

Project 2 - Customer Service Requests Analysis

December 19, 2022

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

1 1. Import a 311 NYC service request.

```
[10]: df = pd.read_csv('NYC311_Service_Requests_from_2010_to_Present.csv')
```

C:\Users\Vinosh\AppData\Local\Temp\ipykernel_9808\2749672019.py:1: DtypeWarning: Columns (48,49) have mixed types. Specify dtype option on import or set low_memory=False.

```
df = pd.read_csv('NYC311_Service_Requests_from_2010_to_Present.csv')
```

```
[11]: df
```

```
[11]:
```

	Unique Key	Created Date	Closed Date	Agency \
0	32310363	12/31/2015 11:59:45 PM	01-01-16 0:55	NYPD
1	32309934	12/31/2015 11:59:44 PM	01-01-16 1:26	NYPD
2	32309159	12/31/2015 11:59:29 PM	01-01-16 4:51	NYPD
3	32305098	12/31/2015 11:57:46 PM	01-01-16 7:43	NYPD
4	32306529	12/31/2015 11:56:58 PM	01-01-16 3:24	NYPD
...
300693	30281872	03/29/2015 12:33:41 AM	NaN	NYPD
300694	30281230	03/29/2015 12:33:28 AM	03/29/2015 02:33:59 AM	NYPD
300695	30283424	03/29/2015 12:33:03 AM	03/29/2015 03:40:20 AM	NYPD
300696	30280004	03/29/2015 12:33:02 AM	03/29/2015 04:38:35 AM	NYPD
300697	30281825	03/29/2015 12:33:01 AM	03/29/2015 04:41:50 AM	NYPD

	Agency Name	Complaint Type \
0	New York City Police Department	Noise - Street/Sidewalk
1	New York City Police Department	Blocked Driveway
2	New York City Police Department	Blocked Driveway
3	New York City Police Department	Illegal Parking
4	New York City Police Department	Illegal Parking
...

300693	New York City Police Department	Noise - Commercial
300694	New York City Police Department	Blocked Driveway
300695	New York City Police Department	Noise - Commercial
300696	New York City Police Department	Noise - Commercial
300697	New York City Police Department	Noise - Commercial

	Descriptor	Location Type	Incident Zip \
0	Loud Music/Party	Street/Sidewalk	10034.0
1	No Access	Street/Sidewalk	11105.0
2	No Access	Street/Sidewalk	10458.0
3	Commercial Overnight Parking	Street/Sidewalk	10461.0
4	Blocked Sidewalk	Street/Sidewalk	11373.0
...
300693	Loud Music/Party	Club/Bar/Restaurant	NaN
300694	Partial Access	Street/Sidewalk	11418.0
300695	Loud Music/Party	Club/Bar/Restaurant	11206.0
300696	Loud Music/Party	Club/Bar/Restaurant	10461.0
300697	Loud Music/Party	Store/Commercial	10036.0

	Incident Address ...	Bridge Highway Name \
0	71 VERMILYEA AVENUE ...	NaN
1	27-07 23 AVENUE ...	NaN
2	2897 VALENTINE AVENUE ...	NaN
3	2940 BAISLEY AVENUE ...	NaN
4	87-14 57 ROAD ...	NaN
...
300693	CRESCENT AVENUE ...	NaN
300694	100-17 87 AVENUE ...	NaN
300695	162 THROOP AVENUE ...	NaN
300696	3151 EAST TREMONT AVENUE ...	NaN
300697	251 WEST 48 STREET ...	NaN

	Bridge Highway Direction	Road Ramp	Bridge Highway Segment \
0	NaN	NaN	NaN
1	NaN	NaN	NaN
2	NaN	NaN	NaN
3	NaN	NaN	NaN
4	NaN	NaN	NaN
...
300693	NaN	NaN	NaN
300694	NaN	NaN	NaN
300695	NaN	NaN	NaN
300696	NaN	NaN	NaN
300697	NaN	NaN	NaN

	Garage Lot Name	Ferry Direction	Ferry Terminal Name	Latitude \
0	NaN	NaN	NaN	40.865682

1	NaN	NaN	NaN	40.775945
2	NaN	NaN	NaN	40.870325
3	NaN	NaN	NaN	40.835994
4	NaN	NaN	NaN	40.733060
...
300693	NaN	NaN	NaN	NaN
300694	NaN	NaN	NaN	40.694077
300695	NaN	NaN	NaN	40.699590
300696	NaN	NaN	NaN	40.837708
300697	NaN	NaN	NaN	40.760583

