

CustSerAnalysis-Part3

November 23, 2022

1 3.Find major types of complaints:

```
[1]: import pandas as pd
```

```
[2]: # read newDF.csv created in Customer service part 2 ipnyb file
```

```
[ ]: newDF = pd.read_csv('/home/labsuser/Applied DS/newDF.csv')
```

```
[16]: newDF.shape
```

```
[16]: (364558, 49)
```

```
[17]: newDF.head()
```

```
[17]: Unnamed: 0 Unique Key Created Date Closed Date Agency \
0 0 32310363 2015-12-31 11:59:45 2016-01-01 12:55:15 NYPD
1 1 32309934 2015-12-31 11:59:44 2016-01-01 01:26:57 NYPD
2 2 32309159 2015-12-31 11:59:29 2016-01-01 04:51:03 NYPD
3 3 32305098 2015-12-31 11:57:46 2016-01-01 07:43:13 NYPD
4 4 32306529 2015-12-31 11:56:58 2016-01-01 03:24:42 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
```

```
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 ...
```

```
School Not Found Bridge Highway Name Bridge Highway Direction Road Ramp \
0 N NaN NaN NaN
```

1	N	NaN	NaN	NaN
2	N	NaN	NaN	NaN
3	N	NaN	NaN	NaN
4	N	NaN	NaN	NaN

	Bridge Highway Segment	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

	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)

[5 rows x 49 columns]

```
[3]: # There is Unnamed: 0 columns which can be dropped
newDF.drop(['Unnamed: 0'], axis=1, inplace=True)
```

```
[19]: newDF.head()
```

```
[19]:
```

	Unique Key	Created Date	Closed Date	Agency \
0	32310363	2015-12-31 11:59:45	2016-01-01 12:55:15	NYPD
1	32309934	2015-12-31 11:59:44	2016-01-01 01:26:57	NYPD
2	32309159	2015-12-31 11:59:29	2016-01-01 04:51:03	NYPD
3	32305098	2015-12-31 11:57:46	2016-01-01 07:43:13	NYPD
4	32306529	2015-12-31 11:56:58	2016-01-01 03:24:42	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

	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

	Incident Address	...	School Not Found	Bridge Highway Name	\
0	71 VERMILYEA AVENUE	...	N	NaN	
1	27-07 23 AVENUE	...	N	NaN	
2	2897 VALENTINE AVENUE	...	N	NaN	
3	2940 BAISLEY AVENUE	...	N	NaN	
4	87-14 57 ROAD	...	N	NaN	

	Bridge Highway Direction	Road Ramp	Bridge Highway Segment	Ferry Direction	\
0	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	
4	NaN	NaN	NaN	NaN	

	Ferry Terminal Name	Latitude	Longitude	\
0	NaN	40.865682	-73.923501	
1	NaN	40.775945	-73.915094	
2	NaN	40.870325	-73.888525	
3	NaN	40.835994	-73.828379	
4	NaN	40.733060	-73.874170	

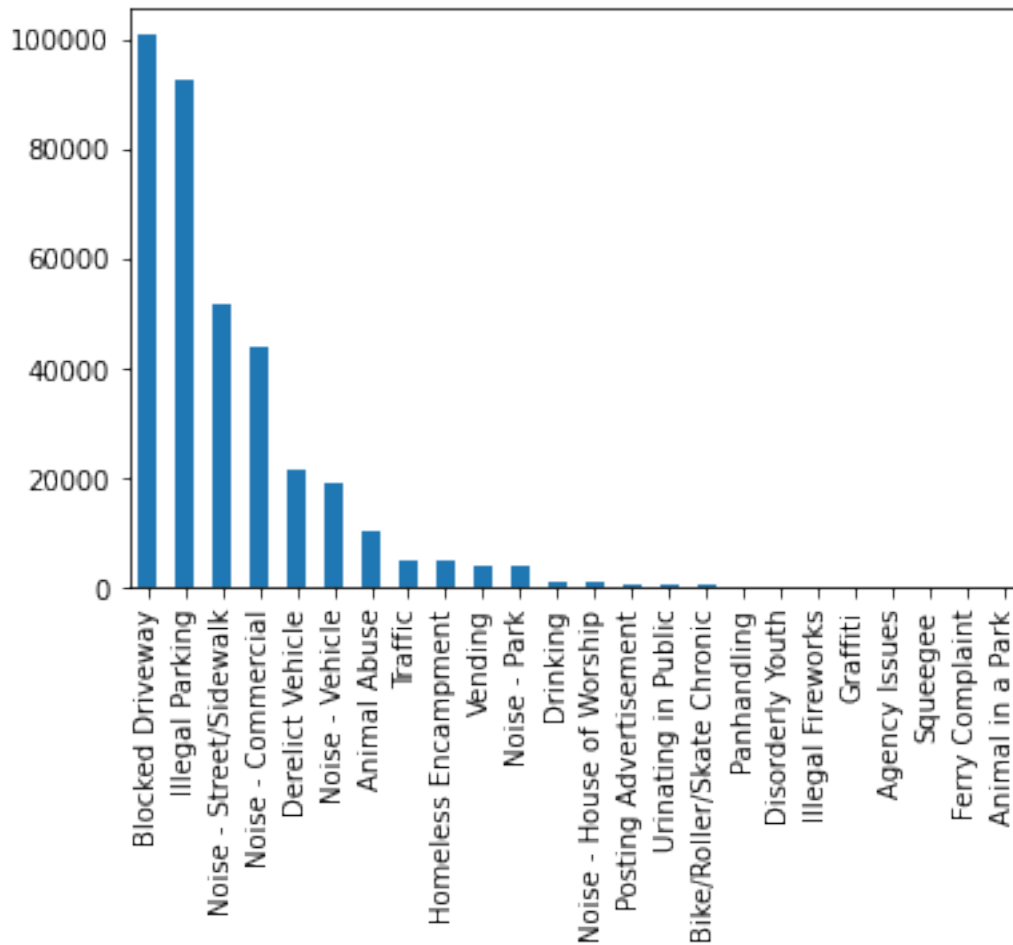
	Location
0	(40.86568153633767, -73.92350095571744)
1	(40.775945312321085, -73.91509393898605)
2	(40.870324522111424, -73.88852464418646)
3	(40.83599404683083, -73.82837939584206)
4	(40.733059618956815, -73.87416975810375)

