Internship project ¶

Title: Data Cleaning Analysis Report

Subtitle: Ensuring Data Quality for Reliable Insights

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1. Data Loading & Data collection

```
In [3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [4]: # Load dataset
df = pd.read_csv("C:\\Users\\FALEYE DOYINSOLA\\project 3 cleaning analysis.cs
```

```
In [5]: # display first few rows
df.head()
```

Out[5]:

	id	name	host_id	host_name	neighbourhood_group	neighbourhood	latitudo
0	2539	Clean & quiet apt home by the park	2787	John	Brooklyn	Kensington	40.64749
1	2595	Skylit Midtown Castle	2845	Jennifer	Manhattan	Midtown	40.75362
2	3647	THE VILLAGE OF HARLEMNEW YORK!	4632	Elisabeth	Manhattan	Harlem	40.80902
3	3831	Cozy Entire Floor of Brownstone	4869	LisaRoxanne	Brooklyn	Clinton Hill	40.68514
4	5022	Entire Apt: Spacious Studio/Loft by central park	7192	Laura	Manhattan	East Harlem	40.7985 ⁻
4		_	_				

2. Data integrity Check

```
In [6]: # check dataset information
df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48895 entries, 0 to 48894
Data columns (total 16 columns):

#	Column	Non-Null Cour	nt Dtype				
0	id	48895 non-nu	ll int64				
1	name	48879 non-nu	ll object				
2	host_id	48895 non-nu	ll int64				
3	host_name	48874 non-nu	ll object				
4	neighbourhood_group	48895 non-nu	ll object				
5	neighbourhood	48895 non-nu	ll object				
6	latitude	48895 non-nu	ll float64				
7	longitude	48895 non-nu	ll float64				
8	room_type	48895 non-nu	ll object				
9	price	48895 non-nu	ll int64				
10	minimum_nights	48895 non-nu	ll int64				
11	number_of_reviews	48895 non-nu	ll int64				
12	last_review	38843 non-nu	ll object				
13	reviews_per_month	38843 non-nu	ll float64				
14	<pre>calculated_host_listings_count</pre>	48895 non-nu	ll int64				
15	availability_365	48895 non-nu	ll int64				
dtyp	<pre>dtypes: float64(3), int64(7), object(6)</pre>						
memo	ry usage: 6.0+ MB						

In [7]: # checking for datatype of the dataset df.dtypes

```
Out[7]: id
                                             int64
        name
                                            object
                                             int64
        host_id
        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
```

```
In [8]: # summary statistics
df.describe()
```

Out[8]:

	id	host_id	latitude	longitude	price	minimum_nights
count	4.889500e+04	4.889500e+04	48895.000000	48895.000000	48895.000000	48895.000000
mean	1.901714e+07	6.762001e+07	40.728949	-73.952170	152.720687	7.029962
std	1.098311e+07	7.861097e+07	0.054530	0.046157	240.154170	20.510550
min	2.539000e+03	2.438000e+03	40.499790	-74.244420	0.000000	1.000000
25%	9.471945e+06	7.822033e+06	40.690100	-73.983070	69.000000	1.000000
50%	1.967728e+07	3.079382e+07	40.723070	-73.955680	106.000000	3.000000
75%	2.915218e+07	1.074344e+08	40.763115	-73.936275	175.000000	5.000000
max	3.648724e+07	2.743213e+08	40.913060	-73.712990	10000.000000	1250.000000

In [9]: # Checking for the rows and columns
df.shape

Out[9]: (48895, 16)

```
In [10]: df.columns
```

```
In [11]: # removal of inrelevant columns
df.drop(['id','host_id'], axis =1 , inplace = True)
```

In [12]: df.head()

Out[12]:

	name	host_name	neighbourhood_group	neighbourhood	latitude	longitude	ro
0	Clean & quiet apt home by the park	John	Brooklyn	Kensington	40.64749	-73.97237	
1	Skylit Midtown Castle	Jennifer	Manhattan	Midtown	40.75362	-73.98377	
2	THE VILLAGE OF HARLEMNEW YORK!	Elisabeth	Manhattan	Harlem	40.80902	-73.94190	
3	Cozy Entire Floor of Brownstone	LisaRoxanne	Brooklyn	Clinton Hill	40.68514	-73.95976	
4	Entire Apt: Spacious Studio/Loft by central park	Laura	Manhattan	East Harlem	40.79851	-73.94399	
4							

In [13]: df.shape

Out[13]: (48895, 14)

2. Missing Data Handling

In [14]:	<pre>df.isnull().sum()</pre>		
Out[14]:	name	16	
	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	
	<pre>calculated_host_listings_count</pre>	0	
	availability_365 dtype: int64	0	

```
# checking for the unique data of 'last_review' column
          df['last_review'].unique()
Out[15]: array(['2018-10-19', '2019-05-21', nan, ..., '2017-12-23', '2018-01-29',
                   '2018-03-29'], dtype=object)
In [16]:
          # Handling the missing data
          df.drop(['last_review','reviews_per_month'], axis =1 , inplace = True)
In [17]: | df.head()
Out[17]:
                       name
                               host_name neighbourhood_group neighbourhood
                                                                               latitude longitude ro
                 Clean & quiet
               apt home by the
                                                      Brooklyn
                                                                   Kensington 40.64749 -73.97237
                                    John
                        park
                Skylit Midtown
           1
                                  Jennifer
                                                     Manhattan
                                                                      Midtown 40.75362 -73.98377
                       Castle
                THE VILLAGE
                         OF
                                 Elisabeth
                                                     Manhattan
                                                                      Harlem 40.80902 -73.94190
              HARLEM....NEW
                      YORK!
                   Cozy Entire
           3
                      Floor of LisaRoxanne
                                                      Brooklyn
                                                                    Clinton Hill 40.68514 -73.95976
                  Brownstone
                   Entire Apt:
                    Spacious
                                   Laura
                                                     Manhattan
                                                                  East Harlem 40.79851 -73.94399
                 Studio/Loft by
                  central park
          df.shape
```

checking for the accuracy of the rows and columns after droping some column

Out[18]: (48895, 12)

