```
# Email: rahulkumarr2080@gmail.com
# Date: 31st July 2021

In [2]: #importing required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from datetime import datetime
plt.rcParams["figure.figsize"] = (20,7)
```

Task 1 - Import a 311 NYC service request.

Author: Rahul ArvindKumar Thakur

In [1]:

In [3]:	<pre># 1)Import a 311 NYC service request. customer_service_request=pd.read_csv('DataSets/311_Service_Requests_from_2010_to_Presen</pre>
	#Giving low Memory Warning so put low_memory=False above

In [4]:	<pre>#Check CSV Imported or not, This code will help to view top 5 Rows customer service request.head()</pre>
	casesmen_ser vice_n eques evineaa()

Out[4]:		Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Type	Incid
	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	100:
	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	111(
	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	104!
	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	1046
	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	1137

5 rows × 53 columns

In [5]: # Check Number of Rows(observations) & Number of Columns(Variables)
customer_service_request.shape

```
In [6]:
          #This means this Dataset has 300698 Rows(observations) and 53 Columns(Variables)
 In [7]:
          Data Preprocessing - Cleaning up the data and make it ready for building
          In [8]:
          # Check the duplicate columns or Variables with duplicate name
          customer service request.columns
Out[8]: 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 [9]:
          # list of empty columns name given in data description they are of no use so we have to
          empty_data_cols = ['Landmark','School Not Found','School or Citywide Complaint','Vehicl
          'Taxi Pick Up Location', 'Bridge Highway Name', 'Bridge Highway Direction', 'Road Ramp',
          'Bridge Highway Segment', 'Garage Lot Name', 'Ferry Direction', 'Ferry Terminal Name']
          unspecified_data_cols =['Park Facility Name','School Name','School Number','School Regi
          'School Phone Number', 'School Address', 'School City', 'School State', 'School Zip']
          # Deleteing both Empty Columns and Unspecified columns from data frame
          customer_service_request= customer_service_request.drop(empty_data_cols, axis=1)
          customer service request= customer service request.drop(unspecified data cols, axis=1)
          customer service request.shape
Out[9]: (300698, 30)
In [10]:
          # Deleting duplicate representation in multiple columns
          # Location with lat long info already avaiable in Lattitude and Longtitude. Deleting lo
          customer_service_request= customer_service_request.drop(['Latitude','Longitude'], axis=
          customer service request.shape
Out[10]: (300698, 28)
In [11]:
          #Removing ID Column 'Unique Key' which doesnt have any significance in analytics model
          customer service request = customer service request.drop(['Unique Key'],axis=1)
```

Out[5]: (300698, 53)

```
# Now we have to identify columns which has null or empty or single value and these col
In [12]:
          # any significance while processing model
          single value columns = []
          for i in customer service request.columns:
              if len(customer_service_request[i].value_counts()) ==1:
                  single value columns.append(i)
                  print('Following columns has single value ', single value columns)
          #Deleting single value columns from dataframe
          customer_service_request= customer_service_request.drop(single_value_columns, axis=1)
         Following columns has single value ['Agency']
         Following columns has single value ['Agency', 'Facility Type']
In [13]:
          # Removing columns which has more than 70% data is null or empty and doesnt have any im
          customer_service_request.isnull().sum().sort_values(ascending=False)
Out[13]: Intersection Street 2
                                            257336
         Intersection Street 1
                                            256840
         Cross Street 2
                                            49779
         Cross Street 1
                                             49279
         Incident Address
                                             44410
         Street Name
                                             44410
         Descriptor
                                              5914
         Y Coordinate (State Plane)
                                              3540
         X Coordinate (State Plane)
                                              3540
         Location
                                              3540
         Address Type
                                              2815
         Incident Zip
                                              2615
         City
                                              2614
         Resolution Action Updated Date
                                              2187
         Closed Date
                                              2164
         Location Type
                                              131
         Due Date
                                                 3
         Status
                                                 0
         Resolution Description
                                                 0
                                                 0
         Community Board
         Borough
                                                 0
         Complaint Type
                                                 0
         Agency Name
                                                 0
         Park Borough
                                                 0
         Created Date
         dtype: int64
In [14]:
          # Above cell listed Intersection Street 2 and Intersection Street 1 has more than 70% d
          # doesnt have much impact on analytics model building.hence dropping these columns
          customer service request =customer service request.drop(['Intersection Street 1','Inter
          customer service request.shape
Out[14]: (300698, 23)
In [15]:
          # Seperating object columns and non object columns from dataframe not required.
          # non object columns are Incident Zip, X Coordinate (State Plane), Y Coordinate (State
          # these columns represented in number format but these are more categorized data than n
          # hence converting these columns as object columns
          customer service request=customer service request.astype(
              {'Incident Zip':'str','X Coordinate (State Plane)':'str','Y Coordinate (State Plane
In [16]:
          #Converting string data to upper case to make model case insentive
```

```
for col in customer_service_request.columns:
    customer_service_request[col] = customer_service_request[col].str.upper()

customer_service_request.head()
```

\cap	4-	$\Gamma 1 \subset I$	
U	uц	TO	

Inc Ad	Incident Zip	Location Type	Descriptor	Complaint Type	Agency Name	Closed Date	Created Date	
VERM AV	10034.0	STREET/SIDEWALK	LOUD MUSIC/PARTY	NOISE - STREET/SIDEWALK	NEW YORK CITY POLICE DEPARTMENT	01-01- 16 0:55	12/31/2015 11:59:45 PM	0
27- AV	11105.0	STREET/SIDEWALK	NO ACCESS	BLOCKED DRIVEWAY	NEW YORK CITY POLICE DEPARTMENT	01-01- 16 1:26	12/31/2015 11:59:44 PM	1
VALE1 VA	10458.0	STREET/SIDEWALK	NO ACCESS	BLOCKED DRIVEWAY	NEW YORK CITY POLICE DEPARTMENT	01-01- 16 4:51	12/31/2015 11:59:29 PM	2
BA AV	10461.0	STREET/SIDEWALK	COMMERCIAL OVERNIGHT PARKING	ILLEGAL PARKING	NEW YORK CITY POLICE DEPARTMENT	01-01- 16 7:43	12/31/2015 11:57:46 PM	3
87- I	11373.0	STREET/SIDEWALK	BLOCKED SIDEWALK	ILLEGAL PARKING	NEW YORK CITY POLICE DEPARTMENT	01-01- 16 3:24	12/31/2015 11:56:58 PM	4