	Longitude	Location
0	-73.923501	(40.86568153633767, -73.92350095571744)
1	-73.915094	(40.775945312321085, -73.91509393898605)
2	-73.888525	(40.870324522111424, -73.88852464418646)
3	-73.828379	(40.83599404683083, -73.82837939584206)
4	-73.874170	(40.733059618956815, -73.87416975810375)
...
300693	NaN	NaN
300694	-73.846087	(40.69407728322387, -73.8460866160573)
300695	-73.944234	(40.69959035300927, -73.94423377144169)
300696	-73.834587	(40.8377075854206, -73.83458731019586)
300697	-73.985922	(40.76058322950115, -73.98592204392392)

[300698 rows x 53 columns]

```
[12]: df.describe()
```

```
[12]:
```

	Unique Key	Incident Zip	X Coordinate (State Plane) \
count	3.006980e+05	298083.000000	2.971580e+05
mean	3.130054e+07	10848.888645	1.004854e+06
std	5.738547e+05	583.182081	2.175338e+04
min	3.027948e+07	83.000000	9.133570e+05
25%	3.080118e+07	10310.000000	9.919752e+05
50%	3.130436e+07	11208.000000	1.003158e+06
75%	3.178446e+07	11238.000000	1.018372e+06
max	3.231065e+07	11697.000000	1.067173e+06

	Y Coordinate (State Plane)	School or Citywide Complaint	Vehicle Type \
count	297158.000000	0.0	0.0
mean	203754.534416	NaN	NaN
std	29880.183529	NaN	NaN
min	121219.000000	NaN	NaN
25%	183343.000000	NaN	NaN
50%	201110.500000	NaN	NaN
75%	224125.250000	NaN	NaN
max	271876.000000	NaN	NaN

	Taxi Company Borough	Taxi Pick Up Location	Garage Lot Name \
count	0.0	0.0	0.0
mean	NaN	NaN	NaN
std	NaN	NaN	NaN
min	NaN	NaN	NaN
25%	NaN	NaN	NaN
50%	NaN	NaN	NaN
75%	NaN	NaN	NaN
max	NaN	NaN	NaN

	Latitude	Longitude
count	297158.000000	297158.000000
mean	40.725885	-73.925630
std	0.082012	0.078454
min	40.499135	-74.254937
25%	40.669796	-73.972142
50%	40.718661	-73.931781
75%	40.781840	-73.876805
max	40.912869	-73.700760

```
[13]: df.shape
```

```
[13]: (300698, 53)
```

2 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.

```
[41]: df['Created Date']=pd.to_datetime(df['Created Date'])
df['Closed Date']=pd.to_datetime(df['Closed Date'])
```

```
[39]: df['Request_Closing_Time']=(df['Closed Date']-df['Created Date'])

Request_Closing_Time=[]
for x in (df['Closed Date']-df['Created Date']):
    close=x.total_seconds()/60
    Request_Closing_Time.append(close)

df['Request_Closing_Time']=Request_Closing_Time
```