```
In [19]: # checking for missing values
         df.isnull().sum()
Out[19]: name
                                             16
         host name
                                             21
         neighbourhood_group
                                              0
         neighbourhood
                                              0
         latitude
                                              0
         longitude
                                              0
         room_type
                                              0
         price
                                              0
         minimum nights
                                              0
         number_of_reviews
                                              0
         calculated_host_listings_count
                                              0
         availability_365
         dtype: int64
In [20]: # drop missing value
         df.dropna(inplace=True)
In [21]: df.isnull().sum()
Out[21]: name
                                             0
         host_name
                                             0
         neighbourhood_group
                                             0
         neighbourhood
                                             0
         latitude
                                             0
         longitude
                                             0
         room_type
         price
                                             0
         minimum_nights
                                             0
         number_of_reviews
                                             0
         calculated_host_listings_count
                                             0
         availability_365
         dtype: int64
```

3. Duplicate Removal

```
In [22]: #check for duplicate
duplicated = df.duplicated()

In [23]: duplicated.sum() # there is no duplicate
Out[23]: 0
```

```
In [24]: df.head()
```

Out[24]:

	name	host_name	neighbourhood_group	neighbourhood	latitude	longitude	ro
0	Clean & quiet apt home by the park	John	Brooklyn	Kensington	40.64749	-73.97237	
1	Skylit Midtown Castle	Jennifer	Manhattan	Midtown	40.75362	-73.98377	
2	THE VILLAGE OF HARLEMNEW YORK!	Elisabeth	Manhattan	Harlem	40.80902	-73.94190	
3	Cozy Entire Floor of Brownstone	LisaRoxanne	Brooklyn	Clinton Hill	40.68514	-73.95976	
4	Entire Apt: Spacious Studio/Loft by central park	Laura	Manhattan	East Harlem	40.79851	-73.94399	

4. Standardization

```
In [25]: #standardized all text column
    textcol = ['name', 'host_name', 'neighbourhood_group', 'neighbourhood', 'room_type']
In [26]: textcol
Out[26]: ['name', 'host_name', 'neighbourhood_group', 'neighbourhood', 'room_type']
In [27]: df[textcol]= df[textcol].apply(lambda col: col.str.strip().str.lower())
```

In [28]: df.head()

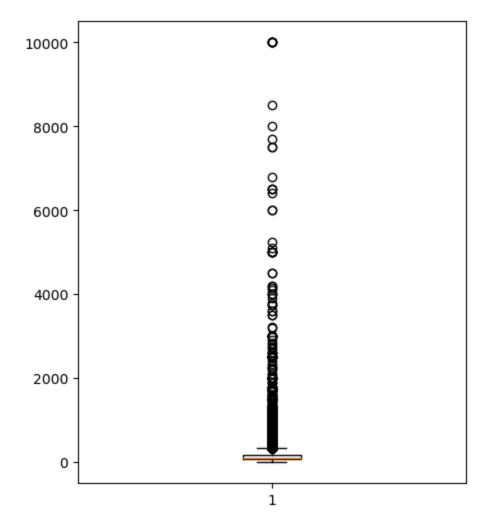
Out[28]:

	name	host_name	neighbourhood_group	neighbourhood	latitude	longitude	room_
0	clean & quiet apt home by the park	john	brooklyn	kensington	40.64749	-73.97237	pr 1
1	skylit midtown castle	jennifer	manhattan	midtown	40.75362	-73.98377	e hom
2	the village of harlemnew york!	elisabeth	manhattan	harlem	40.80902	-73.94190	pr 1
3	cozy entire floor of brownstone	lisaroxanne	brooklyn	clinton hill	40.68514	-73.95976	e hom:
4	entire apt: spacious studio/loft by central park	laura	manhattan	east harlem	40.79851	-73.94399	e home
4							

5. Outlier Detection & Removal

```
In [29]: # to detect or determine the outlier we plot a boxplot
    plt.figure(figsize=(5,6))
    plt.boxplot(df['price'])
    plt.show
```

Out[29]: <function matplotlib.pyplot.show(close=None, block=None)>



```
In [30]: # outlier formular -•
Q1 = df['price'].quantile(0.25)
Q3 = df['price'].quantile(0.75)
IQR = Q3-Q1
```

```
In [31]: # threhold formular
threshold = 1.5
```

```
In [32]: # upperbound and lowerbound formular aferthe outlier
lowerbound = Q1 - threshold * IQR
upperbound = Q3 + threshold * IQR
```

```
In [33]: #remove outlier from df price
df = df[(df['price'] >= lowerbound) & (df['price'] <= upperbound)]
In [34]: # checking for the accuracy of the rows and columns after cleaning</pre>
```

df.shape

```
Out[34]: (45887, 12)
```

```
In [35]: # re-detecting after the removal of outlier
plt.figure(figsize=(5,6))
plt.boxplot(df['price'])
plt.show
```

Out[35]: <function matplotlib.pyplot.show(close=None, block=None)>

