

Customer Service Requests Analysis Project 1

June 25, 2020

1 Customer Service Request Analysis

```
[1]: #import the required libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
[2]: #import the dataset
customer_dataframe = pd.read_csv("311_Service_Requests_from_2010_to_Present.
→csv")
```

C:\Users\Debidutta\anaconda3\lib\site-packages\IPython\core\interactiveshell.py:3063: DtypeWarning: Columns (48,49) have mixed types.Specify dtype option on import or set low_memory=False.
interactivity=interactivity, compiler=compiler, result=result)

```
[3]: #view the first 5 observations
customer_dataframe.head()
```

```
[3]:   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
```

```
      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	...	Bridge Highway Name	Bridge Highway Direction	\
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	
4	87-14 57 ROAD	...	NaN	NaN	

	Road Ramp	Bridge Highway Segment	Garage Lot Name	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 53 columns]

```
[4]: #explore the dataset
customer_dataframe.describe
```

```
[4]: <bound method NDFrame.describe of          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]>

```
[5]: customer_dataframe.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300698 entries, 0 to 300697
Data columns (total 53 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unique Key                           300698 non-null int64
1   Created Date                          300698 non-null object
```

2	Closed Date	298534 non-null	object
3	Agency	300698 non-null	object
4	Agency Name	300698 non-null	object
5	Complaint Type	300698 non-null	object
6	Descriptor	294784 non-null	object
7	Location Type	300567 non-null	object
8	Incident Zip	298083 non-null	float64
9	Incident Address	256288 non-null	object
10	Street Name	256288 non-null	object
11	Cross Street 1	251419 non-null	object
12	Cross Street 2	250919 non-null	object
13	Intersection Street 1	43858 non-null	object
14	Intersection Street 2	43362 non-null	object
15	Address Type	297883 non-null	object
16	City	298084 non-null	object
17	Landmark	349 non-null	object
18	Facility Type	298527 non-null	object
19	Status	300698 non-null	object
20	Due Date	300695 non-null	object
21	Resolution Description	300698 non-null	object
22	Resolution Action Updated Date	298511 non-null	object
23	Community Board	300698 non-null	object
24	Borough	300698 non-null	object
25	X Coordinate (State Plane)	297158 non-null	float64
26	Y Coordinate (State Plane)	297158 non-null	float64
27	Park Facility Name	300698 non-null	object
28	Park Borough	300698 non-null	object
29	School Name	300698 non-null	object
30	School Number	300698 non-null	object
31	School Region	300697 non-null	object
32	School Code	300697 non-null	object
33	School Phone Number	300698 non-null	object
34	School Address	300698 non-null	object
35	School City	300698 non-null	object
36	School State	300698 non-null	object
37	School Zip	300697 non-null	object
38	School Not Found	300698 non-null	object
39	School or Citywide Complaint	0 non-null	float64
40	Vehicle Type	0 non-null	float64
41	Taxi Company Borough	0 non-null	float64
42	Taxi Pick Up Location	0 non-null	float64
43	Bridge Highway Name	243 non-null	object
44	Bridge Highway Direction	243 non-null	object
45	Road Ramp	213 non-null	object
46	Bridge Highway Segment	213 non-null	object
47	Garage Lot Name	0 non-null	float64
48	Ferry Direction	1 non-null	object
49	Ferry Terminal Name	2 non-null	object

```

50 Latitude                297158 non-null float64
51 Longitude               297158 non-null float64
52 Location                297158 non-null object
dtypes: float64(10), int64(1), object(42)
memory usage: 121.6+ MB

```

```

[6]: #convert the columns created date and closed date to pandas datetime format
customer_dataframe["Created Date"] = pd.to_datetime(customer_dataframe['Created_
→Date'])
customer_dataframe["Closed Date"] = pd.to_datetime(customer_dataframe['Closed_
→Date'])
customer_dataframe.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300698 entries, 0 to 300697
Data columns (total 53 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unique Key                           300698 non-null int64
1   Created Date                         300698 non-null datetime64[ns]
2   Closed Date                         298534 non-null datetime64[ns]
3   Agency                              300698 non-null object
4   Agency Name                         300698 non-null object
5   Complaint Type                      300698 non-null object
6   Descriptor                          294784 non-null object
7   Location Type                      300567 non-null object
8   Incident Zip                       298083 non-null float64
9   Incident Address                   256288 non-null object
10  Street Name                        256288 non-null object
11  Cross Street 1                     251419 non-null object
12  Cross Street 2                     250919 non-null object
13  Intersection Street 1              43858 non-null object
14  Intersection Street 2              43362 non-null object
15  Address Type                      297883 non-null object
16  City                              298084 non-null object
17  Landmark                          349 non-null object
18  Facility Type                     298527 non-null object
19  Status                            300698 non-null object
20  Due Date                          300695 non-null object
21  Resolution Description              300698 non-null object
22  Resolution Action Updated Date     298511 non-null object
23  Community Board                   300698 non-null object
24  Borough                           300698 non-null object
25  X Coordinate (State Plane)         297158 non-null float64
26  Y Coordinate (State Plane)         297158 non-null float64
27  Park Facility Name                 300698 non-null object
28  Park Borough                      300698 non-null object
29  School Name                       300698 non-null object

