

```
In [1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
```

```
In [2]: df_nyc_requests = pd.read_csv("311_Service_Requests_from_2010_to_Present.csv")

/home/satheesh/anaconda3/lib/python3.8/site-packages/IPython/core/interacti
veshell.py:3146: DtypeWarning: Columns (48,49) have mixed types.Specify dtype
option on import or set low_memory=False.
  has_raised = await self.run_ast_nodes(code_ast.body, cell_name,
```

```
In [3]: df_nyc_requests.columns
```

```
Out[3]: Index(['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',
              'Intersection Street 1', 'Intersection Street 2', 'Address Type',
              'City', 'Landmark', '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', 'School or Citywide Complaint',
              'Vehicle Type', 'Taxi Company Borough', 'Taxi Pick Up Location',
              'Bridge Highway Name', 'Bridge Highway Direction', 'Road Ramp',
              'Bridge Highway Segment', 'Garage Lot Name', 'Ferry Direction',
              'Ferry Terminal Name', 'Latitude', 'Longitude', 'Location'],
              dtype='object')
```

```
In [4]: df_nyc_requests.head(5)
```

```
Out[4]:
```

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Type
0	32310363	12/31/2015 11:59:45 PM	01-01-16 0:55	NYPD	New York City Police Department	Noise - Street/Sidewalk	Loud Music/Party	Street/Sidewalk
1	32309934	12/31/2015 11:59:44 PM	01-01-16 1:26	NYPD	New York City Police Department	Blocked Driveway	No Access	Street/Sidewalk
2	32309159	12/31/2015 11:59:29 PM	01-01-16 4:51	NYPD	New York City Police Department	Blocked Driveway	No Access	Street/Sidewalk
3	32305098	12/31/2015 11:57:46 PM	01-01-16 7:43	NYPD	New York City Police Department	Illegal Parking	Commercial Overnight Parking	Street/Sidewalk
4	32306529	12/31/2015 11:56:58 PM	01-01-16 3:24	NYPD	New York City Police Department	Illegal Parking	Blocked Sidewalk	Street/Sidewalk

5 rows x 53 columns

```
In [5]: df_nyc_requests.shape
```

Out[5]: (300698, 53)

In [6]: `df_nyc_requests.isnull()`

Out[6]:

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Type	Incident Zip
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
...
300693	False	False	True	False	False	False	False	False	True
300694	False	False	False	False	False	False	False	False	False
300695	False	False	False	False	False	False	False	False	False
300696	False	False	False	False	False	False	False	False	False
300697	False	False	False	False	False	False	False	False	False

300698 rows × 53 columns

In [7]: *#counting the null values that have been observed per column*
`df_nyc_requests.isnull().sum()`

Out[7]:

Unique Key	0
Created Date	0
Closed Date	2164
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	5914
Location Type	131
Incident Zip	2615
Incident Address	44410
Street Name	44410
Cross Street 1	49279
Cross Street 2	49779
Intersection Street 1	256840
Intersection Street 2	257336
Address Type	2815
City	2614
Landmark	300349
Facility Type	2171
Status	0
Due Date	3
Resolution Description	0
Resolution Action Updated Date	2187
Community Board	0
Borough	0
X Coordinate (State Plane)	3540
Y Coordinate (State Plane)	3540
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	300698
Vehicle Type	300698
Taxi Company Borough	300698
Taxi Pick Up Location	300698
Bridge Highway Name	300455
Bridge Highway Direction	300455
Road Ramp	300485
Bridge Highway Segment	300485
Garage Lot Name	300698
Ferry Direction	300697
Ferry Terminal Name	300696
Latitude	3540
Longitude	3540
Location	3540

Conclusion #1:

It is evident that the columns "School or Citywide Complaint", "Vehicle Type", "Taxi Company Borough", "Taxi Pick Up Location", "Garage Lot Name" have no data which can be used for analysis. This is because all of their values are null

The other columns such as "Ferry Direction", "Ferry Terminal Name" have only 1 or 2 records

It's safe to assume that these columns can be dropped and we should be able to proceed with our data analysis

Analysis around Bridge, Landmark and Road ramp for meaningful data

The data from the columns "Bridge Highway Name", "Bridge Highway Direction", "Bridge Highway segment", "Landmark", "Road ramp" all have data that is less than 1% of data. So it would be ideal to drop these columns

Dropping the columns described above

```
In [8]: df_nyc_requests.drop(['Landmark', 'School or Citywide Complaint', 'Vehicle
                             'Taxi Pick Up Location', 'Bridge Highway Name', 'Brid
                             'Road Ramp', 'Bridge Highway Segment', 'Garage Lot Na
                             'Ferry Terminal Name'], axis='columns', inplace=True)
```

```
In [9]: #verifying the shape after dropping the columns
df_nyc_requests.shape
```

```
Out[9]: (300698, 41)
```

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

```
Out[10]: Index(['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',
```

```

',
    'Intersection Street 1', 'Intersection Street 2', 'Address Type',
    'City', 'Facility Type', 'Status', 'Due Date', 'Resolution Descripti
on',
    '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 Addre
ss',
    'School City', 'School State', 'School Zip', 'School Not Found',
    'Latitude', 'Longitude', 'Location'],
dtype='object')

```

```

In [11]: #verifying the first data point after dropping columns
df_nyc_requests.iloc[0]

```

```

Out[11]: Unique Key
32310363
Created Date                                12/31/2015 11:
59:45 PM
Closed Date                                01-01
-16 0:55
Agency
NYPD
Agency Name                                New York City Police De
partment
Complaint Type                                Noise - Street/
Sidewalk
Descriptor                                Loud Mus
ic/Party
Location Type                                Street/
Sidewalk
Incident Zip                                10034
Incident Address                                71 VERMILYE
A AVENUE
Street Name                                VERMILYE
A AVENUE
Cross Street 1                                ACADEM
Y STREET
Cross Street 2                                WEST 20
4 STREET
Intersection Street 1
NaN
Intersection Street 2
NaN
Address Type                                ADDRESS
City                                NEW YORK
Facility Type
Precinct
Status
Closed
Due Date                                01-01
-16 7:59
Resolution Description                                The Police Department responded and upon
arriv...
Resolution Action Updated Date                                01-01
-16 0:55
Community Board                                12 M
ANHATTAN
Borough                                M

```

```

ANHATTAN
X Coordinate (State Plane)      1.0
0541e+06
Y Coordinate (State Plane)
254678
Park Facility Name              Uns
pecified
Park Borough                    M
ANHATTAN
School Name                     Uns
pecified
School Number                   Uns
pecified
School Region                   Uns
pecified
School Code                     Uns
pecified
School Phone Number             Uns
pecified
School Address                  Uns
pecified
School City                     Uns
pecified
School State                    Uns
pecified
School Zip                      Uns
pecified
School Not Found
N
Latitude
40.8657
Longitude
-73.9235
Location                        (40.86568153633767, -73.9235009
5571744)
Name: 0, dtype: object