5 rows × 23 columns

```
In [17]:
```

```
In [18]:
```

#Filling empty/null columns of object columns with max occurance item value
for col in customer_service_request:
 customer_service_request[col]=customer_service_request[col].fillna(
 customer_service_request[col].value_counts().idxmax())
#Checking is there any null cells present in Dataframe
customer_service_request.isnull().sum().sort_values()

```
Out[18]: Created Date 0
Y Coordinate (State Plane) 0
X Coordinate (State Plane) 0
```

```
Borough
         Community Board
                                            0
         Resolution Action Updated Date
                                            0
         Resolution Description
                                            0
         Due Date
                                            0
                                            0
         Status
         City
                                            0
         Park Borough
                                            0
         Address Type
                                            0
         Cross Street 1
                                            0
         Street Name
                                            0
         Incident Address
                                            0
         Incident Zip
                                            0
         Location Type
                                            0
         Descriptor
                                            0
                                            0
         Complaint Type
                                            0
         Agency Name
         Closed Date
                                            0
         Cross Street 2
                                            0
         Location
         dtype: int64
          customer_service_request.Status.value_counts()
Out[19]: CLOSED
                     298471
         OPEN
                       1439
         ASSIGNED
                        786
         DRAFT
                          2
         Name: Status, dtype: int64
          # Converting all the date columns datatype from str object column to datetime object co
          date_cols =['Created Date', 'Closed Date', 'Due Date', 'Resolution Action Updated Date']
          for col in date cols:
              customer_service_request[col] = pd.to_datetime(customer_service_request[col])
          customer service request.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 300698 entries, 0 to 300697
         Data columns (total 23 columns):
          #
              Column
                                               Non-Null Count Dtype
                                               300698 non-null datetime64[ns]
          0
              Created Date
                                               300698 non-null datetime64[ns]
          1
              Closed Date
                                               300698 non-null object
          2
              Agency Name
          3
              Complaint Type
                                               300698 non-null object
                                               300698 non-null object
          4
              Descriptor
          5
              Location Type
                                               300698 non-null object
          6
              Incident Zip
                                               300698 non-null object
                                               300698 non-null object
          7
              Incident Address
          8
              Street Name
                                               300698 non-null object
          9
              Cross Street 1
                                               300698 non-null object
          10 Cross Street 2
                                               300698 non-null object
          11 Address Type
                                               300698 non-null object
          12 City
                                               300698 non-null object
          13 Status
                                               300698 non-null object
          14 Due Date
                                               300698 non-null datetime64[ns]
          15 Resolution Description
                                               300698 non-null object
          16
              Resolution Action Updated Date 300698 non-null datetime64[ns]
                                               300698 non-null object
          17
              Community Board
```

300698 non-null object

0

In [19]:

In [20]:

In [21]:

18

Borough

```
19 X Coordinate (State Plane)
                                                300698 non-null object
           20 Y Coordinate (State Plane)
                                                300698 non-null object
           21 Park Borough
                                                300698 non-null object
           22 Location
                                                300698 non-null object
          dtypes: datetime64[ns](4), object(19)
          memory usage: 52.8+ MB
In [22]:
          closed_status_requests = customer_service_request[customer_service_request['Status']=='
          non_closed_status_requests = customer_service_request[customer_service_request['Status'
          open status requests = customer service request[customer service request['Status']=='OP
          assigned status requests = customer service request[customer service request['Status']=
          draft status requests = customer service request[customer service request['Status']=='D
           print('Closed requests',closed_status_requests.shape)
           print('Not closed requests (OPEN+ASSIGNED+DRAFT)', non closed status requests.shape)
          print('Open requests',open status requests.shape)
          print('Assigned requests',assigned_status_requests.shape)
          print('Drafted requests',draft status requests.shape)
          Closed requests (298471, 23)
         Not closed requests (OPEN+ASSIGNED+DRAFT) (2227, 23)
          Open requests (1439, 23)
          Assigned requests (786, 23)
         Drafted requests (2, 23)
In [23]:
          #Function to range classify observations based on Request_Closing_Time for CLOSED statu
          def segregate_duration_range(duration):
              if duration <=60:</pre>
                   return "00-1 Hrs"
              elif duration >60 and duration<= 120:</pre>
                   return "1-2 Hrs"
              elif duration >120 and duration<= 240:</pre>
                   return "2-4 Hrs"
              elif duration >240 and duration<= 480:</pre>
                   return "4-8 Hrs"
              elif duration >480 and duration < 960:
                   return "8-16 Hrs"
              elif duration >960 and duration < 1440:
                   return "16- 24 Hrs"
              elif duration >1440 and duration< 2880:</pre>
                   return "1-2 Days"
              elif duration >2880 and duration< 5760:</pre>
                   return "2-4 Days"
              elif duration >5760 and duration< 11520:</pre>
                   return "4-8 Days"
              elif duration >11520 :
                   return "More than 8 Days"
```