```

```

30 School Number          300698 non-null object
31 School Region          300697 non-null object
32 School Code            300697 non-null object
33 School Phone Number    300698 non-null object
34 School Address         300698 non-null object
35 School City            300698 non-null object
36 School State           300698 non-null object
37 School Zip             300697 non-null object
38 School Not Found       300698 non-null object
39 School or Citywide Complaint 0 non-null float64
40 Vehicle Type           0 non-null float64
41 Taxi Company Borough   0 non-null float64
42 Taxi Pick Up Location  0 non-null float64
43 Bridge Highway Name     243 non-null object
44 Bridge Highway Direction 243 non-null object
45 Road Ramp              213 non-null object
46 Bridge Highway Segment 213 non-null object
47 Garage Lot Name        0 non-null float64
48 Ferry Direction        1 non-null object
49 Ferry Terminal Name     2 non-null object
50 Latitude               297158 non-null float64
51 Longitude              297158 non-null float64
52 Location               297158 non-null object
dtypes: datetime64[ns](2), float64(10), int64(1), object(40)
memory usage: 121.6+ MB

```

```

[7]: #create a new column 'Request_Closing_Time' as the time elapsed between request
      ↪ creation and request closing
customer_dataframe['Request_Closing_Time'] = customer_dataframe['Closed Date'] -
      ↪ customer_dataframe['Created Date']
customer_dataframe.head()

```

```

[7]: Unique Key      Created Date      Closed Date Agency \
0      32310363 2015-12-31 23:59:45 2016-01-01 00:55:00 NYPD
1      32309934 2015-12-31 23:59:44 2016-01-01 01:26:00 NYPD
2      32309159 2015-12-31 23:59:29 2016-01-01 04:51:00 NYPD
3      32305098 2015-12-31 23:57:46 2016-01-01 07:43:00 NYPD
4      32306529 2015-12-31 23:56:58 2016-01-01 03:24:00 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	...	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	
4	87-14 57 ROAD	...	NaN	NaN	

	Bridge Highway Segment	Garage Lot Name	Ferry Direction	Ferry Terminal Name	\
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	

	Latitude	Longitude	Location	\
0	40.865682	-73.923501	(40.86568153633767, -73.92350095571744)	
1	40.775945	-73.915094	(40.775945312321085, -73.91509393898605)	
2	40.870325	-73.888525	(40.870324522111424, -73.88852464418646)	
3	40.835994	-73.828379	(40.83599404683083, -73.82837939584206)	
4	40.733060	-73.874170	(40.733059618956815, -73.87416975810375)	

	Request_Closing_Time
0	00:55:15
1	01:26:16
2	04:51:31
3	07:45:14
4	03:27:02

[5 rows x 54 columns]