```

```
In [12]: #checking if school name, number region etc is always unspecified
df_nyc_requests["School Name"].value_counts()
```

```
Out[12]: Unspecified      300697
Alley Pond Park - Nature Center      1
Name: School Name, dtype: int64
```

```
In [13]: df_nyc_requests["School Number"].value_counts()
```

```
Out[13]: Unspecified      300697
Q001      1
Name: School Number, dtype: int64
```

```
In [14]: df_nyc_requests["School Region"].value_counts()
```

```
Out[14]: Unspecified      300697
Name: School Region, dtype: int64
```

```
In [15]: df_nyc_requests["School Code"].value_counts()
```

```
Out[15]: Unspecified      300697
Name: School Code, dtype: int64
```

```
In [16]: df_nyc_requests["School Phone Number"].value_counts()
```

```
Out[16]: Unspecified      300697
7182176034      1
```

Name: School Phone Number. dtype: int64

```
In [17]: df_nyc_requests["School Address"].value_counts()
```

```
Out[17]: Unspecified      300697
Grand Central Parkway, near the soccer field      1
Name: School Address, dtype: int64
```

```
In [18]: df_nyc_requests["School City"].value_counts()
```

```
Out[18]: Unspecified      300697
QUEENS      1
Name: School City, dtype: int64
```

```
In [19]: df_nyc_requests["School State"].value_counts()
```

```
Out[19]: Unspecified      300697
NY      1
Name: School State, dtype: int64
```

```
In [20]: df_nyc_requests["School Zip"].value_counts()
```

```
Out[20]: Unspecified      300697
Name: School Zip, dtype: int64
```

```
In [21]: #checking for the type of crime that is committed near/ at a school
school_complaints = df_nyc_requests[df_nyc_requests["School Name"]!="Unspe
school_complaints
```

```
Out[21]:
```

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Type
--	---------------	-----------------	----------------	--------	----------------	-------------------	------------	------------------

283132	30427220	04/18/2015 09:44:55 AM	05-02-15 10:35	NYPD	New York City Police Department	Animal in a Park	Animal Waste	Park
--------	----------	------------------------------	-------------------	------	---------------------------------------	---------------------	-----------------	------

1 rows x 41 columns

```
In [22]: school_complaints.iloc[0]
```

```
Out[22]: Unique Key
30427220
Created Date
44:55 AM
Closed Date
15 10:35
Agency
NYPD
Agency Name
New York City Police De
partment
Complaint Type
Animal i
n a Park
Descriptor
Anim
al Waste
Location Type
Park
Incident Zip
NaN
Incident Address
```

NaN	
Street Name	
NaN	
Cross Street 1	
NaN	
Cross Street 2	
NaN	
Intersection Street 1	
NaN	
Intersection Street 2	
NaN	
Address Type	
NaN	
City	
QUEENS	
Facility Type	
NaN	
Status	
Closed	
Due Date	05-02
-15 9:44	
Resolution Description	The condition was determined to be an iss
ue ap...	
Resolution Action Updated Date	05-02-
15 10:35	
Community Board	0 Uns
pecified	
Borough	Uns
pecified	
X Coordinate (State Plane)	
NaN	
Y Coordinate (State Plane)	
NaN	
Park Facility Name	Alley Pond Park - Natur
e Center	
Park Borough	Uns
pecified	
School Name	Alley Pond Park - Natur
e Center	
School Number	
Q001	
School Region	
NaN	
School Code	
NaN	
School Phone Number	71
82176034	
School Address	Grand Central Parkway, near the socc
er field	
School City	
QUEENS	
School State	
NY	
School Zip	
NaN	
School Not Found	
N	
Latitude	
NaN	
Longitude	
NaN	
Location	
NaN	
..	

Conclusion #2

There is only 1 occurrence of a complaint which is associated with a school in Queens. It's safe to assume that the New York Police department has been keeping all school areas well guarded/ protected

```
In [23]: #Examining the Unspecified for Park Facility Name
df_nyc_requests["Park Facility Name"].value_counts()
```

```
Out[23]: Unspecified          300697
Alley Pond Park - Nature Center      1
Name: Park Facility Name, dtype: int64
```

```
In [24]: #keeping this as a separate data set for understanding the nature of the c
park_complaints = df_nyc_requests[df_nyc_requests["Park Facility Name"]!=
```

```
In [25]: park_complaints.iloc[0]
```

```
Out[25]: Unique Key
30427220
Created Date          04/18/2015 09:
44:55 AM
Closed Date          05-02-
15 10:35
Agency
NYPD
Agency Name          New York City Police De
partment
Complaint Type          Animal i
n a Park
Descriptor          Anim
al Waste
Location Type
Park
Incident Zip
NaN
Incident Address
NaN
Street Name
NaN
Cross Street 1
NaN
Cross Street 2
NaN
Intersection Street 1
NaN
Intersection Street 2
NaN
Address Type
NaN
City
QUEENS
Facility Type
NaN
Status
Closed
Due Date          05-02-
-15 9:44
Resolution Description          The condition was determined to be an iss
ue ap...
Resolution Action Updated Date          05-02-
```



```

15 10:35
Community Board                                0 Uns
pecified
Borough                                         Uns
pecified
X Coordinate (State Plane)
NaN
Y Coordinate (State Plane)
NaN
Park Facility Name                            Alley Pond Park - Natur
e Center
Park Borough                                  Uns
pecified
School Name                                    Alley Pond Park - Natur
e Center
School Number
Q001
School Region
NaN
School Code
NaN
School Phone Number                           71
82176034
School Address                                Grand Central Parkway, near the socc
er field
School City
QUEENS
School State
NY
School Zip
NaN
School Not Found
N
Latitude
NaN
Longitude
NaN
Location
NaN

```

The complaint from the school is of the park and there are no other part related complaints where the facility name is explicitly mentioned

Examining distribution of complaints

```

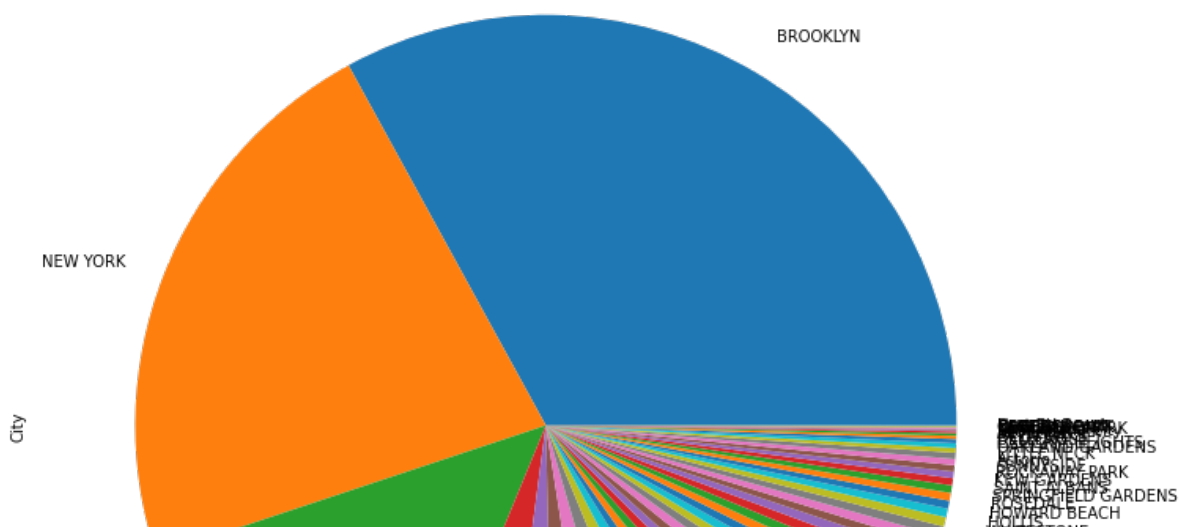
In [26]: #Distribution by city
df_nyc_requests["City"].value_counts().plot(kind='pie', figsize=(12,12))

