import the required libraries

```
In [1]: import pandas as pd
import numpy as np

In [2]: #loading the dataset
data = pd.read_csv('/home/tamanna/Downloads/AB_NYC_2019.csv')

In [3]: data
```

Out[3]:		id	name	host_id	host_name	neighbourhood_group	neighbou
	0	2539	Clean & quiet apt home by the park	2787	John	Brooklyn	Kens
	1	2595	Skylit Midtown Castle	2845	Jennifer	Manhattan	Mi
	2	3647	THE VILLAGE OF HARLEMNEW YORK!	4632	Elisabeth	Manhattan	F
	3	3831	Cozy Entire Floor of Brownstone	4869	LisaRoxanne	Brooklyn	Clint
	4	5022	Entire Apt: Spacious Studio/Loft by central park	7192	Laura	Manhattan	East ⊦
	•••						
	48890	36484665	Charming one bedroom - newly renovated rowhouse	8232441	Sabrina	Brooklyn	Be Stuy
	48891	36485057	Affordable room in Bushwick/East Williamsburg	6570630	Marisol	Brooklyn	Bus
	48892	36485431	Sunny Studio at Historical Neighborhood	23492952	Ilgar & Aysel	Manhattan	F
	48893	36485609	43rd St. Time Square-cozy single bed	30985759	Taz	Manhattan	Hell's K
	48894	36487245	Trendy duplex in the very heart of Hell's Kitchen	68119814	Christophe	Manhattan	Hell's K

48895 rows × 16 columns

Exploring the data

In [4]: data.head()

Out[4]:		id	I	name	host_id	host_nan	ne neighbou	ırhood_group	neighbou	ırhood	lati
	0	2539	Clean & apt home		2787	Jol	nn	Brooklyn	Kens	ington	40.6
	1	2595	Skylit Mi	dtown Castle	2845	Jennif	er	Manhattan	Mi	dtown	40.7
	2	3647	HARLEM	OF	4632	Elisabe	th	Manhattan	ŀ	Harlem	40.8
	3	3831	F	Entire loor of nstone	4869	LisaRoxanı	ne	Brooklyn	Clint	on Hill	40.6
	4	5022	Sp Studio/I	re Apt: acious Loft by al park	7192	Lau	ra	Manhattan	East H	Harlem	40.7
In [5]:	da	ta.ta	ail()								
Out[5]:			id		name	host_id	host_name	neighbourho	od_group	neighb	ourho
	48	890	36484665	be re	ning one droom - newly novated whouse	8232441	Sabrina		Brooklyn		Bedfo
	48	3891	36485057	Ви	fordable room in ishwick/ East imsburg	6570630	Marisol		Brooklyn	I	Bushv
	48	8892	36485431	at H	y Studio istorical oorhood	23492952	Ilgar & Aysel	М	1anhattan		Harl
	48	8893	36485609	Squa	St. Time are-cozy ngle bed	30985759	Taz	Μ	1anhattan	Hell'	s Kitcl
	48	8894	36487245	in	y duplex the very of Hell's Kitchen	68119814	Christophe	N	1anhattan	Hell'	s Kitcl
<pre>In [6]: pd.concat([data.head(), data.tail()])</pre>											

Out[6]:		id	name	host_id	host_name	neighbourhood_group	neighbou
	0	2539	Clean & quiet apt home by the park	2787	John	Brooklyn	Kens
	1	2595	Skylit Midtown Castle	2845	Jennifer	Manhattan	Mi
	2	3647	THE VILLAGE OF HARLEMNEW YORK!	4632	Elisabeth	Manhattan	F
	3	3831	Cozy Entire Floor of Brownstone	4869	LisaRoxanne	Brooklyn	Clint
	4	5022	Entire Apt: Spacious Studio/Loft by central park	7192	Laura	Manhattan	East H
	48890	36484665	Charming one bedroom - newly renovated rowhouse	8232441	Sabrina	Brooklyn	Be Stuy
	48891	36485057	Affordable room in Bushwick/East Williamsburg	6570630	Marisol	Brooklyn	Bus
	48892	36485431	Sunny Studio at Historical Neighborhood	23492952	Ilgar & Aysel	Manhattan	F
	48893	36485609	43rd St. Time Square-cozy single bed	30985759	Taz	Manhattan	Hell's K
	48894	36487245	Trendy duplex in the very heart of Hell's Kitchen	68119814	Christophe	Manhattan	Hell's K
In [7]:	data.i	nfo()					

In [7]: data.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 48895 non-null int64 host id 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

In [8]: data.shape

Out[8]: (48895, 16)

In [9]: data.describe()

Out[9]:

	id	host_id	latitude	longitude	price	minimı
count	4.889500e+04	4.889500e+04	48895.000000	48895.000000	48895.000000	4889
mean	1.901714e+07	6.762001e+07	40.728949	-73.952170	152.720687	
std	1.098311e+07	7.861097e+07	0.054530	0.046157	240.154170	
min	2.539000e+03	2.438000e+03	40.499790	-74.244420	0.000000	
25%	9.471945e+06	7.822033e+06	40.690100	-73.983070	69.000000	
50%	1.967728e+07	3.079382e+07	40.723070	-73.955680	106.000000	
75%	2.915218e+07	1.074344e+08	40.763115	-73.936275	175.000000	
max	3.648724e+07	2.743213e+08	40.913060	-73.712990	10000.000000	125

