | -73.97925 | United States | US | True |
| ## 3 | 40.76850 | -73.96034 | United States | US | False |
| ## 4 | 40.76147 | -73.99152 | United States | US | True |
| ## 5 | 40.80192 | -73.95827 | United States | US | True |
| ## 6 | 40.71125 | -73.95613 | United States | US | True |
| ## 7 | 40.77728 | -73.97818 | United States | US | False |
| ## 8 | 40.85099 | -73.92822 | United States | US | False |
| ## 9 | 40.66847 | -73.94875 | United States | US | False |
| ## 10 | 40.83177 | -73.95000 | United States | US | True |

| ## | cancellation_policy | room type | Construction year | price | service fee |
|-------|---------------------|-----------------|-------------------|-------|-------------|
| ## 1 | moderate | Private room | 2013 | \$999 | \$200 |
| ## 2 | strict | Entire home/apt | 2021 | \$996 | \$199 |
| ## 3 | moderate | Private room | 2008 | \$996 | \$199 |
| ## 4 | strict | Entire home/apt | 2016 | \$993 | \$199 |
| ## 5 | flexible | Private room | 2022 | \$993 | \$199 |
| ## 6 | strict | Private room | 2011 | \$99 | \$20 |
| ## 7 | moderate | Entire home/apt | 2008 | \$989 | \$198 |
| ## 8 | flexible | Private room | 2021 | \$986 | \$197 |
| ## 9 | flexible | Entire home/apt | 2013 | \$984 | \$197 |
| ## 10 | strict | Private room | 2003 | \$983 | \$197 |

| ## | minimum nights | number of reviews | last review | reviews per month |
|------|----------------|-------------------|-------------|-------------------|
| ## 1 | 1 | 27 | 10/8/2017 | 0.28 |
| ## 2 | 2 | 16 | 12/30/2018 | 0.24 |
| ## 3 | 7 | 93 | 6/7/2019 | 1.06 |
| ## 4 | 2 | 72 | 7/31/2018 | 0.82 |
| ## 5 | 1 | 0 | | NA |
| ## 6 | 3 | 31 | 3/1/2019 | 0.31 |
| ## 7 | 13 | 38 | 2/15/2019 | 0.39 |

```
## 8          60          4    7/3/2015          0.06
## 9           5          1    9/4/2014          0.02
## 10          1         56   7/16/2018          0.64
##  review rate number calculated host listings count availability 365
## 1           1           4           181
## 2          NA          NA           146
## 3           3           1           270
## 4           4           1           125
## 5           3           1           160
## 6           2           1           419
## 7           3           1           105
## 8          NA           1           355
## 9           3           1           212
## 10          5           1           425
```

```
##
house_rules
```

```
## 1
```

```
## 2
```

```
## 3 House Rules 1. Check-in is 4 pm local time. If the unit is ready earlier, we'll let you know. Check-out is normally 11 am local time, but we'd be happy to extend it as long as we don't have a cleaning scheduled. Just let us know. 2. All bookings require a security deposit of at least $300, which will be refunded within 7 days of your check-out. 3. For security measures we require all guests to provide proof of identification through ID verification on our own website. In order to check in we'll need a photo of your ID. 4. Our cancellation policy is as shown on our ad and defined by the site you are booking through. 5. Don't let $300 go up in smoke. There's no smoking allowed in any Flatbook and a $300 fine for breaking this rule. 6. Unfortunately we don't allow pets in any of our apartments. 7. Unless you're staying in one of our specialty apartments, we don't allow parties or excessive noise. 8. If we find the place very messy, we have to charge an extra $40 for every extra hour o
```

```
## 4
```

```
Respect our place please.
```

```
## 5
```

```
No smoking in the apartment, in the foyer or outside the building at any time. No pets. No parties.
```

```
## 6
```

```
Maximum of 2 guests staying in a bedroom. No smoking anywhere in or outside the property. No pets anywhere in or outside the property. No parties anywhere in or outside the property. No parking on our property or in the lot. No frying food. No extreme temperature settings. Any broken house rules, including evidence of cigarette butts, ashes, odors, stained linens, damaged furniture, or missing items will result in loss of deposit. All doors must be locked, and all lights, fans, air conditioners, and heaters should be turned off each time you leave. We kindly ask that you respect our property and our neighbors in this wonderful community.
```

```
## 7
```

```
No drugs, no smoking, no pets (our dog is enough pet for us). Make sure the front door is shut securely when you come and go, as our dog Lucy likes to sneak out and explore when we're not looking! No additional guests or visitors without prior permission. If you'd like to have guests over please ask us first.
```

```
## 8 Check in/out: We can meet outside of my building at check-in between 4:00pm-6:00pm. If you can give me an estimated ETA, that would be helpful! Please leave your keys in the studio when you check-out Thursday afternoon. I am very flexible with check-out between 12:00pm-3:00pm. Go up 1 flight of stairs, and it will be the first door to the right. The door locks automatically, but there is a deadbolt and lock. You do not have to deadbolt the door every time you leave
```