```

```

Out[26]: <AxesSubplot:ylabel='City'>

```



Conclusion #3

Brooklyn, New York, Bronx and Staten Island contribute to over 70% of the total complaints received by the New York Police Department. This could be owing to the larger amount of population in these areas

Examining complaints based on date

```
In [27]: #converting the available dates to date-time format
df_nyc_requests["Created Date"] = pd.to_datetime(df_nyc_requests["Created Date"])
df_nyc_requests["Closed Date"] = pd.to_datetime(df_nyc_requests["Closed Date"])
df_nyc_requests["Due Date"] = pd.to_datetime(df_nyc_requests["Due Date"])
df_nyc_requests["Resolution Action Updated Date"] = pd.to_datetime(df_nyc_requests["Resolution Action Updated Date"])
```

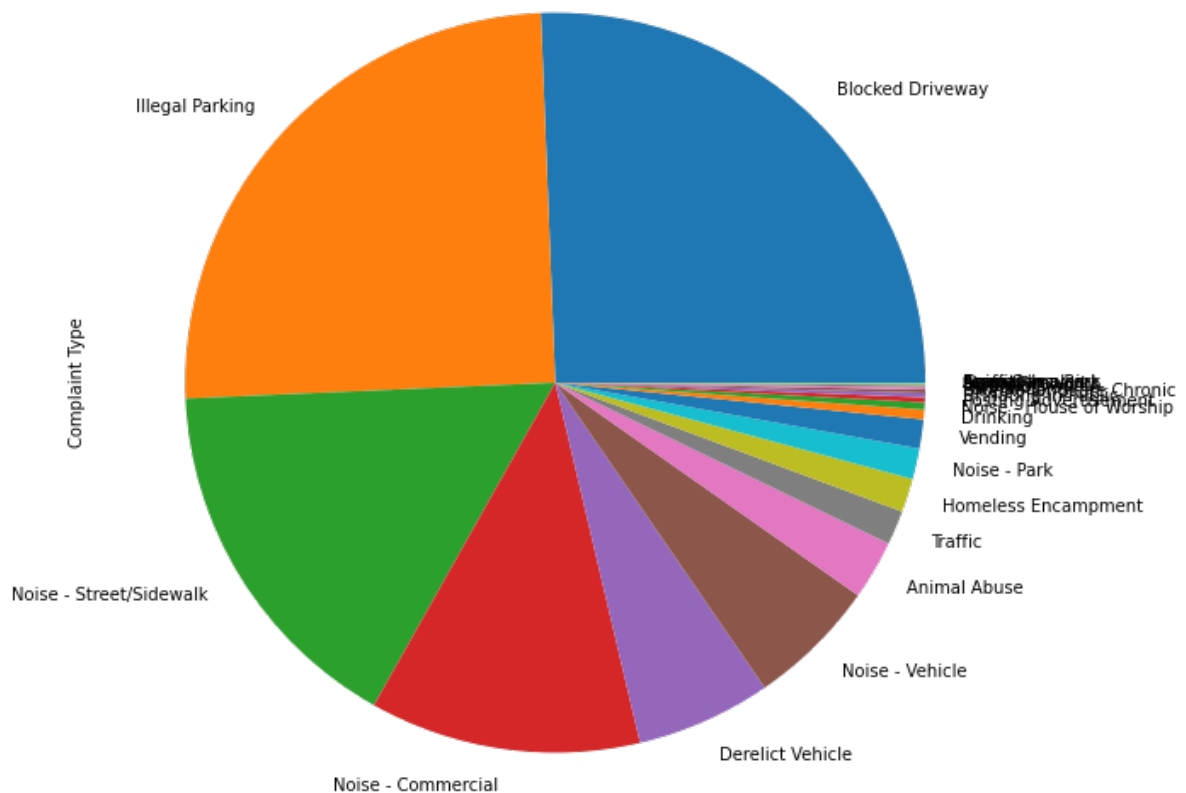
```
In [28]: df_nyc_requests.iloc[0]
```

```
Out[28]: Unique Key
32310363
Created Date          2015-12-31
23:59:45
Closed Date          2016-01-01
00:55:00
Agency
NYPD
Agency Name          New York City Police De
partment
Complaint Type        Noise - Street/
Sidewalk
Descriptor            Loud Mus
ic/Party
Location Type        Street/
Sidewalk
Incident Zip
10034
Incident Address      71 VERMILYE
A AVENUE
Street Name          VERMILYE
A AVENUE
Cross Street 1       ACADEM
Y STREET
```

Cross Street 2	WEST 20
4 STREET	
Intersection Street 1	
NaN	
Intersection Street 2	
NaN	
Address Type	
ADDRESS	
City	
NEW YORK	
Facility Type	
Precinct	
Status	
Closed	
Due Date	2016-01-01
07:59:00	
Resolution Description	The Police Department responded and upon
arriv...	
Resolution Action Updated Date	2016-01-01
00:55:00	
Community Board	12 M
ANHATTAN	
Borough	M
ANHATTAN	
X Coordinate (State Plane)	1.0
0541e+06	
Y Coordinate (State Plane)	
254678	
Park Facility Name	Uns
pecified	
Park Borough	M
ANHATTAN	
School Name	Uns
pecified	
School Number	Uns
pecified	
School Region	Uns
pecified	
School Code	Uns
pecified	
School Phone Number	Uns
pecified	
School Address	Uns
pecified	
School City	Uns
pecified	
School State	Uns
pecified	
School Zip	Uns
pecified	
School Not Found	
N	
Latitude	
40.8657	
Longitude	
-73.9235	
Location	(40.86568153633767, -73.9235009
5571744)	

```
In [29]: #complaint types split up
df_nyc_requests["Complaint Type"].value_counts().plot(kind='pie', figsize=
```

```
Out[29]: <AxesSubplot:ylabel='Complaint Type'>
```



Conclusion #4

Blocked Driveway and Illegal Parking forms over 50% of the total complaints to the NYPD which is an indication that parking is generally a huge problem.