Task 2 - Read or convert the columns 'Created Date' and Closed Date' to datetime datatype and create a new column 'Request_Closing_Time' as the time elapsed between request creation and request closing. (Hint: Explore the package/module datetime)

```
closed_status_requests['Closed-MM-YYYY']=''
# request_closing_time_index = closed_status_requests.columns.get_loc('Request_Closing_for index, csr in closed_status_requests.iterrows():
    Request_Closing_Time = int(((csr['Closed Date']-csr['Created Date']).total_seconds(
    closed_status_requests.at[index,'Request_Closing_Time'] = Request_Closing_Time
    closed_status_requests.at[index,'Resolved_Duration_Grp'] = segregate_duration_range
    closed_status_requests.head()
```

```
Both Created Date and Closed Date is available for closed CSRs 
<ipython-input-25-308fd748276a>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy closed_status_requests['Request_Closing_Time']=0 cipython-input-25-308fd748276a>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy closed_status_requests['Resolved_Duration_Grp']='' <ipython-input-25-308fd748276a>:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy closed_status_requests['Closed-MM-YYYY']=''

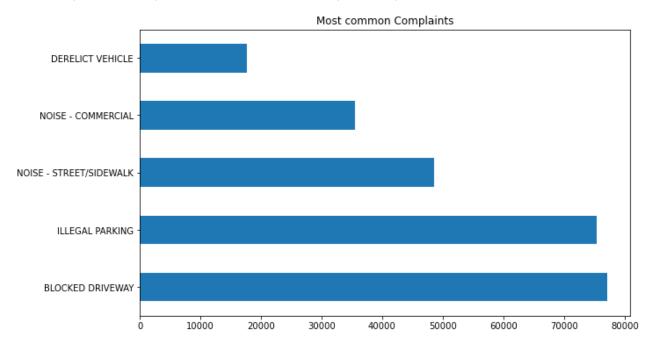
Out[25]:

	Created Date	Closed Date	Agency Name	Complaint Type	Descriptor	Location Type	Incident Zip	Incid Addı
0	2015- 12-31 23:59:45	2016- 01-01 00:55:00	NEW YORK CITY POLICE DEPARTMENT	NOISE - STREET/SIDEWALK	LOUD MUSIC/PARTY	STREET/SIDEWALK	10034.0	VERMIL' AVEI
1	2015- 12-31 23:59:44	2016- 01-01 01:26:00	NEW YORK CITY POLICE DEPARTMENT	BLOCKED DRIVEWAY	NO ACCESS	STREET/SIDEWALK	11105.0	27-07 AVEI
2	2015- 12-31 23:59:29	2016- 01-01 04:51:00	NEW YORK CITY POLICE DEPARTMENT	BLOCKED DRIVEWAY	NO ACCESS	STREET/SIDEWALK	10458.0	2 Valent Aver
3	2015- 12-31 23:57:46	2016- 01-01 07:43:00	NEW YORK CITY POLICE DEPARTMENT	ILLEGAL PARKING	COMMERCIAL OVERNIGHT PARKING	STREET/SIDEWALK	10461.0	2 BAIS AVEI
4	2015- 12-31 23:56:58	2016- 01-01 03:24:00	NEW YORK CITY POLICE DEPARTMENT	ILLEGAL PARKING	BLOCKED SIDEWALK	STREET/SIDEWALK	11373.0	87-1 ² RC

5 rows × 26 columns

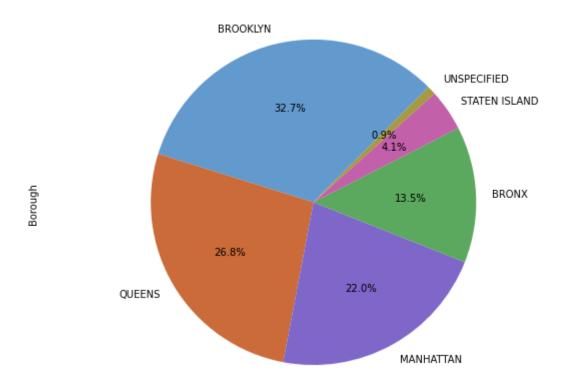
Task 3 - Provide major insights/patterns that you can offer in a visual format (graphs or tables); at least 4 major conclusions that you can come up with after generic data mining.

Out[26]: <AxesSubplot:title={'center':'Most common Complaints'}>



1) This graph shows that Blocked Driveway and Illegal Parking are the common complaints

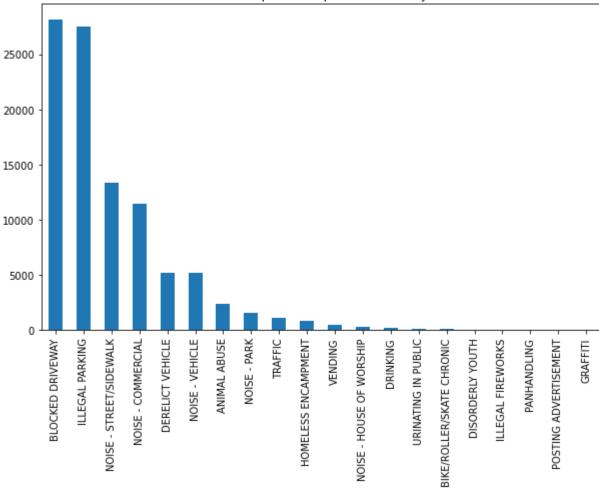
Complaints distribution across Boroughs



2) This Conclusion indicates that Brooklyn has more complaints with (32.7%) and after that Queens (26.8%)

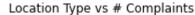
```
In [28]:
          #Conclusion - 3 Analysis for Brooklyn borough which has highest number of complains
          df_Brooklyn = customer_service_request[customer_service_request['Borough']=='BROOKLYN']
In [29]:
          (df_Brooklyn['Complaint Type'].value_counts()).head(25).plot(kind='bar',figsize=(10,6),
Out[29]: <AxesSubplot:title={'center':'Most Frequent Complaints in Brooklyn'}>
```

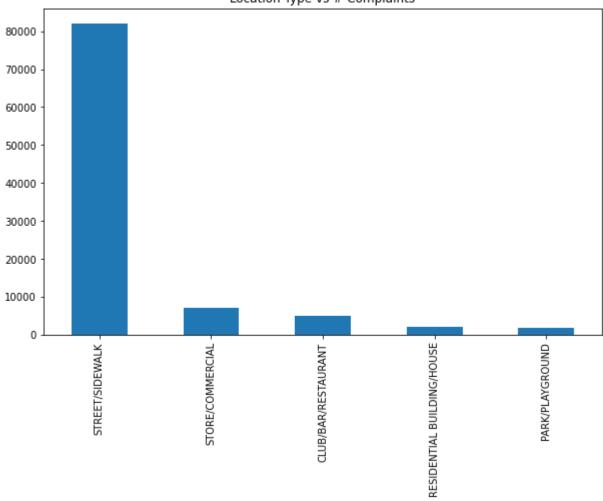




3) Blocked Driveway and Illegal Parking is the Most Frequent Complaints in Brooklyn

Out[30]: <AxesSubplot:title={'center':'Location Type vs # Complaints'}>

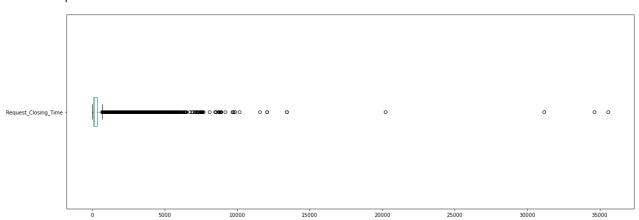




4) Street/Sidewalk has more complaints

#There are some outliers that means some of the requests taken longer time than the avg
closed_status_requests.Request_Closing_Time.plot(kind='box', vert=False)

Out[31]: <AxesSubplot:>



*Task 4 - Order the complaint types based on the average 'Request_Closing_Time', grouping them for different locations.

```
df = df.sort_values(by=['City','Request_Closing_Time'],ascending=False)
df.shape
```

Out[32]: (747, 1)

