

# Customer Service Requests Analysis

## 311 Service Requests from 2010 to Present

**DESCRIPTION:** You have been asked to perform data analysis of service request (311) calls from New York City. You have also been asked to utilize data wrangling techniques to understand the pattern in the data and visualize the major types of complaints

### INITIALLY IMPORT ALL LIBRARY

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
import warnings
warnings.filterwarnings('ignore')
```

### IMPORT THE DATA SET BY USING PANDAS LIBRARY

```
calls = pd.read_csv('311_Service_Requests_from_2010_to_Present.csv')
```

### 1.UNDERSTAND THE DATASET:

**\*\* calls.head():** This will display the first few rows of the dataset\*\*

**\*\* calls.shape:** This will give you the number of rows and columns in the dataset\*\*

**\*\* calls.info ():** This will display information about each column, such as its data type and the number of non-null values\*\*

**\*\* calls.describe():** This will give you a summary of statistics for each numeric column in the dataset\*\*

**>>> calls.head(): This will display the first few rows of the dataset <<<**

```
calls.head(2)
```

	Unique Key	Created Date	Closed Date
Agency \			
0	32310363	12/31/2015 11:59:45 PM	01/01/2016 12:55:15 AM NYPD

```
1      32309934  12/31/2015 11:59:44 PM  01/01/2016 01:26:57 AM  NYPD
```

	Agency Name	Complaint Type
Descriptor \		
0 New York City Police Department	Noise - Street/Sidewalk	Loud Music/Party
1 New York City Police Department	Blocked Driveway	No Access

	Location Type	Incident Zip	Incident Address ... \
0	Street/Sidewalk	10034.0	71 VERMILYEA AVENUE ...
1	Street/Sidewalk	11105.0	27-07 23 AVENUE ...

	Bridge Highway Name	Bridge Highway Direction	Road Ramp \
0	NaN	NaN	NaN
1	NaN	NaN	NaN

	Bridge Highway Segment	Garage Lot Name	Ferry Direction	Ferry Terminal Name \
0	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN

	Latitude	Longitude	Location
0	40.865682	-73.923501	(40.86568153633767, -73.92350095571744)
1	40.775945	-73.915094	(40.775945312321085, -73.91509393898605)

```
[2 rows x 53 columns]
```

**>>> calls.shape: This will give us the number of rows and columns in the dataset <<<**

```
calls.shape
(364558, 53)
```

**>>> calls.info():This will display information about each column,such as its data type and the number of non-null values<<<**

```
calls.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 364558 entries, 0 to 364557
Data columns (total 53 columns):
```

#	Column	Non-Null Count	Dtype
0	Unique Key	364558 non-null	int64
1	Created Date	364558 non-null	object
2	Closed Date	362177 non-null	object
3	Agency	364558 non-null	object
4	Agency Name	364558 non-null	object
5	Complaint Type	364558 non-null	object
6	Descriptor	358057 non-null	object
7	Location Type	364425 non-null	object
8	Incident Zip	361560 non-null	float64
9	Incident Address	312859 non-null	object
10	Street Name	312859 non-null	object
11	Cross Street 1	307370 non-null	object
12	Cross Street 2	306753 non-null	object
13	Intersection Street 1	51120 non-null	object
14	Intersection Street 2	50512 non-null	object
15	Address Type	361306 non-null	object
16	City	361561 non-null	object
17	Landmark	375 non-null	object
18	Facility Type	362169 non-null	object
19	Status	364558 non-null	object
20	Due Date	364555 non-null	object
21	Resolution Description	364558 non-null	object
22	Resolution Action Updated Date	362156 non-null	object
23	Community Board	364558 non-null	object
24	Borough	364558 non-null	object
25	X Coordinate (State Plane)	360528 non-null	float64
26	Y Coordinate (State Plane)	360528 non-null	float64
27	Park Facility Name	364558 non-null	object
28	Park Borough	364558 non-null	object
29	School Name	364558 non-null	object
30	School Number	364558 non-null	object
31	School Region	364557 non-null	object
32	School Code	364557 non-null	object
33	School Phone Number	364558 non-null	object
34	School Address	364558 non-null	object
35	School City	364558 non-null	object
36	School State	364558 non-null	object
37	School Zip	364557 non-null	object
38	School Not Found	364558 non-null	object
39	School or Citywide Complaint	0 non-null	float64
40	Vehicle Type	0 non-null	float64
41	Taxi Company Borough	0 non-null	float64
42	Taxi Pick Up Location	0 non-null	float64
43	Bridge Highway Name	297 non-null	object
44	Bridge Highway Direction	297 non-null	object

```

45 Road Ramp                262 non-null    object
46 Bridge Highway Segment   262 non-null    object
47 Garage Lot Name          0 non-null      float64
48 Ferry Direction          1 non-null      object
49 Ferry Terminal Name       2 non-null      object
50 Latitude                 360528 non-null float64
51 Longitude                360528 non-null float64
52 Location                 360528 non-null object
dtypes: float64(10), int64(1), object(42)
memory usage: 147.4+ MB

```

**>>> calls.describe(): This will give us a summary of statistics for each numeric column in the dataset <<<**

```
calls.describe()
```

	Unique Key	Incident Zip	X Coordinate (State Plane) \
count	3.645580e+05	361560.000000	3.605280e+05
mean	3.106595e+07	10858.496659	1.005043e+06
std	7.331531e+05	578.263114	2.196362e+04
min	2.960737e+07	83.000000	9.133570e+05
25%	3.049938e+07	10314.000000	9.919460e+05
50%	3.108795e+07	11209.000000	1.003470e+06
75%	3.167433e+07	11238.000000	1.019134e+06
max	3.231065e+07	11697.000000	1.067186e+06

	Y Coordinate (State Plane)	School or Citywide Complaint
Vehicle Type \		
count	360528.000000	0.0
0.0		
mean	203425.305782	NaN
NaN		
std	29842.192857	NaN
NaN		
min	121185.000000	NaN
NaN		
25%	182945.000000	NaN
NaN		
50%	201023.000000	NaN
NaN		
75%	222790.000000	NaN
NaN		
max	271876.000000	NaN
NaN		

	Taxi Company Borough	Taxi Pick Up Location	Garage Lot Name \
count	0.0	0.0	0.0

mean	NaN	NaN	NaN
std	NaN	NaN	NaN
min	NaN	NaN	NaN
25%	NaN	NaN	NaN
50%	NaN	NaN	NaN
75%	NaN	NaN	NaN
max	NaN	NaN	NaN

	Latitude	Longitude
count	360528.000000	360528.000000
mean	40.724980	-73.924946
std	0.081907	0.079213
min	40.499040	-74.254937
25%	40.668742	-73.972253
50%	40.718406	-73.930643
75%	40.778166	-73.874098
max	40.912869	-73.700715

## 1.1 UNDERSTANDING THE SHAPE OF THE DATASET

**\*\* To identify the shape of a dataset, we can use the. shape attribute in pandas \*\***

**\*\* This will output a tuple containing the number of rows and columns in the dataset in the format (rows, columns) \*\***

**\*\* My output is (364558, 53)\*\***

```
calls.shape
(364558, 53)
```

## 1.2 IDENTIFY VARIABLES WITH NULL VALUES

**\*\* To identify variables with null values in a dataset, you can use the isnull() function in pandas \*\***

**\*\* This function returns a Boolean Data Frame showing which values are missing (True) and which values are present (False) \*\***

**\*\* There are two type we can identify\*\***

**\*\* 1.calls.isnull().sum()\*\***

**\*\* 2.calls.isna().sum()\*\***

**\*\* Total null values from data set also we can find by using this attribute \*\***

```
** calls.isna().sum().sum()**
```

```
null_counts=calls.isnull().sum()
```

```
print("Variables with null values:\n\n",null_counts[null_counts > 0])
```

Variables with null values:

Closed Date	2381
Descriptor	6501
Location Type	133
Incident Zip	2998
Incident Address	51699
Street Name	51699
Cross Street 1	57188
Cross Street 2	57805
Intersection Street 1	313438
Intersection Street 2	314046
Address Type	3252
City	2997
Landmark	364183
Facility Type	2389
Due Date	3
Resolution Action Updated Date	2402
X Coordinate (State Plane)	4030
Y Coordinate (State Plane)	4030
School Region	1
School Code	1
School Zip	1
School or Citywide Complaint	364558
Vehicle Type	364558
Taxi Company Borough	364558
Taxi Pick Up Location	364558
Bridge Highway Name	364261
Bridge Highway Direction	364261
Road Ramp	364296
Bridge Highway Segment	364296
Garage Lot Name	364558
Ferry Direction	364557
Ferry Terminal Name	364556
Latitude	4030
Longitude	4030
Location	4030
dtype:	int64

```
calls.isna().sum()
```

Unique Key	0
Created Date	0
Closed Date	2381
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	6501
Location Type	133
Incident Zip	2998
Incident Address	51699
Street Name	51699
Cross Street 1	57188
Cross Street 2	57805
Intersection Street 1	313438
Intersection Street 2	314046
Address Type	3252
City	2997
Landmark	364183
Facility Type	2389
Status	0
Due Date	3
Resolution Description	0
Resolution Action Updated Date	2402
Community Board	0
Borough	0
X Coordinate (State Plane)	4030
Y Coordinate (State Plane)	4030
Park Facility Name	0
Park Borough	0
School Name	0
School Number	0
School Region	1
School Code	1
School Phone Number	0
School Address	0
School City	0
School State	0
School Zip	1
School Not Found	0
School or Citywide Complaint	364558
Vehicle Type	364558
Taxi Company Borough	364558
Taxi Pick Up Location	364558
Bridge Highway Name	364261
Bridge Highway Direction	364261
Road Ramp	364296
Bridge Highway Segment	364296
Garage Lot Name	364558

```
Ferry Direction          364557
Ferry Terminal Name      364556
Latitude                 4030
Longitude                4030
Location                 4030
dtype: int64
```

```
calls.isna().sum().sum()