The top 4 complaint areas "Blocked Driveway", "Illegal Parking", "Noise - Street/Sidewalk", "Noise - Commercial" form over 75% of the complaints which is an indication that Noise is the second largest contributor towards complaints to the NYPD

```
In [30]: #Creating a column to find out resolution time
df_nyc_requests["Expected Resolution Time"] = df_nyc_requests["Due Date"]
df_nyc_requests["Actual Resolution Time"] = df_nyc_requests["Closed Date"]
```

```
In [31]: df_nyc_requests.iloc[3]
```

```
Out[31]: Unique Key      32305098
Created Date      2015-12-31 23:57:46
Closed Date      2016-01-01 07:43:00
Agency      NYPD
Agency Name      New York City Police Department
Complaint Type      Illegal Parking
Descriptor      Commercial Overnight Parking
Location Type      Street/
```

Sidewalk	
Incident Zip	
10461	
Incident Address	2940 BAISLE
Y AVENUE	
Street Name	BAISLE
Y AVENUE	
Cross Street 1	EDISO
N AVENUE	
Cross Street 2	
B STREET	
Intersection Street 1	
NaN	
Intersection Street 2	
NaN	
Address Type	
ADDRESS	
City	
BRONX	
Facility Type	
Precinct	
Status	
Closed	
Due Date	2016-01-01
07:57:00	
Resolution Description	The Police Department responded to the co
mplai...	
Resolution Action Updated Date	2016-01-01
07:43:00	
Community Board	
10 BRONX	
Borough	
BRONX	
X Coordinate (State Plane)	1.0
3174e+06	
Y Coordinate (State Plane)	
243899	
Park Facility Name	Uns
pecified	
Park Borough	
BRONX	
School Name	Uns
pecified	
School Number	Uns
pecified	
School Region	Uns
pecified	
School Code	Uns
pecified	
School Phone Number	Uns
pecified	
School Address	Uns
pecified	
School City	Uns
pecified	
School State	Uns
pecified	
School Zip	Uns
pecified	
School Not Found	
N	
Latitude	
40.836	
Longitude	

```
-73.8284
Location (40.83599404683083, -73.8283793
9584206)
Expected Resolution Time 0 days
07:59:14
Actual Resolution Time 0 days
07:45:14
```

```
In [32]: #Creating an SLA breach column to understand if the SLA has been breached
df_nyc_requests["SLA Breach"] = df_nyc_requests["Actual Resolution Time"] >
```

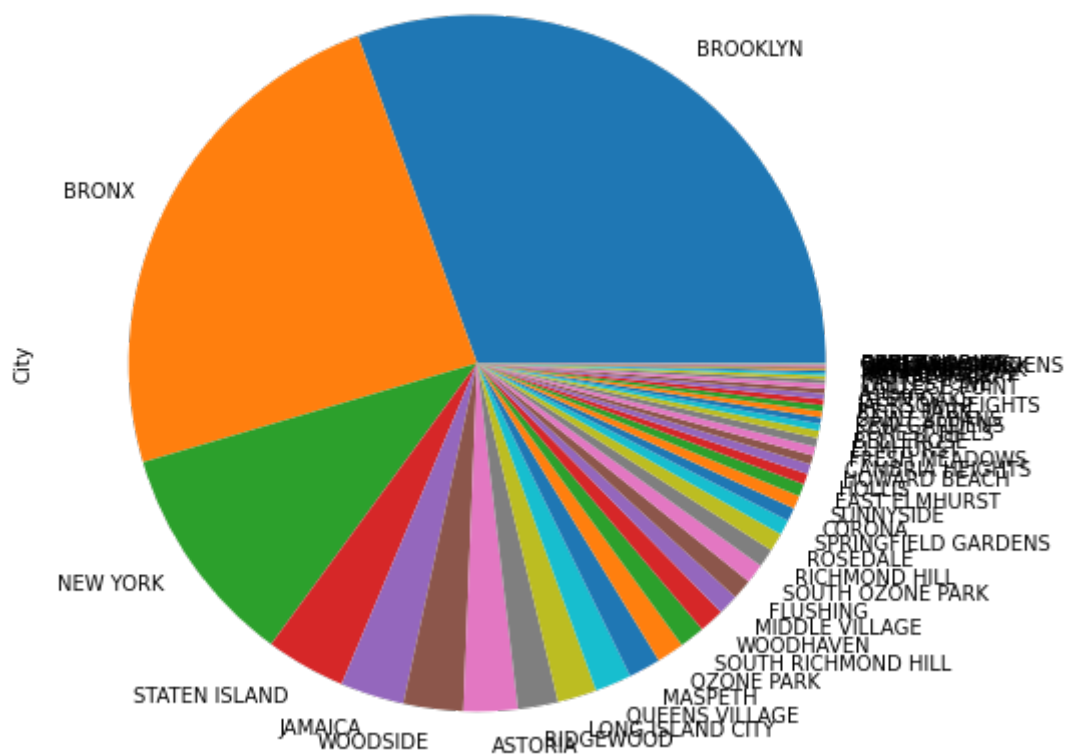
```
In [33]: df_nyc_requests["SLA Breach"].value_counts()
```

```
Out[33]: False    262118
         True     38580
         Name: SLA Breach, dtype: int64
```

```
In [34]: df_sla_breach = df_nyc_requests[df_nyc_requests["SLA Breach"]==True]
```

```
In [35]: df_sla_breach["City"].value_counts().plot(kind='pie', figsize=(8,8))
```

```
Out[35]: <AxesSubplot:ylabel='City'>
```



Conclusion #5

Brooklyn, Bronx and New York form the top 60% of the SLA breaches where the resolution time that was expected has been exceeded

```
In [36]: df_nyc_requests.columns
```

```
Out[36]: Index(['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
',
    'Intersection Street 1', 'Intersection Street 2', 'Address Type',
    'City', 'Facility Type', 'Status', 'Due Date', 'Resolution Descripti
on',
    '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 Addre
ss',
    'School City', 'School State', 'School Zip', 'School Not Found',
    'Latitude', 'Longitude', 'Location', 'Expected Resolution Time',
    'Actual Resolution Time', 'SLA Breach'],
dtype='object')

```

In [37]: `df_nyc_requests.iloc[12]`

```

Out[37]: Unique Key
32306612
Created Date                2015-12-31
23:48:03
Closed Date                2016-01-01
02:17:00
Agency
NYPD
Agency Name                New York City Police De
partment
Complaint Type              Noise - Street/
Sidewalk
Descriptor                  Loud Mus
ic/Party
Location Type              Street/
Sidewalk
Incident Zip
10461
Incident Address            1701 PILGRI
M AVENUE
Street Name                PILGRI
M AVENUE
Cross Street 1              ROBERT
S AVENUE
Cross Street 2              WESTCHESTE
R AVENUE
Intersection Street 1
NaN
Intersection Street 2
NaN
Address Type
ADDRESS
City
BRONX
Facility Type
Precinct
Status
Closed
Due Date                2016-01-01
07:48:00
Resolution Description      The Police Department responded to the co
mplai...
Resolution Action Updated Date  2016-01-01
02:18:00
Community Board
10 BRONX