e. TV - Press and hold the "All On" button to turn on TV Towels are on the top shelf in the bathroom, and the bed is all set! I do have a portable A/C in the second window on the right. Since I live on the second floor, it does not get too warm, but if you do need it, please remember to turn it off if you are leaving the building. Trash/recycling days are Sundays and Thursdays. There's trash bags on the third shelf in the bathroom closet to the left. When you are checking out on Thursday, please take the trash out r

```
## 9
```

```
## 10
```

```
NO Smoking
```

```
## license
```

```
## 1 NA
```

```
## 2 NA
```

```
## 3 NA
```

```
## 4 NA
```

```
## 5 NA
```

```
## 6 NA
```

```
## 7 NA
```

```
## 8 NA
```

```
## 9 NA
```

```
## 10 NA
```

Task 2: Rename Some of the Column Names

In this task, we will rename some of the column names in the Airbnb dataset to make them clearer.

The following changes will be made: - NAME will be renamed to listing_name - price will be renamed to listing_price - host id will be renamed to host_id_number

```
# Rename selected column names
airbnb_renamed <- airbnb_data %>%
  rename(
    listing_name = NAME,
    listing_price = price,
    host_id_number = `host id`
  )

# View new column names
colnames(airbnb_renamed)
```



```
## [1] "id" "listing_name"
## [3] "host_id_number" "host_identity_verified"
## [5] "host name" "neighbourhood group"
## [7] "neighbourhood" "lat"
## [9] "long" "country"
## [11] "country code" "instant_bookable"
## [13] "cancellation_policy" "room type"
## [15] "Construction year" "listing_price"
## [17] "service fee" "minimum nights"
## [19] "number of reviews" "last review"
## [21] "reviews per month" "review rate number"
## [23] "calculated host listings count" "availability 365"
## [25] "house_rules" "license"
```

Conclusion

In this assignment, we successfully performed basic data manipulation tasks using R Programming on the Airbnb dataset.

We reordered the dataset rows in descending order based on the `price` column and renamed selected column names to improve clarity.

New calculated variable

Ritvik Pande

2025-03-31

```
# Load the ds
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)

# Removing the dollar sign and converting to numeric for both columns
airbnb_data$price <- as.numeric(gsub("$", "", airbnb_data$price))
airbnb_data$service.fee <- as.numeric(gsub("$", "", airbnb_data$service.fee))

# Performing mathematical calculation and adding to new variable
airbnb_data$total_unit_price <- airbnb_data$price + airbnb_data$service.fee

# Printing the result
# Check the result
head(airbnb_data[, c("price", "service.fee", "total_unit_price")])
```

```
##   price service.fee total_unit_price
## 1   966         193         1159
## 2   142          28          170
## 3   620         124          744
## 4   368          74          442
## 5   204          41          245
## 6   577         115          692
```

Airbnb Data Analysis - Training set using random number generator engine & Summary statistics

Aneesh

2025-03-31

Introduction:

This document demonstrates how to create a training set from the Airbnb dataset using a random number generator engine and provides summary statistics of the dataset.