5262284
```

## 2. PERFORM BASIC DATA EXPLORATORY ANALYSIS

### 2.1 UTILIZE MISSING VALUE TREATMENT

**\*\* Total 14 columns are having almost Null values (Unnecessary Column)\*\***

**\*\* we should remove those columns before treating null value using dropna or fillna\*\***

**\*\* Because if use dropna all rows are deleting\*\***

**\*\* Example calls['Garage Lot Name'] has 364558 null values if we use dropna pandas attribute all row were deleting \*\***

**\*\* Eventually we could not able to Imputing missing values because unnecessarily added with our dataset \*\***

#### *Step- 1*

**>>> So below some self-study technique I used**

**\*>>> Cut-off values should occur or must be remove the column**

**>>> Total row \* 90% = result**

**>>> 364558 \* 90% = 328102.2 data must be occur in columns or otherwise we will remove the column to get better result**

**>>> Example 'Taxi Company Borough' has na values 300698 absolutly looks useless column**

```
calls.drop(['Intersection Street 1'],axis=1,inplace=True)
calls.drop(['Intersection Street 2'],axis=1,inplace=True)
calls.drop(['Landmark'],axis=1,inplace=True)
calls.drop(['School or Citywide Complaint'],axis=1,inplace=True)
calls.drop(['Vehicle Type'],axis=1,inplace=True)
calls.drop(['Taxi Company Borough'],axis=1,inplace=True)
```



```
calls.drop(['Taxi Pick Up Location'],axis=1,inplace=True)
calls.drop(['Bridge Highway Name'],axis=1,inplace=True)
calls.drop(['Bridge Highway Direction'],axis=1,inplace=True)
calls.drop(['Road Ramp'],axis=1,inplace=True)
calls.drop(['Bridge Highway Segment'],axis=1,inplace=True)
calls.drop(['Garage Lot Name'],axis=1,inplace=True)
calls.drop(['Ferry Direction'],axis=1,inplace=True)
calls.drop(['Ferry Terminal Name'],axis=1,inplace=True)
```

*# deleted some useless columns eariler it was 53 columns now only 39 useful columns*

```
calls.shape
```

```
(364558, 39)
```

```
calls.isna().sum()
```

Unique Key	0
Created Date	0
Closed Date	2381
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	6501
Location Type	133
Incident Zip	2998
Incident Address	51699
Street Name	51699
Cross Street 1	57188
Cross Street 2	57805
Address Type	3252
City	2997
Facility Type	2389
Status	0
Due Date	3
Resolution Description	0
Resolution Action Updated Date	2402
Community Board	0
Borough	0
X Coordinate (State Plane)	4030
Y Coordinate (State Plane)	4030
Park Facility Name	0
Park Borough	0
School Name	0
School Number	0
School Region	1
School Code	1

```

School Phone Number      0
School Address           0
School City              0
School State             0
School Zip               1
School Not Found         0
Latitude                 4030
Longitude                 4030
Location                 4030
dtype: int64

```

*focusing Na values for filling values*

**>>> I divided my Data set into two types one Numerical columns and another one is Categorical columns**

**>>> for Categorical Na values I am using filna by mode function (attribute from pandas)**

**>>> for Numerical Na values I am using filna by Interpolate function (attribute from pandas)**

*# this helps to understand how many columns are numerical and other statistics details*  
calls.describe()

	Unique Key	Incident Zip	X Coordinate (State Plane)	\
count	3.645580e+05	361560.000000	3.605280e+05	
mean	3.106595e+07	10858.496659	1.005043e+06	
std	7.331531e+05	578.263114	2.196362e+04	
min	2.960737e+07	83.000000	9.133570e+05	
25%	3.049938e+07	10314.000000	9.919460e+05	
50%	3.108795e+07	11209.000000	1.003470e+06	
75%	3.167433e+07	11238.000000	1.019134e+06	
max	3.231065e+07	11697.000000	1.067186e+06	

	Y Coordinate (State Plane)	Latitude	Longitude
count	360528.000000	360528.000000	360528.000000
mean	203425.305782	40.724980	-73.924946
std	29842.192857	0.081907	0.079213
min	121185.000000	40.499040	-74.254937
25%	182945.000000	40.668742	-73.972253
50%	201023.000000	40.718406	-73.930643
75%	222790.000000	40.778166	-73.874098
max	271876.000000	40.912869	-73.700715

```
# Seperating Na values by (Numerical and Categerical) by using
describe () function
# apart from statistical columns I considerd time columns and
categorical columns
```

```
numerical_Na_Values = calls[['Incident Zip','X Coordinate (State
Plane)','Y Coordinate (State Plane)','Latitude','Longitude']]
```

```
categorical_Na_Values = calls[['Closed Date','Descriptor','Location
Type','Incident Address', 'Street Name', 'Cross Street 1',
                                'Cross Street 2','Address Type',
'City', 'Facility Type','Due Date','Resolution Description',
                                'School Region', 'School
Code','Location']]
```

```
numerical_Na_Values.head(3)
```

	Incident Zip	X Coordinate (State Plane)	Y Coordinate (State Plane)
0	10034.0	1005409.0	254678.0
1	11105.0	1007766.0	221986.0
2	10458.0	1015081.0	256380.0

	Latitude	Longitude
0	40.865682	-73.923501
1	40.775945	-73.915094
2	40.870325	-73.888525

```
categorical_Na_Values.head(1)
```

	Closed Date	Descriptor	Location Type
0	01/01/2016 12:55:15 AM	Loud Music/Party	Street/Sidewalk

	Incident Address	Street Name	Cross Street 1	Cross Street 2
0	71 VERMILYEA AVENUE	VERMILYEA AVENUE	ACADEMY STREET	WEST 204 STREET

	Address Type	City	Facility Type	Due Date
0	ADDRESS	NEW YORK	Precinct	01/01/2016 07:59:45 AM

	Resolution Description	School Region
0	The Police Department responded and upon arriv...	Unspecified

	School Code	Location
0	Unspecified	(40.86568153633767, -73.92350095571744)

**Closed date has Na values It is really import column to find the Average Response Time if using Interpolate function for numerical could not able get accurate result so that in Closed Date Na value removed from the data set**

```
calls = calls.dropna(subset=['Closed Date'])
```

```
calls.head(1)
```

	Unique Key	Created Date	Closed Date	
0	32310363	12/31/2015 11:59:45 PM	01/01/2016 12:55:15 AM	NYPD

	Agency Name	Complaint Type
0	New York City Police Department	Noise - Street/Sidewalk Loud Music/Party

	Location Type	Incident Zip	Incident Address	...	School Code
0	Street/Sidewalk	10034.0	71 VERMILYEA AVENUE	...	Unspecified

	School Phone Number	School Address	School City	School State
0	Unspecified	Unspecified	Unspecified	Unspecified

	School Not Found	Latitude	Longitude
0	N	40.865682	-73.923501

	Location
0	(40.86568153633767, -73.92350095571744)

```
[1 rows x 39 columns]
```

```
# Categorical columns are filling by mode function
```

```
calls['Descriptor']=
calls['Descriptor'].fillna(calls['Descriptor'].mode()[0])
calls['Location Type']= calls['Location Type'].fillna(calls['Location
Type'].mode()[0])
calls['Street Name']= calls['Street Name'].fillna(calls['Street
```

```

Name'].mode()[0])
calls['Incident Address']= calls['Incident
Address'].fillna(calls['Incident Address'].mode()[0])
calls['Cross Street 1']= calls['Cross Street 1'].fillna(calls['Cross
Street 1'].mode()[0])
calls['Cross Street 2']= calls['Cross Street 2'].fillna(calls['Cross
Street 2'].mode()[0])
calls['Address Type']= calls['Address Type'].fillna(calls['Address
Type'].mode()[0])
calls['City']= calls['City'].fillna(calls['City'].mode()[0])
calls['Facility Type']= calls['Facility Type'].fillna(calls['Facility
Type'].mode()[0])
calls['Due Date']= calls['Due Date'].fillna(calls['Due Date'].mode()
[0])
calls['Resolution Action Updated Date']= calls['Resolution Action
Updated Date'].fillna(calls['Resolution Action Updated Date'].mode()
[0])
calls['Resolution Description']= calls['Resolution
Description'].fillna(calls['Resolution Description'].mode()[0])
calls['School Region']= calls['School Region'].fillna(calls['School
Region'].mode()[0])
calls['School Code']= calls['School Code'].fillna(calls['School
Code'].mode()[0])
calls['School Zip']= calls['School Zip'].fillna(calls['School
Zip'].mode()[0])
calls['Location']= calls['Location'].fillna(calls['Location'].mode()
[0])

```

```

calls.shape

(362177, 39)

```

```
calls.isna().sum()
```

Unique Key	0
Created Date	0
Closed Date	0
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	0
Location Type	0
Incident Zip	675
Incident Address	0
Street Name	0
Cross Street 1	0
Cross Street 2	0

Address Type	0
City	0
Facility Type	0
Status	0
Due Date	0
Resolution Description	0
Resolution Action Updated Date	0
Community Board	0
Borough	0
X Coordinate (State Plane)	1707
Y Coordinate (State Plane)	1707
Park Facility Name	0
Park Borough	0
School Name	0
School Number	0
School Region	0
School Code	0
School Phone Number	0
School Address	0
School City	0
School State	0
School Zip	0
School Not Found	0
Latitude	1707
Longitude	1707
Location	0
dtype: int64	

```
# Numerical columns are filling by interpolate with linear method
calls.interpolate(method='linear', inplace=True)
```

```
calls.shape
(362177, 39)
```

```
calls.isna().sum()
```

Unique Key	0
Created Date	0
Closed Date	0
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	0
Location Type	0
Incident Zip	0

```

Incident Address      0
Street Name           0
Cross Street 1        0
Cross Street 2        0
Address Type          0
City                  0
Facility Type         0
Status                0
Due Date              0
Resolution Description 0
Resolution Action Updated Date 0
Community Board       0
Borough               0
X Coordinate (State Plane) 0
Y Coordinate (State Plane) 0
Park Facility Name    0
Park Borough          0
School Name           0
School Number         0
School Region         0
School Code           0
School Phone Number   0
School Address        0
School City           0
School State          0
School Zip            0
School Not Found      0
Latitude              0
Longitude             0
Location              0
dtype: int64

```