```

```

Borough
BRONX
X Coordinate (State Plane) 1.0
3029e+06
Y Coordinate (State Plane)
247376
Park Facility Name Uns
pecified
Park Borough
BRONX
School Name Uns
pecified
School Number Uns
pecified
School Region Uns
pecified
School Code Uns
pecified
School Phone Number Uns
pecified
School Address Uns
pecified
School City Uns
pecified
School State Uns
pecified
School Zip Uns
pecified
School Not Found
N
Latitude
40.8455
Longitude
-73.8336
Location (40.845545043640215, -73.8335847
1831198)
Expected Resolution Time 0 days
07:59:57
Actual Resolution Time 0 days
02:28:57
SLA Breach
False
Name: 12 dtype: object

```

```
In [38]: df_nyc_requests["Facility Type"].value_counts()
```

```
Out[38]: Precinct    298527
Name: Facility Type, dtype: int64
```

Resolution time and Location Analysis

```
In [39]: #creating a new dataframe with only location related information and resolution time
df_nyc_location = df_nyc_requests[["Unique Key", "Agency Name", "Complaint Type", "Resolution Time"]]
```

```
In [40]: df_nyc_location.iloc[0]
```

```
Out[40]: Unique Key    32310363
Agency Name    New York City Police Department
Complaint Type    Noise - Street/Sidewalk
Descriptor    Loud Music/Party
Location Type    Street/Sidewalk
Incident Zip    10034
Incident Address    71 VERMILYEA AVENUE
```



```

Street Name          VERMILYEA AVENUE
Cross Street 1       ACADEMY STREET
Cross Street 2       WEST 204 STREET
Address Type         ADDRESS
City                 NEW YORK
Status               Closed
Actual Resolution Time 0 days 00:55:15
Name: 0, dtype: object

```

```

In [41]: #dropping not closed complaints for analysis purposes and examining the data
df_nyc_location = df_nyc_location[df_nyc_location["Status"] == "Closed"]
df_nyc_location.shape

```

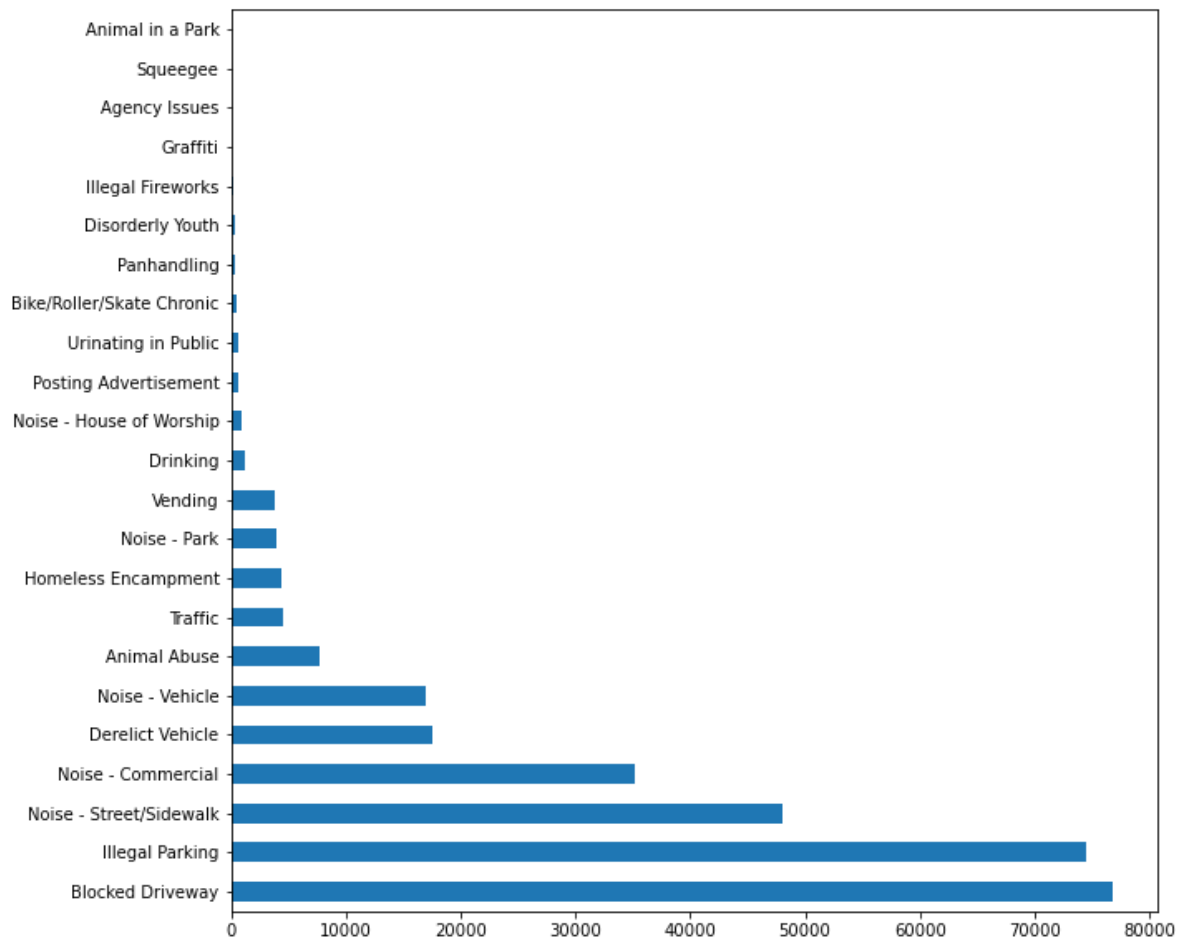
Out[41]: (298471, 14)

```

In [42]: #complaint distribution
df_nyc_location["Complaint Type"].value_counts().plot(kind = 'barh', figsize=(15, 10))

```

Out[42]: <AxesSubplot:>



```

In [43]: #viewing total complaints duration
df_nyc_location.groupby(by="Complaint Type")["Actual Resolution Time"].sum

```

```

Out[43]: Complaint Type
Agency Issues          1 days 07:33:43
Animal Abuse          1687 days 00:59:09
Animal in a Park        14 days 00:50:05
Bike/Roller/Skate Chronic  66 days 12:58:48
Blocked Driveway      15169 days 05:21:46
Derelict Vehicle       5395 days 05:35:38
Disorderly Youth        42 days 09:45:18
Drinking               205 days 03:50:13
Graffiti              33 days 16:05:29

```

Homeless Encampment	800 days 05:18:32
Illegal Fireworks	19 days 07:52:17
Illegal Parking	13973 days 04:08:46
Noise - Commercial	4621 days 17:23:24
Noise - House of Worship	123 days 14:34:27
Noise - Park	571 days 03:34:58
Noise - Street/Sidewalk	6900 days 13:20:22
Noise - Vehicle	2547 days 00:00:34
Panhandling	55 days 13:41:39
Posting Advertisement	53 days 06:28:08
Squeegee	0 days 16:10:57
Traffic	645 days 20:57:11
Urinating in Public	89 days 10:59:07
Vending	634 days 09:54:17