```
[8]: #move the column Request_Closing_Time to the 3rd index
customer_dataframe.insert(3, 'Request_Closing_Time', customer_dataframe.
    ↳pop('Request_Closing_Time'))
customer_dataframe.head(10)
```

	Unique Key	Created Date	Closed Date	Request_Closing_Time	\
0	32310363	2015-12-31 23:59:45	2016-01-01 00:55:00	00:55:15	
1	32309934	2015-12-31 23:59:44	2016-01-01 01:26:00	01:26:16	
2	32309159	2015-12-31 23:59:29	2016-01-01 04:51:00	04:51:31	
3	32305098	2015-12-31 23:57:46	2016-01-01 07:43:00	07:45:14	
4	32306529	2015-12-31 23:56:58	2016-01-01 03:24:00	03:27:02	

5	32306554	2015-12-31	23:56:30	2016-01-01	01:50:00	01:53:30
6	32306559	2015-12-31	23:55:32	2016-01-01	01:53:00	01:57:28
7	32307009	2015-12-31	23:54:05	2016-01-01	01:42:00	01:47:55
8	32308581	2015-12-31	23:53:58	2016-01-01	08:27:00	08:33:02
9	32308391	2015-12-31	23:53:58	2016-01-01	01:17:00	01:23:02

	Agency	Agency Name	Complaint Type	\
0	NYPD	New York City Police Department	Noise - Street/Sidewalk	
1	NYPD	New York City Police Department	Blocked Driveway	
2	NYPD	New York City Police Department	Blocked Driveway	
3	NYPD	New York City Police Department	Illegal Parking	
4	NYPD	New York City Police Department	Illegal Parking	
5	NYPD	New York City Police Department	Illegal Parking	
6	NYPD	New York City Police Department	Illegal Parking	
7	NYPD	New York City Police Department	Blocked Driveway	
8	NYPD	New York City Police Department	Illegal Parking	
9	NYPD	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	...	
4	Blocked Sidewalk	Street/Sidewalk	11373.0	...	
5	Posted Parking Sign Violation	Street/Sidewalk	11215.0	...	
6	Blocked Hydrant	Street/Sidewalk	10032.0	...	
7	No Access	Street/Sidewalk	10457.0	...	
8	Posted Parking Sign Violation	Street/Sidewalk	11415.0	...	
9	No Access	Street/Sidewalk	11219.0	...	

	Bridge Highway Name	Bridge Highway Direction	Road Ramp	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	
5	NaN	NaN	NaN	
6	NaN	NaN	NaN	
7	NaN	NaN	NaN	
8	NaN	NaN	NaN	
9	NaN	NaN	NaN	

	Bridge Highway Segment	Garage Lot Name	Ferry Direction	Ferry Terminal Name	\
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
5	NaN	NaN	NaN	NaN
6	NaN	NaN	NaN	NaN
7	NaN	NaN	NaN	NaN
8	NaN	NaN	NaN	NaN
9	NaN	NaN	NaN	NaN

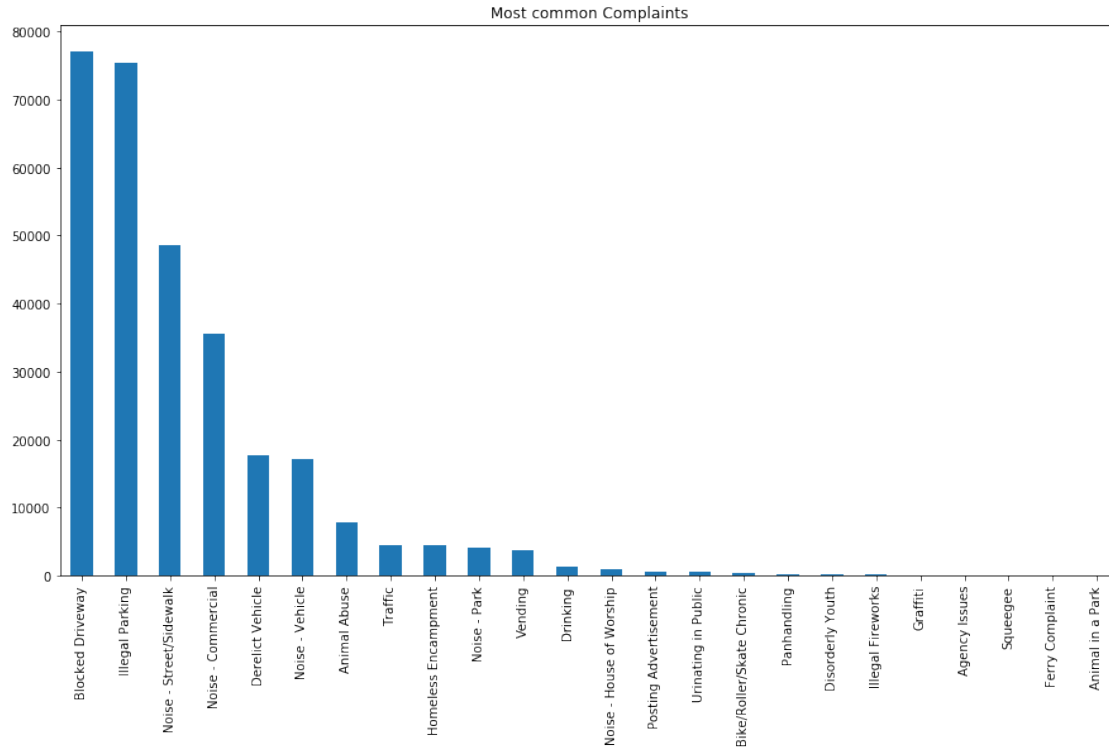
	Latitude	Longitude	Location
0	40.865682	-73.923501	(40.86568153633767, -73.92350095571744)
1	40.775945	-73.915094	(40.775945312321085, -73.91509393898605)
2	40.870325	-73.888525	(40.870324522111424, -73.88852464418646)
3	40.835994	-73.828379	(40.83599404683083, -73.82837939584206)
4	40.733060	-73.874170	(40.733059618956815, -73.87416975810375)
5	40.660823	-73.992568	(40.66082272389114, -73.99256786342693)
6	40.840848	-73.937375	(40.840847591440415, -73.9373750864581)
7	40.837503	-73.902905	(40.83750262540012, -73.90290517326568)
8	40.704977	-73.832605	(40.704977164399935, -73.8326047502584)
9	40.623793	-73.999539	(40.623793065806524, -73.99953890121567)

[10 rows x 54 columns]