```
calls.head(1)
```

	Unique Key	Created Date	Closed Date	
Agency \				
0	32310363	12/31/2015 11:59:45 PM	01/01/2016 12:55:15 AM	NYPD

	Agency Name	Complaint Type
Descriptor \		
0	New York City Police Department	Noise - Street/Sidewalk Loud Music/Party

	Location Type	Incident Zip	Incident Address	...	School
Code \					
0	Street/Sidewalk	10034.0	71 VERMILYEA AVENUE	...	
	Unspecified				

```

School Phone Number School Address School City School State
School Zip \
0 Unspecified Unspecified Unspecified Unspecified
Unspecified

```

```

School Not Found Latitude Longitude \
0 N 40.865682 -73.923501

```

```

Location
0 (40.86568153633767, -73.92350095571744)

```

```
[1 rows x 39 columns]
```

## 2.2 Analyze the date column and remove the entries if it has an incorrect timeline

```
** There are two columns have date format stored values**
```

```
** 1. Created Date**
```

```
** 2. Closed Date**
```

```
# 1. Created Date
```

```
# step-1 Convert the date column to datetime format
```

```
calls['Created Date'] = pd.to_datetime(calls['Created Date'],
errors='coerce')
```

```
# step - 2 Analyze the date column
```

```
print(calls['Created Date'].describe())
```

```

count          362177
unique          359655
top    2015-08-08 22:16:46
freq              3
first    2015-01-01 00:00:50
last     2015-12-31 23:59:45
Name: Created Date, dtype: object

```

```
# step-3- Checking the the outlier ['string or non identical time'] by
conditional formatting
```

```
# In that data set Created Date column start with 2015-01-01
```



```
00:00:50 ,  
# if less than the start data or incorrect date format should be  
delete from the dataset as outlier
```

```
incorrect_dates = calls[calls['Created Date'] < '2015-01-01 00:00:50']
```

```
# No outlier rows as per output result  
incorrect_dates
```

```
Empty DataFrame  
Columns: [Unique Key, Created Date, Closed Date, Agency, Agency Name,  
Complaint Type, Descriptor, Location Type, Incident Zip, Incident  
Address, Street Name, Cross Street 1, Cross Street 2, Address Type,  
City, Facility Type, Status, Due Date, Resolution Description,  
Resolution Action Updated Date, Community Board, Borough, X Coordinate  
(State Plane), Y Coordinate (State Plane), Park Facility Name, Park  
Borough, School Name, School Number, School Region, School Code,  
School Phone Number, School Address, School City, School State, School  
Zip, School Not Found, Latitude, Longitude, Location]  
Index: []
```

```
[0 rows x 39 columns]
```

```
# step-3- Checking the the outlier ['string or non identical time'] by  
conditional formatting  
# In that data set Created Date column end with 2015-12-31 23:59:50 ,  
# if greater than the start data or incorrect date format should be  
delete from the dataset as outlier  
incorrect_dates = calls[calls['Created Date'] >= '2015-12-31  
23:59:50']
```

```
# No outlier rows as per output result
```

```
incorrect_dates
```

```
Empty DataFrame  
Columns: [Unique Key, Created Date, Closed Date, Agency, Agency Name,  
Complaint Type, Descriptor, Location Type, Incident Zip, Incident  
Address, Street Name, Cross Street 1, Cross Street 2, Address Type,  
City, Facility Type, Status, Due Date, Resolution Description,  
Resolution Action Updated Date, Community Board, Borough, X Coordinate  
(State Plane), Y Coordinate (State Plane), Park Facility Name, Park  
Borough, School Name, School Number, School Region, School Code,  
School Phone Number, School Address, School City, School State, School  
Zip, School Not Found, Latitude, Longitude, Location]
```

Index: []

[0 rows x 39 columns]

*# 2. Closed Date*

*# step 1 - Convert the date column to datetime format*

```
calls['Closed Date'] = pd.to_datetime(calls['Closed Date'],
errors='coerce')
```

*# step - 2 Analyze the date column*

```
print(calls['Closed Date'].describe())
```

```
count          362177
unique          339837
top    2015-09-10 07:12:49
freq              3
first    2015-01-01 00:20:33
last     2016-01-03 16:22:52
Name: Closed Date, dtype: object
```

*# step-3- Checking the the outlier ['string or non identical time']  
by conditional formatting*

*# In that data set Created Date column end with 2015-01-01 00:20:33 ,  
# if less than the start data or incorrect date format should be  
delete from the dataset as outlier*

```
incorrect_dates = calls[calls['Closed Date'] < '2015-01-01 00:20:33']
```

*# No outlier rows as per output result*

```
incorrect_dates
```

Empty DataFrame

Columns: [Unique Key, Created Date, Closed Date, Agency, Agency Name, Complaint Type, Descriptor, Location Type, Incident Zip, Incident Address, Street Name, Cross Street 1, Cross Street 2, Address Type, City, Facility Type, Status, Due Date, Resolution Description, Resolution Action Updated Date, Community Board, Borough, X Coordinate (State Plane), Y Coordinate (State Plane), Park Facility Name, Park Borough, School Name, School Number, School Region, School Code, School Phone Number, School Address, School City, School State, School Zip, School Not Found, Latitude, Longitude, Location]

Index: []

[0 rows x 39 columns]

```
# step-3- Checking the the outlier ['string or non identical time']
by conditional formatting
# In that data set Created Date column end with 2016-01-03 16:22:53 ,
# if greater than the start data or incorrect date format should be
delete from the dataset as outlier
incorrect_dates = calls[calls['Closed Date'] >= '2016-01-03 16:22:53']
```

```
# No outlier rows as per output result
incorrect_dates
```

```
Empty DataFrame
Columns: [Unique Key, Created Date, Closed Date, Agency, Agency Name,
Complaint Type, Descriptor, Location Type, Incident Zip, Incident
Address, Street Name, Cross Street 1, Cross Street 2, Address Type,
City, Facility Type, Status, Due Date, Resolution Description,
Resolution Action Updated Date, Community Board, Borough, X Coordinate
(State Plane), Y Coordinate (State Plane), Park Facility Name, Park
Borough, School Name, School Number, School Region, School Code,
School Phone Number, School Address, School City, School State, School
Zip, School Not Found, Latitude, Longitude, Location]
Index: []
```

```
[0 rows x 39 columns]
```

### 2.1.1 Draw a frequency plot for city-wise complaints

```
** Sloved this task by two ways**
```

```
** Histplot from seaborn and Bar Plot from Matplot.pyplot **
```

```
# font Style for x, y labels and tile
font_style = {'family': 'Arial', 'size': 20, 'weight': 'bold',
'style': 'italic'}
```

```
# Suitable figure size by seaborn
sns.set(rc={'figure.figsize':(25,8.5)})
```

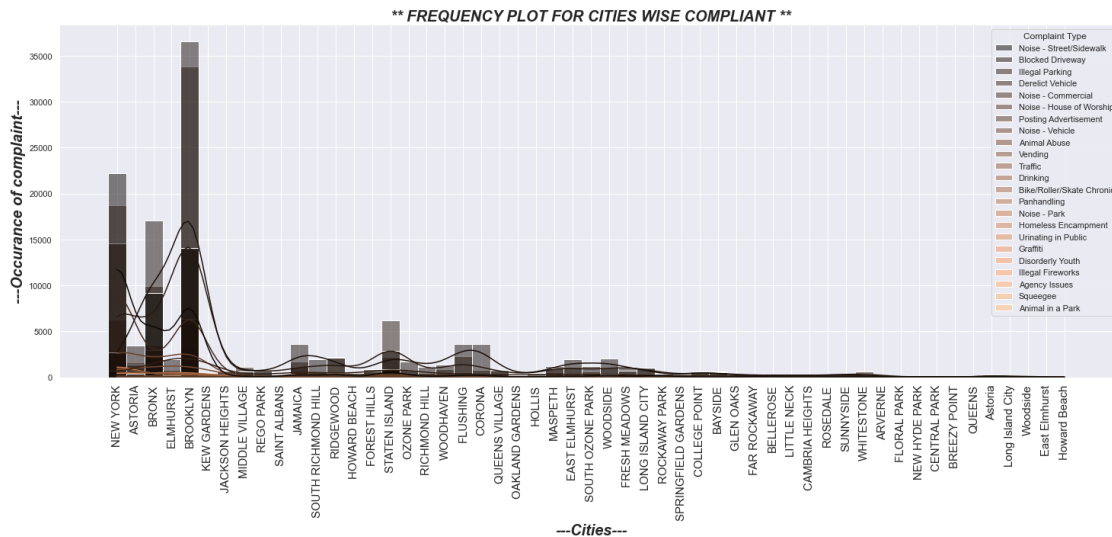
```
# Histplot can helps to make frequency plot by two categorical columns
```

```
sns.histplot(data=calls, x='City', hue='Complaint Type',
palette='copper', alpha=0.5, kde=True)
```

```
# pLot represenation by labels
plt.xlabel('---Cities---',fontdict=font_style)
plt.ylabel('---Occurance of complaint---',fontdict=font_style)
```

```
plt.title('** FREQUENCY PLOT FOR CITIES WISE COMPLIANT
**',fontdict=font_style)
```

```
# the below comment used for showing the x values by vertically
plt.xticks(rotation='vertical', ha='center',size=15)
plt.show()
```



# There is one more type we can create a plot the ferquency occurance for city wise complaint