[5 rows x 48 columns]

```
[27]: # 1. Plot a bar graph of count vs. complaint types
```

```
[4]: newDF['Complaint Type'].value_counts().plot(kind='bar')
```

```
[4]: <AxesSubplot:>
```



```
[ ]: #Find the top 10 types of complaints
```

```
[22]: top15TypeDF=newDF['Complaint Type'].value_counts()[:15]
```

```
[23]: top15TypeDF
```

```
[23]: Blocked Driveway      100881
      Illegal Parking      92679
      Noise - Street/Sidewalk  51692
      Noise - Commercial    44109
      Derelict Vehicle      21661
      Noise - Vehicle       19352
      Animal Abuse         10541
      Traffic              5198
      Homeless Encampment   4879
      Vending              4192
      Noise - Park         4109
```

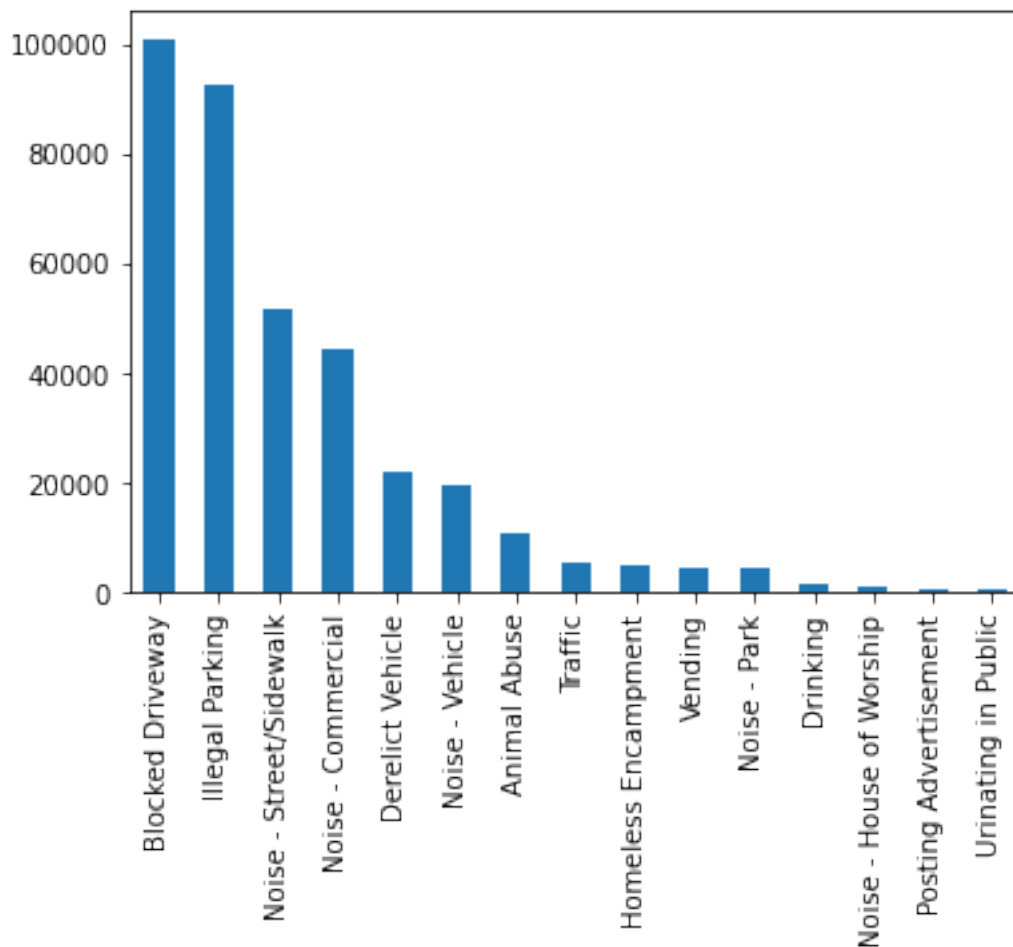
```

Drinking          1409
Noise - House of Worship  1070
Posting Advertisement  681
Urinating in Public  641
Name: Complaint Type, dtype: int64

```

```
[35]: top15TypeDF.plot(kind='bar')
```

```
[35]: <AxesSubplot:>
```



```
[39]: # Display the types of complaints in each city in a separate dataset

df_type_city=newDF.groupby(['City','Complaint Type'])['Unique Key'].count().
↳unstack()
```

```
[40]: df_type_city
```