```
[9]: #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

      #insight 1
      #The most received complaint type
      (customer_dataframe['Complaint Type'].value_counts()).plot(kind='bar',
                                                                    figsize=(15,8), title = 'Most common Complaints')
```

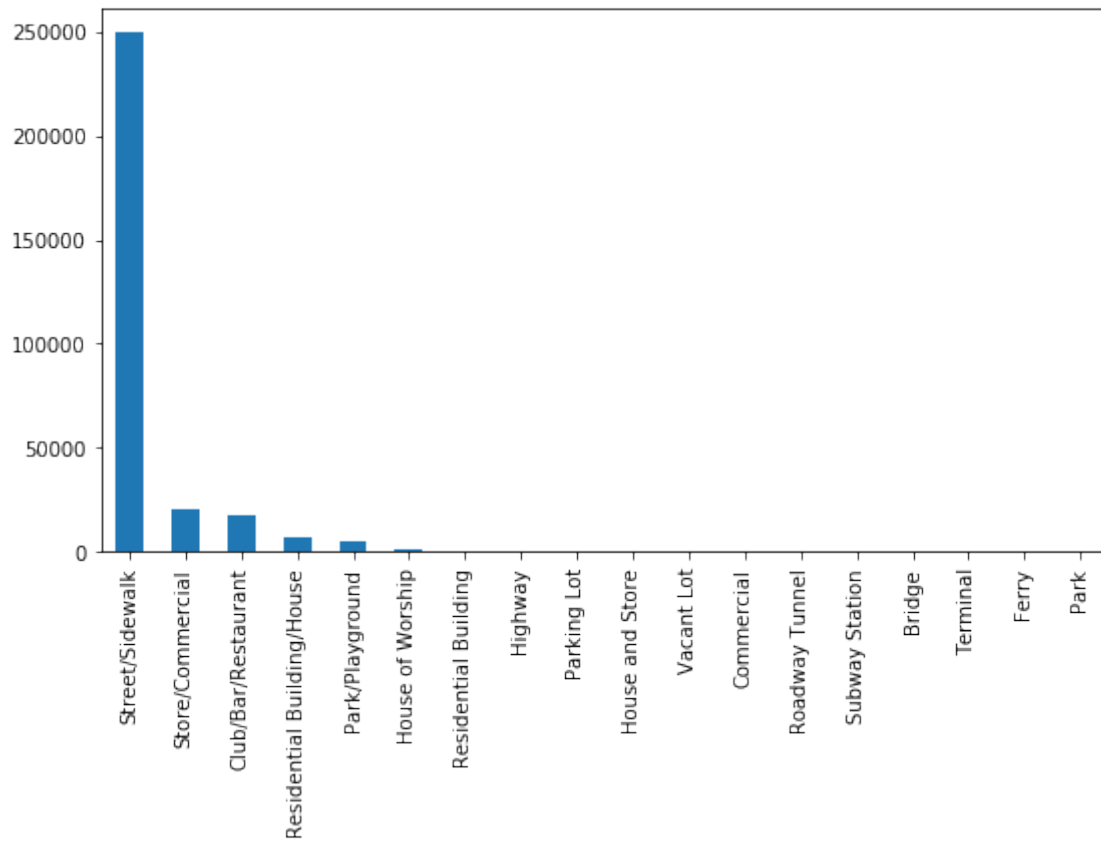
[9]: <matplotlib.axes._subplots.AxesSubplot at 0x1add4df4e48>



Insight 1 So most of the people call 311 to complain about Blocked Driveway, followed by Illegal Parking and so on.

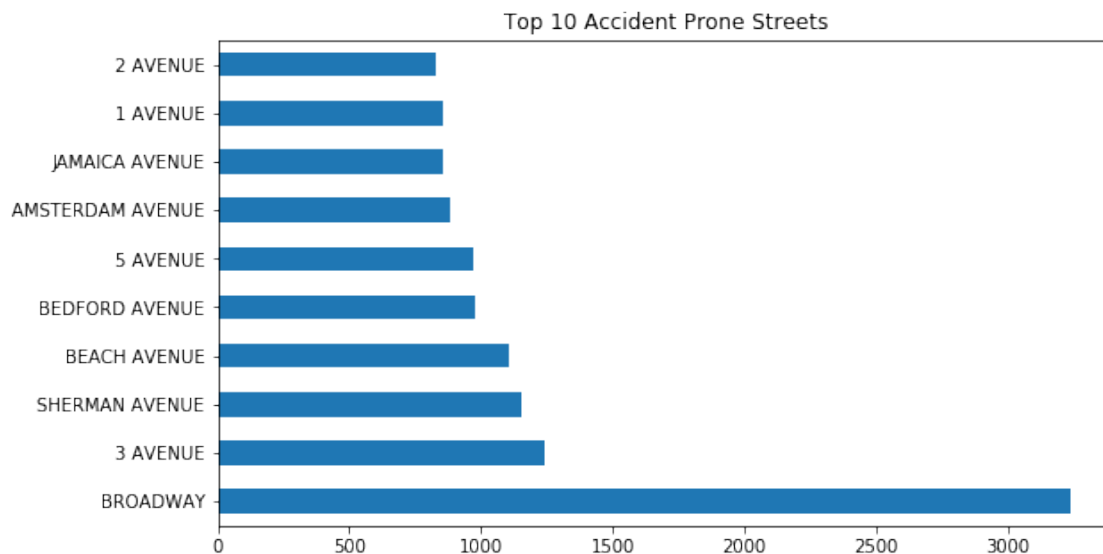
```
[10]: (customer_dataframe['Location Type'].value_counts()).plot(kind='bar',
      figsize=(9,5), title = '')
```