```
value_count= calls['City'].value_counts()

font_style = {'family': 'Arial', 'size': 20, 'weight': 'bold',
'style': 'italic'}
plt.figure(figsize=(25,8.5))
plt.plot(value_count,marker = 'o',ms=15, mec='red',
mfc='yellow',c='purple',lw=5)
plt.grid(axis='y',ls='solid',color = 'k',lw=0.5,alpha=0.5)
plt.xlabel('---Cities---',fontdict=font_style)
plt.ylabel('---Occurance of complaint---',fontdict=font_style)
plt.title('** FREQUENCY PLOT FOR CITIES WISE COMPLIANT
**',fontdict=font_style)
plt.xticks(rotation='vertical', ha='center',size=15)
plt.show()
```



1	0	1	0	0	0	0	0
---	---	---	---	---	---	---	---

	BROOKLYN	CAMBRIA HEIGHTS	CENTRAL PARK	...	SAINT ALBANS	\
0	0	0	0	...	0	
1	0	0	0	...	0	

	SOUTH OZONE PARK	SOUTH RICHMOND HILL	SPRINGFIELD GARDENS	STATEN ISLAND	\
0	0	0	0	0	
0					
1	0	0	0	0	
0					

	SUNNYSIDE	WHITESTONE	WOODHAVEN	WOODSIDE	Woodside
0	0	0	0	0	0
1	0	0	0	0	0

[2 rows x 53 columns]

*# step2- adding only Brooklyn column into original data set*

```
calls['BROOKLYN_STATE'] = var1['BROOKLYN']
```

*# step3 -Created a brand new variable and used conditional as Brooklyn as 1*

*# now we have all columns under Brooklyn city*

```
select_ones_Brooklyn = calls[calls['BROOKLYN_STATE'] >=1]
```

```
select_ones_Brooklyn.head(3)
```

	Unique Key	Created Date	Closed Date	Agency	\
5	32306554	2015-12-31 23:56:30	2016-01-01 01:50:11	NYPD	
9	32308391	2015-12-31 23:53:58	2016-01-01 01:17:40	NYPD	
13	32305074	2015-12-31 23:47:58	2016-01-01 08:18:47	NYPD	

	Agency Name	Complaint Type	\
5	New York City Police Department	Illegal Parking	
9	New York City Police Department	Blocked Driveway	
13	New York City Police Department	Illegal Parking	

	Descriptor	Location Type	Incident Zip	\
5	Posted Parking Sign Violation	Street/Sidewalk	11215.0	
9	No Access	Street/Sidewalk	11219.0	
13	Posted Parking Sign Violation	Street/Sidewalk	11208.0	

	Incident Address	...	School Phone Number	School Address	School
City \					
5	260 21 STREET	...	Unspecified	Unspecified	
Unspecified					
9	1408 66 STREET	...	Unspecified	Unspecified	
Unspecified					
13	38 COX PLACE	...	Unspecified	Unspecified	
Unspecified					

	School State	School Zip	School Not Found	Latitude	Longitude	\
5	Unspecified	Unspecified	N	40.660823	-73.992568	
9	Unspecified	Unspecified	N	40.623793	-73.999539	
13	Unspecified	Unspecified	N	40.687511	-73.874505	

	Location	BROOKLYN_STATE
5	(40.66082272389114, -73.99256786342693)	1
9	(40.623793065806524, -73.99953890121567)	1
13	(40.68751060232221, -73.87450451131276)	1

[3 rows x 40 columns]

select\_ones\_Brooklyn.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 119523 entries, 5 to 364546
Data columns (total 40 columns):
```

#	Column	Non-Null Count	Dtype
0	Unique Key	119523 non-null	int64
1	Created Date	119523 non-null	datetime64[ns]
2	Closed Date	119523 non-null	datetime64[ns]
3	Agency	119523 non-null	object
4	Agency Name	119523 non-null	object
5	Complaint Type	119523 non-null	object
6	Descriptor	119523 non-null	object
7	Location Type	119523 non-null	object
8	Incident Zip	119523 non-null	float64
9	Incident Address	119523 non-null	object
10	Street Name	119523 non-null	object
11	Cross Street 1	119523 non-null	object
12	Cross Street 2	119523 non-null	object
13	Address Type	119523 non-null	object
14	City	119523 non-null	object
15	Facility Type	119523 non-null	object
16	Status	119523 non-null	object
17	Due Date	119523 non-null	object
18	Resolution Description	119523 non-null	object
19	Resolution Action Updated Date	119523 non-null	object

```

20 Community Board          119523 non-null object
21 Borough                  119523 non-null object
22 X Coordinate (State Plane) 119523 non-null float64
23 Y Coordinate (State Plane) 119523 non-null float64
24 Park Facility Name        119523 non-null object
25 Park Borough              119523 non-null object
26 School Name                119523 non-null object
27 School Number              119523 non-null object
28 School Region              119523 non-null object
29 School Code                119523 non-null object
30 School Phone Number        119523 non-null object
31 School Address             119523 non-null object
32 School City                119523 non-null object
33 School State               119523 non-null object
34 School Zip                 119523 non-null object
35 School Not Found           119523 non-null object
36 Latitude                   119523 non-null float64
37 Longitude                  119523 non-null float64
38 Location                   119523 non-null object
39 BROOKLYN_STATE             119523 non-null uint8
dtypes: datetime64[ns](2), float64(5), int64(1), object(31), uint8(1)
memory usage: 36.6+ MB

```

```

# font Style for x, y labels and tile
font_style = {'family': 'Arial', 'size': 20, 'weight': 'bold',
'style': 'italic'}

plt.rcParams['figure.figsize'] = [15, 6]

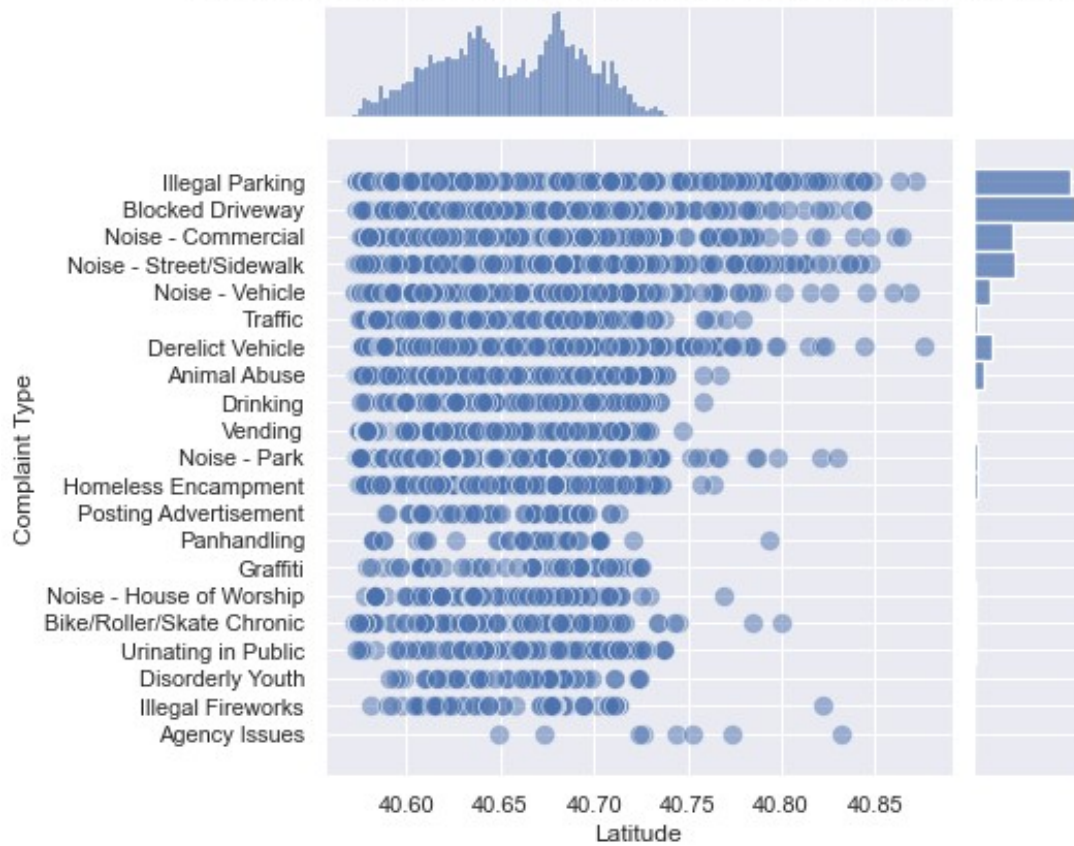
# JointPlot can helps to create scatter plot by One Numerical col and
one categorical columns
sns.jointplot(x='Latitude', y = 'Complaint
Type', data=select_ones_Brooklyn, kind='scatter', palette
='husl', s=100, alpha=0.5)

# pLot represenation by labels
plt.xlabel('---Latitude---', size=20)
plt.ylabel('---Occurance of complaint Type---', size=20)
plt.title('** SCATTER PLOT FOR COMPLIANT CONCENTARTION ACROSS BROOKLYN
**', x=-3, y=1.2)
plt.xticks(rotation=90)
plt.show()