[40]: Complaint Type	Animal Abuse	Blocked Driveway	Derelict Vehicle	\
City				
ARVERNE	46.0	50.0	32.0	
ASTORIA	170.0	3436.0	426.0	
Astoria	NaN	159.0	14.0	
BAYSIDE	53.0	514.0	231.0	
BELLEROSE	15.0	138.0	120.0	
BREEZY POINT	2.0	3.0	3.0	
BRONX	1971.0	17063.0	2403.0	
BROOKLYN	3191.0	36447.0	6259.0	
CAMBRIA HEIGHTS	15.0	177.0	148.0	
CENTRAL PARK	NaN	NaN	NaN	
COLLEGE POINT	35.0	597.0	223.0	
CORONA	104.0	3597.0	72.0	
EAST ELMHURST	85.0	1925.0	136.0	
ELMHURST	59.0	1992.0	94.0	
East Elmhurst	NaN	NaN	2.0	
FAR ROCKAWAY	111.0	383.0	215.0	
FLORAL PARK	7.0	33.0	74.0	
FLUSHING	191.0	3641.0	532.0	
FOREST HILLS	78.0	873.0	71.0	
FRESH MEADOWS	66.0	682.0	347.0	
GLEN OAKS	5.0	48.0	57.0	
HOLLIS	39.0	442.0	162.0	
HOWARD BEACH	51.0	215.0	172.0	
Howard Beach	NaN	1.0	NaN	
JACKSON HEIGHTS	50.0	703.0	41.0	
JAMAICA	317.0	3621.0	1133.0	
KEW GARDENS	26.0	429.0	16.0	
LITTLE NECK	21.0	174.0	73.0	
LONG ISLAND CITY	40.0	1052.0	220.0	
Long Island City	NaN	55.0	4.0	
MASPETH	56.0	1000.0	510.0	
MIDDLE VILLAGE	36.0	663.0	366.0	
NEW HYDE PARK	1.0	76.0	14.0	
NEW YORK	1941.0	2707.0	695.0	
OAKLAND GARDENS	29.0	177.0	117.0	
OZONE PARK	72.0	1681.0	479.0	
QUEENS	1.0	3.0	2.0	
QUEENS VILLAGE	90.0	772.0	478.0	
REGO PARK	33.0	780.0	94.0	
RICHMOND HILL	55.0	1100.0	201.0	
RIDGEWOOD	154.0	2162.0	507.0	
ROCKAWAY PARK	33.0	80.0	19.0	
ROSEDALE	44.0	270.0	247.0	
SAINT ALBANS	43.0	318.0	248.0	
SOUTH OZONE PARK	74.0	1202.0	425.0	

SOUTH RICHMOND HILL	40.0	1946.0	356.0
SPRINGFIELD GARDENS	42.0	330.0	267.0
STATEN ISLAND	786.0	2845.0	2184.0
SUNNYSIDE	40.0	278.0	17.0
WHITESTONE	43.0	279.0	279.0
WOODHAVEN	57.0	1364.0	369.0
WOODSIDE	111.0	2038.0	298.0
Woodside	NaN	27.0	8.0

Complaint Type	Disorderly Youth	Drinking	Graffiti \
City			
ARVERNE	2.0	1.0	1.0
ASTORIA	5.0	43.0	4.0
Astoria	NaN	NaN	NaN
BAYSIDE	2.0	1.0	3.0
BELLEROSE	2.0	1.0	NaN
BREEZY POINT	NaN	1.0	NaN
BRONX	66.0	206.0	15.0
BROOKLYN	79.0	291.0	60.0
CAMBRIA HEIGHTS	NaN	NaN	NaN
CENTRAL PARK	NaN	NaN	NaN
COLLEGE POINT	1.0	1.0	2.0
CORONA	6.0	34.0	4.0
EAST ELMHURST	1.0	9.0	3.0
ELMHURST	2.0	13.0	1.0
East Elmhurst	NaN	NaN	NaN
FAR ROCKAWAY	1.0	4.0	NaN
FLORAL PARK	1.0	1.0	NaN
FLUSHING	2.0	47.0	6.0
FOREST HILLS	1.0	1.0	3.0
FRESH MEADOWS	NaN	2.0	NaN
GLEN OAKS	NaN	NaN	NaN
HOLLIS	1.0	3.0	NaN
HOWARD BEACH	1.0	4.0	NaN
Howard Beach	NaN	NaN	NaN
JACKSON HEIGHTS	NaN	10.0	1.0
JAMAICA	9.0	40.0	3.0
KEW GARDENS	NaN	1.0	NaN
LITTLE NECK	2.0	1.0	NaN
LONG ISLAND CITY	2.0	8.0	3.0
Long Island City	NaN	NaN	NaN
MASPETH	2.0	9.0	1.0
MIDDLE VILLAGE	NaN	2.0	NaN
NEW HYDE PARK	NaN	NaN	NaN
NEW YORK	81.0	321.0	25.0
OAKLAND GARDENS	1.0	2.0	NaN
OZONE PARK	4.0	20.0	NaN

QUEENS	NaN	NaN	NaN
QUEENS VILLAGE	NaN	5.0	1.0
REGO PARK	NaN	4.0	1.0
RICHMOND HILL	NaN	10.0	1.0
RIDGEWOOD	3.0	10.0	3.0
ROCKAWAY PARK	4.0	23.0	NaN
ROSEDALE	NaN	2.0	2.0
SAINT ALBANS	1.0	3.0	NaN
SOUTH OZONE PARK	2.0	14.0	2.0
SOUTH RICHMOND HILL	2.0	25.0	NaN
SPRINGFIELD GARDENS	NaN	6.0	NaN
STATEN ISLAND	25.0	188.0	6.0
SUNNYSIDE	2.0	12.0	1.0
WHITESTONE	1.0	3.0	1.0
WOODHAVEN	NaN	4.0	NaN
WOODSIDE	1.0	15.0	4.0
Woodside	NaN	NaN	NaN