Name: Actual Resolution Time. dtype: timedelta64[ns]

```
In [44]: df_nyc_location.groupby(by="Complaint Type")["Actual Resolution Time"].count
```

```
Out[44]: Complaint Type
Agency Issues          6
Animal Abuse          7766
Animal in a Park         1
Bike/Roller/Skate Chronic  424
Blocked Driveway       76793
Derelict Vehicle       17585
Disorderly Youth        286
Drinking              1275
Graffiti             113
Homeless Encampment    4410
Illegal Fireworks      168
Illegal Parking       74515
Noise - Commercial    35245
Noise - House of Worship  929
Noise - Park         4021
Noise - Street/Sidewalk 48068
Noise - Vehicle       17032
Panhandling           305
Posting Advertisement   647
Squeegee              4
Traffic              4493
Urinating in Public     592
Vending              3793
Name: Actual Resolution Time, dtype: int64
```

```
In [45]: #Converting all the datetime into minutes for statistical analysis
df_nyc_location['Actual Resolution Time'] = (df_nyc_location['Actual Resolution Time'] - df_nyc_location['Actual Resolution Time'].min()) / (df_nyc_location['Actual Resolution Time'].max() - df_nyc_location['Actual Resolution Time'].min()) * 1440
```

```
In [46]: df_nyc_location["Location Type"].value_counts()
```

```
Out[46]: Street/Sidewalk          247503
Store/Commercial                 20183
Club/Bar/Restaurant              17227
Residential Building/House       6953
Park/Playground                 4751
House of Worship                 927
Residential Building             227
Highway                         214
Parking Lot                     117
House and Store                  93
Vacant Lot                      77
Commercial                     62
Roadway Tunnel                  35
Subway Station                  34
```

Bridge	2
Park	1

```
In [47]: #Grouping NYC data by City
grouped_nyc = df_nyc_location.groupby(['City', 'Complaint Type'])
```

```
In [48]: grouped_nyc['Actual Resolution Time'].mean()
```

```
Out[48]: City      Complaint Type
ARVERNE  Animal Abuse      128.894737
          Blocked Driveway  151.200000
          Derelict Vehicle  177.740741
          Disorderly Youth  215.000000
          Drinking          14.000000
          ...
Woodside  Blocked Driveway  384.181818
          Derelict Vehicle  298.000000
          Illegal Parking   312.830000
          Noise - Commercial 143.000000
          Noise - Street/Sidewalk 204.400000
Name: Actual Resolution Time, Length: 764, dtype: float64
```

```
In [49]: grouped_nyc['Actual Resolution Time'].sum()
```

```
Out[49]: City      Complaint Type
ARVERNE  Animal Abuse      4898
          Blocked Driveway  5292
          Derelict Vehicle  4799
          Disorderly Youth   430
          Drinking          14
          ...
Woodside  Blocked Driveway  4226
          Derelict Vehicle   596
          Illegal Parking   31283
          Noise - Commercial  286
          Noise - Street/Sidewalk 1022
Name: Actual Resolution Time, Length: 764, dtype: int64
```

```
In [50]: grouped_nyc['Actual Resolution Time'].count()
```

```
Out[50]: City      Complaint Type
ARVERNE  Animal Abuse      38
          Blocked Driveway  35
          Derelict Vehicle  27
          Disorderly Youth   2
          Drinking          1
          ...
Woodside  Blocked Driveway      11
          Derelict Vehicle      2
          Illegal Parking    100
          Noise - Commercial    2
          Noise - Street/Sidewalk 5
Name: Actual Resolution Time, Length: 764, dtype: int64
```

```
In [51]: city_mean_df = grouped_nyc['Actual Resolution Time'].mean()
```

```
In [52]: city_mean_df
```

```
Out[52]: City      Complaint Type
ARVERNE  Animal Abuse      128.894737
          Blocked Driveway  151.200000
          Derelict Vehicle  177.740741
          Disorderly Youth  215.000000
```

```

                Drinking                14.000000
                ...
Woodside Blocked Driveway            384.181818
        Derelict Vehicle            298.000000
        Illegal Parking            312.830000
        Noise - Commercial            143.000000
        Noise - Street/Sidewalk        204.400000
Name: Actual Resolution Time. Length: 764. dtype: float64

```

Hypothesis Definition and Testing

Question 1: Whether the average response time across complaint types is similar or not (overall)

Hypothesis #1. Are complaints resolved in the same mean time across categories?

Null Hypothesis: Mean resolution time across different complaints handled by the NYC police department are same

Alternate Hypothesis: Mean resolution time across different complaints handled by the NYC police department are not the same

Preparing the dataset required for analysis

```
In [67]: #taking a subset including only the city name, complaint type, caseid and
df_nyc_analysis = df_nyc_location[['City', 'Complaint Type', 'Location Type']]
df_nyc_analysis.head(5)
```

```
Out[67]:
```

	City	Complaint Type	Location Type	Actual Resolution Time
0	NEW YORK	Noise - Street/Sidewalk	Street/Sidewalk	55
1	ASTORIA	Blocked Driveway	Street/Sidewalk	86
2	BRONX	Blocked Driveway	Street/Sidewalk	291
3	BRONX	Illegal Parking	Street/Sidewalk	465
4	ELMHURST	Illegal Parking	Street/Sidewalk	207

```
In [85]: #taking the required subset for the ANOVA test for complaint type and reso
df_nyc_complaints = df_nyc_analysis[['Complaint Type', 'Actual Resolution Time']]
df_nyc_complaints = df_nyc_complaints.rename(columns = {
    'Complaint Type': 'Complaint',
    'Actual Resolution Time': 'Time'
}, inplace = False)
```

```
In [86]: df_nyc_complaints.head(5)
```

```
Out[86]:
```

	Complaint	Time
0	Noise - Street/Sidewalk	55
1	Blocked Driveway	86

	Complaint	Time
2	Blocked Driveway	291
2	Illegal Parking	465

```
In [89]: df_nyc_complaints.groupby('Complaint').mean()
```

```
Out[89]:
```

	Complaint	Time
	Agency Issues	315.333333
	Animal Abuse	312.512490
	Animal in a Park	20210.000000
	Bike/Roller/Skate Chronic	225.693396
	Blocked Driveway	284.142643
	Derelict Vehicle	441.499119
	Disorderly Youth	213.167832
	Drinking	231.391373
	Graffiti	428.752212
	Homeless Encampment	261.000907
	Illegal Fireworks	165.470238
	Illegal Parking	269.724982
	Noise - Commercial	188.524159
	Noise - House of Worship	191.287406
	Noise - Park	204.239741
	Noise - Street/Sidewalk	206.425917
	Noise - Vehicle	215.039220
	Panhandling	262.072131
	Posting Advertisement	118.262751
	Squeegee	242.500000
	Traffic	206.695526
	Urinating in Public	217.302365
	Vending	240.542578

On the outset, it seems that the mean time to solve different types of complaints is not the same. We will use the ANOVA test to analyse if this is the case

```
In [78]: import statsmodels.api as sm
from statsmodels.formula.api import ols
```

```
In [87]: lm = ols('Time~Complaint', data = df_nyc_complaints).fit()
table = sm.stats.anova_lm(lm)
print(table)
```

	df	sum_sq	mean_sq	F	PR(>F)
Complaint	22.0	1.454406e+09	6.610938e+07	513.960687	0.0
Residual	298448.0	3.838856e+10	1.286273e+05	NaN	NaN

The p value that is derived from the ANOVA test is very insignificant even with a confidence interval of 0.05 and so we reject the null hypothesis.