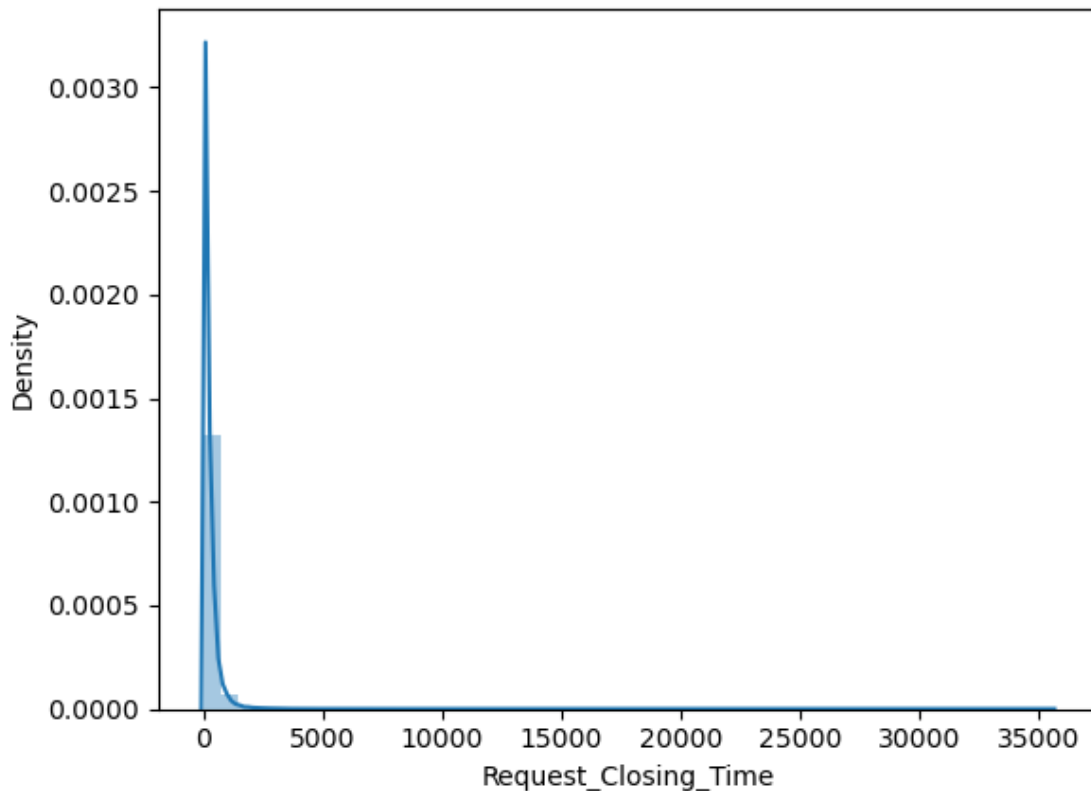
```
[40]: df['Request_Closing_Time'].head()
```

```
[40]: 0      55.250000
      1      86.266667
      2     291.516667
      3     465.233333
      4     207.033333
      Name: Request_Closing_Time, dtype: float64
```

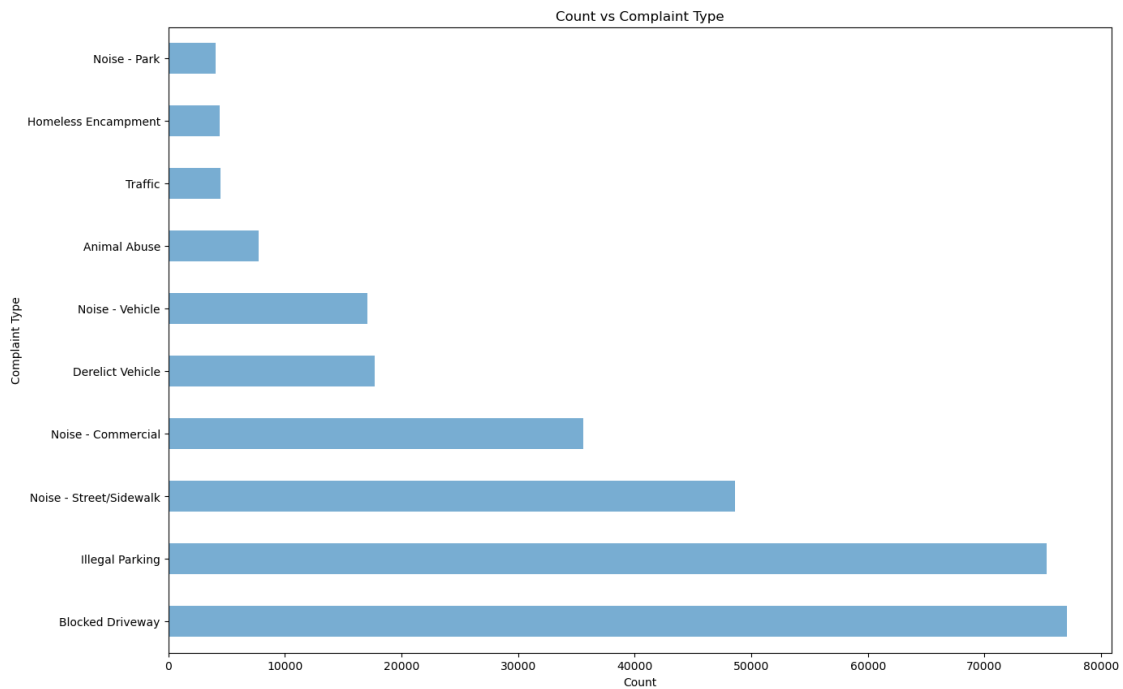
3. Provide major insights/patterns that you can offer in a visual format (graphs or tables); at least 4 major conclusions that you can come up with after generic data mining

```
[23]: sns.distplot(df['Request_Closing_Time'])
      plt.show()
```