Complaint Type	Homeless Encampment	Illegal Parking	Noise - Commercial \
City			
ARVERNE	4.0	62.0	2.0
ASTORIA	32.0	1340.0	1653.0
Astoria	NaN	277.0	311.0
BAYSIDE	2.0	638.0	47.0
BELLEROSE	1.0	132.0	38.0
BREEZY POINT	NaN	16.0	4.0
BRONX	275.0	9889.0	2945.0
BROOKLYN	948.0	33533.0	13860.0
CAMBRIA HEIGHTS	6.0	113.0	19.0
CENTRAL PARK	NaN	5.0	NaN
COLLEGE POINT	3.0	449.0	38.0
CORONA	26.0	791.0	281.0
EAST ELMHURST	2.0	1093.0	41.0
ELMHURST	34.0	760.0	85.0
East Elmhurst	NaN	28.0	NaN
FAR ROCKAWAY	16.0	339.0	59.0
FLORAL PARK	NaN	72.0	3.0
FLUSHING	26.0	2250.0	222.0
FOREST HILLS	18.0	627.0	163.0
FRESH MEADOWS	6.0	1158.0	21.0
GLEN OAKS	NaN	95.0	84.0
HOLLIS	9.0	181.0	54.0
HOWARD BEACH	3.0	384.0	258.0
Howard Beach	NaN	NaN	NaN
JACKSON HEIGHTS	11.0	241.0	619.0
JAMAICA	93.0	1698.0	552.0
KEW GARDENS	5.0	276.0	203.0

LITTLE NECK	NaN	322.0	77.0
LONG ISLAND CITY	10.0	988.0	269.0
Long Island City	NaN	64.0	19.0
MASPETH	11.0	1234.0	58.0
MIDDLE VILLAGE	5.0	1104.0	13.0
NEW HYDE PARK	NaN	32.0	4.0
NEW YORK	3060.0	14553.0	18692.0
OAKLAND GARDENS	1.0	337.0	2.0
OZONE PARK	8.0	774.0	125.0
QUEENS	2.0	10.0	6.0
QUEENS VILLAGE	19.0	669.0	49.0
REGO PARK	6.0	640.0	82.0
RICHMOND HILL	30.0	489.0	249.0
RIDGEWOOD	26.0	2235.0	491.0
ROCKAWAY PARK	4.0	337.0	72.0
ROSEDALE	4.0	326.0	28.0
SAINT ALBANS	11.0	237.0	36.0
SOUTH OZONE PARK	5.0	602.0	82.0
SOUTH RICHMOND HILL	12.0	596.0	223.0
SPRINGFIELD GARDENS	7.0	291.0	38.0
STATEN ISLAND	77.0	6224.0	784.0
SUNNYSIDE	12.0	167.0	238.0
WHITESTONE	NaN	631.0	21.0
WOODHAVEN	10.0	896.0	209.0
WOODSIDE	38.0	1083.0	256.0
Woodside	NaN	124.0	2.0

Complaint Type	Noise - House of Worship	...	Noise - Vehicle	\
City		...		
ARVERNE	14.0	...	10.0	
ASTORIA	21.0	...	236.0	
Astoria	NaN	...	NaN	
BAYSIDE	3.0	...	24.0	
BELLEROSE	1.0	...	11.0	
BREEZY POINT	NaN	...	1.0	
BRONX	90.0	...	3556.0	
BROOKLYN	389.0	...	5966.0	
CAMBRIA HEIGHTS	2.0	...	100.0	
CENTRAL PARK	NaN	...	NaN	
COLLEGE POINT	2.0	...	140.0	
CORONA	3.0	...	110.0	
EAST ELMHURST	25.0	...	82.0	
ELMHURST	6.0	...	69.0	
East Elmhurst	NaN	...	NaN	
FAR ROCKAWAY	1.0	...	83.0	
FLORAL PARK	NaN	...	2.0	
FLUSHING	5.0	...	147.0	

FOREST HILLS	1.0	...	70.0
FRESH MEADOWS	NaN	...	97.0
GLEN OAKS	NaN	...	4.0
HOLLIS	215.0	...	52.0
HOWARD BEACH	1.0	...	10.0
Howard Beach	NaN	...	NaN
JACKSON HEIGHTS	2.0	...	75.0
JAMAICA	15.0	...	337.0
KEW GARDENS	1.0	...	23.0
LITTLE NECK	NaN	...	8.0
LONG ISLAND CITY	NaN	...	124.0
Long Island City	NaN	...	NaN
MASPETH	2.0	...	26.0
MIDDLE VILLAGE	NaN	...	45.0
NEW HYDE PARK	NaN	...	2.0
NEW YORK	222.0	...	6295.0
OAKLAND GARDENS	NaN	...	7.0
OZONE PARK	4.0	...	81.0
QUEENS	1.0	...	2.0
QUEENS VILLAGE	2.0	...	54.0
REGO PARK	1.0	...	60.0
RICHMOND HILL	NaN	...	69.0
RIDGEWOOD	2.0	...	249.0
ROCKAWAY PARK	NaN	...	29.0
ROSEDALE	2.0	...	25.0
SAINT ALBANS	1.0	...	50.0
SOUTH OZONE PARK	5.0	...	97.0
SOUTH RICHMOND HILL	3.0	...	93.0
SPRINGFIELD GARDENS	1.0	...	48.0
STATEN ISLAND	18.0	...	424.0
SUNNYSIDE	NaN	...	53.0
WHITESTONE	NaN	...	31.0
WOODHAVEN	3.0	...	81.0
WOODSIDE	4.0	...	136.0
Woodside	NaN	...	NaN