Hence the mean time to resolve across various types of complaints is not equal

```
In [90]: #examining if the animal in the park is the outlier and if removed the mean
df_nyc_animals = df_nyc_complaints[df_nyc_complaints['Complaint']=='Animal']
df_nyc_animals.value_counts()
```

```
Out[90]: Complaint      Time
Animal in a Park  20210      1
dtype: int64
```

```
In [92]: #removing this datapoint alone to examine the dataset again using the ANOVA
df_nyc_noanimals = df_nyc_complaints[df_nyc_complaints['Complaint']!='Animal']
```

```
In [94]: df_nyc_noanimals.groupby('Complaint').mean()
```

```
Out[94]:
```

	Time
Complaint	
Agency Issues	315.333333
Animal Abuse	312.512490
Bike/Roller/Skate Chronic	225.693396
Blocked Driveway	284.142643
Derelict Vehicle	441.499119
Disorderly Youth	213.167832
Drinking	231.391373
Graffiti	428.752212
Homeless Encampment	261.000907
Illegal Fireworks	165.470238
Illegal Parking	269.724982
Noise - Commercial	188.524159
Noise - House of Worship	191.287406
Noise - Park	204.239741
Noise - Street/Sidewalk	206.425917
Noise - Vehicle	215.039220
Panhandling	262.072131
Posting Advertisement	118.262751
Squeegee	242.500000
Traffic	206.695526
Urinating in Public	217.302365
Vending	240.542578

```
In [93]: no_animals = ols('Time~Complaint', data = df_nyc_noanimals).fit()
table_noanimals = sm.stats.anova_lm(no_animals)
print(table_noanimals)
```

	df	sum_sq	mean_sq	F	PR(>F)
Complaint	21.0	1.056344e+09	5.030210e+07	391.068618	0.0
Residual	298448.0	3.838856e+10	1.286273e+05	NaN	NaN

Even after removing the possible outlier of 'Animal in a Park', the p value that is derived from the ANOVA test is very insignificant even with a confidence interval of 0.05 and so we reject the null hypothesis.

Hence the mean time to resolve across various types of complaints is not equal

ANOVA Conclusion (Hypothesis #1): The mean time to resolve various types of complaints is not similar

Examining the same across the prominent cities of Brooklyn and Bronx

Hypothesis #2

Null Hypothesis: Mean resolution time across different complaints across the city 'BROOKLYN' are same

Alternate Hypothesis: Mean resolution time across different complaints across the city 'BROOKLYN' are not the same

```
In [106... #getting the necessary dataset
df_brooklyn = df_nyc_analysis[df_nyc_analysis['City']=='BROOKLYN']
df_brooklyn.head(5)
```

	City	Complaint Type	Location Type	Actual Resolution Time
5	BROOKLYN	Illegal Parking	Street/Sidewalk	113
9	BROOKLYN	Blocked Driveway	Street/Sidewalk	83
13	BROOKLYN	Illegal Parking	Street/Sidewalk	510
17	BROOKLYN	Noise - Commercial	Club/Bar/Restaurant	51
18	BROOKLYN	Noise - Commercial	Club/Bar/Restaurant	176

```
In [107... df_brooklyn = df_brooklyn.rename(columns = {
    'Complaint Type': 'Complaint',
    'Actual Resolution Time': 'Time'
}, inplace = False)
```

```
In [108... df_brooklyn.drop(['City', 'Location Type'], axis = 1)
```

	Complaint	Time
5	Illegal Parking	113
9	Blocked Driveway	83
13	Illegal Parking	510
17	Noise - Commercial	51

	Complaint	Time
18	Noise - Commercial	176
...
300681	Noise - Street/Sidewalk	174
300682	Noise - Commercial	385
300683	Noise - Street/Sidewalk	175
300689	Noise - Street/Sidewalk	218
300695	Noise - Commercial	187

00075 rows × 3 columns

```
In [100]: df_brooklyn['Complaint'].value_counts()
```

```
Out[100]: Blocked Driveway      28139
Illegal Parking      27454
Noise - Street/Sidewalk  13354
Noise - Commercial    11458
Derelict Vehicle      5179
Noise - Vehicle       5176
Animal Abuse          2393
Noise - Park          1554
Traffic               1085
Homeless Encampment    855
Vending               514
Noise - House of Worship  340
Drinking              257
Urinating in Public   136
Bike/Roller/Skate Chronic  111
Disorderly Youth       72
Illegal Fireworks      61
Panhandling            49
Posting Advertisement   45
Graffiti              43
Name: Complaint, dtype: int64
```

```
In [101]: df_brooklyn.groupby('Complaint').mean()
```

```
Out[101]:
```

	Time
Complaint	
Animal Abuse	289.633932
Bike/Roller/Skate Chronic	299.972973
Blocked Driveway	264.310743
Derelict Vehicle	356.535238
Disorderly Youth	248.680556
Drinking	212.124514
Graffiti	494.232558
Homeless Encampment	281.561404
Illegal Fireworks	140.180328
Illegal Parking	255.982298
Noise - Commercial	178.859749

	Time
Complaint	
Noise - House of Worship	183.873529
Noise - Park	188.453668
Noise - Street/Sidewalk	197.460761
Noise - Vehicle	196.709428
Panhandling	258.673469
Posting Advertisement	201.422222
Traffic	186.470046

In [102...

```
#ANOVA test for brooklyn
brooklyn = ols('Time~Complaint', data = df_brooklyn).fit()
table_brooklyn = sm.stats.anova_lm(brooklyn)
print(table_brooklyn)
```

	df	sum_sq	mean_sq	F	PR(>F)
Complaint	19.0	1.903002e+08	1.001580e+07	86.69383	0.0
Residual	98255.0	1.135147e+10	1.155307e+05	NaN	NaN

Examining the brooklyn dataset, the p value that is derived from the ANOVA test is very insignificant even with a confidence interval of 0.05 and so we reject the null hypothesis.

Hence the mean time to resolve across various types of complaints in brooklyn is not equal

ANOVA Conclusion (Hypothesis #2): The mean time to resolve various types of complaints in brooklyn is not similar

Hypothesis #3:

Null Hypothesis: Mean resolution time across different complaints across the city 'BRONX' are same

Alternate Hypothesis: Mean resolution time across different complaints across the city 'BRONX' are not the same

In [105...

```
#getting the necessary dataset
df_bronx = df_nyc_analysis[df_nyc_analysis['City']=='BRONX']
df_bronx.head(5)
```

Out[105...