```
[10]: <matplotlib.axes._subplots.AxesSubplot at 0x1add7621c48>
```



```
[11]: (customer_dataframe['Street Name'].value_counts().head(10)).plot(kind='barh',
      figsize=(9,5), title = 'Top 10 Accident Prone Streets')
```

```
[11]: <matplotlib.axes._subplots.AxesSubplot at 0x1add73f82c8>
```



Insight 2 Broadway witnesses the most accidents

```
[12]: #Categorise the complaint service time based on the data of Request_Closing_Time  
#Below 2 hours - Fast,  
#Between 2 to 4 hours - Acceptable,  
#Between 4 to 6 - Slow,  
#More than 6 hours - Very Slow  
  
customer_dataframe['Request_Closing_Time'].dtype
```

```
[12]: dtype('<m8[ns]')
```

```
[13]: #testing the data  
customer_dataframe['Request_Closing_Time'].iloc[247].days
```

```
[13]: 0
```

```
[14]: def toHrs(timeDel):  
        days = timeDel.days  
        hrs = round(timeDel.seconds/3600, 2)  
        res = (days*24)+hrs  
        return res
```

```
[15]: customer_dataframe['Request_Closing_Time'] =  
        ↪customer_dataframe['Request_Closing_Time'].apply(toHrs)
```

```
[16]: customer_dataframe['Request_Closing_Time'].head()
```

```
[16]: 0    0.92  
      1    1.44  
      2    4.86  
      3    7.75  
      4    3.45  
      Name: Request_Closing_Time, dtype: float64
```

```
[17]: #making a function to categorise the Request_Closing_Time  
def Response_Time(time_hrs):  
    if (time_hrs < 2):  
        return 'Fast'  
    elif (time_hrs >2 and time_hrs < 4 ):  
        return 'Average'  
    elif (time_hrs >4 and time_hrs < 6 ):  
        return 'Slow'  
    else:
```

```
return 'Very Slow'
```

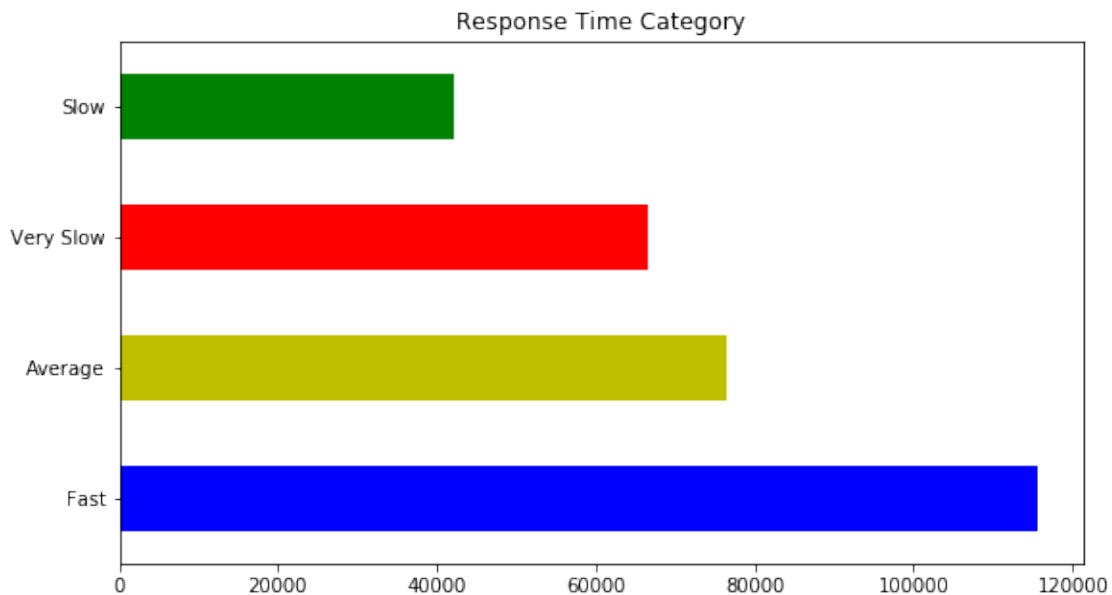
```
[18]: #creating a new column Response_time_category to store the categories of
      ↳Request_Closing_Time
customer_dataframe['Response_time_category'] =
      ↳customer_dataframe['Request_Closing_Time'].apply(Response_Time)
```

```
[19]: customer_dataframe['Response_time_category'].head(10)
```

```
[19]: 0      Fast
      1      Fast
      2      Slow
      3  Very Slow
      4  Average
      5      Fast
      6      Fast
      7      Fast
      8  Very Slow
      9      Fast
      Name: Response_time_category, dtype: object
```

```
[20]: (customer_dataframe['Response_time_category'].value_counts()).
      ↳plot(kind='barh',figsize=(9,5), color=list('byrg'), title = 'Response Time_
      ↳Category')
```