Complaint Type City	Panhandling	Traffic	Urinating in Public	Vending \
ARVERNE	1.0	1.0	1.0	1.0
ASTORIA	2.0	60.0	10.0	57.0
Astoria	NaN	NaN	NaN	NaN
BAYSIDE	NaN	9.0	NaN	2.0
BELLEROSE	1.0	9.0	1.0	NaN
BREEZY POINT	NaN	NaN	NaN	NaN
BRONX	20.0	427.0	54.0	433.0
BROOKLYN	49.0	1258.0	155.0	575.0
CAMBRIA HEIGHTS	NaN	7.0	NaN	NaN

CENTRAL PARK	NaN	NaN	NaN	NaN
COLLEGE POINT	NaN	16.0	NaN	1.0
CORONA	1.0	14.0	7.0	65.0
EAST ELMHURST	NaN	24.0	6.0	9.0
ELMHURST	3.0	18.0	10.0	25.0
East Elmhurst	NaN	NaN	NaN	NaN
FAR ROCKAWAY	NaN	11.0	1.0	10.0
FLORAL PARK	NaN	NaN	NaN	NaN
FLUSHING	2.0	59.0	12.0	37.0
FOREST HILLS	6.0	65.0	2.0	10.0
FRESH MEADOWS	1.0	15.0	1.0	1.0
GLEN OAKS	NaN	3.0	2.0	19.0
HOLLIS	NaN	11.0	2.0	NaN
HOWARD BEACH	2.0	9.0	NaN	5.0
Howard Beach	NaN	NaN	NaN	NaN
JACKSON HEIGHTS	1.0	13.0	3.0	86.0
JAMAICA	3.0	632.0	37.0	24.0
KEW GARDENS	NaN	10.0	3.0	1.0
LITTLE NECK	NaN	20.0	1.0	NaN
LONG ISLAND CITY	2.0	83.0	3.0	31.0
Long Island City	NaN	NaN	NaN	NaN
MASPETH	NaN	71.0	2.0	7.0
MIDDLE VILLAGE	NaN	14.0	NaN	NaN
NEW HYDE PARK	NaN	NaN	NaN	NaN
NEW YORK	206.0	1770.0	264.0	2639.0
OAKLAND GARDENS	NaN	6.0	NaN	2.0
OZONE PARK	7.0	21.0	4.0	1.0
QUEENS	NaN	2.0	1.0	NaN
QUEENS VILLAGE	1.0	27.0	5.0	2.0
REGO PARK	NaN	16.0	1.0	3.0
RICHMOND HILL	NaN	8.0	5.0	15.0
RIDGEWOOD	NaN	50.0	9.0	9.0
ROCKAWAY PARK	NaN	7.0	1.0	2.0
ROSEDALE	NaN	25.0	NaN	19.0
SAINT ALBANS	NaN	14.0	1.0	2.0
SOUTH OZONE PARK	NaN	36.0	2.0	5.0
SOUTH RICHMOND HILL	NaN	12.0	1.0	24.0
SPRINGFIELD GARDENS	2.0	12.0	3.0	1.0
STATEN ISLAND	13.0	229.0	19.0	25.0
SUNNYSIDE	NaN	17.0	2.0	15.0
WHITESTONE	NaN	32.0	NaN	1.0
WOODHAVEN	1.0	7.0	2.0	6.0
WOODSIDE	NaN	45.0	8.0	15.0
Woodside	NaN	NaN	NaN	NaN

Complaint Type	Bike/Roller/Skate Chronic	Illegal Fireworks	\
City			

ARVERNE	NaN	NaN
ASTORIA	16.0	4.0
Astoria	NaN	NaN
BAYSIDE	NaN	NaN
BELLEROSE	1.0	1.0
BREEZY POINT	NaN	NaN
BRONX	22.0	24.0
BROOKLYN	124.0	61.0
CAMBRIA HEIGHTS	NaN	1.0
CENTRAL PARK	NaN	NaN
COLLEGE POINT	NaN	NaN
CORONA	NaN	NaN
EAST ELMHURST	1.0	NaN
ELMHURST	2.0	1.0
East Elmhurst	NaN	NaN
FAR ROCKAWAY	NaN	NaN
FLORAL PARK	NaN	NaN
FLUSHING	3.0	2.0
FOREST HILLS	6.0	1.0
FRESH MEADOWS	NaN	NaN
GLEN OAKS	NaN	NaN
HOLLIS	NaN	NaN
HOWARD BEACH	1.0	4.0
Howard Beach	NaN	NaN
JACKSON HEIGHTS	2.0	1.0
JAMAICA	3.0	4.0
KEW GARDENS	NaN	NaN
LITTLE NECK	NaN	NaN
LONG ISLAND CITY	3.0	NaN
Long Island City	NaN	NaN
MASPETH	1.0	1.0
MIDDLE VILLAGE	1.0	NaN
NEW HYDE PARK	NaN	NaN
NEW YORK	254.0	38.0
OAKLAND GARDENS	2.0	NaN
OZONE PARK	1.0	1.0
QUEENS	NaN	NaN
QUEENS VILLAGE	NaN	5.0
REGO PARK	NaN	NaN
RICHMOND HILL	NaN	4.0
RIDGEWOOD	3.0	2.0
ROCKAWAY PARK	NaN	NaN
ROSEDALE	2.0	NaN
SAINT ALBANS	NaN	NaN
SOUTH OZONE PARK	1.0	1.0
SOUTH RICHMOND HILL	1.0	2.0
SPRINGFIELD GARDENS	NaN	1.0