```

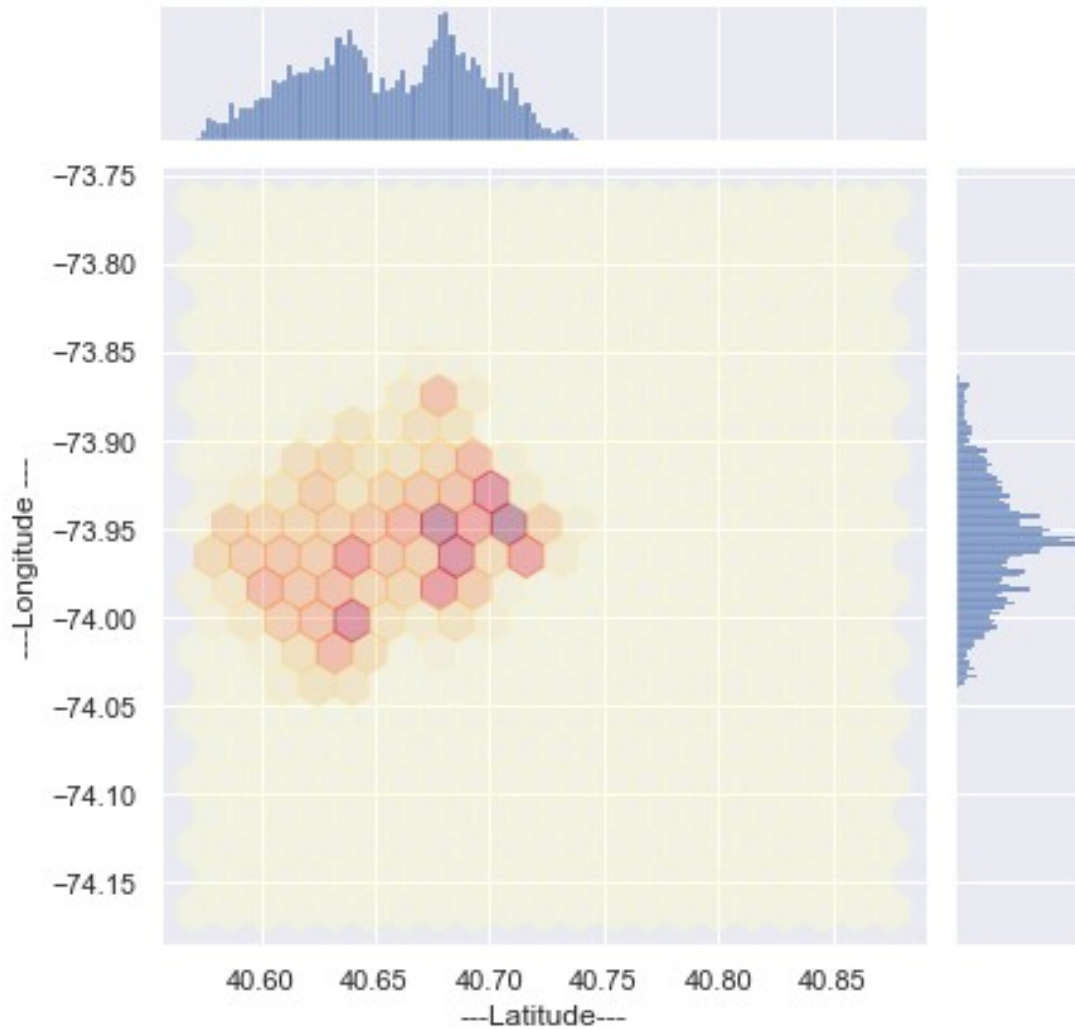


**\*\* SCATTER PLOT FOR COMPLIANT CONCENTARTION ACROSS BROOKLYN \*\***



```
# Hexa bin will works only in float type value columns
sns.jointplot(x='Latitude', y='Longitude', data=select_ones_Brooklyn,
kind='hex', gridsize=20, cmap='YlOrRd', alpha=0.345)
plt.xlabel('---Latitude---', size=12)
plt.ylabel('---Longitude ---', size=12)
plt.title('**HEXBIN PLOT FOR COMPLIANT CONCENTARTION ACROSS BROOKLYN
**', y=1.2)
plt.show()
```

**\*\*HEXBIN PLOT FOR COMPLIANT CONCENTRATION ACROSS BROOKLYN \*\***



### 3. Find major types of complaints\*\*

#### 3.1 Plot Bar Graph of Count vs Compliant Types

**\*\* To plot a bar graph of count versus categorical column values \*\***

**\*\* We can use seaborn's countplot() function \*\***

```
# font Style for x, y labels and tile
font_style = {'family': 'Arial', 'size': 20, 'weight': 'bold',
'style': 'italic'}
```

```
# Suitable figure size by seaborn
sns.set(rc={'figure.figsize':(25,8.5)})
```

*# countplot can helps to make frequency plot by one categorical columns*

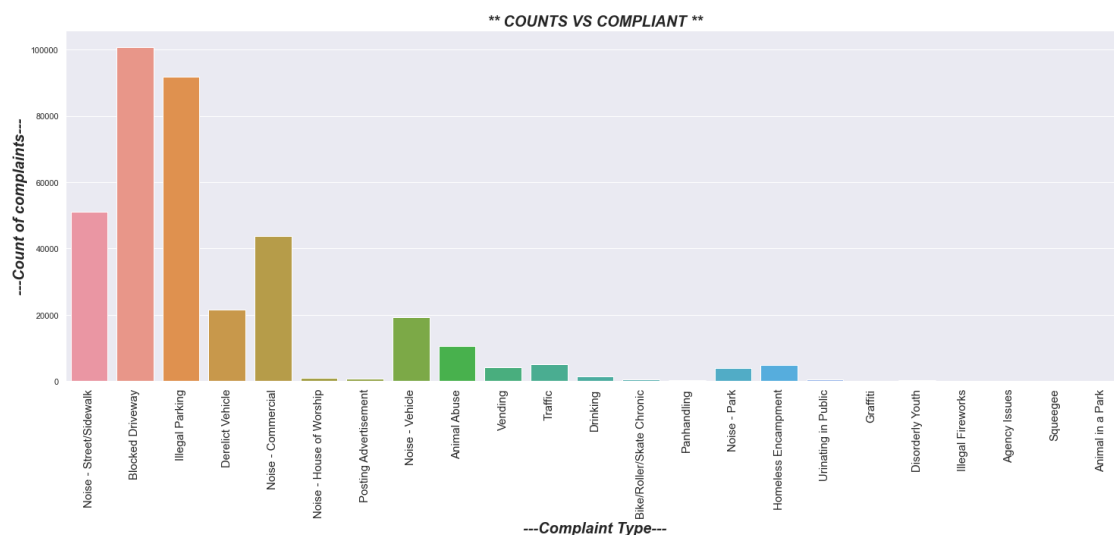
```
sns.countplot(x="Complaint Type", data=calls)
```

*# pLot representation by labels*

```
plt.xlabel('---Complaint Type---',fontdict=font_style)
plt.ylabel('---Count of complaints---',fontdict=font_style)
plt.title('** COUNTS VS COMPLIANT **',fontdict=font_style)
```

*# the below comment used for showing the x values by vertically*

```
plt.xticks(rotation='vertical', ha='center',size=15)
plt.show()
```



## 3.2 Find the top 10 types of complaints

**\*\* We can also use the nlargest() function to find the top n types from a categorical column in a pandas dataframe \*\***

```
top_10_values = calls['Complaint Type'].value_counts().nlargest(10)
```

```
top_10_values
```

```
Blocked Driveway      100624
Illegal Parking       91716
Noise - Street/Sidewalk 51139
Noise - Commercial    43751
Derelict Vehicle      21518
Noise - Vehicle       19301
Animal Abuse          10530
Traffic               5196
Homeless Encampment   4879
```

Vending 4185  
Name: Complaint Type, dtype: int64

### 3.3 Display the types of complaints in each city in separate dataset

**\*\* Steps Involved for this task\*\***

**\*\* step 1- we need to create group by for city column\*\***

**\*\* step 2- Create a dictionary of DataFrames, with one DataFrame for each category\*\***

**\*\* step 3- Print the unique subcategories for each category\*\***

**\*\* step 4 -filling na values by using numpy.np\*\***

**\*\* (could not able create dataframe due to same length value error) \*\***

**\*\* step 5 - Creating Dataframe from dictionary key value pair along with nan by pandas as new data set\*\***

*# step 1- we need to create group by for city column*

groups = calls.groupby('City')

*# step 2- Create a dictionary of DataFrames, with one DataFrame for each category*

data = {category: group['Complaint Type'].unique() for category, group in groups}

*# step 3- Print the unique subcategories for each category*

for category, subcategories in data.items():  
 print(f'{category}:')  
 print(subcategories)

ARVERNE:

['Illegal Parking' 'Noise - Commercial' 'Animal Abuse' 'Blocked Driveway'

'Derelict Vehicle' 'Noise - Street/Sidewalk' 'Homeless Encampment'

'Urinating in Public' 'Noise - Vehicle' 'Noise - House of Worship'

'Panhandling' 'Vending' 'Disorderly Youth' 'Noise - Park' 'Drinking'

'Graffiti' 'Traffic']

ASTORIA:

['Blocked Driveway' 'Noise - Commercial' 'Noise - Vehicle'

'Illegal Parking' 'Noise - Street/Sidewalk' 'Bike/Roller/Skate

Chronic'

'Animal Abuse' 'Derelict Vehicle' 'Drinking' 'Homeless Encampment'

'Urinating in Public' 'Traffic' 'Noise - House of Worship' 'Vending'

'Noise - Park' 'Illegal Fireworks' 'Disorderly Youth' 'Panhandling'

'Graffiti' 'Posting Advertisement']

Astoria:

['Illegal Parking' 'Noise - Street/Sidewalk' 'Derelict Vehicle'  
'Blocked Driveway' 'Noise - Commercial']

BAYSIDE:

['Blocked Driveway' 'Derelict Vehicle' 'Illegal Parking'  
'Noise - Street/Sidewalk' 'Noise - Vehicle' 'Animal Abuse'  
'Noise - Commercial' 'Homeless Encampment' 'Graffiti' 'Noise - Park'  
'Traffic' 'Disorderly Youth' 'Noise - House of Worship' 'Vending'  
'Drinking']

BELLEROSE:

['Derelict Vehicle' 'Blocked Driveway' 'Noise - Street/Sidewalk'  
'Noise - Commercial' 'Illegal Parking' 'Animal Abuse' 'Traffic'  
'Homeless Encampment' 'Urinating in Public' 'Noise - Vehicle'  
'Bike/Roller/Skate Chronic' 'Disorderly Youth' 'Panhandling'  
'Noise - Park' 'Illegal Fireworks' 'Noise - House of Worship'  
'Posting Advertisement' 'Drinking']

BREEZY POINT:

['Noise - Street/Sidewalk' 'Blocked Driveway' 'Animal Abuse'  
'Noise - Commercial' 'Noise - Vehicle' 'Illegal Parking' 'Drinking'  
'Derelict Vehicle']

BRONX:

['Blocked Driveway' 'Illegal Parking' 'Noise - Street/Sidewalk'  
'Noise - Vehicle' 'Noise - Commercial' 'Derelict Vehicle' 'Animal  
Abuse'  
'Vending' 'Traffic' 'Drinking' 'Bike/Roller/Skate Chronic' 'Noise -  
Park'  
'Homeless Encampment' 'Noise - House of Worship' 'Disorderly Youth'  
'Urinating in Public' 'Panhandling' 'Graffiti' 'Illegal Fireworks'  
'Posting Advertisement']

BROOKLYN:

['Illegal Parking' 'Blocked Driveway' 'Noise - Commercial'  
'Noise - Street/Sidewalk' 'Noise - Vehicle' 'Traffic' 'Derelict  
Vehicle'  
'Animal Abuse' 'Drinking' 'Vending' 'Noise - Park' 'Homeless  
Encampment'  
'Posting Advertisement' 'Panhandling' 'Graffiti'  
'Noise - House of Worship' 'Bike/Roller/Skate Chronic'  
'Urinating in Public' 'Disorderly Youth' 'Illegal Fireworks'  
'Agency Issues']

CAMBRIA HEIGHTS:

['Derelict Vehicle' 'Blocked Driveway' 'Noise - Commercial'  
'Noise - Vehicle' 'Illegal Parking' 'Homeless Encampment'  
'Noise - Street/Sidewalk' 'Animal Abuse' 'Traffic' 'Illegal  
Fireworks'  
'Noise - House of Worship']

CENTRAL PARK:

['Noise - Street/Sidewalk' 'Illegal Parking']

COLLEGE POINT:

['Illegal Parking' 'Blocked Driveway' 'Derelict Vehicle'  
'Noise - Street/Sidewalk' 'Noise - Vehicle' 'Noise - Commercial'  
'Traffic' 'Animal Abuse' 'Noise - Park' 'Homeless Encampment']

'Vending'  
'Graffiti' 'Disorderly Youth' 'Drinking' 'Noise - House of Worship']

CORONA:

['Blocked Driveway' 'Illegal Parking' 'Urinating in Public'  
'Noise - Commercial' 'Noise - Street/Sidewalk' 'Vending' 'Animal Abuse']

'Noise - Vehicle' 'Drinking' 'Derelict Vehicle'  
'Noise - House of Worship' 'Homeless Encampment' 'Traffic'  
'Disorderly Youth' 'Noise - Park' 'Posting Advertisement'  
'Panhandling'  
'Graffiti']

EAST ELMHURST:

['Noise - House of Worship' 'Blocked Driveway' 'Illegal Parking'  
'Derelict Vehicle' 'Traffic' 'Noise - Commercial'  
'Noise - Street/Sidewalk' 'Noise - Vehicle' 'Animal Abuse' 'Vending'  
'Homeless Encampment' 'Urinating in Public' 'Drinking'  
'Posting Advertisement' 'Graffiti' 'Noise - Park' 'Disorderly Youth'  
'Bike/Roller/Skate Chronic']

ELMHURST:

['Illegal Parking' 'Blocked Driveway' 'Noise - Vehicle' 'Drinking'  
'Noise - Commercial' 'Noise - Street/Sidewalk' 'Traffic'  
'Homeless Encampment' 'Derelict Vehicle' 'Animal Abuse' 'Vending'  
'Noise - Park' 'Urinating in Public' 'Noise - House of Worship'  
'Panhandling' 'Bike/Roller/Skate Chronic' 'Illegal Fireworks'  
'Disorderly Youth' 'Posting Advertisement' 'Graffiti']

East Elmhurst:

['Illegal Parking' 'Derelict Vehicle']

FAR ROCKAWAY:

['Blocked Driveway' 'Illegal Parking' 'Animal Abuse' 'Noise - Vehicle'  
'Noise - Street/Sidewalk' 'Derelict Vehicle' 'Noise - Commercial'  
'Homeless Encampment' 'Noise - House of Worship' 'Noise - Park'  
'Traffic'  
'Vending' 'Drinking' 'Disorderly Youth' 'Urinating in Public']

FLORAL PARK:

['Illegal Parking' 'Derelict Vehicle' 'Blocked Driveway' 'Drinking'  
'Animal Abuse' 'Noise - Commercial' 'Noise - Vehicle'  
'Noise - Street/Sidewalk' 'Disorderly Youth']

FLUSHING:

['Blocked Driveway' 'Illegal Parking' 'Derelict Vehicle' 'Traffic'  
'Vending' 'Noise - Vehicle' 'Noise - Commercial'  
'Noise - Street/Sidewalk' 'Animal Abuse' 'Drinking' 'Homeless

Encampment'

'Noise - Park' 'Posting Advertisement' 'Graffiti' 'Illegal Fireworks'  
'Urinating in Public' 'Panhandling' 'Noise - House of Worship'  
'Disorderly Youth' 'Bike/Roller/Skate Chronic']

FOREST HILLS:

['Illegal Parking' 'Noise - Commercial' 'Blocked Driveway'  
'Noise - Vehicle' 'Animal Abuse' 'Noise - Street/Sidewalk'  
'Derelict Vehicle' 'Traffic' 'Vending' 'Bike/Roller/Skate Chronic'  
'Noise - Park' 'Panhandling' 'Homeless Encampment']

'Posting Advertisement' 'Graffiti' 'Disorderly Youth'  
'Urinating in Public' 'Drinking' 'Illegal Fireworks'  
'Noise - House of Worship']

FRESH MEADOWS:

['Blocked Driveway' 'Illegal Parking' 'Noise - Vehicle' 'Derelict Vehicle'  
'Animal Abuse' 'Noise - Street/Sidewalk' 'Traffic' 'Noise - Park'  
'Noise - Commercial' 'Homeless Encampment' 'Urinating in Public'  
'Drinking' 'Panhandling' 'Vending']

GLEN OAKS:

['Illegal Parking' 'Derelict Vehicle' 'Noise - Commercial'  
'Blocked Driveway' 'Noise - Park' 'Noise - Vehicle'  
'Noise - Street/Sidewalk' 'Traffic' 'Vending' 'Urinating in Public'  
'Animal Abuse']

HOLLIS:

['Blocked Driveway' 'Illegal Parking' 'Derelict Vehicle' 'Noise - Vehicle'  
'Noise - House of Worship' 'Noise - Street/Sidewalk' 'Noise - Commercial'  
'Animal Abuse' 'Traffic' 'Homeless Encampment' 'Drinking' 'Noise - Park'  
'Urinating in Public' 'Disorderly Youth']

HOWARD BEACH:

['Illegal Parking' 'Blocked Driveway' 'Noise - Commercial'  
'Derelict Vehicle' 'Animal Abuse' 'Panhandling' 'Traffic'  
'Noise - Street/Sidewalk' 'Noise - Park' 'Vending' 'Noise - Vehicle'  
'Homeless Encampment' 'Disorderly Youth' 'Illegal Fireworks'  
'Drinking'  
'Noise - House of Worship' 'Bike/Roller/Skate Chronic']

Howard Beach:

['Blocked Driveway']

JACKSON HEIGHTS:

['Blocked Driveway' 'Noise - House of Worship' 'Noise - Commercial'  
'Illegal Parking' 'Traffic' 'Noise - Street/Sidewalk'  
'Homeless Encampment' 'Vending' 'Noise - Park' 'Noise - Vehicle'  
'Animal Abuse' 'Drinking' 'Bike/Roller/Skate Chronic' 'Derelict Vehicle'  
'Urinating in Public' 'Illegal Fireworks' 'Panhandling'  
'Posting Advertisement' 'Graffiti']

JAMAICA:

['Blocked Driveway' 'Illegal Parking' 'Animal Abuse' 'Derelict Vehicle'  
'Traffic' 'Noise - Commercial' 'Homeless Encampment' 'Vending'  
'Noise - Street/Sidewalk' 'Noise - Vehicle' 'Urinating in Public'  
'Noise - Park' 'Posting Advertisement' 'Drinking'  
'Bike/Roller/Skate Chronic' 'Noise - House of Worship' 'Disorderly Youth'  
'Illegal Fireworks' 'Panhandling' 'Graffiti']

KEW GARDENS:

['Illegal Parking' 'Animal Abuse' 'Blocked Driveway' 'Noise - Vehicle']

'Derelict Vehicle' 'Noise - Commercial' 'Traffic'  
'Noise - Street/Sidewalk' 'Drinking' 'Vending' 'Homeless Encampment'  
'Urinating in Public' 'Noise - House of Worship' 'Posting  
Advertisement']

LITTLE NECK:

['Blocked Driveway' 'Illegal Parking' 'Traffic' 'Derelict Vehicle'  
'Noise - Vehicle' 'Noise - Commercial' 'Animal Abuse'  
'Urinating in Public' 'Noise - Street/Sidewalk' 'Noise - Park'  
'Disorderly Youth' 'Drinking' 'Posting Advertisement']

LONG ISLAND CITY:

['Illegal Parking' 'Blocked Driveway' 'Noise - Commercial'  
'Noise - Street/Sidewalk' 'Derelict Vehicle' 'Vending' 'Noise - Park'  
'Noise - Vehicle' 'Traffic' 'Drinking' 'Animal Abuse' 'Graffiti'  
'Urinating in Public' 'Posting Advertisement' 'Homeless Encampment'  
'Panhandling' 'Disorderly Youth' 'Bike/Roller/Skate Chronic']

Long Island City:

['Illegal Parking' 'Noise - Commercial' 'Derelict Vehicle'  
'Blocked Driveway' 'Noise - Street/Sidewalk']

MASPETH:

['Illegal Parking' 'Blocked Driveway' 'Urinating in Public'  
'Noise - Street/Sidewalk' 'Noise - Commercial' 'Derelict Vehicle'  
'Traffic' 'Drinking' 'Animal Abuse' 'Noise - Vehicle'  
'Homeless Encampment' 'Bike/Roller/Skate Chronic' 'Noise - Park'  
'Disorderly Youth' 'Vending' 'Noise - House of Worship'  
'Illegal Fireworks' 'Graffiti']

MIDDLE VILLAGE:

['Derelict Vehicle' 'Illegal Parking' 'Blocked Driveway' 'Noise -  
Vehicle'  
'Animal Abuse' 'Noise - Street/Sidewalk' 'Traffic' 'Noise -  
Commercial'  
'Homeless Encampment' 'Noise - Park' 'Bike/Roller/Skate Chronic'  
'Drinking']

NEW HYDE PARK:

['Derelict Vehicle' 'Blocked Driveway' 'Illegal Parking' 'Noise -  
Vehicle'  
'Animal Abuse' 'Noise - Commercial']

NEW YORK:

['Noise - Street/Sidewalk' 'Illegal Parking' 'Noise - House of  
Worship'  
'Noise - Commercial' 'Blocked Driveway' 'Vending' 'Noise - Vehicle'  
'Panhandling' 'Animal Abuse' 'Noise - Park' 'Homeless Encampment'  
'Traffic' 'Derelict Vehicle' 'Drinking' 'Urinating in Public'  
'Bike/Roller/Skate Chronic' 'Graffiti' 'Posting Advertisement'  
'Disorderly Youth' 'Illegal Fireworks' 'Squeegee']

OAKLAND GARDENS:

['Blocked Driveway' 'Illegal Parking' 'Noise - Vehicle' 'Derelict  
Vehicle'  
'Noise - Street/Sidewalk' 'Traffic' 'Animal Abuse' 'Noise - Park'  
'Vending' 'Drinking' 'Bike/Roller/Skate Chronic' 'Homeless  
Encampment']



'Disorderly Youth' 'Noise - Commercial']

OZONE PARK:

['Blocked Driveway' 'Illegal Parking' 'Animal Abuse' 'Derelict Vehicle'  
 'Homeless Encampment' 'Noise - Vehicle' 'Noise - Commercial'  
 'Traffic'  
 'Noise - Street/Sidewalk' 'Drinking' 'Panhandling' 'Urinating in Public'  
 'Bike/Roller/Skate Chronic' 'Disorderly Youth' 'Noise - House of Worship'  
 'Noise - Park' 'Illegal Fireworks' 'Posting Advertisement' 'Vending']

QUEENS:

['Noise - Commercial' 'Illegal Parking' 'Noise - Street/Sidewalk'  
 'Noise - House of Worship' 'Derelict Vehicle' 'Blocked Driveway'  
 'Traffic' 'Homeless Encampment' 'Noise - Vehicle' 'Urinating in Public'  
 'Animal in a Park' 'Animal Abuse']

QUEENS VILLAGE:

['Animal Abuse' 'Blocked Driveway' 'Illegal Parking' 'Noise - Vehicle'  
 'Derelict Vehicle' 'Noise - Commercial' 'Panhandling'  
 'Homeless Encampment' 'Noise - Street/Sidewalk' 'Traffic'  
 'Urinating in Public' 'Vending' 'Noise - House of Worship' 'Drinking'  
 'Noise - Park' 'Illegal Fireworks' 'Graffiti' 'Posting Advertisement']

REGO PARK:

['Blocked Driveway' 'Derelict Vehicle' 'Illegal Parking' 'Noise - Vehicle'  
 'Noise - Commercial' 'Drinking' 'Noise - Park' 'Noise - Street/Sidewalk'  
 'Traffic' 'Animal Abuse' 'Graffiti' 'Noise - House of Worship'  
 'Homeless Encampment' 'Vending' 'Urinating in Public']

RICHMOND HILL:

['Blocked Driveway' 'Illegal Parking' 'Drinking' 'Noise - Vehicle'  
 'Derelict Vehicle' 'Noise - Commercial' 'Animal Abuse'  
 'Noise - Street/Sidewalk' 'Homeless Encampment' 'Vending'  
 'Posting Advertisement' 'Traffic' 'Noise - Park' 'Illegal Fireworks'  
 'Graffiti' 'Urinating in Public']

RIDGEWOOD:

['Blocked Driveway' 'Illegal Parking' 'Noise - Vehicle'  
 'Noise - Commercial' 'Derelict Vehicle' 'Noise - Street/Sidewalk'  
 'Animal Abuse' 'Drinking' 'Vending' 'Noise - Park' 'Traffic'  
 'Bike/Roller/Skate Chronic' 'Homeless Encampment' 'Urinating in Public'  
 'Illegal Fireworks' 'Graffiti' 'Noise - House of Worship'  
 'Posting Advertisement' 'Disorderly Youth']

ROCKAWAY PARK:

['Blocked Driveway' 'Animal Abuse' 'Illegal Parking'  
 'Noise - Street/Sidewalk' 'Noise - Commercial' 'Vending' 'Drinking'  
 'Noise - Vehicle' 'Derelict Vehicle' 'Homeless Encampment' 'Noise - Park']

'Traffic' 'Disorderly Youth' 'Urinating in Public']

ROSEDALE:

['Animal Abuse' 'Illegal Parking' 'Blocked Driveway' 'Derelict Vehicle'

'Traffic' 'Noise - Vehicle' 'Vending' 'Noise - Street/Sidewalk'

'Drinking' 'Noise - Commercial' 'Homeless Encampment' 'Noise - Park'

'Graffiti' 'Bike/Roller/Skate Chronic' 'Noise - House of Worship']

SAINT ALBANS:

['Blocked Driveway' 'Illegal Parking' 'Derelict Vehicle' 'Animal Abuse'

'Noise - Vehicle' 'Noise - Commercial' 'Noise - Street/Sidewalk'

'Traffic' 'Vending' 'Disorderly Youth' 'Drinking' 'Homeless Encampment'

'Noise - Park' 'Urinating in Public' 'Noise - House of Worship']

SOUTH OZONE PARK:

['Blocked Driveway' 'Illegal Parking' 'Derelict Vehicle' 'Noise - Vehicle'

'Noise - Street/Sidewalk' 'Animal Abuse' 'Noise - Commercial'

'Traffic'

'Noise - House of Worship' 'Urinating in Public' 'Drinking'

'Homeless Encampment' 'Noise - Park' 'Vending' 'Illegal Fireworks'

'Bike/Roller/Skate Chronic' 'Disorderly Youth' 'Posting Advertisement'

'Graffiti']

SOUTH RICHMOND HILL:

['Blocked Driveway' 'Illegal Parking' 'Noise - Commercial'

'Derelict Vehicle' 'Vending' 'Noise - Street/Sidewalk' 'Drinking'

'Noise - Vehicle' 'Traffic' 'Animal Abuse' 'Homeless Encampment'

'Noise - House of Worship' 'Bike/Roller/Skate Chronic' 'Disorderly Youth'

'Illegal Fireworks' 'Noise - Park' 'Urinating in Public']

SPRINGFIELD GARDENS:

['Illegal Parking' 'Blocked Driveway' 'Animal Abuse' 'Derelict Vehicle'

'Traffic' 'Noise - Vehicle' 'Drinking' 'Noise - Street/Sidewalk'

'Posting Advertisement' 'Noise - Commercial' 'Noise - House of Worship'

'Homeless Encampment' 'Panhandling' 'Noise - Park' 'Urinating in Public'

'Vending' 'Illegal Fireworks']

STATEN ISLAND:

['Posting Advertisement' 'Noise - Commercial' 'Illegal Parking'

'Blocked Driveway' 'Bike/Roller/Skate Chronic' 'Derelict Vehicle'

'Drinking' 'Animal Abuse' 'Noise - Vehicle' 'Homeless Encampment'

'Noise - Street/Sidewalk' 'Disorderly Youth' 'Noise - House of Worship'

'Urinating in Public' 'Traffic' 'Vending' 'Panhandling' 'Noise - Park'

'Illegal Fireworks' 'Graffiti']

SUNNYSIDE:

```

['Blocked Driveway' 'Noise - Commercial' 'Noise - Street/Sidewalk'
 'Illegal Parking' 'Noise - Vehicle' 'Vending' 'Traffic'
 'Bike/Roller/Skate Chronic' 'Derelict Vehicle' 'Noise - Park'
 'Drinking'
 'Animal Abuse' 'Homeless Encampment' 'Urinating in Public'
 'Posting Advertisement' 'Graffiti' 'Disorderly Youth']
WHITESTONE:
['Illegal Parking' 'Blocked Driveway' 'Derelict Vehicle'
 'Noise - Street/Sidewalk' 'Traffic' 'Noise - Commercial' 'Animal
 Abuse'
 'Bike/Roller/Skate Chronic' 'Noise - Vehicle' 'Noise - Park'
 'Illegal Fireworks' 'Drinking' 'Disorderly Youth' 'Graffiti'
 'Vending']
WOODHAVEN:
['Illegal Parking' 'Blocked Driveway' 'Noise - Vehicle' 'Animal Abuse'
 'Noise - Commercial' 'Derelict Vehicle' 'Noise - Street/Sidewalk'
 'Noise - House of Worship' 'Homeless Encampment' 'Noise - Park'
 'Traffic'
 'Urinating in Public' 'Vending' 'Drinking' 'Bike/Roller/Skate
 Chronic'
 'Panhandling']
WOODSIDE:
['Blocked Driveway' 'Illegal Parking' 'Derelict Vehicle'
 'Noise - Street/Sidewalk' 'Animal Abuse' 'Noise - Commercial'
 'Vending'
 'Noise - Park' 'Noise - Vehicle' 'Drinking' 'Traffic'
 'Bike/Roller/Skate Chronic' 'Urinating in Public' 'Graffiti'
 'Homeless Encampment' 'Noise - House of Worship' 'Illegal Fireworks'
 'Disorderly Youth']
Woodside:
['Illegal Parking' 'Blocked Driveway' 'Derelict Vehicle'
 'Noise - Street/Sidewalk' 'Noise - Commercial']

```

*# for checkin purpose*

```
data['BROOKLYN']
```

```

array(['Illegal Parking', 'Blocked Driveway', 'Noise - Commercial',
      'Noise - Street/Sidewalk', 'Noise - Vehicle', 'Traffic',
      'Derelict Vehicle', 'Animal Abuse', 'Drinking', 'Vending',
      'Noise - Park', 'Homeless Encampment', 'Posting Advertisement',
      'Panhandling', 'Graffiti', 'Noise - House of Worship',
      'Bike/Roller/Skate Chronic', 'Urinating in Public',
      'Disorderly Youth', 'Illegal Fireworks', 'Agency Issues'],
      dtype=object)

```

*# step 4 -filling na values by using numpy.np*

*#(could not able create dataframe due to same length value error)*

*# by using for loop and numpy.nan and*

```
max_len = max(len(v) for v in data.values())

for k in data.keys():
    if len(data[k]) < max_len:
        data[k] = np.pad(data[k], (0, max_len - len(data[k])),
                           mode='constant', constant_values=np.nan)
```