```
[20]: <matplotlib.axes._subplots.AxesSubplot at 0x1add73c0488>
```



```
[21]: customer_dataframe['Response_time_category'].value_counts()
```

```
[21]: Fast          115550
Average          76539
Very Slow       66557
Slow            42052
Name: Response_time_category, dtype: int64
```

Insight 3 So, it is seen that the NYC is able to solve most of the received complaints within 2 hours

```
[22]: #Finding on which months most complaints were received
customer_dataframe['Created Date'].dtype
```

```
[22]: dtype('<M8[ns]')
```

```
[23]: #testing the data
customer_dataframe['Created Date'].iloc[247].month
```

```
[23]: 12
```

```
[24]: #a function to return month name from the month number
monthName = pd.Series({1:'Jan', 2:'Feb', 3:'Mar', 4:'Apr', 5:'May', 6:'Jun',
                        7:'Jul', 8:'Aug', 9:'Sep', 10:'Oct', 11:'Nov', 12:'Dec'})
def getMonth(timeDel):
    month = timeDel.month
    return monthName[month]
```

```
[25]: #testing the function
print(getMonth(customer_dataframe['Created Date'].iloc[247]))
```

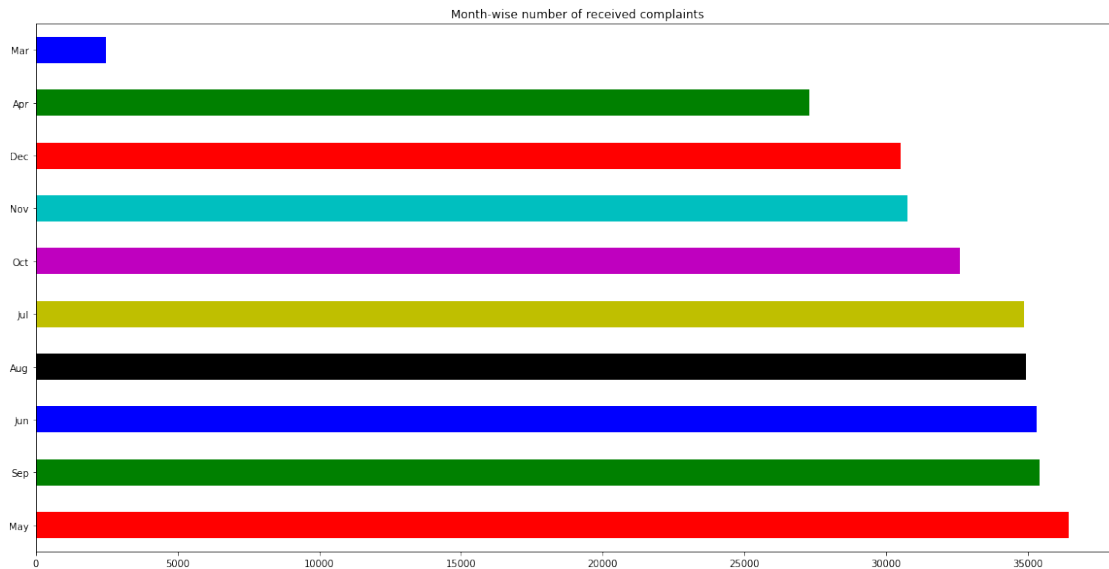
Dec

```
[26]: #applying the function to the entire column and storing the values in a new
      ↪column Created_Month
customer_dataframe['Created_Month'] = customer_dataframe['Created Date'].
      ↪apply(getMonth)
customer_dataframe['Created_Month'].head()
```

```
[26]: 0    Dec
1    Dec
2    Dec
3    Dec
4    Dec
Name: Created_Month, dtype: object
```

```
[27]: (customer_dataframe['Created_Month'].value_counts()).plot(kind='barh',
↳figsize=(20,10), color=list('rgbkymc'), title = 'Month-wise number of
↳received complaints')
```

```
[27]: <matplotlib.axes._subplots.AxesSubplot at 0x1ade7ebf648>
```



Insight 4 The month of May has received the highest number of complaints according to this database in the year 2015

```
[28]: #Order the complaint types based on the average 'Request_Closing_Time',
↳grouping them for different locations.
#grouping the data city wise
customer_dataframe['City'].head()
```

```
[28]: 0    NEW YORK
1    ASTORIA
2    BRONX
3    BRONX
4    ELMHURST
Name: City, dtype: object
```

```
[29]: #checking null values
customer_dataframe['City'].isnull().sum()
```

```
[29]: 2614
```

```
[30]: customer_dataframe['City'].fillna('Not Available')
```



```
[30]: 0          NEW YORK
      1          ASTORIA
      2          BRONX
      3          BRONX
      4          ELMHURST
      ...
      300693    Not Available
      300694    RICHMOND HILL
      300695          BROOKLYN
      300696          BRONX
      300697          NEW YORK
      Name: City, Length: 300698, dtype: object
```