Create Training and Test Sets:

```
library(dplyr) # Load dplyr if you haven't already

## Warning: package 'dplyr' was built under R version 4.4.3

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

# Load dataset
Airbnb_Open_Data <- read.csv("Airbnb_Open_Data.csv", check.names = FALSE)

# Clean the 'price' column and create 'price_numeric'
Airbnb_Open_Data <- Airbnb_Open_Data %>%
  mutate(price_numeric = gsub("$,", "", price),
         price_numeric = as.numeric(price_numeric))

# Determine the training set size (e.g., 70% for training)
set.seed(123) # For reproducibility
train_size <- floor(0.70 * nrow(Airbnb_Open_Data))
train_indices <- sample(1:nrow(Airbnb_Open_Data), train_size)

# Create the training set
```

```
train_data <- Airbnb_Open_Data[train_indices, ]

# Create the test set
test_data <- Airbnb_Open_Data[-train_indices, ]

# Verify the number of rows in each set
cat("Number of rows in training set:", nrow(train_data), "\n")
```

```
## Number of rows in training set: 71819
```

```
cat("Number of rows in test set:", nrow(test_data), "\n")
```

```
## Number of rows in test set: 30780
```

```
cat("Number of rows in test set:", nrow(test_data), "\n")
```

```
## Number of rows in test set: 30780
```

Summary Statistics:

```
##          id          NAME          host id
## Min.      : 1001254  Length:102599  Min.      :1.236e+08
## 1st Qu.:15085814  Class :character  1st Qu.:2.458e+10
## Median :29136603  Mode  :character  Median :4.912e+10
## Mean    :29146235                      Mean    :4.925e+10
## 3rd Qu.:43201198                      3rd Qu.:7.400e+10
## Max.    :57367417                      Max.    :9.876e+10
##
## host_identity_verified host name      neighbourhood group
## Length:102599          Length:102599  Length:102599
## Class :character       Class :character  Class :character
## Mode  :character       Mode  :character  Mode  :character
##
##
##
## neighbourhood      lat      long      country
## Length:102599      Min.    :40.50  Min.    : -74.25  Length:102599
## Class :character    1st Qu.:40.69  1st Qu.: -73.98  Class :character
## Mode  :character    Median :40.72  Median : -73.95  Mode  :character
##                      Mean    :40.73  Mean    : -73.95
##                      3rd Qu.:40.76  3rd Qu.: -73.93
##                      Max.    :40.92  Max.    : -73.71
##                      NA's    :8      NA's    :8
##
## country code      instant_bookable cancellation_policy room type
## Length:102599      Mode :logical  Length:102599  Length:102599
## Class :character    FALSE:51474  Class :character  Class :character
## Mode  :character    TRUE :51020  Mode  :character  Mode  :character
##                      NA's :105
##
##
```

```
##
## Construction year      price      service fee      minimum nights
## Min.      :2003      Length:102599      Length:102599      Min.      : -1223.000
## 1st Qu.   :2007      Class :character      Class :character      1st Qu.   :  2.000
## Median    :2012      Mode  :character      Mode  :character      Median    :  3.000
## Mean      :2012                                     Mean     :  8.136
## 3rd Qu.   :2017                                     3rd Qu.  :  5.000
## Max.      :2022                                     Max.     : 5645.000
## NA's      :214                                     NA's     : 409
## number of reviews last review      reviews per month review rate number
## Min.      :  0.00      Length:102599      Min.      : 0.010      Min.      :1.000
## 1st Qu.   :  1.00      Class :character      1st Qu.   : 0.220      1st Qu.   :2.000
## Median    :  7.00      Mode  :character      Median    : 0.740      Median    :3.000
## Mean      : 27.48                                     Mean     : 1.374      Mean     :3.279
## 3rd Qu.   : 30.00                                     3rd Qu.  : 2.000      3rd Qu.  :4.000
## Max.      :1024.00                                     Max.     :90.000      Max.     :5.000
## NA's      :183                                     NA's     :15879      NA's     :326
## calculated host listings count availability 365 house_rules
## Min.      :  1.000                                     Min.      : -10.0      Length:102599
## 1st Qu.   :  1.000                                     1st Qu.   :  3.0      Class :character
## Median    :  1.000                                     Median    : 96.0      Mode  :character
## Mean      :  7.937                                     Mean     : 141.1
## 3rd Qu.   :  2.000                                     3rd Qu.  : 269.0
## Max.      :332.000                                     Max.     :3677.0
## NA's      :319                                     NA's     :448
## license      price_numeric
## Length:102599      Min.      :  50.0
## Class :character      1st Qu.   : 340.0
## Mode  :character      Median    : 624.0
##                                     Mean     : 625.3
##                                     3rd Qu.  : 913.0
##                                     Max.     :1200.0
##                                     NA's     :247
```

Description of Create Training, Test Sets and Summary Statistics :

The code first loads the Airbnb dataset and prepares it for analysis by cleaning the price data. It then splits the data into training and test sets using a random selection process. The training set will be used to build a machine learning model, while the test set will be used to evaluate its performance. Finally, the summary() function is used to display descriptive statistics for all columns in the original dataset, providing an overview of its distribution and characteristics.