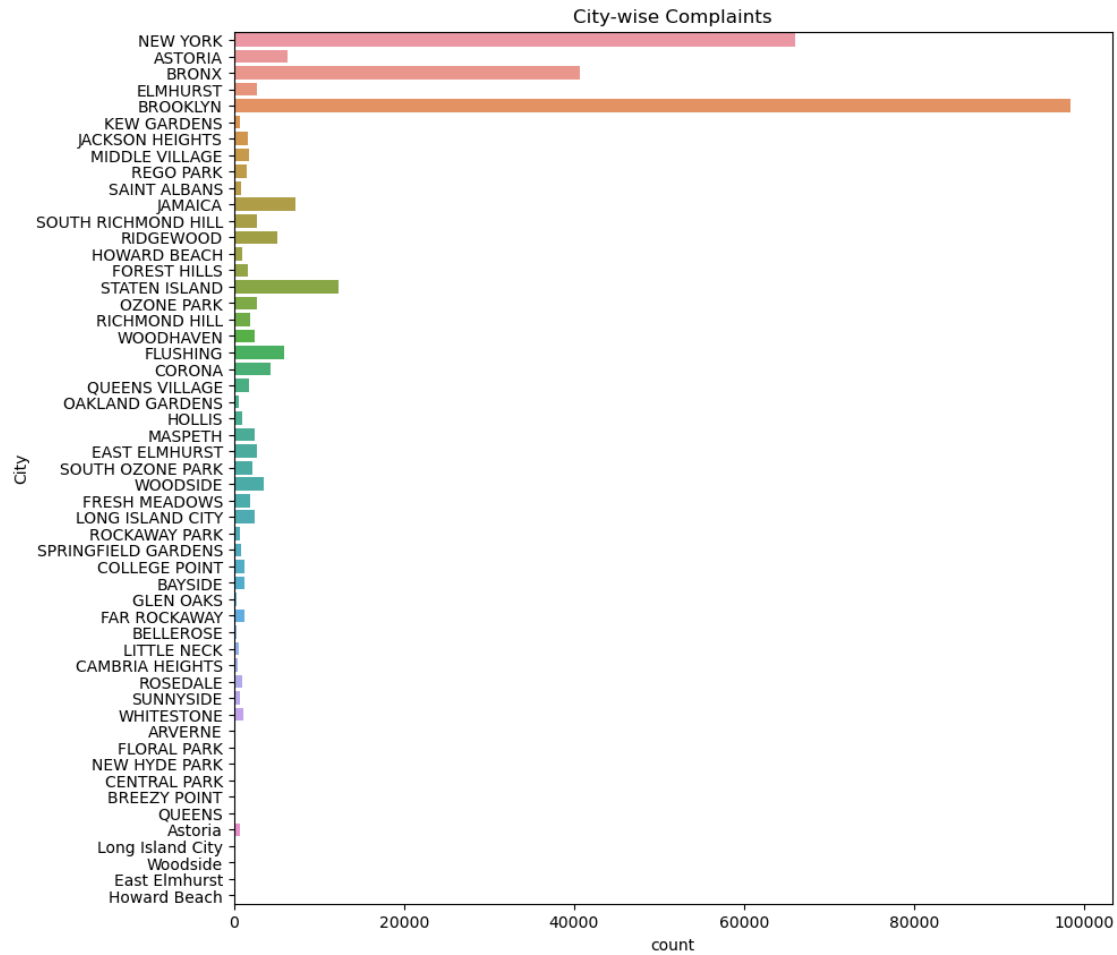
```
C:\Users\Vinosh\anaconda3\lib\site-packages\seaborn\distributions.py:2619:
FutureWarning: `distplot` is a deprecated function and will be removed in a
future version. Please adapt your code to use either `displot` (a figure-level
function with similar flexibility) or `histplot` (an axes-level function for
histograms).
  warnings.warn(msg, FutureWarning)
```