```
In [33]: #Storing the sorted results
    #df.to_csv('sorted_data.csv')
    with pd.option_context('display.max_rows', None, 'display.max_columns', None):
        print(df)
```

		Request_Closing_Time
City	Complaint Type	
WOODSIDE	BIKE/ROLLER/SKATE CHRONIC	728.500000
	DERELICT VEHICLE	560.562249
	GRAFFITI	539.333333
	ANIMAL ABUSE	506.101449
	VENDING	437.800000
	ILLEGAL PARKING	422.168517
	NOISE - PARK	404.710526
	HOMELESS ENCAMPMENT	402.696970
	NOISE - COMMERCIAL	398.445498
	NOISE - STREET/SIDEWALK	393.329412
	BLOCKED DRIVEWAY	388.239063
	URINATING IN PUBLIC	385.125000
	NOISE - VEHICLE	328.590476
	DRINKING	328.466667
	TRAFFIC	289.923077
	NOISE - HOUSE OF WORSHIP	284.000000
	ILLEGAL FIREWORKS	148.000000
	DISORDERLY YOUTH	73.000000
WOODHAVEN	DERELICT VEHICLE	446.672078
	HOMELESS ENCAMPMENT	426.000000
	ILLEGAL PARKING	343.778267
	BLOCKED DRIVEWAY	331.081209
	NOISE - STREET/SIDEWALK	314.000000
	ANIMAL ABUSE	297.844444
	NOISE - COMMERCIAL	233.205714
	URINATING IN PUBLIC	204.500000
	NOISE - VEHICLE	203.932432
	NOISE - HOUSE OF WORSHIP	198.333333
	DRINKING	171.666667
	VENDING	170.166667
	TRAFFIC	109.666667
	NOISE - PARK	82.666667
	BIKE/ROLLER/SKATE CHRONIC	74.500000
WHITESTONE	GRAFFITI	528.000000
	NOISE - COMMERCIAL	266.562500
	ILLEGAL FIREWORKS	262.000000
	DRINKING	217.500000
	NOISE - PARK	217.500000
	NOISE - STREET/SIDEWALK	207.030303
	DERELICT VEHICLE	204.555066
	NOISE - VEHICLE	201.642857
	DISORDERLY YOUTH	195.000000
	ILLEGAL PARKING	190.266667
	BLOCKED DRIVEWAY	188.913462
	BIKE/ROLLER/SKATE CHRONIC	188.500000
	ANIMAL ABUSE	168.107143
	TRAFFIC	162.823529
	VENDING	140.000000
SUNNYSIDE	ANIMAL ABUSE	692.800000
	DERELICT VEHICLE	589.100000

	VENDING	583.000000
	POSTING ADVERTISEMENT	435.500000
	HOMELESS ENCAMPMENT	431.090909
	BLOCKED DRIVEWAY	417.752427
	ILLEGAL PARKING	399.663934
	NOISE - COMMERCIAL	397.099379
	NOISE - PARK	387.533333
	NOISE - STREET/SIDEWALK	373.646154
	TRAFFIC	356.875000
	NOISE - VEHICLE	295.041667
	DRINKING	276.000000
	DISORDERLY YOUTH	224.000000
	URINATING IN PUBLIC	204.000000
	BIKE/ROLLER/SKATE CHRONIC	72.000000
	GRAFFITI	35.000000
STATEN ISLAND	GRAFFITI	573.500000
	DERELICT VEHICLE	302.134278
	HOMELESS ENCAMPMENT	298.464789
	ANIMAL ABUSE	297.874327
	PANHANDLING	280.000000
	VENDING	256.000000
	BIKE/ROLLER/SKATE CHRONIC	244.428571
	BLOCKED DRIVEWAY	243.966853
	DISORDERLY YOUTH	233.304348
	ILLEGAL PARKING	230.471853
	ILLEGAL FIREWORKS	223.900000
	TRAFFIC	214.190955
	DRINKING	209.251429
	NOISE - VEHICLE	193.946629
	NOISE - COMMERCIAL	179.722304
	NOISE - STREET/SIDEWALK	177.844363
	NOISE - PARK	175.805970
	NOISE - HOUSE OF WORSHIP	
	URINATING IN PUBLIC POSTING ADVERTISEMENT	148.428571
CDDINGETELD CARDENG	POSTING ADVERTISEMENT	92.225243
SPRINGFIELD GARDENS	ANIMAL ABUSE	1170.000000 891.375000
	DERELICT VEHICLE	702.411483
	TRAFFIC	620.363636
	ILLEGAL PARKING	550.428571
	BLOCKED DRIVEWAY	541.862595
	VENDING	411.000000
	PANHANDLING	381.000000
	ILLEGAL FIREWORKS	333.000000
	NOISE - VEHICLE	302.476190
	DRINKING	290.333333
	URINATING IN PUBLIC	287.000000
	NOISE - STREET/SIDEWALK	252.789474
	HOMELESS ENCAMPMENT	226.600000
	NOISE - COMMERCIAL	223.138889
	NOISE - HOUSE OF WORSHIP	222.000000
	NOISE - PARK	100.000000
SOUTH RICHMOND HILL	BIKE/ROLLER/SKATE CHRONIC	1854.000000
	DERELICT VEHICLE	712.311419
	TRAFFIC	346.272727
	DRINKING	345.739130
	ILLEGAL PARKING	342.190476
	HOMELESS ENCAMPMENT	319.909091
	VENDING	315.750000
	BLOCKED DRIVEWAY	291.661499
	ANIMAL ABUSE	261.807692
	NOISE - STREET/SIDEWALK	247.032967
	NOISE - COMMERCIAL	232.065657
	NOISE - VEHICLE	216.160494
	NOISE - PARK	215.500000

	DISORDERLY YOUTH	183.500000
	NOISE - HOUSE OF WORSHIP	119.000000
	ILLEGAL FIREWORKS	28.500000
SOUTH OZONE PARK	DERELICT VEHICLE	623.606145
	NOISE - PARK	342.250000
	DRINKING	292.538462
	ILLEGAL PARKING	281.447368
	BLOCKED DRIVEWAY	265.338641
	BIKE/ROLLER/SKATE CHRONIC	243.000000
	NOISE - VEHICLE	227.282353
	NOISE - STREET/SIDEWALK	219.371429
	TRAFFIC	213.071429
	VENDING	210.800000
	ANIMAL ABUSE	209.218182
	NOISE - COMMERCIAL	200.142857
	DISORDERLY YOUTH	160.500000
	NOISE - HOUSE OF WORSHIP	
	HOMELESS ENCAMPMENT	109.750000
	URINATING IN PUBLIC	101.500000
	POSTING ADVERTISEMENT	77.000000
	ILLEGAL FIREWORKS	30.000000
SAINT ALBANS	HOMELESS ENCAMPMENT	453.875000
SAINT ALDANS	ANIMAL ABUSE	441.933333
	URINATING IN PUBLIC	390.000000
	DERELICT VEHICLE	
		353.722772
	BLOCKED DRIVEWAY	280.872951
	NOISE - HOUSE OF WORSHIP ILLEGAL PARKING	278.000000 250.712707
	TRAFFIC	237.727273
	NOISE - VEHICLE	
		207.195122
	NOISE - STREET/SIDEWALK	
	NOISE - COMMERCIAL	185.758621
	VENDING	143.000000
	DISORDERLY YOUTH	105.000000
	DRINKING	57.333333
DOCEDAL E	NOISE - PARK	49.000000
ROSEDALE	HOMELESS ENCAMPMENT	1131.250000
	ANIMAL ABUSE	875.333333
	DERELICT VEHICLE	870.149038
	TRAFFIC	631.913043
	ILLEGAL PARKING	613.191336
	BIKE/ROLLER/SKATE CHRONIC	511.500000
	NOISE - COMMERCIAL	465.280000
	BLOCKED DRIVEWAY	447.170616
	VENDING	407.625000
	NOISE - STREET/SIDEWALK	352.083333
	NOISE - VEHICLE	350.600000
	NOISE - PARK	346.376812
	DRINKING	330.000000
	NOISE - HOUSE OF WORSHIP	270.000000
	GRAFFITI	9.000000
ROCKAWAY PARK	TRAFFIC	213.142857
	DISORDERLY YOUTH	205.000000
	VENDING	176.500000
	DERELICT VEHICLE	163.555556
	ILLEGAL PARKING	156.958991
	BLOCKED DRIVEWAY	144.728571
	DRINKING	137.950000
	ANIMAL ABUSE	136.700000
	NOISE - VEHICLE	135.185185
	NOISE - COMMERCIAL	120.555556
	NOISE - STREET/SIDEWALK	109.698413
	HOMELESS ENCAMPMENT	102.750000
	NOISE - PARK	61.500000
	URINATING IN PUBLIC	43.000000

RIDGEWOOD	DERELICT VEHICLE	454.278788
	ANIMAL ABUSE	385.418803
	HOMELESS ENCAMPMENT	346.347826