Conclusion:

In conclusion, the code performs essential data preparation steps on the Airbnb dataset. The price data is cleaned, and the dataset is split into training and testing sets to facilitate model development and evaluation. The summary statistics offer key insights into the dataset's characteristics.

Statistical Functions

Ritvik Pande

2025-03-31

```
# Load the ds
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)
```

```
# Converting to numeric
minimum.nights_1 <- as.numeric(airbnb_data$minimum.nights)
head(minimum.nights_1)
```

```
## [1] 10 30 3 30 10 3
```

```
# Mean of minimum nights
mean_1 <- mean(minimum.nights_1, na.rm = TRUE)
print(paste("Mean: ", mean_1))
```

```
## [1] "Mean: 8.13584499461787"
```

```
# Median of minimum nights
median_1 <- median(minimum.nights_1, na.rm = TRUE)
print(paste("Median: ", median_1))
```

```
## [1] "Median: 3"
```

```
# Mode of minimum nights
mode_1 <- function(x) {
  uniq_x <- unique(x)
  uniq_x[which.max(tabulate(match(x, uniq_x)))]
}
```

```
mode_1 <- mode_1(minimum.nights_1)
print(paste("Mode: ", mode_1))
```

```
## [1] "Mode: 1"
```

```
# Range of minimum nights
range_1 <- range(minimum.nights_1, na.rm = TRUE)
range_1_value <- diff(range_1)
print(paste("Range: ", range_1_value))
```

```
## [1] "Range: 6868"
```

Airbnb Data Analysis - Scatter Plot & Bar Plot

Deepak Kumar

2025-03-31

Introduction:

This document explores two aspects of the Airbnb dataset: the relationship between the price of listings and the number of reviews they receive, and the comparison of average prices across different room types.

Scatter Plot: Price vs Reviews per Month:

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.4.3
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 4.4.3
```

```
##
```

```
## Attaching package: 'dplyr'
```

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

```
##
```

```
## filter, lag
```

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

```
##
```

```
## intersect, setdiff, setequal, union
```

```
library(scales)
```

```
## Warning: package 'scales' was built under R version 4.4.3
```

```
# Load dataset
```

```
Airbnb_Open_Data <- read.csv("Airbnb_Open_Data.csv", check.names = FALSE)
```

```
# Clean the 'price' column and create 'price_numeric'
```

```
Airbnb_Open_Data <- Airbnb_Open_Data %>%  
  mutate(price_numeric = gsub("$", "", price),
```

```

price_numeric = as.numeric(price_numeric))

# create the scatter plot
ggplot(Airbnb_Open_Data, aes(x = price_numeric, y = `reviews per month`, color = `room type`)) +
  geom_jitter(alpha = 0.2, width = 10, height = 1) +
  labs(title = "Scatter plot of Price vs Reviews per Month",
       x = "Price ($)",
       y = "Reviews per Month",
       color = "Room Type") +
  theme_minimal() +
  scale_color_manual(values = c("Entire home/apt" = "red",
                                "Hotel room" = "blue",
                                "Private room" = "green",
                                "Shared room" = "black")) +