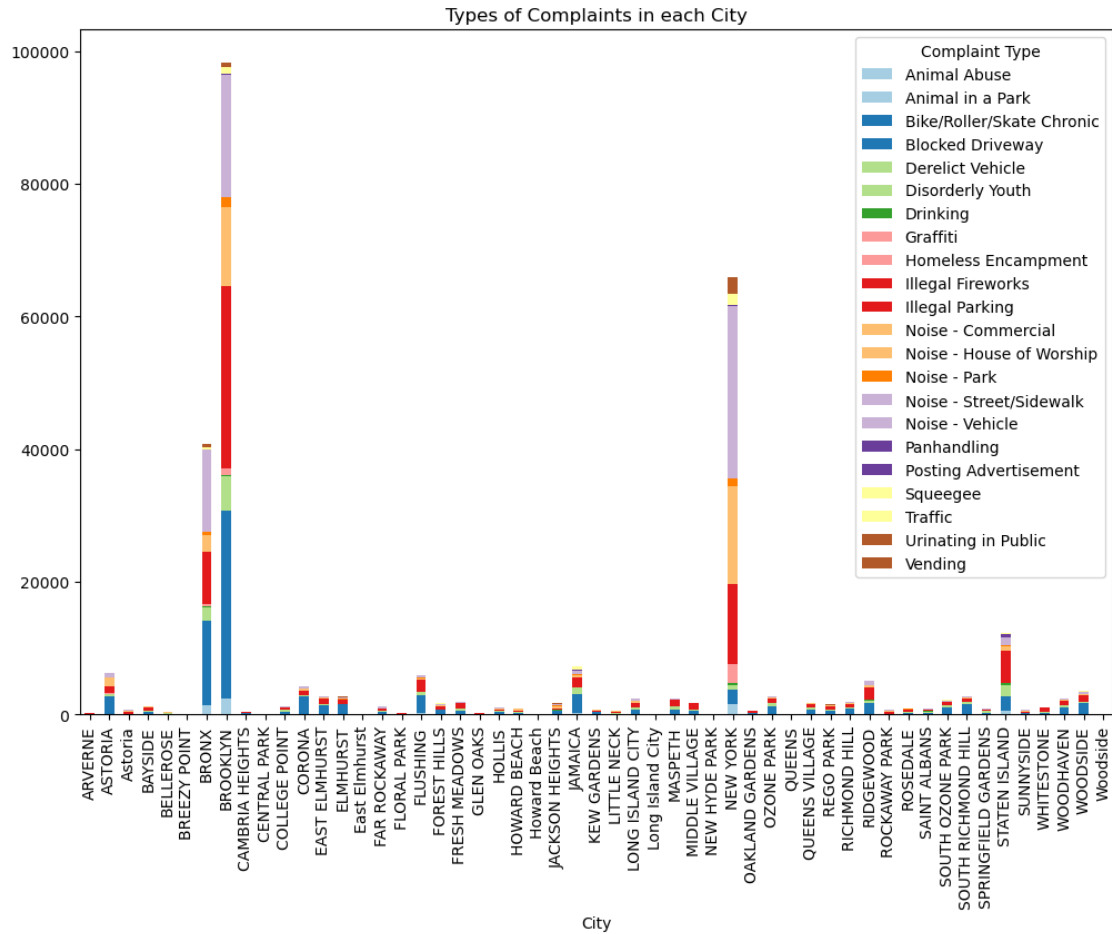
```
[28]: df['Complaint Type'].value_counts()[:10].plot(kind='barh',alpha=0.6,figsize=(15,10))
plt.xlabel('Count')
plt.ylabel('Complaint Type')
plt.title('Count vs Complaint Type')
plt.show()
```



```
[27]: plt.figure(figsize=(10,10))
sns.countplot(y=df['City'])
plt.title('City-wise Complaints')
plt.show()
```