	ILLEGAL PARKING	265.799674
	NOISE - VEHICLE	260.714286
	DRINKING	260.600000
	URINATING IN PUBLIC	248.625000
	NOISE - STREET/SIDEWALK	247.411085
	BLOCKED DRIVEWAY	240.255759
	DISORDERLY YOUTH	239.666667
	NOISE - COMMERCIAL	220.690955
	TRAFFIC	217.761905
	NOISE - PARK	215.357143
	BIKE/ROLLER/SKATE CHRONIC	208.333333
	NOISE - HOUSE OF WORSHIP	184.000000
	VENDING	182.625000
	ILLEGAL FIREWORKS	64.500000
	GRAFFITI	59.000000
	POSTING ADVERTISEMENT	14.000000
RICHMOND HILL	DERELICT VEHICLE	575.240964
	HOMELESS ENCAMPMENT	484.785714
	URINATING IN PUBLIC	376.800000
	VENDING	348.076923
	ILLEGAL PARKING	346.421320
	ANIMAL ABUSE	332.093750
	NOISE - STREET/SIDEWALK	308.632184
	BLOCKED DRIVEWAY	302.847302
	NOISE - COMMERCIAL	245.064815
	NOISE - VEHICLE	244.312500
	ILLEGAL FIREWORKS	208.750000
	TRAFFIC	195.571429
	POSTING ADVERTISEMENT	180.000000
	DRINKING	176.888889
	GRAFFITI	145.000000
DECO DADI/	NOISE - PARK	134.250000
REGO PARK	NOISE - HOUSE OF WORSHIP	558.000000
	URINATING IN PUBLIC	335.000000
	VENDING	306.333333 272.716049
	DERELICT VEHICLE	
	ANIMAL ABUSE TRAFFIC	265.153846 225.571429
	NOISE - PARK	224.772727
	BLOCKED DRIVEWAY	215.875614
	ILLEGAL PARKING	201.718808
	NOISE - VEHICLE	176.534884
	NOISE - STREET/SIDEWALK	155.982456
	DRINKING	153.250000
	NOISE - COMMERCIAL	143.253165
	HOMELESS ENCAMPMENT	91.333333
	GRAFFITI	56.000000
QUEENS VILLAGE	GRAFFITI	3200.000000
Q022/13 V222/102	DERELICT VEHICLE	964.313514
	VENDING	875.500000
	ANIMAL ABUSE	768.954545
	NOISE - COMMERCIAL	602.000000
	ILLEGAL PARKING	595.823529
	BLOCKED DRIVEWAY	571.994872
	HOMELESS ENCAMPMENT	562.666667
	PANHANDLING	543.000000
	NOISE - VEHICLE	532.414634
	TRAFFIC	425.615385
	URINATING IN PUBLIC	380.000000
	NOISE - STREET/SIDEWALK	334.590909
	DRINKING	280.600000
	NOISE - PARK	211.000000

	ILLEGAL FIREWORKS	197.000000
	POSTING ADVERTISEMENT	183.000000
	NOISE - HOUSE OF WORSHIP	150.500000
QUEENS	ANIMAL IN A PARK	20210.000000
	DERELICT VEHICLE	521.000000
	HOMELESS ENCAMPMENT	432.000000
	BLOCKED DRIVEWAY	239.500000
	NOISE - STREET/SIDEWALK	219.000000
	ILLEGAL PARKING	205.000000
	NOISE - HOUSE OF WORSHIP	156.000000
	TRAFFIC	126.500000
	NOISE - VEHICLE	79.500000
	NOISE - COMMERCIAL	79.166667
	URINATING IN PUBLIC	21.000000
OZONE PARK	DERELICT VEHICLE	640.380952
	BIKE/ROLLER/SKATE CHRONIC	431.000000
	ILLEGAL PARKING	306.898223
	POSTING ADVERTISEMENT	298.000000
	ANIMAL ABUSE	297.500000
	BLOCKED DRIVEWAY	297.184420
	PANHANDLING	274.571429
	URINATING IN PUBLIC	263.500000
	TRAFFIC	260.263158
	VENDING	256.000000
	DRINKING	236.421053
	NOISE - COMMERCIAL	234.217391
	NOISE - STREET/SIDEWALK	229.080292
	NOISE - VEHICLE	200.126761
	NOISE - PARK	174.055556
	DISORDERLY YOUTH	131.750000
	HOMELESS ENCAMPMENT	117.500000
	NOISE - HOUSE OF WORSHIP	50.000000
	ILLEGAL FIREWORKS	19.000000
OAKLAND GARDENS	HOMELESS ENCAMPMENT	1719.000000
	VENDING	227.000000
	DERELICT VEHICLE	222.802326
	ANIMAL ABUSE	165.526316
	NOISE - PARK	164.714286
	BLOCKED DRIVEWAY	152.477273
	DRINKING	146.000000
	TRAFFIC	144.333333
	NOISE - VEHICLE	138.600000
	ILLEGAL PARKING	137.273764
	NOISE - STREET/SIDEWALK	98.421053
	DISORDERLY YOUTH	85.000000
NEU VODY	BIKE/ROLLER/SKATE CHRONIC	67.000000
NEW YORK	GRAFFITI	303.545455
	DERELICT VEHICLE	255.636872 242.500000
	SQUEEGEE	
	ANIMAL ABUSE HOMELESS ENCAMPMENT	220.807869 220.795381
	BLOCKED DRIVEWAY	213.216425
	PANHANDLING	208.155440
	ILLEGAL PARKING	203.007341
	VENDING	198.110555
	DRINKING	183.159322
	POSTING ADVERTISEMENT	177.048780
	NOISE - PARK	176.380165
	URINATING IN PUBLIC	173.338645
	BIKE/ROLLER/SKATE CHRONIC	173.226667
	NOISE - COMMERCIAL	163.794211
	NOISE - STREET/SIDEWALK	163.690022
	TRAFFIC	158.785252
	NOISE - VEHICLE	157.015682
	DISORDERLY YOUTH	144.130435

	NOISE - HOUSE OF WORSHIP	138.020619
NEW 10/05 DADY	ILLEGAL FIREWORKS	102.944444
NEW HYDE PARK	DERELICT VEHICLE BLOCKED DRIVEWAY	467.714286 464.018868
	ILLEGAL PARKING	455.178571
	NOISE - VEHICLE	200.000000
	ANIMAL ABUSE	115.000000
MIDDLE VILLAGE	BIKE/ROLLER/SKATE CHRONIC	940.000000
	DERELICT VEHICLE	494.395270
	ANIMAL ABUSE	412.545455
	NOISE - COMMERCIAL	306.200000
	ILLEGAL PARKING NOISE - STREET/SIDEWALK	299.583808 293.378378
	NOISE - STREET/SIDEWALK	284.214286
	NOISE - PARK	272.500000
	BLOCKED DRIVEWAY	261.838074
	HOMELESS ENCAMPMENT	260.800000
	TRAFFIC	211.416667
	DRINKING	74.000000
MASPETH	NOISE - PARK	661.666667
	BIKE/ROLLER/SKATE CHRONIC NOISE - HOUSE OF WORSHIP	530.000000 474.000000
	DERELICT VEHICLE	463.519630
	ANIMAL ABUSE	436.444444
	HOMELESS ENCAMPMENT	430.600000
	ILLEGAL PARKING	329.421107
	NOISE - COMMERCIAL	295.269231
	BLOCKED DRIVEWAY	288.632514
	VENDING	277.833333
	NOISE - VEHICLE NOISE - STREET/SIDEWALK	275.789474 252.280992
	DISORDERLY YOUTH	240.500000
	TRAFFIC	224.890909
	URINATING IN PUBLIC	221.000000
	DRINKING	139.333333
	ILLEGAL FIREWORKS	97.000000
LONG ISLAND CITY		626.854271
	VENDING	556.500000
	ILLEGAL PARKING HOMELESS ENCAMPMENT	438.230769 420.700000
	BIKE/ROLLER/SKATE CHRONIC	405.333333
	ANIMAL ABUSE	363.266667
	BLOCKED DRIVEWAY	357.853598
	DISORDERLY YOUTH	329.000000
	NOISE - PARK	319.685185
	NOISE - VEHICLE	303.457944
	NOISE - COMMERCIAL URINATING IN PUBLIC	267.024194 251.333333
	NOISE - STREET/SIDEWALK	248.389262
	PANHANDLING	233.500000
	GRAFFITI	210.000000
	TRAFFIC	202.763889
	DRINKING	181.571429
	POSTING ADVERTISEMENT	49.000000
LITTLE NECK	DISORDERLY YOUTH	260.500000
	DERELICT VEHICLE URINATING IN PUBLIC	215.508197 184.000000
	ILLEGAL PARKING	170.710843
	BLOCKED DRIVEWAY	144.743802
	ANIMAL ABUSE	137.866667
	POSTING ADVERTISEMENT	134.000000
	TRAFFIC	131.764706
	NOISE - VEHICLE	128.600000
	NOISE - STREET/SIDEWALK DRINKING	118.625000 94.000000