STATEN ISLAND	10.0	11.0
SUNNYSIDE	2.0	NaN
WHITESTONE	4.0	1.0
WOODHAVEN	2.0	NaN
WOODSIDE	5.0	1.0
Woodside	NaN	NaN

Complaint Type City	Posting Advertisement	Squeegee	Animal in a Park
ARVERNE	NaN	NaN	NaN
ASTORIA	3.0	NaN	NaN
Astoria	NaN	NaN	NaN
BAYSIDE	NaN	NaN	NaN
BELLEROSE	1.0	NaN	NaN
BREEZY POINT	NaN	NaN	NaN
BRONX	18.0	NaN	NaN
BROOKLYN	58.0	NaN	NaN
CAMBRIA HEIGHTS	NaN	NaN	NaN
CENTRAL PARK	NaN	NaN	NaN
COLLEGE POINT	NaN	NaN	NaN
CORONA	1.0	NaN	NaN
EAST ELMHURST	1.0	NaN	NaN
ELMHURST	1.0	NaN	NaN
East Elmhurst	NaN	NaN	NaN
FAR ROCKAWAY	NaN	NaN	NaN
FLORAL PARK	NaN	NaN	NaN
FLUSHING	1.0	NaN	NaN
FOREST HILLS	4.0	NaN	NaN
FRESH MEADOWS	NaN	NaN	NaN
GLEN OAKS	NaN	NaN	NaN
HOLLIS	NaN	NaN	NaN
HOWARD BEACH	NaN	NaN	NaN
Howard Beach	NaN	NaN	NaN
JACKSON HEIGHTS	1.0	NaN	NaN
JAMAICA	8.0	NaN	NaN
KEW GARDENS	1.0	NaN	NaN
LITTLE NECK	1.0	NaN	NaN
LONG ISLAND CITY	2.0	NaN	NaN
Long Island City	NaN	NaN	NaN
MASPETH	NaN	NaN	NaN
MIDDLE VILLAGE	NaN	NaN	NaN
NEW HYDE PARK	NaN	NaN	NaN
NEW YORK	49.0	4.0	NaN
OAKLAND GARDENS	NaN	NaN	NaN
OZONE PARK	3.0	NaN	NaN
QUEENS	NaN	NaN	1.0
QUEENS VILLAGE	1.0	NaN	NaN

REGO PARK	NaN	NaN	NaN
RICHMOND HILL	2.0	NaN	NaN
RIDGEWOOD	1.0	NaN	NaN
ROCKAWAY PARK	NaN	NaN	NaN
ROSEDALE	NaN	NaN	NaN
SAINT ALBANS	NaN	NaN	NaN
SOUTH OZONE PARK	1.0	NaN	NaN
SOUTH RICHMOND HILL	NaN	NaN	NaN
SPRINGFIELD GARDENS	2.0	NaN	NaN
STATEN ISLAND	517.0	NaN	NaN
SUNNYSIDE	3.0	NaN	NaN
WHITESTONE	NaN	NaN	NaN
WOODHAVEN	NaN	NaN	NaN
WOODSIDE	NaN	NaN	NaN
Woodside	NaN	NaN	NaN

[53 rows x 22 columns]

```
[25]: #Visualize the major types of complaints in each city
#get top 10 complaints in top 10 cities
top10_types = newDF['Complaint Type'].value_counts()[:10]
top10_city = newDF['City'].value_counts()[:10]
```

```
[26]: #data whose complaintType is in the top 10 complaint types
top10_typesDF = newDF[newDF['Complaint Type'].isin(top10_types.index)]
```

```
[27]: #data whose city and Types are in teh top 10 cities with top 10 complaints
top10cityTypeDF = top10_typesDF[top10_typesDF['City'].isin(top10_city.index)]
```

```
[28]: top10cityTypeDF.shape
```

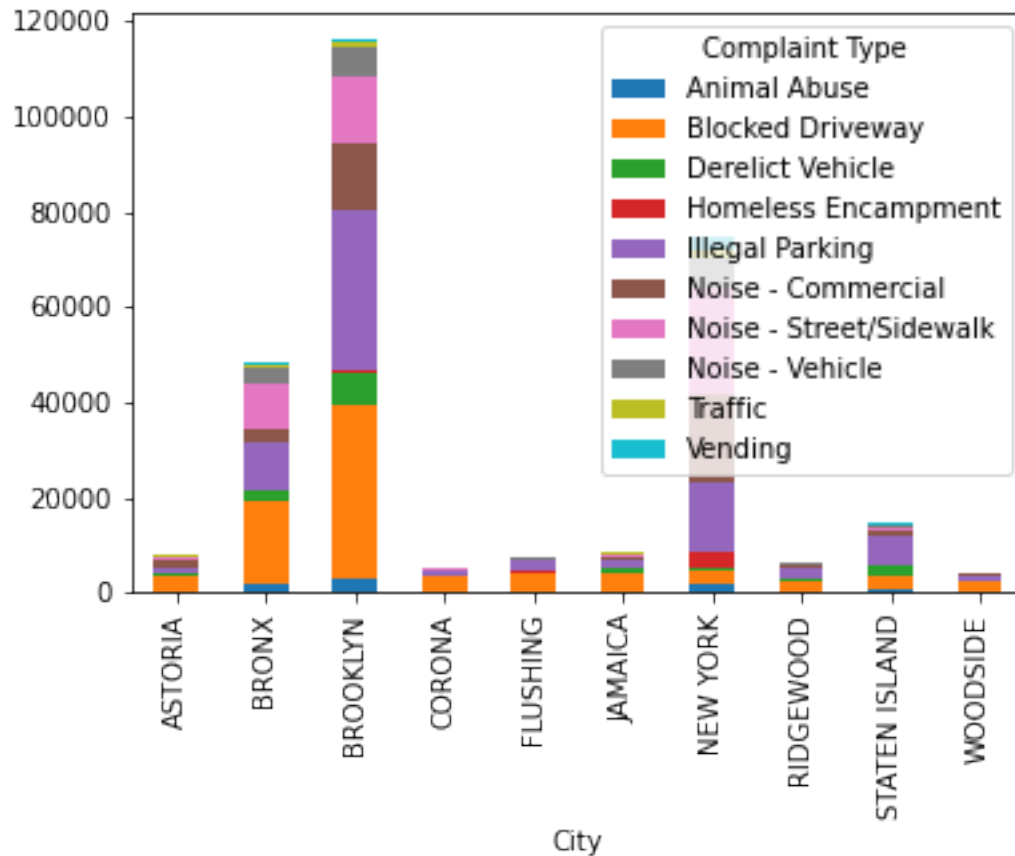
```
[28]: (293051, 50)
```

```
[49]: #for categorical data lets do a cross tab
cityComplaints=pd.
    ↳crosstab(index=top10cityTypeDF['City'],columns=top10cityTypeDF['Complaint_
    ↳Type'])
```

```
[51]: import matplotlib.pyplot as plt
plt.figure(figsize=(12,10))
cityComplaints.plot(kind='bar',stacked=True)
```

```
[51]: <AxesSubplot:xlabel='City'>
```

```
<Figure size 864x720 with 0 Axes>
```