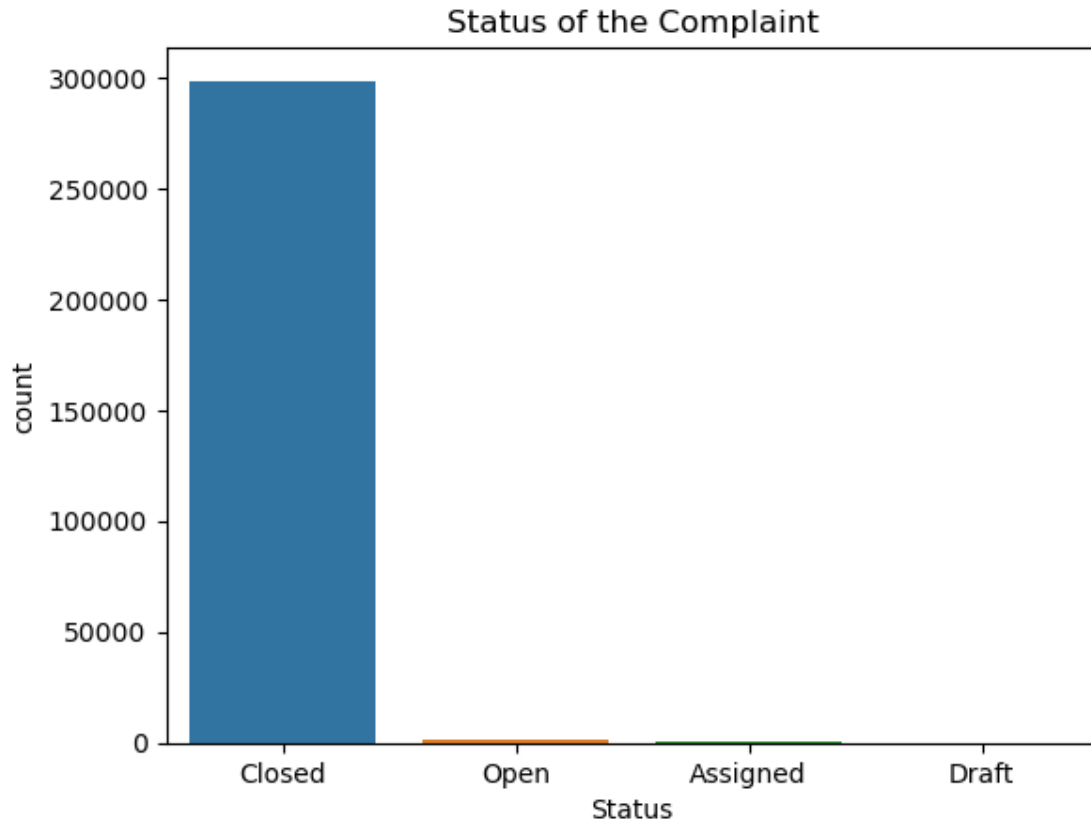


```
[31]: crosstab=pd.crosstab(index=df['City'],columns=df['Complaint Type'])
crosstab.plot(kind='bar',figsize=(12,8),stacked=True,colormap='Paired')
plt.title('Types of Complaints in each City')
plt.show()
```



```
[34]: sns.countplot(df['Status'])
plt.title('Status of the Complaint')
plt.show()
```

C:\Users\Vinosh\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning: Pass the following variable as a keyword arg: x. From version
0.12, the only valid positional argument will be `data`, and passing other
arguments without an explicit keyword will result in an error or
misinterpretation.
warnings.warn(



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

```
[36]: df['Location Type'].unique()
```

```
[36]: array(['Street/Sidewalk', 'Club/Bar/Restaurant', 'Store/Commercial',
        'House of Worship', 'Residential Building/House',
        'Residential Building', 'Park/Playground', 'Vacant Lot',
        'House and Store', 'Highway', 'Commercial', 'Roadway Tunnel',
        'Subway Station', 'Parking Lot', 'Bridge', 'Terminal', nan,
        'Ferry', 'Park'], dtype=object)
```

```
[37]: pd.DataFrame(df.groupby('Location Type')['Request_Closing_Time'].mean()).
        ↪sort_values('Request_Closing_Time')
```

```
[37]:
```

Location Type	Request_Closing_Time
Subway Station	142.250980
Club/Bar/Restaurant	186.074330

House of Worship	191.833279
Store/Commercial	198.089073
Park/Playground	207.137129
Highway	223.424221
Bridge	229.158333
Roadway Tunnel	266.525714
Street/Sidewalk	268.515306
Residential Building	289.089941
House and Store	300.795699
Residential Building/House	309.505679
Parking Lot	320.130342
Commercial	320.566129
Vacant Lot	448.435498
Park	20210.083333
Ferry	NaN
Terminal	NaN

5 5. Perform a statistical test for the following:

Please note: For the below statements you need to state the Null and Alternate and then provide a statistical test to accept or reject the Null Hypothesis along with the corresponding 'p-value'.