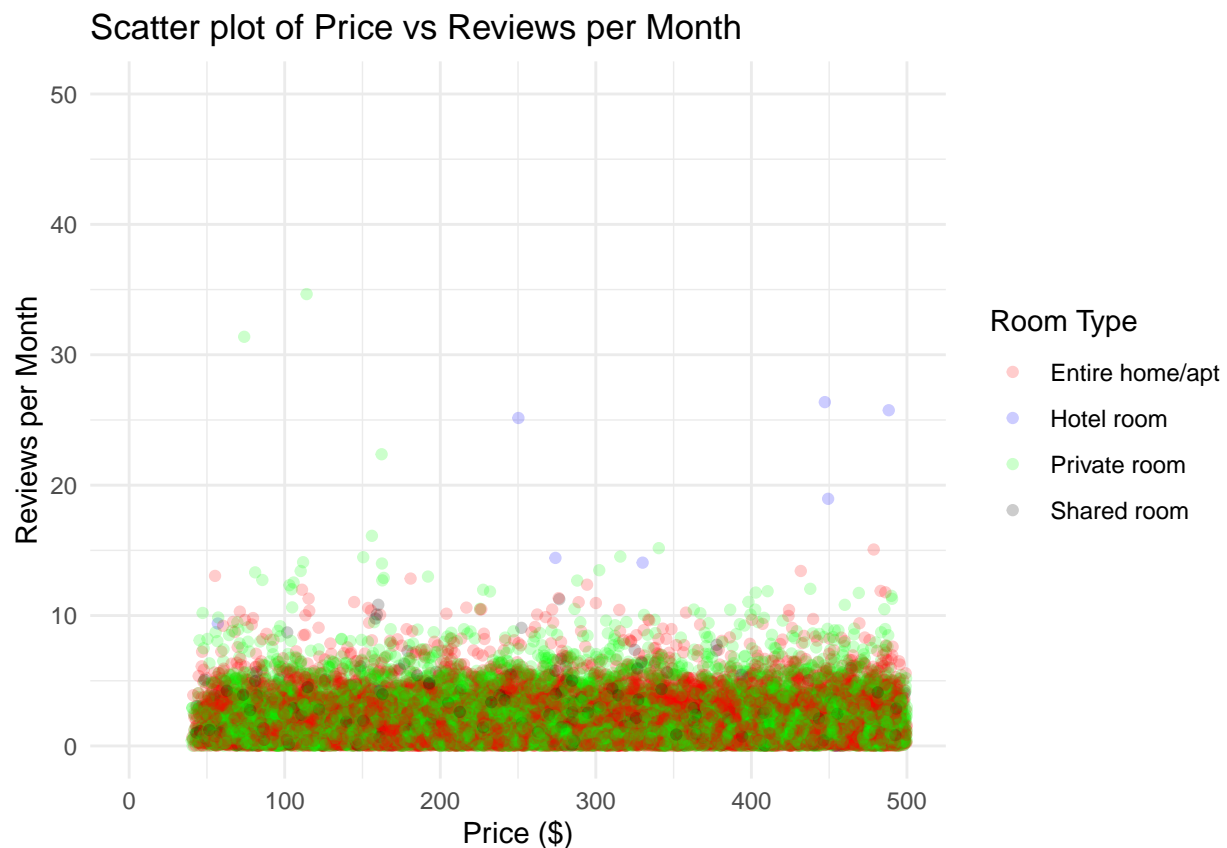
  xlim(0, 500) +
  ylim(0, 50)

```

```

## Warning: Removed 75361 rows containing missing values or values outside the scale range
## ('geom_point()').

```

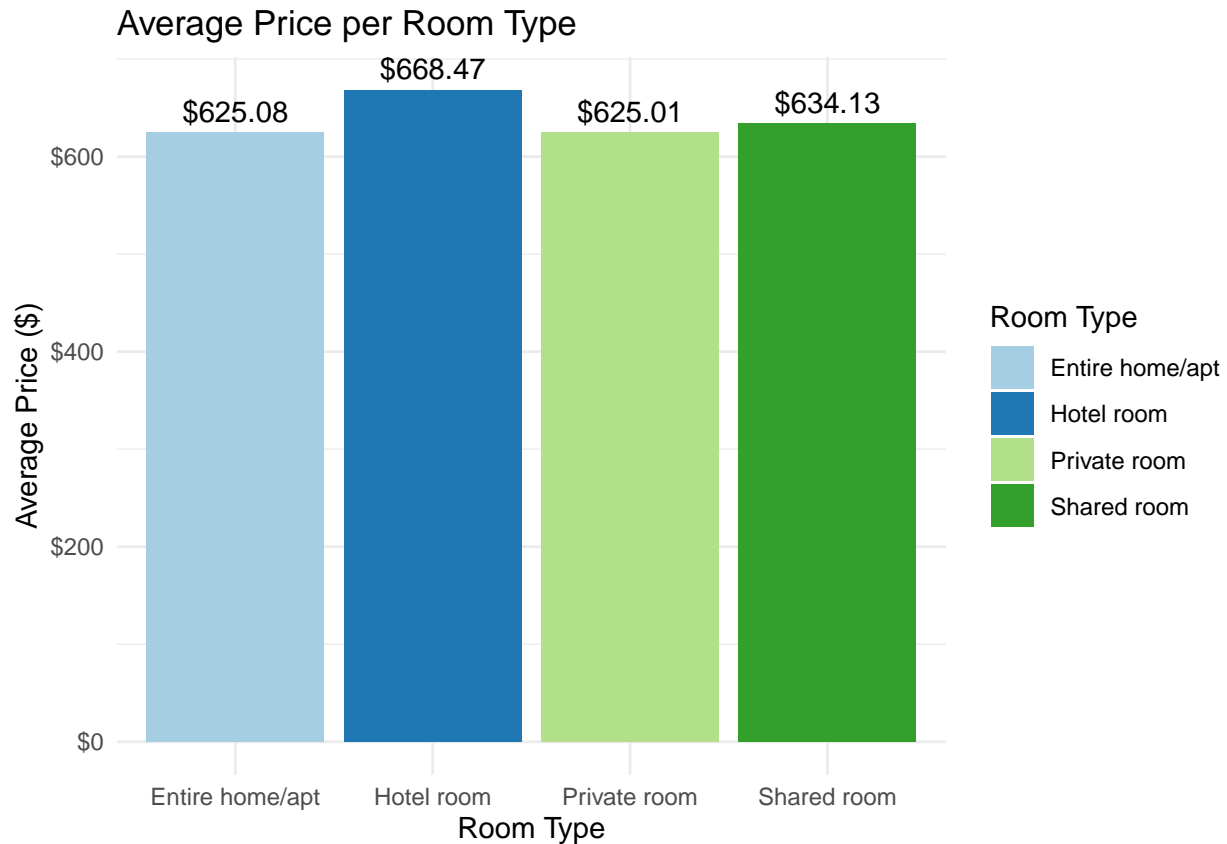


Description of the Scatter Plot:

This scatter plot visualizes the relationship between the price of Airbnb listings and the number of reviews they've received, focusing on listings priced between \$0 and \$500 and with 0 to 50 reviews per month. Each

point represents an individual Airbnb listing, and the color of the point indicates the type of room being offered: red for entire homes/apartments, blue for hotel rooms, green for private rooms, and black for shared rooms. The points are slightly jittered to help distinguish listings with similar price and review counts, especially in densely populated areas of the plot.

Bar Plot: Average Price per Room Type:



Description of the Bar Plot:

This bar plot displays the average price for different types of Airbnb listings: Entire home/apt, Hotel room, Private room, and Shared room. The height of each bar represents the average price in dollars, with the exact average price shown as a label above each bar. The colors differentiate the room types, as indicated by the legend on the right. This visualization allows for a quick comparison of the average listing price across different accommodation types.

Conclusion:

This analysis provides insights into the relationship between listing price and reviews, as well as the average prices for different room types on Airbnb.

Pearson Correlation

Ritvik Pande

2025-03-31

```
# Load the ds
setwd("C:/Users/ritvi/Downloads/Airbnb data")
airbnb_data <- read.csv("Airbnb_Open_Data.csv", stringsAsFactors = FALSE)
```

```
# Converting to numeric
Construction.year_1 <- as.numeric(airbnb_data$Construction.year)
head(Construction.year_1)
```

```
## [1] 2020 2007 2005 2005 2009 2013
```

```
review.rate.number_1 <- as.numeric(airbnb_data$review.rate.number)
head(review.rate.number_1)
```

```
## [1] 4 4 5 4 3 3
```

```
# Pearson correlation between property Construction year and Review ratings
correlation_value <- cor(Construction.year_1, review.rate.number_1, use = "complete.obs", method = "pearson")

# Print the correlation value
print(paste("Pearson Correlation: ", correlation_value))
```

```
## [1] "Pearson Correlation: 0.00475278600332765"
```