

datacleaning-1

September 18, 2024

Data Cleaning-New York City Airbnb

The purpose of this data cleaning project is to refine and enhance the New York City Airbnb Open Data to ensure it is accurate, reliable, and ready for insightful analysis.

```
[1]: #importing libraries
```

```
[2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
```

```
[3]: Airbnb_dataset=pd.read_csv('AB_NYC_2019.csv')
```

```
[4]: Airbnb_dataset.head(10)
```

```
[4]:
```

	id		name	host_id	\
0	2539		Clean & quiet apt home by the park	2787	
1	2595		Skylit Midtown Castle	2845	
2	3647		THE VILLAGE OF HARLEM...NEW YORK !	4632	
3	3831		Cozy Entire Floor of Brownstone	4869	
4	5022	Entire Apt: Spacious Studio/Loft by central park		7192	
5	5099	Large Cozy 1 BR Apartment In Midtown East		7322	
6	5121		BlissArtsSpace!	7356	
7	5178		Large Furnished Room Near B'way	8967	
8	5203		Cozy Clean Guest Room - Family Apt	7490	
9	5238		Cute & Cozy Lower East Side 1 bdrm	7549	

	host_name	neighbourhood_group	neighbourhood	latitude	longitude	\
0	John	Brooklyn	Kensington	40.64749	-73.97237	
1	Jennifer	Manhattan	Midtown	40.75362	-73.98377	
2	Elisabeth	Manhattan	Harlem	40.80902	-73.94190	
3	LisaRoxanne	Brooklyn	Clinton Hill	40.68514	-73.95976	
4	Laura	Manhattan	East Harlem	40.79851	-73.94399	
5	Chris	Manhattan	Murray Hill	40.74767	-73.97500	
6	Garon	Brooklyn	Bedford-Stuyvesant	40.68688	-73.95596	
7	Shunichi	Manhattan	Hell's Kitchen	40.76489	-73.98493	

8	MaryEllen	Manhattan	Upper West Side	40.80178	-73.96723
9	Ben	Manhattan	Chinatown	40.71344	-73.99037

	room_type	price	minimum_nights	number_of_reviews	last_review	\
0	Private room	149	1	9	2018-10-19	
1	Entire home/apt	225	1	45	2019-05-21	
2	Private room	150	3	0	NaN	
3	Entire home/apt	89	1	270	2019-07-05	
4	Entire home/apt	80	10	9	2018-11-19	
5	Entire home/apt	200	3	74	2019-06-22	
6	Private room	60	45	49	2017-10-05	
7	Private room	79	2	430	2019-06-24	
8	Private room	79	2	118	2017-07-21	
9	Entire home/apt	150	1	160	2019-06-09	

	reviews_per_month	calculated_host_listings_count	availability_365
0	0.21	6	365
1	0.38	2	355
2	NaN	1	365
3	4.64	1	194
4	0.10	1	0
5	0.59	1	129
6	0.40	1	0
7	3.47	1	220
8	0.99	1	0
9	1.33	4	188

```
[5]: print("Number of rows and columns",Airbnb_dataset.shape)
```

Number of rows and columns (48895, 16)

```
[6]: Airbnb_dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 48895 entries, 0 to 48894
```

```
Data columns (total 16 columns):
```

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	id	48895 non-null	int64
1	name	48879 non-null	object
2	host_id	48895 non-null	int64
3	host_name	48874 non-null	object
4	neighbourhood_group	48895 non-null	object
5	neighbourhood	48895 non-null	object
6	latitude	48895 non-null	float64
7	longitude	48895 non-null	float64
8	room_type	48895 non-null	object
9	price	48895 non-null	int64

```

10  minimum_nights          48895 non-null  int64
11  number_of_reviews       48895 non-null  int64
12  last_review             38843 non-null  object
13  reviews_per_month       38843 non-null  float64
14  calculated_host_listings_count  48895 non-null  int64
15  availability_365         48895 non-null  int64
dtypes: float64(3), int64(7), object(6)
memory usage: 6.0+ MB