	DIVINITING	34. WUUUUU

KEW GARDENS	NOISE - COMMERCIAL NOISE - PARK URINATING IN PUBLIC DERELICT VEHICLE VENDING BLOCKED DRIVEWAY ILLEGAL PARKING NOISE - COMMERCIAL TRAFFIC NOISE - VEHICLE ANIMAL ABUSE NOISE - STREET/SIDEWALK HOMELESS ENCAMPMENT NOISE - HOUSE OF WORSHIP	80.921053 64.000000 431.333333 431.142857 373.000000 367.479233 284.896226 231.115854 227.400000 217.333333 189.684211 154.700000 113.800000 69.000000
JAMAICA	DRINKING HOMELESS ENCAMPMENT DERELICT VEHICLE VENDING POSTING ADVERTISEMENT GRAFFITI NOISE - HOUSE OF WORSHIP URINATING IN PUBLIC ILLEGAL PARKING DISORDERLY YOUTH ANIMAL ABUSE BLOCKED DRIVEWAY TRAFFIC DRINKING	61.000000 484.379747 476.499475 450.350000 329.000000 324.666667 320.769231 306.242424 304.430683 304.375000 303.253275 298.505682 282.105357 264.441176
JACKSON HEIGHTS		235.344262 220.649701 207.509934 190.578947 174.50000 156.250000 104.000000 343.444444
HOWARD BEACH	HOMELESS ENCAMPMENT NOISE - PARK NOISE - STREET/SIDEWALK NOISE - VEHICLE PANHANDLING POSTING ADVERTISEMENT NOISE - HOUSE OF WORSHIP ILLEGAL FIREWORKS DERELICT VEHICLE NOISE - HOUSE OF WORSHIP NOISE - COMMERCIAL ILLEGAL PARKING ANIMAL ABUSE BLOCKED DRIVEWAY DISORDERLY YOUTH NOISE - STREET/SIDEWALK NOISE - PARK PANHANDLING TRAFFIC NOISE - VEHICLE	156.454545 144.875000 139.018433 132.431034 101.000000 83.000000 66.0000000 42.0000000 688.318841 408.000000 349.782946 345.145907 264.354839 254.857143 251.000000 236.428571 232.500000 198.000000 198.000000 185.333333 184.6000000

	HOMELESS ENCAMPMENT	158.666667
	BIKE/ROLLER/SKATE CHRONIC	158.000000
	DRINKING	120.000000
	ILLEGAL FIREWORKS	110.333333
	VENDING	108.000000
HOLLIS		
HULLIS	DERELICT VEHICLE	693.629371
	DRINKING	437.333333
	ILLEGAL PARKING	407.278146
	NOISE - COMMERCIAL	402.520000
	BLOCKED DRIVEWAY	287.406433
	ANIMAL ABUSE	261.969697
		256.777778
		242.839572
		237.000000
	DISORDERLY YOUTH	232.000000
	TRAFFIC	228.727273
	NOISE - PARK	210.588235
	NOISE - VEHICLE	191.148936
	NOISE - STREET/SIDEWALK	169.658537
GLEN OAKS	DERELICT VEHICLE	
GLEN UAKS		914.428571
	ANIMAL ABUSE	779.400000
	NOISE - STREET/SIDEWALK	714.500000
	NOISE - VEHICLE	685.250000
	BLOCKED DRIVEWAY	677.566667
	URINATING IN PUBLIC	660.500000
	ILLEGAL PARKING	537.675676
	NOISE - COMMERCIAL	366.679487
	TRAFFIC	332.666667
	VENDING	290.388889
	NOISE - PARK	263.891892
FRESH MEADOWS	HOMELESS ENCAMPMENT	353.600000
		270.415808
		237.675299
	DRINKING	223.500000
	NOISE - PARK	204.750000
	ANIMAL ABUSE	196.511111
	VENDING	157.000000
	ILLEGAL PARKING	156.692308
	NOISE - STREET/SIDEWALK	147.857143
		144.142857
	NOISE - VEHICLE	128.488636
	TRAFFIC	97.538462
	URINATING IN PUBLIC	96.000000
	PANHANDLING	94.000000
FOREST HILLS	POSTING ADVERTISEMENT	351.666667
	PANHANDLING	348.600000
	BIKE/ROLLER/SKATE CHRONIC	289.200000
	DISORDERLY YOUTH	249.000000
	DRINKING	246.000000
	NOISE - HOUSE OF WORSHIP	234.000000
	DERELICT VEHICLE	223.250000
	BLOCKED DRIVEWAY	222.594268
	HOMELESS ENCAMPMENT	210.277778
	ILLEGAL PARKING	199.158416
	GRAFFITI	196.333333
	ANIMAL ABUSE	195.488889
	VENDING	164.400000
	NOISE - VEHICLE	142.719298
	NOISE - STREET/SIDEWALK	138.936842
	TRAFFIC	120.116667
	NOISE - COMMERCIAL	112.531915
	NOISE - PARK	92.050000
	URINATING IN PUBLIC	82.000000
	ILLEGAL FIREWORKS	40.000000
FLUSHING	POSTING ADVERTISEMENT	368.000000
I FOSITING	I ODITING WOMEN LIDELIENI	200.000

	BIKE/ROLLER/SKATE CHRONIC	343.666667
	DERELICT VEHICLE	221.268182
	NOISE - HOUSE OF WORSHIP	218.000000
	ANIMAL ABUSE	213.711268
	NOISE - VEHICLE	201.310078
	DRINKING	182.775000
	HOMELESS ENCAMPMENT	180.000000
	ILLEGAL PARKING	178.616184
	BLOCKED DRIVEWAY	176.793770
	NOISE - PARK	173.120690
	NOISE - COMMERCIAL	170.417143
	NOISE - STREET/SIDEWALK	169.520000
	VENDING	146.666667
	DISORDERLY YOUTH	128.000000
	TRAFFIC	123.382979
	GRAFFITI	103.250000
	URINATING IN PUBLIC	73.166667
	PANHANDLING	68.500000
	ILLEGAL FIREWORKS	41.500000
FLORAL PARK	ANIMAL ABUSE	1594.000000
	DERELICT VEHICLE	999.017857
	ILLEGAL PARKING	554.984375
	DRINKING	494.000000
	BLOCKED DRIVEWAY	461.750000
	NOISE - STREET/SIDEWALK	413.666667
	NOISE - COMMERCIAL	261.666667
	DISORDERLY YOUTH	195.000000
	NOISE - VEHICLE	117.000000
FAR ROCKAWAY	DERELICT VEHICLE	219.614973
	DISORDERLY YOUTH	201.000000
	NOISE - STREET/SIDEWALK	181.992647
	HOMELESS ENCAMPMENT	180.642857
	VENDING	168.000000
	ANIMAL ABUSE	162.741573
	ILLEGAL PARKING	162.040678
	BLOCKED DRIVEWAY	157.795775
	DRINKING	144.750000
	TRAFFIC	142.900000
	NOISE - VEHICLE	130.506494
	NOISE - COMMERCIAL	115.270833
	NOISE - PARK	90.086957
	URINATING IN PUBLIC	90.000000
	NOISE - HOUSE OF WORSHIP	68.000000
ELMHURST	DERELICT VEHICLE	288.833333
	BIKE/ROLLER/SKATE CHRONIC	277.500000
	VENDING	239.285714
	ANIMAL ABUSE	231.921053
	HOMELESS ENCAMPMENT	220.218750
	BLOCKED DRIVEWAY	205.694329
	PANHANDLING	197.666667
	ILLEGAL PARKING	196.440323
	DRINKING	183.307692
	NOISE - COMMERCIAL	167.086420
	TRAFFIC	157.428571
	NOISE - VEHICLE	157.297872
	URINATING IN PUBLIC	152.000000
	NOISE - STREET/SIDEWALK	150.263393
	NOISE - PARK	138.617647
	NOISE - HOUSE OF WORSHIP	112.800000
	ILLEGAL FIREWORKS	59.000000
	DISORDERLY YOUTH	51.000000
	POSTING ADVERTISEMENT	44.000000
EAST ELMHURST	GRAFFITI	458.333333
	DISORDERLY YOUTH	414.000000
	HOMELESS ENCAMPMENT	378.000000

	DERELICT VEHICLE	342.921053
	ANIMAL ABUSE	243.016949
	VENDING	241.000000
	BLOCKED DRIVEWAY	223.690341
	ILLEGAL PARKING	204.479190
	TRAFFIC	161.050000
	NOISE - PARK	159.000000
	NOISE - STREET/SIDEWALK	148.158879
	NOISE - COMMERCIAL	140.857143
	NOISE - VEHICLE	129.786885
	URINATING IN PUBLIC	125.200000
	DRINKING	111.111111
	POSTING ADVERTISEMENT	111.000000