*#step 5 - Creating Dataframe from dictionary key value pair along with nan by pandas as new data set*

```
City_wise_complaint = pd.DataFrame(data)
```

```
City_wise_complaint.head(3)
```

	ARVERNE	ASTORIA	Astoria \
0	Illegal Parking	Blocked Driveway	Illegal Parking
1	Noise - Commercial	Noise - Commercial	Noise - Street/Sidewalk
2	Animal Abuse	Noise - Vehicle	Derelict Vehicle
	BAYSIDE	BELLEROSE	BREEZY POINT
\			
0	Blocked Driveway	Derelict Vehicle	Noise - Street/Sidewalk
1	Derelict Vehicle	Blocked Driveway	Blocked Driveway
2	Illegal Parking	Noise - Street/Sidewalk	Animal Abuse
	BRONX	BROOKLYN	CAMBRIA HEIGHTS \
0	Blocked Driveway	Illegal Parking	Derelict Vehicle
1	Illegal Parking	Blocked Driveway	Blocked Driveway
2	Noise - Street/Sidewalk	Noise - Commercial	Noise - Commercial
	CENTRAL PARK ...	SAINT ALBANS	SOUTH OZONE PARK \
0	Noise - Street/Sidewalk ...	Blocked Driveway	Blocked Driveway
1	Illegal Parking ...	Illegal Parking	Illegal Parking
2	NaN ...	Derelict Vehicle	Derelict Vehicle
	SOUTH RICHMOND HILL	SPRINGFIELD GARDENS	STATEN ISLAND \
0	Blocked Driveway	Illegal Parking	Posting Advertisement
1	Illegal Parking	Blocked Driveway	Noise - Commercial
2	Noise - Commercial	Animal Abuse	Illegal Parking
	SUNNYSIDE	WHITESTONE	WOODHAVEN \
0	Blocked Driveway	Illegal Parking	Illegal Parking
1	Noise - Commercial	Blocked Driveway	Blocked Driveway
2	Noise - Street/Sidewalk	Derelict Vehicle	Noise - Vehicle
	WOODSIDE	Woodside	

```
0  Blocked Driveway    Illegal Parking
1  Illegal Parking    Blocked Driveway
2  Derelict Vehicle    Derelict Vehicle
```

```
[3 rows x 53 columns]
```

#### 4. Visualize the major types of complaints in each city\*\*

\*\* pd.crosstab() is a pandas function that is used to create a cross-tabulation \*\*

\*\* Table between two or more categorical variables\*\*

\*\* It is a convenient way to summarize and analyze the relationships between categorical variables in a dataset\*\*

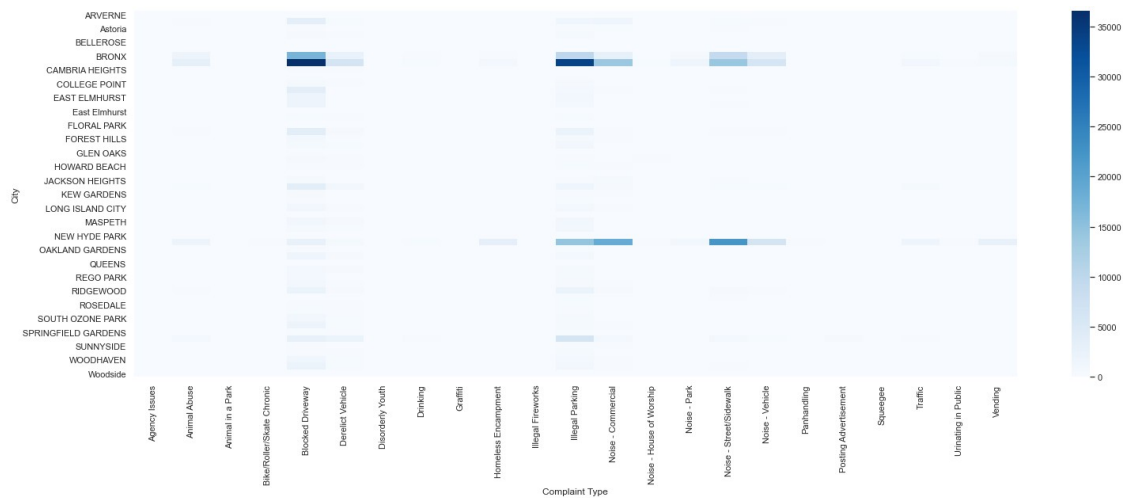
```
Heat_map_visual_Type_of_complaint = pd.crosstab(calls['City'],
calls['Complaint Type'])
```

```
calls['Complaint Type'].value_counts()
```

```
Blocked Driveway          100624
Illegal Parking           91716
Noise - Street/Sidewalk   51139
Noise - Commercial        43751
Derelict Vehicle          21518
Noise - Vehicle           19301
Animal Abuse              10530
Traffic                   5196
Homeless Encampment       4879
Vending                   4185
Noise - Park              4089
Drinking                  1404
Noise - House of Worship  1068
Posting Advertisement      679
Urinating in Public        641
Bike/Roller/Skate Chronic  475
Panhandling               325
Disorderly Youth          315
Illegal Fireworks         172
Graffiti                  157
Agency Issues             8
Squeegee                  4
Animal in a Park          1
Name: Complaint Type, dtype: int64
```

```
sns.heatmap(Heat_map_visual_Type_of_complaint, cmap='Blues')
```

<AxesSubplot:xlabel='Complaint Type', ylabel='City'>



## 5. Check if the average response time across various types of Complaints

**\*\* steps involved for this task \*\***

**\*\* step1 - Subtraction for create date column and closed Date column \*\***

**\*\* step 2 - convert the 'Time Difference' column to seconds \*\***

**\*\* (because without converting the out put was in days only or it was float point values)\*\***

**\*\* step 3 - First I grouped by Complaint type column and then I used aggregate function for chekin average response\*\***

**\*\* step 4 - pd.to\_timedelta() is a pandas function\*\***

**\*\* ( for easy readable like human readable \*\***

**\*\* it is used to convert a numeric value (in seconds, minutes, hours, or days) to a timedelta object\*\***

**\*\* Timedelta objects represent a duration of time\*\***

**\*\* similar to datetime objects which represent a specific point in time \*\***

*# step1 - Subtraction for create date column and closed Date column*

`calls['Time_Difference']= calls['Closed Date'] - calls['Created Date']`

*# for chekin purpose*

`calls.head(1)`

Unique Key	Created Date	Closed Date	Agency	\
0 32310363	2015-12-31 23:59:45	2016-01-01 00:55:15	NYPD	

Descriptor	Agency Name	Complaint Type
0 New York City Police Department	Noise - Street/Sidewalk	Loud Music/Party

Location Type	Incident Zip	Incident Address	... School
0 Street/Sidewalk	10034.0	71 VERMILYEA AVENUE	...
Unspecified			

School City	School State	School Zip	School Not Found
0 Unspecified	Unspecified	Unspecified	N 40.865682

Longitude	Location
0 -73.923501	(40.86568153633767, -73.92350095571744)
	0

Time\_Difference  
0 0 days 00:55:30

[1 rows x 41 columns]

```
# step 2 - convert the 'Time Difference' column to seconds
# because without converting the out put was in days only or it was
# float point values
calls['Time_Difference'] = calls['Time_Difference'].dt.total_seconds()
```

```
# step 3 - First I grouped by Complaint type column and then I used
# aggregate function for chekin average response
```

```
avg_time_difference = calls.groupby('Complaint Type')
['Time_Difference'].mean()
```

```
# step 4 - pd.to_timedelta() is a pandas function
# for easy readable like human readable
# it is used to convert a numeric value (in seconds, minutes, hours,
# or days) to a timedelta object.
# Timedelta objects represent a duration of time,
```

```
# similar to datetime objects which represent a specific point in time.
```

```
avg_time_difference = pd.to_timedelta(avg_time_difference, unit='s')
```

```
avg_time_difference
```

```
Complaint Type
Agency Issues          0 days 05:04:49.125000
Animal Abuse           0 days 05:00:32.556030389
Animal in a Park        14 days 00:50:34
Bike/Roller/Skate Chronic 0 days 03:38:43.688421053
Blocked Driveway        0 days 04:30:32.521515741
Derelict Vehicle        0 days 07:02:39.600102240
Disorderly Youth        0 days 03:26:03.749206349
Drinking               0 days 03:50:21.300569801
Graffiti              0 days 06:27:56.343949045
Homeless Encampment     0 days 04:17:31.384505022
Illegal Fireworks       0 days 02:48:33.482558140
Illegal Parking         0 days 04:20:50.435670984
Noise - Commercial      0 days 03:04:45.760531188
Noise - House of Worship 0 days 03:09:51.087078652
Noise - Park           0 days 03:23:46.055514796
Noise - Street/Sidewalk 0 days 03:23:51.295410548
Noise - Vehicle         0 days 03:29:21.800010362
Panhandling            0 days 04:24:13.550769231
Posting Advertisement    0 days 02:01:26.256259205
Squeegee               0 days 04:02:40.250000
Traffic                0 days 03:25:09.120092379
Urinating in Public     0 days 03:35:59.293291732
Vending                0 days 03:59:26.278375149
Name: Time_Difference, dtype: timedelta64[ns]
```

```
>>> Using Plot for Checking Average Time Response<<<
```

```
font_style = {'family': 'Arial', 'size': 20, 'weight': 'bold',
'style': 'italic'}
plt.figure(figsize=(25,8.5))
plt.plot(avg_time_difference,marker = 'o',ms=15, mec='red',
mfc='yellow',c='purple',lw=5)
plt.grid(axis='y',ls='solid',color = 'k',lw=0.5,alpha=0.5)
plt.xlabel('---Occurance of complaint---',fontdict=font_style)
plt.ylabel('--- Average Time ---',fontdict=font_style)
plt.title('** AVERAGE RESPONSE TIME ACROSS VARIOUS TYPES OF COMPLAINTS
**',fontdict=font_style)
plt.xticks(rotation='vertical', ha='center',size=15)
plt.show()
```