```

Data cleaning

1)checking for data types

```
[7]: Airbnb_dataset.dtypes
```

```

[7]: id                int64
     name              object
     host_id           int64
     host_name         object
     neighbourhood_group  object
     neighbourhood     object
     latitude          float64
     longitude         float64
     room_type         object
     price             int64
     minimum_nights    int64
     number_of_reviews int64
     last_review       object
     reviews_per_month float64
     calculated_host_listings_count  int64
     availability_365  int64
     dtype: object

```

column “last_review” has a datatype object which needs to be converted to date

```

[8]: # Convert 'last_review' column to datetime
     Airbnb_dataset['last_review'] = pd.to_datetime(Airbnb_dataset['last_review'],
     ↪errors='coerce')

```

```
[9]: Airbnb_dataset.dtypes
```

```

[9]: id                int64
     name              object
     host_id           int64
     host_name         object
     neighbourhood_group  object
     neighbourhood     object
     latitude          float64
     longitude         float64

```

```

room_type                object
price                    int64
minimum_nights            int64
number_of_reviews         int64
last_review              datetime64[ns]
reviews_per_month         float64
calculated_host_listings_count int64
availability_365          int64
dtype: object

```

2)checking for duplicate values

```
[10]: Airbnb_dataset.duplicated().sum()
```

```
[10]: 0
```

no duplicates in dataset

3)checking for consistency

The strings in the dataset are inconsistent in case, so we need to convert them all to lowercase for uniformity.

```
[11]: Airbnb_dataset = Airbnb_dataset.map(lambda x: x.lower() if isinstance(x, str)
↪else x)

Airbnb_dataset.head()
```

```
[11]:
```

	id	name	host_id	\
0	2539	clean & quiet apt home by the park	2787	
1	2595	skylit midtown castle	2845	
2	3647	the village of harlem...new york !	4632	
3	3831	cozy entire floor of brownstone	4869	
4	5022	entire apt: spacious studio/loft by central park	7192	

	host_name	neighbourhood_group	neighbourhood	latitude	longitude	\
0	john	brooklyn	kensington	40.64749	-73.97237	
1	jennifer	manhattan	midtown	40.75362	-73.98377	
2	elisabeth	manhattan	harlem	40.80902	-73.94190	
3	lisaroxanne	brooklyn	clinton hill	40.68514	-73.95976	
4	laura	manhattan	east harlem	40.79851	-73.94399	

	room_type	price	minimum_nights	number_of_reviews	last_review	\
0	private room	149	1	9	2018-10-19	
1	entire home/apt	225	1	45	2019-05-21	
2	private room	150	3	0	NaT	
3	entire home/apt	89	1	270	2019-07-05	
4	entire home/apt	80	10	9	2018-11-19	

	reviews_per_month	calculated_host_listings_count	availability_365
0	0.21	6	365
1	0.38	2	355
2	NaN	1	365
3	4.64	1	194
4	0.10	1	0

4)checking for null values

```
[12]: Airbnb_dataset.isna().sum()
```

```
[12]: id          0
      name        16
      host_id      0
      host_name    21
      neighbourhood_group  0
      neighbourhood  0
      latitude      0
      longitude     0
      room_type     0
      price         0
      minimum_nights  0
      number_of_reviews  0
      last_review   10052
      reviews_per_month  10052
      calculated_host_listings_count  0
      availability_365  0
      dtype: int64
```

from above we see that columns “name”, “host_name”, “last_review”, “reviews_per_month” have null values. dropping critical null values can impact loss of valuable information, inconsistent results.

so we can replace it with most repeating values as it is a large data set.

```
[13]: # correcting name column
      # Calculate the mode for each room_type
      mode_names = Airbnb_dataset.groupby('room_type')['name'].apply(lambda x: x.
      ↪mode()[0] if not x.mode().empty else np.nan)

      print(mode_names)
```

```
room_type
entire home/apt          home away from home
private room             private room
shared room      amazing cozy and warm male room on manhattan iv
Name: name, dtype: object
```

```
[14]: # Function to replace null values with the mode
def replace_null_with_mode(row):
    if pd.isna(row['name']):
        return mode_names[row['room_type']]
    return row['name']

# Apply the function to the DataFrame
Airbnb_dataset['name'] = Airbnb_dataset.apply(replace_null_with_mode, axis=1)

Airbnb_dataset['room_type'].isna().sum()
```