	NOISE - HOUSE OF WORSHIP	
	BIKE/ROLLER/SKATE CHRONIC	15.000000
CORONA	DERELICT VEHICLE	274.842105
CORONA	NOISE - HOUSE OF WORSHIP	225.000000
	HOMELESS ENCAMPMENT	219.894737
	ANIMAL ABUSE	219.868852
	DRINKING	218.545455
	ILLEGAL PARKING	201.371212
	BLOCKED DRIVEWAY	198.939515
		185.774194
	VENDING	
	DISORDERLY YOUTH	172.333333
	NOISE - VEHICLE	159.350000
	NOISE - COMMERCIAL	157.004032
	NOISE - STREET/SIDEWALK	142.126582
	NOISE - PARK	139.000000
	TRAFFIC	136.916667
	URINATING IN PUBLIC	117.000000
	POSTING ADVERTISEMENT	92.000000
	PANHANDLING	70.000000
	GRAFFITI	44.000000
COLLEGE POINT	GRAFFITI	711.000000
	VENDING	279.000000
	ANIMAL ABUSE	278.428571
	NOISE - COMMERCIAL	238.114286
	DERELICT VEHICLE	210.809783
	BLOCKED DRIVEWAY	202.501149
	NOISE - PARK	190.500000
	ILLEGAL PARKING	187.198864
	NOISE - STREET/SIDEWALK	180.666667
	NOISE - VEHICLE	158.732824
	TRAFFIC	137.357143
	HOMELESS ENCAMPMENT	86.333333
	DISORDERLY YOUTH	36.000000
CENTRAL PARK	ILLEGAL PARKING	275.000000
	NOISE - STREET/SIDEWALK	195.747368
CAMBRIA HEIGHTS	HOMELESS ENCAMPMENT	1367.000000
	DERELICT VEHICLE	966.347826
	ANIMAL ABUSE	680.909091
	ILLEGAL PARKING	674.328947
	TRAFFIC	512.333333
	BLOCKED DRIVEWAY	461.517007
	NOISE - VEHICLE	414.792208
	NOISE - STREET/SIDEWALK	276.160000
	NOISE - COMMERCIAL	228.250000
	NOISE - HOUSE OF WORSHIP	158.000000
	ILLEGAL FIREWORKS	91.000000
BROOKLYN	GRAFFITI	494.232558
	DERELICT VEHICLE	360.920191
	AGENCY ISSUES	315.333333
	BIKE/ROLLER/SKATE CHRONIC	300.884956
	ANIMAL ABUSE	289.598162
	HOMELESS ENCAMPMENT	281.410047

	VENDING	271.295146
	BLOCKED DRIVEWAY	264.795297
	PANHANDLING	258.673469
	ILLEGAL PARKING	258.533008
	DISORDERLY YOUTH	248.680556
	URINATING IN PUBLIC	233.698529
	DRINKING	214.246154
	POSTING ADVERTISEMENT	201.422222
	NOISE - STREET/SIDEWALK	198.943916
	NOISE - VEHICLE	196.810691
	NOISE - PARK	189.916720
	TRAFFIC	187.090156
	NOISE - HOUSE OF WORSHIP	183.873529
	NOISE - COMMERCIAL ILLEGAL FIREWORKS	180.791753
DDONV		140.180328
BRONX	PANHANDLING DERELICT VEHICLE	852.684211 553.342725
	GRAFFITI	533.666667
	HOMELESS ENCAMPMENT	446.222672
	ANIMAL ABUSE	439.826855
	VENDING	409.245383
	ILLEGAL PARKING	394.493445
	BLOCKED DRIVEWAY	375.418948
	DRINKING	347.297872
	ILLEGAL FIREWORKS	336.333333
	NOISE - VEHICLE	333.379676
	URINATING IN PUBLIC	323.196078
	NOISE - STREET/SIDEWALK	313.278965
	TRAFFIC	295.121127
	NOISE - PARK	281.606947
	NOISE - COMMERCIAL	281.542951
	NOISE - HOUSE OF WORSHIP	273.265823
	DISORDERLY YOUTH	254.015873
	POSTING ADVERTISEMENT	213.250000
	BIKE/ROLLER/SKATE CHRONIC	207.300000
BREEZY POINT	DERELICT VEHICLE	428.333333
	ILLEGAL PARKING	236.266667
	DRINKING	158.000000
	ANIMAL ABUSE	156.500000
	NOISE - COMMERCIAL	152.000000
	BLOCKED DRIVEWAY	79.333333
	NOISE - VEHICLE	79.000000
	NOISE - STREET/SIDEWALK	60.000000
BELLEROSE	HOMELESS ENCAMPMENT	2348.000000
	DERELICT VEHICLE	1029.764045
	ANIMAL ABUSE	763.142857
	BLOCKED DRIVEWAY	605.726316
	NOISE - STREET/SIDEWALK	543.846154
	ILLEGAL PARKING	491.849057
	URINATING IN PUBLIC	452.000000
	PANHANDLING	449.000000
	NOISE - COMMERCIAL	404.243243
	ILLEGAL FIREWORKS	400.000000
	TRAFFIC	345.142857
	BIKE/ROLLER/SKATE CHRONIC DRINKING	294.000000 235.000000
	NOISE - VEHICLE	154.700000
	POSTING ADVERTISEMENT	135.000000
	NOISE - HOUSE OF WORSHIP	131.000000
	DISORDERLY YOUTH	111.000000
	NOISE - PARK	84.000000
BAYSIDE	GRAFFITI	272.666667
DAIJIDE	NOISE - HOUSE OF WORSHIP	212.000007
	DERELICT VEHICLE	201.242424
	NOISE - PARK	196.250000

	ANIMAL ABUSE	196.135135
	DISORDERLY YOUTH	178.000000
	HOMELESS ENCAMPMENT	172.500000
	BLOCKED DRIVEWAY	153.718085
	ILLEGAL PARKING	153.606238
	NOISE - COMMERCIAL	133.725000
	DRINKING	114.000000
	VENDING	112.500000
	NOISE - VEHICLE	102.312500
	NOISE - STREET/SIDEWALK	91.666667
	TRAFFIC	91.222222
ASTORIA	GRAFFITI	845.000000
	DERELICT VEHICLE	574.206612
	POSTING ADVERTISEMENT	352.000000
	TRAFFIC	324.297872
	ANIMAL ABUSE	299.664000
	VENDING	295.851852
	HOMELESS ENCAMPMENT	294.781250
	BLOCKED DRIVEWAY	288.893197
	ILLEGAL PARKING	288.491803
	DRINKING	283.000000
	URINATING IN PUBLIC	277.111111
	NOISE - STREET/SIDEWALK	210.402000
	NOISE - VEHICLE	210.220588
	NOISE - COMMERCIAL	191.805663
	NOISE - PARK	179.409836
	DISORDERLY YOUTH	173.666667
	ILLEGAL FIREWORKS	166.250000
	NOISE - HOUSE OF WORSHIP	120.947368
	BIKE/ROLLER/SKATE CHRONIC	104.133333
	PANHANDLING	69.000000
ARVERNE	DISORDERLY YOUTH	215.000000
AITVEITIVE	DERELICT VEHICLE	177.740741
	BLOCKED DRIVEWAY	151.200000
	ILLEGAL PARKING	138.724138
	NOISE - COMMERCIAL	136.500000
	ANIMAL ABUSE	128.894737
	NOISE - STREET/SIDEWALK	119.275862
	NOISE - VEHICLE	111.142857
	HOMELESS ENCAMPMENT	108.250000
	NOISE - HOUSE OF WORSHIP	93.272727
	GRAFFITI	92.000000
	NOISE - PARK	77.000000
	PANHANDLING	62.000000
	URINATING IN PUBLIC	41.000000
	VENDING	29.000000
	7 E110 E110	

Task 5 - Perform a statistical test for the following:

DRINKING

In [34]:

#Please note: For the below statements you need to state the Null and Alternate and the #a statistical test to accept or reject the Null Hypothesis along with the correspondin # a. Whether the average response time across complaint types is similar or # b. Are the type of complaint or service requested and location related?

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In [35]:

#a. Whether the average response time across complaint types is similar or not (overall #Here complaint type is categorical data and response time is numerical data. Hence, nu closed_status_requests.Request_Closing_Time.groupby(closed_status_requests['Complaint T #Mean of response time based on Complaint type