2 Check if the average response time across various types of complaints

```
[8]: #measure response time
#first convert the Date columns to datetime object
# lets convert them to dateTime objects
newDF['Created Date'] = pd.to_datetime(newDF['Created Date']);
newDF['Closed Date'] = pd.to_datetime(newDF['Closed Date']);
newDF['Resolution Action Updated Date'] = pd.to_datetime(newDF['Resolution_
→Action Updated Date']);
newDF['Due Date'] = pd.to_datetime(newDF['Due Date']);
```

```
[ ]: newDF.dtypes
```

```
[9]: import numpy as np
```

```
[10]: newDF['Response Time'] = newDF['Closed Date']-newDF['Created Date']
```

```
[12]: newDF['RespTime_Hrs']=newDF['Response Time']/np.timedelta64(1,'h')
```

```
[ ]: newDF.head()
```

3 Identify significant variables by performing a statistical analysis using p-values and chi-square values (Optional)

```
[13]: newDF['Complaint Type'].value_counts().head(5)
```

```
[13]: Blocked Driveway          100881
      Illegal Parking          92679
      Noise - Street/Sidewalk   51692
      Noise - Commercial       44109
      Derelict Vehicle         21661
      Name: Complaint Type, dtype: int64
```

```
[14]: # Check if the average response time across various types of complaints is same
      ↪ or not
newDF[['Complaint Type','RespTime_Hrs']]
# h0: avg response time for all the complaints in population for type a=
#      avg response time for type b =avg response time for type c=..
# avg response time for type e.

# ha: at least one statement is not true
# In our scenario lets take top 5 complaint Types as samples and analyse
      ↪ average response times
```

```
[14]:
```

	Complaint Type	RespTime_Hrs
0	Noise - Street/Sidewalk	24.925000
1	Blocked Driveway	13.453611
2	Blocked Driveway	16.859444
3	Illegal Parking	19.757500
4	Illegal Parking	15.462222
...
364553	Illegal Parking	-1.703611
364554	Noise - Vehicle	-9.657222
364555	Noise - Street/Sidewalk	0.317500
364556	Blocked Driveway	-9.318611
364557	Blocked Driveway	-9.216667

```
[364558 rows x 2 columns]
```

```
[15]: df1=newDF[newDF['Complaint Type']!='Blocked Driveway']['RespTime_Hrs'].dropna()
      ↪ # lest remove NAN values from samples
```



```
[16]: df2=newDF[newDF['Complaint Type']=='Illegal Parking']['RespTime_Hrs'].dropna()
      ↪# lest remove NAN values from samples

[17]: df3=newDF[newDF['Complaint Type']=='Noise - Street/Sidewalk']['RespTime_Hrs'].
      ↪dropna() # lest remove NAN values from samples

[18]: df4=newDF[newDF['Complaint Type']=='Noise - Commercial']['RespTime_Hrs'].
      ↪dropna() # lest remove NAN values from samples

[19]: df5=newDF[newDF['Complaint Type']=='Derelict Vehicle']['RespTime_Hrs'].dropna()
      ↪# lest remove NAN values from samples

[84]: #Null Hypothesis claim = average response time is same for all the complaints

[20]: from scipy import stats

[21]: # lets do #anova test test to perform null hypothesis
      stats.f_oneway(df1,df2,df3,df4,df5)

[21]: F_onewayResult(statistic=304.95779134697534, pvalue=2.649697471012518e-262)

[83]: #here p value is 2.64*** ~-262 ~ 0
      # alpha(statistics) = 304.9577
      #Since pvalue is less than alpha(risk) factor we will reject(do not accept null
      ↪hypothesis)
```

3.1 chi sq test

```
[ ]: # # any relationship between city and complaint type

[29]: top10cityTypeDF
```

```
[29]:
```

	Unique Key	Created Date	Closed Date	Agency	\
0	32310363	2015-12-31 11:59:45	2016-01-01 12:55:15	NYPD	
1	32309934	2015-12-31 11:59:44	2016-01-01 01:26:57	NYPD	
2	32309159	2015-12-31 11:59:29	2016-01-01 04:51:03	NYPD	
3	32305098	2015-12-31 11:57:46	2016-01-01 07:43:13	NYPD	
5	32306554	2015-12-31 11:56:30	2016-01-01 01:50:11	NYPD	
...	
364551	29607567	2015-01-01 12:06:02	2015-01-01 12:43:41	NYPD	
364552	29610051	2015-01-01 12:05:05	2015-01-01 01:22:10	NYPD	
364554	29608392	2015-01-01 12:04:28	2015-01-01 02:25:02	NYPD	
364555	29607589	2015-01-01 12:01:30	2015-01-01 12:20:33	NYPD	
364556	29610889	2015-01-01 12:01:29	2015-01-01 02:42:22	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
5	New York City Police Department	Illegal Parking
...
364551	New York City Police Department	Noise - Street/Sidewalk
364552	New York City Police Department	Noise - Street/Sidewalk
364554	New York City Police Department	Noise - Vehicle
364555	New York City Police Department	Noise - Street/Sidewalk
364556	New York City Police Department	Blocked Driveway