[14]: 0

```
[15]: # correcting host_name column
# Calculate the mode for each room_type
mode_names1 = Airbnb_dataset.groupby('room_type')['host_name'].apply(lambda x:
    ↪x.mode()[0] if not x.mode().empty else np.nan)

print(mode_names1)
```

```
room_type
entire home/apt    sonder (nyc)
private room      david
shared room       sergii
Name: host_name, dtype: object
```

```
[16]: # Function to replace null values with the mode
def replace_null_with_mode(row):
    if pd.isna(row['host_name']):
        return mode_names1[row['room_type']]
    return row['host_name']

# Apply the function to the DataFrame
Airbnb_dataset['host_name'] = Airbnb_dataset.apply(replace_null_with_mode,
    ↪axis=1)

Airbnb_dataset['room_type'].isna().sum()
```

[16]: 0

```
[17]: # Calculate the mode of the last_review column
mode_last_review = Airbnb_dataset['last_review'].mode().iloc[0] if not
    ↪Airbnb_dataset['last_review'].mode().empty else np.nan

# Replace NaN values in last_review column with the mode
Airbnb_dataset['last_review'] = Airbnb_dataset['last_review'].
    ↪fillna(mode_last_review)
```

```
[18]: # Calculate the mode of the reviews_per_month column
mode_reviews_per_month = Airbnb_dataset['reviews_per_month'].mode().iloc[0] if_
↳not Airbnb_dataset['reviews_per_month'].mode().empty else np.nan

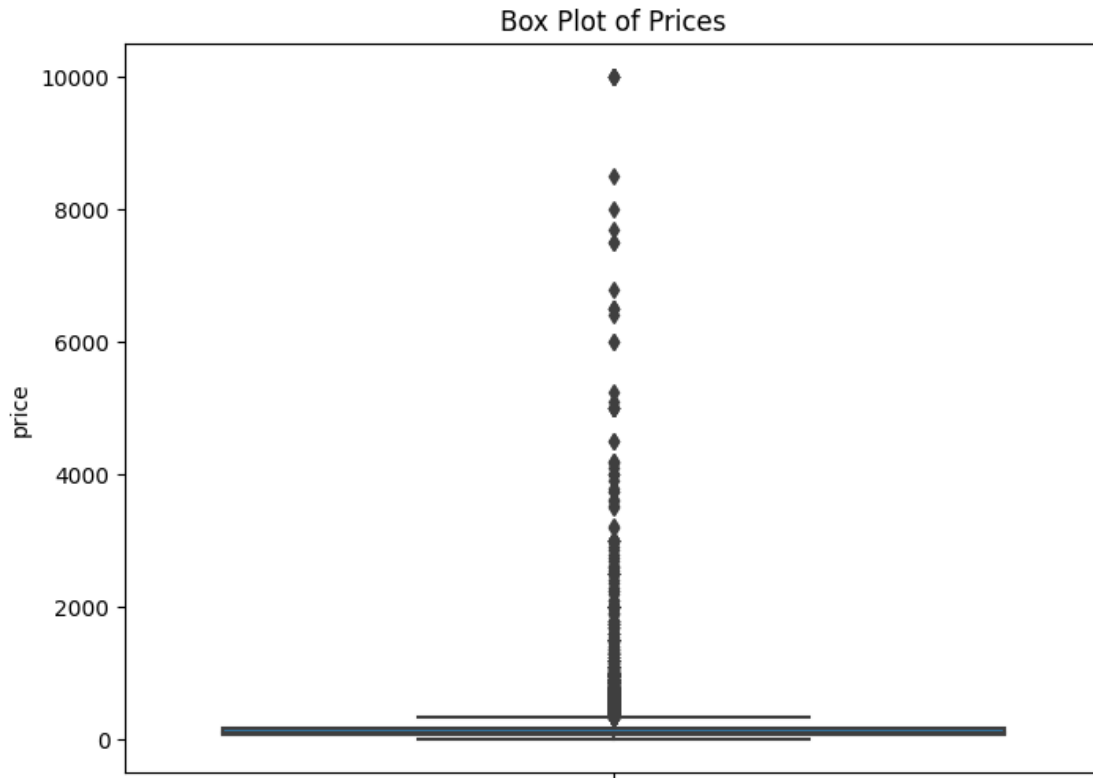
# Replace NaN values in reviews_per_month column with the mode
Airbnb_dataset['reviews_per_month'] = Airbnb_dataset['reviews_per_month'].
↳fillna(mode_reviews_per_month)
```

```
[19]: Airbnb_dataset.isna().sum()
```

```
[19]: id                0
      name              0
      host_id          0
      host_name        0
      neighbourhood_group 0
      neighbourhood    0
      latitude         0
      longitude        0
      room_type        0
      price            0
      minimum_nights   0
      number_of_reviews 0
      last_review      0
      reviews_per_month 0
      calculated_host_listings_count 0
      availability_365  0
      dtype: int64
```