```
ANIMAL ABUSE
                                        312.512490
         ANIMAL IN A PARK
                                      20210.000000
         BIKE/ROLLER/SKATE CHRONIC
                                    225.693396
         BLOCKED DRIVEWAY
                                        284.142643
                                        441.499119
         DERELICT VEHICLE
         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
         SOUEEGEE
                                        242.500000
         TRAFFIC
                                        206.695526
         URINATING IN PUBLIC
                                        217.302365
         VENDING
                                        240.542578
         Name: Request_Closing_Time, dtype: float64
In [36]:
          # import anova
          from scipy.stats import f oneway
In [37]:
          #Programatically adding compaint types Request closing time params for f oneway functio
          result = closed status requests.groupby('Complaint Type')['Request Closing Time'].apply
          f_oneway(*result)
Out[37]: F_onewayResult(statistic=513.9606866025152, pvalue=0.0)
In [38]:
          #Response Summary: ● Whether the average response time across complaint types is simila
          # f oneway(for each complaint type request closing time avg)
          # Response F_onewayResult(statistic=513.9606866025152, pvalue=0.0)
          #Since pvalue = 0.0 is less than 0.05, Reject Null
          # Null - No Significant difference in average request closing time by complaint type
          # Al - Significant difference in average request closing time by complaint type
          #Finally:Reject null There is relationship between avg response time for each complaint
In [39]:
          #5.b. Are the type of complaint or service requested and location related?
          #Both complaint type and locations are categorical data. Hence p score should be calucl
          #Import chi2 contigency module
          from scipy.stats import chi2 contingency
In [40]:
          chi2_contingency(pd.crosstab(closed_status_requests['Complaint Type'], closed_status_re
Out[40]: (119069.84639856996,
          0.0,
          1034,
          array([[4.42254021e-03, 1.41641902e-01, 2.45048933e-02, ...,
                  2.20724962e-02, 4.94922455e-02, 7.36352946e-02],
                 [5.72424122e+00, 1.83331835e+02, 3.17175002e+01, ...,
                  2.85691675e+01, 6.40594631e+01, 9.53086162e+01],
```

315.333333

AGENCY ISSUES

```
[7.37090036e-04, 2.36069836e-02, 4.08414888e-03, ...,
                  3.67874936e-03, 8.24870758e-03, 1.22725491e-02],
                 [3.31174553e+00, 1.06066177e+02, 1.83500809e+01, ...,
                  1.65286209e+01, 3.70614432e+01, 5.51405631e+01],
                 [4.36357301e-01, 1.39753343e+01, 2.41781614e+00, ...,
                  2.17781962e+00, 4.88323489e+00, 7.26534906e+00],
                 [2.79578250e+00, 8.95412888e+01, 1.54911767e+01, ...,
                  1.39534963e+01, 3.12873478e+01, 4.65497787e+01]]))
In [41]:
          #Response summary: Are the type of complaint or service requested and location related?
          #Response Summary: chi2 contigency(cross tabe between complaint type and city)
          #Response: (119069.84639856996, (ch2 stat),
          # 0.0 (pvalue),
          # 1034,
          #Since pvalue = 0.0 is less than 0.05, Reject Null
          # Null - No Significant difference in city and complaint type
          # Al - Significant difference in city and complaint type
          #Finally: Reject null, means There is relationship between complaint type location of i
```

Thank You