	City	Complaint Type	Location Type	Actual Resolution Time
2	BRONX	Blocked Driveway	Street/Sidewalk	291
3	BRONX	Illegal Parking	Street/Sidewalk	465
7	BRONX	Blocked Driveway	Street/Sidewalk	107
11	BRONX	Blocked Driveway	Street/Sidewalk	667
12	BRONX	Noise - Street/Sidewalk	Street/Sidewalk	148

```
In [109... df_bronx = df_bronx.rename(columns = {  
    'Complaint Type': 'Complaint',  
    'Actual Resolution Time': 'Time'  
}, inplace = False)
```

```
In [110... df_bronx.drop(['City', 'Location Type'], axis = 1)
```

```
Out[110...      Complaint  Time  
2      Blocked Driveway  291  
3      Illegal Parking  465  
7      Blocked Driveway  107  
11     Blocked Driveway  667  
12  Noise - Street/Sidewalk  148  
...      ...      ...  
300643      Illegal Parking  176  
300652      Blocked Driveway  105  
300656      Blocked Driveway  450  
300690      Illegal Parking  486  
300696      Noise - Commercial  245
```

40690 rows × 2 columns

```
In [111... df_bronx['Complaint'].value_counts()
```

```
Out[111... Blocked Driveway      12751  
Noise - Street/Sidewalk      8890  
Illegal Parking      7857  
Noise - Vehicle      3395  
Noise - Commercial      2433  
Derelict Vehicle      1952  
Animal Abuse      1415  
Noise - Park      547  
Vending      379  
Traffic      355  
Homeless Encampment      247  
Drinking      188  
Noise - House of Worship      79  
Disorderly Youth      63  
Urinating in Public      51  
Illegal Fireworks      24  
Bike/Roller/Skate Chronic      20  
Panhandling      19  
Posting Advertisement      16  
Graffiti      9  
Name: Complaint, dtype: int64
```

```
In [112... df_bronx.groupby('Complaint').mean()
```

```
Out[112...      Time  
Complaint  
Animal Abuse  439.826855
```

	Time
Complaint	
Bike/Roller/Skate Chronic	207.300000
Blocked Driveway	375.418948
Derelict Vehicle	553.342725
Disorderly Youth	254.015873
Drinking	347.297872
Graffiti	533.666667
Homeless Encampment	446.222672
Illegal Fireworks	336.333333
Illegal Parking	394.493445
Noise - Commercial	281.542951
Noise - House of Worship	273.265823
Noise - Park	281.606947
Noise - Street/Sidewalk	313.278965
Noise - Vehicle	333.379676
Panhandling	852.684211
Posting Advertisement	213.250000
Traffic	295.121127

In [113]...

```
#ANOVA test for bronx
bronx = ols('Time~Complaint', data = df_bronx).fit()
table_bronx = sm.stats.anova_lm(bronx)
print(table_bronx)
```

	df	sum_sq	mean_sq	F	PR(>F)
Complaint	19.0	1.446558e+08	7.613462e+06	36.908754	3.333669e-135
Residual	40670.0	8.389324e+09	2.062779e+05	NaN	NaN

Examining the bronx dataset, the p value that is derived from the ANOVA test is very insignificant even with a confidence interval of 0.05 and so we reject the null hypothesis.

Hence the mean time to resolve across various types of complaints in bronx is not equal

ANOVA Conclusion (Hypothesis #3): The mean time to resolve various types of complaints in bronx is not similar

Answer 1: The mean time to resolve various types of complaints is not similar; it is also not similar when we examine it across the top cities as well

Question 2: Are the type of complaint or service requested and location related?

```
In [126... #taking subset of the nyc_complaints dataset required for analysis
df_nyc_services = df_nyc_location[['City', 'Complaint Type', 'Location Type']]
```

```
In [127... df_nyc_services.head(5)
```

```
Out[127...
      City      Complaint Type  Location Type
0  NEW YORK  Noise - Street/Sidewalk  Street/Sidewalk
1   ASTORIA   Blocked Driveway  Street/Sidewalk
2   BRONX     Blocked Driveway  Street/Sidewalk
3   BRONX      Illegal Parking  Street/Sidewalk
4  ELMHURST      Illegal Parking  Street/Sidewalk
```

```
In [128... #renaming columns for ease of use
df_nyc_services = df_nyc_services.rename(columns = {
    'Complaint Type': 'Complaint',
    'Location Type': 'Location'
}, inplace = False)
```

```
In [129... df_nyc_services.head(5)
```

```
Out[129...
      City      Complaint      Location
0  NEW YORK  Noise - Street/Sidewalk  Street/Sidewalk
1   ASTORIA   Blocked Driveway  Street/Sidewalk
2   BRONX     Blocked Driveway  Street/Sidewalk
3   BRONX      Illegal Parking  Street/Sidewalk
4  ELMHURST      Illegal Parking  Street/Sidewalk
```

```
In [130... #dropping the city column & using a label encoder for the complaints and l
df_nyc_nocity = df_nyc_services.drop('City', axis=1)
```

```
In [131... from sklearn.preprocessing import LabelEncoder
number = LabelEncoder()
df_nyc_nocity['Complaint'] = number.fit_transform(df_nyc_nocity['Complaint'])
df_nyc_nocity['Location'] = number.fit_transform(df_nyc_nocity['Location'])
```

```
In [132... df_nyc_nocity.head(5)
```

```
Out[132...
      Complaint  Location
0           15         13
1            4         13
2            4         13
```

	Complaint	Location
3	11	13

Hypothesis #4

Null Hypothesis: There is no relationship between the complaint type and the location

Alternate Hypothesis: There exists some relationship between the complaint type and the location

```
In [133... #spearman's rank correlation to check if the  
from scipy.stats import spearmanr
```

```
In [134... data1 = df_nyc_nocity['Complaint']  
data2 = df_nyc_nocity['Location']
```

```
In [137... #converting the pandas series to numpy array  
data1.to_numpy()  
data2.to_numpy()
```

```
Out[137... array([13, 13, 13, ..., 1, 1, 12])
```

```
In [138... stat, p = spearmanr(data1, data2)
```

```
In [139... print('stat=%.3f, p=%.3f' % (stat, p))  
stat=-0.143, p=0.000
```

```
In [140... if p > 0.05:  
    print('The two variables are probably independent')  
else:  
    print('The two variables are probably dependent')
```

The two variables are probably dependent

According to spearman's rank correlation, the two variables are probably interdependent

Spearman's rank correlation result:

There seems to exist some form of correlation which might be monotonic in nature between the complaints raised and the location

Performing a Chi-squared test to confirm

```
In [141... from scipy.stats import chi2_contingency
```

```
In [142... table = [data1,data2]
```

```
In [143... stat_chi, p_chi, dof, expected = chi2_contingency(table)  
print('stat=%.3f, p=%.3f' % (stat, p))  
if p_chi > 0.05:  
    print('Probably independent')  
else:  
    print('Probably dependent')
```

```
stat=-0.143, p=0.000  
Probably dependent
```

In []:

There seems to exist some form of correlation between the complaints raised and the location

Answer 2: There is a some correlation between the complaints raised and the location. This is explained by the low value of p in the spearman's correlation