Handling duplicate values

```
In [10]: data.duplicated().sum()
```

Out[10]: 0

```
In [11]: data.drop_duplicates(inplace = True)
```

Handling Missing Values

```
In [12]:
         data.isnull().sum()
Out[12]: id
                                                  0
                                                16
          name
          host id
                                                 0
          host name
                                                21
          neighbourhood group
                                                  0
          neighbourhood
                                                  0
          latitude
                                                  0
          longitude
                                                  0
                                                  0
          room type
          price
                                                  0
          minimum nights
                                                  0
          number of reviews
                                                  0
          last review
                                             10052
                                             10052
          reviews per month
          calculated host listings count
                                                  0
                                                  0
          availability 365
          dtype: int64
In [13]: # First Method imputing mean value
         data['reviews per month'].fillna(data['reviews per month'].mean(), inplace =
In [14]: # Second Method dropping na values
         data.dropna(inplace = True)
In [15]: data.isnull().sum()
Out[15]: id
                                             0
                                             0
          name
          host id
                                             0
          host name
                                             0
          neighbourhood group
                                             0
          neighbourhood
          latitude
                                             0
          longitude
          room type
                                             0
                                             0
          price
          minimum nights
          number_of_reviews
          last review
          reviews per month
                                             0
          calculated host listings count
                                             0
          availability 365
                                             0
          dtype: int64
```

Detecting and Removing Outliers

```
In [16]: # z-score and interquartile range both are the criteria to identify an outli
# ways to find outlier -
#using scatter plot - distribution of x and y
```

```
#box plot
         #using z score
         #using igr range
         #Interquantile range steps
         #75% - 25%
         #steps
         #arrange the data in increasing order
         #calculate first(q1) and third quatile(q3)
         #find interquartile range(q3-q1)
         #find lower bound q1*1.5
         #find upper bound q3*1.5
         #anything outside lower or upper bound is an outlier.
In [17]: | des stat = data['price'].describe(percentiles = [.25, .75])
In [18]: des stat
                   38821.000000
Out[18]: count
                     142.332526
          mean
                     196.994756
          std
         min
                       0.000000
          25%
                      69.000000
          50%
                     101.000000
          75%
                     170.000000
          max
                   10000.000000
         Name: price, dtype: float64
In [19]: # find ist and 3rd quantile
         q1, q3 = np.percentile(data['price'], [25, 75])
In [20]: | print(q1, q3)
        69.0 170.0
In [21]: # find IOR
         iqr value = q3 - q1
         print(iqr value)
        101.0
In [22]: lower bound val = q1 - (1.5 * iqr value)
         upper bound val = q3 + (1.5 * iqr value)
In [23]: print(lower_bound_val, upper_bound_val)
        -82.5 321.5
In [24]: new data = data[(data['price'] >= lower bound val) & (data['price'] <= upper</pre>
In [25]: new data
```

Out[25]:

	id	name	host_id	host_name	neighbourhood_group	neighbourho
0	2539	Clean & quiet apt home by the park	2787	John	Brooklyn	Kensingt
1	2595	Skylit Midtown Castle	2845	Jennifer	Manhattan	Midto
3	3831	Cozy Entire Floor of Brownstone	4869	LisaRoxanne	Brooklyn	Clinton I
4	5022	Entire Apt: Spacious Studio/Loft by central park	7192	Laura	Manhattan	East Harl
5	5099	Large Cozy 1 BR Apartment In Midtown East	7322	Chris	Manhattan	Murray l
•••						
48782	36425863	Lovely Privet Bedroom with Privet Restroom	83554966	Rusaa	Manhattan	Upper East Si
48790	36427429	No.2 with queen size bed	257683179	НАі	Queens	Flushi
48799	36438336	Seas The Moment	211644523	Ben	Staten Island	Great K
48805	36442252	1B-1B apartment near by Metro	273841667	Blaine	Bronx	Mott Hav
48852	36455809	Cozy Private Room in Bushwick, Brooklyn	74162901	Christine	Brooklyn	Bushw

36744 rows × 16 columns

outliers removed

Standardization

```
In [26]: #Standardization
    from sklearn.preprocessing import StandardScaler
In [27]: scaling = StandardScaler()
In [28]: column_to_standardize = data['reviews_per_month'].values.reshape(-1, 1)
In [29]: standradize_values = scaling.fit_transform(column_to_standardize)
In [30]: data['reviews_per_month'] = standradize_values.flatten()
In [31]: data
```

Out[31]:

	id	name	host_id	host_name	neighbourhood_group	neighbourho
0	2539	Clean & quiet apt home by the park	2787	John	Brooklyn	Kensingt
1	2595	Skylit Midtown Castle	2845	Jennifer	Manhattan	Midto
3	3831	Cozy Entire Floor of Brownstone	4869	LisaRoxanne	Brooklyn	Clinton I
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48805	36442252	1B-1B apartment near by Metro	273841667	Blaine	Bronx	Mott Hav
48852	36455809	Cozy Private Room in Bushwick, Brooklyn	74162901	Christine	Brooklyn	Bushw

38821 rows × 16 columns

Now data is cleaned to do further process.