```
[31]: #grouping the dataom
      df_grouped_data = customer_dataframe[['City', 'Complaint_
      ↳Type', 'Request_Closing_Time']].groupby(['City', 'Complaint Type']).mean()
```

```
[32]: df_grouped_data
```

```
[32]:                                     Request_Closing_Time
City    Complaint Type
ARVERNE  Animal Abuse                2.153158
         Blocked Driveway            2.526000
         Derelict Vehicle            2.968889
         Disorderly Youth            3.595000
         Drinking                   0.240000
...
Woodside Blocked Driveway            6.405455
         Derelict Vehicle            4.965000
         Illegal Parking            5.219500
         Noise - Commercial         2.390000
         Noise - Street/Sidewalk     3.410000
```

[764 rows x 1 columns]

```
[33]: #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?
```

```
[34]: import scipy.stats as stats
      from math import sqrt
```

```
[35]: #Performing ANOVA test for the first one

# H0 : All Complaint Type average response time is similar
# H1 : Not similar

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

```
[35]: Blocked Driveway          77044
      Illegal Parking          75361
      Noise - Street/Sidewalk  48612
      Noise - Commercial      35577
      Derelict Vehicle         17718
      Noise - Vehicle          17083
      Animal Abuse             7778
      Traffic                  4498
      Homeless Encampment      4416
      Noise - Park             4042
      Vending                  3802
      Drinking                 1280
      Noise - House of Worship  931
      Posting Advertisement     650
      Urinating in Public       592
      Bike/Roller/Skate Chronic  427
      Panhandling               307
      Disorderly Youth          286
      Illegal Fireworks         168
      Graffiti                 113
      Agency Issues              6
      Squeegee                  4
      Ferry Complaint           2
      Animal in a Park          1
      Name: Complaint Type, dtype: int64
```

```
[36]: top5_complaints_type = customer_dataframe['Complaint Type'].value_counts()[:5]
      top5_complaints_type
```

```
[36]: Blocked Driveway          77044
      Illegal Parking          75361
      Noise - Street/Sidewalk  48612
      Noise - Commercial      35577
      Derelict Vehicle         17718
      Name: Complaint Type, dtype: int64
```

```
[37]: top5_complaints_type_names = top5_complaints_type.index
      top5_complaints_type_names
```

```
[37]: Index(['Blocked Driveway', 'Illegal Parking', 'Noise - Street/Sidewalk',
          'Noise - Commercial', 'Derelict Vehicle'],
          dtype='object')
```

```
[38]: complaint_wise_response_time = customer_dataframe[['Complaint Type',
→ 'Request_Closing_Time']].groupby(['Complaint Type']).mean()
complaint_wise_response_time
```

```
[38]:
```

Complaint Type	Request_Closing_Time
Agency Issues	5.258333
Animal Abuse	5.213240
Animal in a Park	336.830000
Bike/Roller/Skate Chronic	3.766486
Blocked Driveway	4.740904
Derelict Vehicle	7.364152
Disorderly Youth	3.558846
Drinking	3.861859
Ferry Complaint	NaN
Graffiti	7.151062
Homeless Encampment	4.365609
Illegal Fireworks	2.761190
Illegal Parking	4.501152
Noise - Commercial	3.147167
Noise - House of Worship	3.193283
Noise - Park	3.410721
Noise - Street/Sidewalk	3.445222
Noise - Vehicle	3.588984
Panhandling	4.372820
Posting Advertisement	1.975941
Squeegee	4.047500
Traffic	3.448645
Urinating in Public	3.626503
Vending	4.013897

```
[39]: customer_dataframe[customer_dataframe['Complaint Type'] == 'Ferry Complaint']
```

```
[39]:
```

	Unique Key	Created Date	Closed Date	Request_Closing_Time	\
161610	31227876	2015-08-03 08:28:00	NaT	NaN	
192974	31015799	2015-07-06 09:12:00	NaT	NaN	

	Agency	Agency Name	Complaint Type	\
161610	NYPD	New York City Police Department	Ferry Complaint	
192974	NYPD	New York City Police Department	Ferry Complaint	

	Descriptor	Location Type	Incident	Zip	...	Road Ramp	\
161610	Homeless Issue	Terminal	NaN	...		NaN	

192974	Disruptive Passenger	Ferry	NaN	...	NaN
--------	----------------------	-------	-----	-----	-----

	Bridge Highway Segment	Garage Lot Name	Ferry Direction	\
161610	NaN	NaN	NaN	
192974	NaN	NaN	Manhattan Bound	

	Ferry Terminal Name	Latitude	Longitude	Location	\
161610	St. George Terminal (Staten Island)	NaN	NaN	NaN	
192974	Barberi	NaN	NaN	NaN	