5) Outlier detection and handling (price column)

```
[20]: plt.figure(figsize=(8, 6))
      sns.boxplot(y=Airbnb_dataset['price'])
      plt.title('Box Plot of Prices')
      plt.show()
```



```
[21]: Airbnb_dataset.shape
```

```
[21]: (48895, 16)
```

```
[22]: # Calculate Z-scores
z_scores = stats.zscore(Airbnb_dataset['price'])
df_no_outliers = Airbnb_dataset[(z_scores < 3) & (z_scores > -3)]

print("\nDataFrame After Removing Outliers:")
print(df_no_outliers)
```

DataFrame After Removing Outliers:

	id	name	host_id \
0	2539	clean & quiet apt home by the park	2787
1	2595	skylit midtown castle	2845
2	3647	the village of harlem...new york !	4632
3	3831	cozy entire floor of brownstone	4869
4	5022	entire apt: spacious studio/loft by central park	7192
...
48890	36484665	charming one bedroom - newly renovated rowhouse	8232441
48891	36485057	affordable room in bushwick/east williamsburg	6570630

48892	36485431	sunny studio at historical neighborhood	23492952
48893	36485609	43rd st. time square-cozy single bed	30985759
48894	36487245	trendy duplex in the very heart of hell's kitchen	68119814

	host_name	neighbourhood_group	neighbourhood	latitude	\
0	john	brooklyn	kensington	40.64749	
1	jennifer	manhattan	midtown	40.75362	
2	elisabeth	manhattan	harlem	40.80902	
3	lisaroxanne	brooklyn	clinton hill	40.68514	
4	laura	manhattan	east harlem	40.79851	
...	
48890	sabrina	brooklyn	bedford-stuyvesant	40.67853	
48891	marisol	brooklyn	bushwick	40.70184	
48892	ilgar & aysel	manhattan	harlem	40.81475	
48893	taz	manhattan	hell's kitchen	40.75751	
48894	christophe	manhattan	hell's kitchen	40.76404	

	longitude	room_type	price	minimum_nights	number_of_reviews	\
0	-73.97237	private room	149	1	9	
1	-73.98377	entire home/apt	225	1	45	
2	-73.94190	private room	150	3	0	
3	-73.95976	entire home/apt	89	1	270	
4	-73.94399	entire home/apt	80	10	9	
...	
48890	-73.94995	private room	70	2	0	
48891	-73.93317	private room	40	4	0	
48892	-73.94867	entire home/apt	115	10	0	
48893	-73.99112	shared room	55	1	0	
48894	-73.98933	private room	90	7	0	

	last_review	reviews_per_month	calculated_host_listings_count	\
0	2018-10-19	0.21	6	
1	2019-05-21	0.38	2	
2	2019-06-23	0.02	1	
3	2019-07-05	4.64	1	
4	2018-11-19	0.10	1	
...	
48890	2019-06-23	0.02	2	
48891	2019-06-23	0.02	2	
48892	2019-06-23	0.02	1	
48893	2019-06-23	0.02	6	
48894	2019-06-23	0.02	1	

	availability_365
0	365
1	355
2	365
3	194

4	0
...	...
48890	9
48891	36
48892	27
48893	2
48894	23

[48507 rows x 16 columns]

```
[23]: plt.figure(figsize=(8, 6))
sns.boxplot(y=df_no_outliers['price'])
plt.title('Box Plot of Prices')
plt.show()
```