	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
5	Posted Parking Sign Violation	Street/Sidewalk	11215.0
...
364551	Loud Music/Party	Street/Sidewalk	10453.0
364552	Loud Music/Party	Street/Sidewalk	10002.0
364554	Car/Truck Horn	Street/Sidewalk	10468.0
364555	Loud Music/Party	Street/Sidewalk	10031.0
364556	No Access	Street/Sidewalk	10466.0

	Incident Address	...	Bridge Highway Direction	Road Ramp \
0	71 VERMILYEA AVENUE	...	NaN	NaN
1	27-07 23 AVENUE	...	NaN	NaN
2	2897 VALENTINE AVENUE	...	NaN	NaN
3	2940 BAISLEY AVENUE	...	NaN	NaN
5	260 21 STREET	...	NaN	NaN
...
364551	NaN	...	NaN	NaN
364552	NaN	...	NaN	NaN
364554	2555 SEDGWICK AVENUE	...	NaN	NaN
364555	508 WEST 139 STREET	...	NaN	NaN
364556	931 EAST 226 STREET	...	NaN	NaN

	Bridge Highway Segment	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
5	NaN	NaN	NaN	40.660823
...
364551	NaN	NaN	NaN	40.848639

364552	NaN	NaN	NaN	40.721235
364554	NaN	NaN	NaN	40.867830
364555	NaN	NaN	NaN	40.821647
364556	NaN	NaN	NaN	40.886361

	Longitude	Location	Response Time	\
0	-73.923501	(40.86568153633767, -73.92350095571744)	1 days 00:55:30	
1	-73.915094	(40.775945312321085, -73.91509393898605)	0 days 13:27:13	
2	-73.888525	(40.870324522111424, -73.88852464418646)	0 days 16:51:34	
3	-73.828379	(40.83599404683083, -73.82837939584206)	0 days 19:45:27	
5	-73.992568	(40.66082272389114, -73.99256786342693)	0 days 13:53:41	
...	
364551	-73.911679	(40.84863947227845, -73.91167941229558)	0 days 00:37:39	
364552	-73.987770	(40.72123468734571, -73.98777023226815)	-1 days +13:17:05	
364554	-73.907178	(40.86782963689454, -73.90717786644662)	-1 days +14:20:34	
364555	-73.950873	(40.821646626438095, -73.95087342885292)	0 days 00:19:03	
364556	-73.853290	(40.88636077906953, -73.85329048666742)	-1 days +14:40:53	

	RespTime_Hrs
0	24.925000
1	13.453611
2	16.859444
3	19.757500
5	13.894722
...	...
364551	0.627500
364552	-10.715278
364554	-9.657222
364555	0.317500
364556	-9.318611

[293051 rows x 50 columns]

```
[32]: obj=pd.
      ↪crosstab(index=top10cityTypeDF['City'],columns=top10cityTypeDF['Complaint_
      ↪Type'])
```

```
[33]: obj
```

```
[33]: Complaint Type  Animal Abuse  Blocked Driveway  Derelict Vehicle  \
City
ASTORIA              170              3436              426
BRONX                1971             17063             2403
BROOKLYN            3191             36447             6259
CORONA               104              3597              72
FLUSHING             191              3641             532
JAMAICA              317              3621             1133
```

NEW YORK	1941	2707	695
RIDGEWOOD	154	2162	507
STATEN ISLAND	786	2845	2184
WOODSIDE	111	2038	298

Complaint Type	Homeless Encampment	Illegal Parking	Noise - Commercial	\
City				
ASTORIA	32	1340	1653	
BRONX	275	9889	2945	
BROOKLYN	948	33533	13860	
CORONA	26	791	281	
FLUSHING	26	2250	222	
JAMAICA	93	1698	552	
NEW YORK	3060	14553	18692	
RIDGEWOOD	26	2235	491	
STATEN ISLAND	77	6224	784	
WOODSIDE	38	1083	256	

Complaint Type	Noise - Street/Sidewalk	Noise - Vehicle	Traffic	Vending
City				
ASTORIA	409	236	60	57
BRONX	9146	3556	427	433
BROOKLYN	13984	5966	1258	575
CORONA	243	110	14	65
FLUSHING	241	147	59	37
JAMAICA	365	337	632	24
NEW YORK	22252	6295	1770	2639
RIDGEWOOD	448	249	50	9
STATEN ISLAND	888	424	229	25
WOODSIDE	261	136	45	15

```
[35]: # h0: the city and the complaint types are independent of each other # Null
      ↪Hypothesis
```

```
# ha:the city and the complaint types are not independent of each other #
      ↪alternate Hypothesis
```

```
[36]: chi_sq,pvalue,degrees,exp_value= stats.chi2_contingency(obj)
```

```
[47]: doHypothesisTest(pvalue)
```

Dependent (reject H0)

```
[45]: def doHypothesisTest(pval):
      alpha = 0.05 # 50% risk factor
      if(pval <= alpha):
          print('Dependent (reject H0)')
```

```
else:  
    print('Independent (H0 holds true)')
```

```
[ ]:
```