Whether the average response time across complaint types is similar or not (overall), Are the type of complaint or service requested and location related?

```
[53]: from scipy import stats
      from scipy.stats import chi2_contingency

      import statsmodels.api as sm
      from statsmodels.formula.api import ols
```

```
[42]: new_df=df.loc[:,(df.isnull().sum()/df.shape[0]*100)<=50]
```

```
[43]: print('Old DataFrame Shape :',df.shape)
      print('New DataFrame Shape : ',new_df.shape)
```

```
Old DataFrame Shape : (300698, 54)
New DataFrame Shape : (300698, 40)
```

```
[47]: rem=[]
      for x in new_df.columns.tolist():
          if new_df[x].nunique()<=3:
              print(x+ ' '*10+' : ',new_df[x].unique())
              rem.append(x)
```

```
Agency          : ['NYPD']
Agency Name      : ['New York City Police Department' 'NYPD' 'Internal
Affairs Bureau']
Facility Type     : ['Precinct' nan]
```

```
Park Facility Name      : ['Unspecified' 'Alley Pond Park - Nature
Center']
School Name            : ['Unspecified' 'Alley Pond Park - Nature Center']
School Number          : ['Unspecified' 'Q001']
School Region          : ['Unspecified' nan]
School Code            : ['Unspecified' nan]
School Phone Number    : ['Unspecified' '7182176034']
School Address         : ['Unspecified' 'Grand Central Parkway, near the
soccer field']
School City            : ['Unspecified' 'QUEENS']
School State           : ['Unspecified' 'NY']
School Zip             : ['Unspecified' nan]
School Not Found       : ['N']
```

```
[48]: new_df.drop(rem,axis=1,inplace=True)
```

C:\Users\Vinosh\AppData\Local\Temp\ipykernel_9808\2260080562.py:1:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
new_df.drop(rem,axis=1,inplace=True)
```

```
[49]: new_df.shape
```

```
[49]: (300698, 26)
```

```
[50]: rem1=["Unique Key","Incident Address","Descriptor","Street Name","Cross Street_
↵1","Cross Street 2","Due Date","Resolution Description","Resolution Action_
↵Updated Date","Community Board","X Coordinate (State Plane)","Y Coordinate_
↵(State Plane)","Park Borough","Latitude","Longitude","Location"]

new_df.drop(rem1,axis=1,inplace=True)
```

C:\Users\Vinosh\AppData\Local\Temp\ipykernel_9808\1160912026.py:3:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
new_df.drop(rem1,axis=1,inplace=True)
```

```
[51]: new_df.head()
```

```
[51]:      Created Date      Closed Date      Complaint Type \
0 2015-12-31 23:59:45 2016-01-01 00:55:00 Noise - Street/Sidewalk
1 2015-12-31 23:59:44 2016-01-01 01:26:00 Blocked Driveway
2 2015-12-31 23:59:29 2016-01-01 04:51:00 Blocked Driveway
```

```

3 2015-12-31 23:57:46 2016-01-01 07:43:00      Illegal Parking
4 2015-12-31 23:56:58 2016-01-01 03:24:00      Illegal Parking

```

	Location Type	Incident Zip	Address Type	City	Status	Borough \
0	Street/Sidewalk	10034.0	ADDRESS	NEW YORK	Closed	MANHATTAN
1	Street/Sidewalk	11105.0	ADDRESS	ASTORIA	Closed	QUEENS
2	Street/Sidewalk	10458.0	ADDRESS	BRONX	Closed	BRONX
3	Street/Sidewalk	10461.0	ADDRESS	BRONX	Closed	BRONX
4	Street/Sidewalk	11373.0	ADDRESS	ELMHURST	Closed	QUEENS