	Response_time_category	Created_Month
161610	Very Slow	Aug
192974	Very Slow	Jul

[2 rows x 56 columns]

```
[40]: complaint_wise_response_time.dropna(inplace =True)
      complaint_wise_response_time
```

```
[40]: Request_Closing_Time
Complaint Type
Agency Issues          5.258333
Animal Abuse           5.213240
Animal in a Park       336.830000
Bike/Roller/Skate Chronic  3.766486
Blocked Driveway       4.740904
Derelict Vehicle       7.364152
Disorderly Youth       3.558846
Drinking               3.861859
Graffiti              7.151062
Homeless Encampment    4.365609
Illegal Fireworks      2.761190
Illegal Parking        4.501152
Noise - Commercial     3.147167
Noise - House of Worship 3.193283
Noise - Park           3.410721
Noise - Street/Sidewalk 3.445222
Noise - Vehicle        3.588984
Panhandling            4.372820
Posting Advertisement   1.975941
Squeegee              4.047500
Traffic                3.448645
Urinating in Public    3.626503
Vending                4.013897
```

```
[41]:
```

```
sample_data = customer_dataframe.loc[customer_dataframe['Complaint Type'].
↳isin(top5_complaints_type_names), ['Complaint Type', 'Request_Closing_Time']]
sample_data
```

```
[41]:
```

	Complaint Type	Request_Closing_Time
0	Noise - Street/Sidewalk	0.92
1	Blocked Driveway	1.44
2	Blocked Driveway	4.86
3	Illegal Parking	7.75
4	Illegal Parking	3.45
...
300693	Noise - Commercial	NaN
300694	Blocked Driveway	2.01
300695	Noise - Commercial	3.12
300696	Noise - Commercial	4.09
300697	Noise - Commercial	4.15

[254312 rows x 2 columns]

```
[42]: sample_data.isnull().sum()
```

```
[42]: Complaint Type      0
Request_Closing_Time    2059
dtype: int64
```

```
[43]: sample_data.dropna(how='any', inplace=True)
sample_data.isnull().sum()
```

```
[43]: Complaint Type      0
Request_Closing_Time    0
dtype: int64
```

```
[44]: sample_data.shape
```

```
[44]: (252253, 2)
```

```
[45]: s1 = sample_data[sample_data['Complaint Type'] ==
↳top5_complaints_type_names[0]].Request_Closing_Time
s2 = sample_data[sample_data['Complaint Type'] ==
↳top5_complaints_type_names[1]].Request_Closing_Time
s3 = sample_data[sample_data['Complaint Type'] ==
↳top5_complaints_type_names[2]].Request_Closing_Time
s4 = sample_data[sample_data['Complaint Type'] ==
↳top5_complaints_type_names[3]].Request_Closing_Time
s5 = sample_data[sample_data['Complaint Type'] ==
↳top5_complaints_type_names[4]].Request_Closing_Time
s1.head()
```

```
[45]: 1      1.44
      2      4.86
      7      1.80
      9      1.38
     10      7.80
      Name: Request_Closing_Time, dtype: float64
```

```
[46]: print(s1.isnull().sum())
      print(s2.isnull().sum())
      print(s3.isnull().sum())
      print(s4.isnull().sum())
      print(s5.isnull().sum())
```

```
0
0
0
0
0
```

```
[47]: stats.f_oneway(s1, s2, s3, s4, s5)
```

```
[47]: F_onewayResult(statistic=1799.598683238952, pvalue=0.0)
```

We can see pvalue is less than 0.05 so we reject null hypothesis and so all complaint types average response time is not similar.

```
[48]: # Trying ChiSquare Test for second one :-
      #Are the type of complaint or service requested and location related?

      # H0 : 2 categories - Complain Type and Location is independent means not
      ↪related
      # Ha : 2 categories - Complain Type and Location is dependent means related
```

```
[49]: top5_location = customer_dataframe['City'].value_counts()[:5]
      top5_location
```

```
[49]: BROOKLYN      98307
      NEW YORK      65994
      BRONX         40702
      STATEN ISLAND  12343
      JAMAICA        7296
      Name: City, dtype: int64
```

```
[50]: top5_location_names = top5_location.index
      top5_location_names
```

```
[50]: Index(['BROOKLYN', 'NEW YORK', 'BRONX', 'STATEN ISLAND', 'JAMAICA'],
      dtype='object')
```

```
[51]: sample_data_location_c_type = customer_dataframe.
      ↪loc[(customer_dataframe['Complaint Type'].isin(top5_complaints_type_names))
      ↪& (customer_dataframe['City'].isin(top5_location_names)), ['Complaint Type',
      ↪'City']]
      sample_data_location_c_type.head()
```

```
[51]:
```

	Complaint Type	City
0	Noise - Street/Sidewalk	NEW YORK
2	Blocked Driveway	BRONX
3	Illegal Parking	BRONX
5	Illegal Parking	BROOKLYN
6	Illegal Parking	NEW YORK

```
[52]: ch2, p_value, df, exp_frq = stats.chi2_contingency(pd.
      ↪crosstab(sample_data_location_c_type['Complaint Type'],
      ↪sample_data_location_c_type['City']))
```

```
[53]: print(ch2)
      print(p_value)
```

```
40522.79928349593
```

```
0.0
```

We can see that the pvalue is less than 0.05, so we reject the null hypothesis. That means the complaint type and location are not independent.

THE END! THANK YOU!