	Request_Closing_Time
0	55.250000
1	86.266667
2	291.516667
3	465.233333
4	207.033333

```

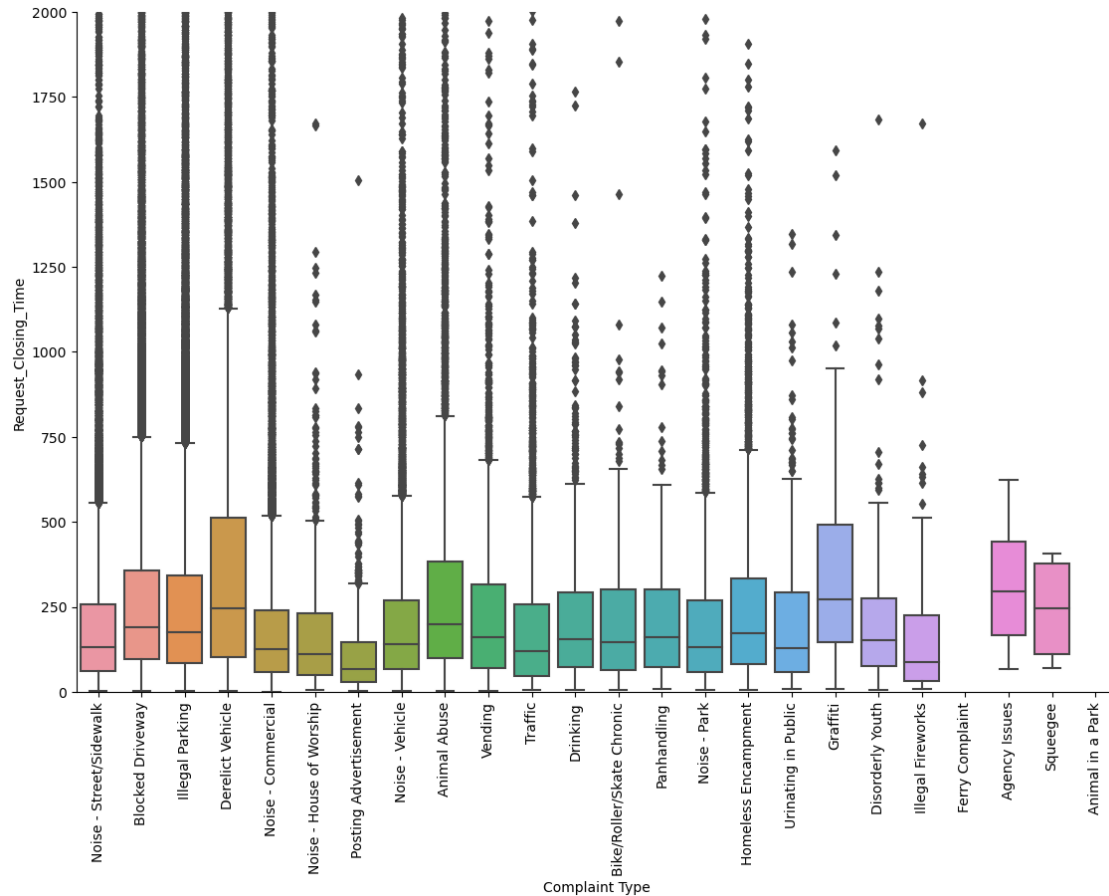
[52]: g=sns.catplot(x='Complaint_
      ↪Type',y='Request_Closing_Time',kind='box',data=new_df)
      g.fig.set_figheight(8)
      g.fig.set_figwidth(15)
      plt.xticks(rotation=90)
      plt.ylim((0,2000))

```

```

[52]: (0.0, 2000.0)

```



H0: there is no significant different in mean of Request_Closing_Time for different Complaint

H1:there is significant different in mean of Request_Closing_Time for different Complaint

```
[55]: anova_df=pd.DataFrame()
anova_df['Request_Closing_Time']=new_df['Request_Closing_Time']
anova_df['Complaint']=new_df['Complaint Type']

anova_df.dropna(inplace=True)
anova_df.head()
```

```
[55]: Request_Closing_Time      Complaint
0          55.250000 Noise - Street/Sidewalk
1          86.266667 Blocked Driveway
2         291.516667 Blocked Driveway
3         465.233333 Illegal Parking
4         207.033333 Illegal Parking
```

```
[57]: lm=ols('Request_Closing_Time~Complaint',data=anova_df).fit()
table=sm.stats.anova_lm(lm)
table
```

```
[57]:
```

	df	sum_sq	mean_sq	F	PR(>F)
Complaint	22.0	1.455049e+09	6.613860e+07	514.177089	0.0
Residual	298511.0	3.839747e+10	1.286300e+05	NaN	NaN

Since p value for the Complaint is less than 0.01 thus we accept alternate hypothesis

H0:Complaint Type and Location Type are independent

H1:Complaint Type and Location Type are related

```
[60]: chi_sq=pd.DataFrame()
chi_sq['Location Type']=new_df['Location Type']
chi_sq['Complaint Type']=new_df['Complaint Type']

chi_sq.dropna(inplace=True)
```

```
[61]: data_crosstab = pd.crosstab( chi_sq['Location Type'],chi_sq['Complaint Type'])
```

```
[62]: stat, p, dof, expected = chi2_contingency(data_crosstab)

alpha = 0.05
if p <= alpha:
    print('Dependent (reject H0)')
else:
    print('Independent (H0 holds true)')
```

Dependent (reject H0)

Since p value for the chi square test is less than 0.05(LOS) we can conclude that Complaint Type is dependent on Location Type

```
[ ]:
```