# Data Trend 0-5

# April 17, 2017

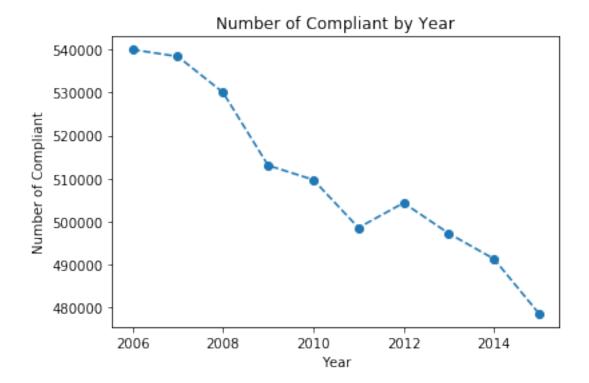
# 0.1 Some Trends

# 0.1.1 1. Number of Crimes by Year

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

from ipykernel import kernelapp as app

Code: *col5\_report\_year.py* This code returns number of crimes by year.



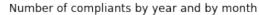
**Trend:** Overall, the crime decreased year over year. Our goal is to identify when, where, and why the crime decreased.

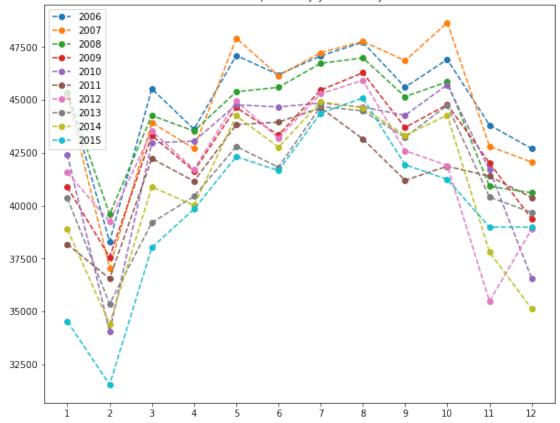
# 0.1.2 2. Number of crimes by year and by month

Code: *col5\_report\_year\_month.py* This code returns number of crimes by each year and each month.

```
In [3]: fig = plt.figure(figsize=(10, 8))
    df = pd.read_table('col5_report_year_month.out',header=-1)
    df = df.sort([0,1], ascending=True)
    for i in df[0].unique():
        plt.plot(list(df[df[0]==i][1]),list(df[df[0]==i][2]),linestyle='--', marker='o',laber
    plt.xticks(list(range(1,13)))
    plt.legend(loc="upper left")
    plt.title("Number of compliants by year and by month")
    plt.show()
```

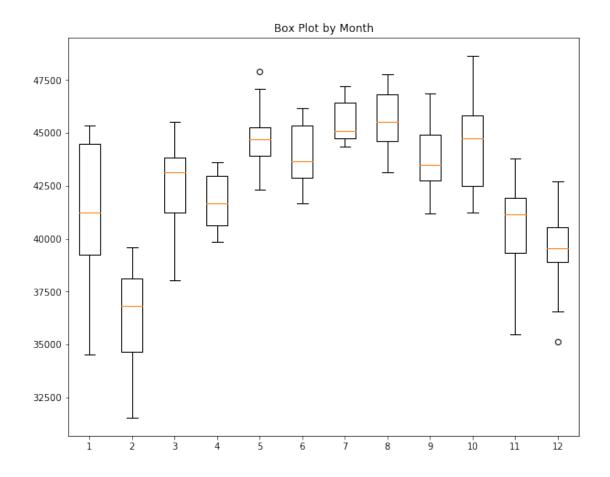
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It seems that Jan and Feb in 2015 has some abnormal number of crimes compared to the other month.

I also plot a boxplot and clearly Feburay has the lowest number of crimes on average (partialy is becasue of a shorter month). It seems there is an outliar in May and December.

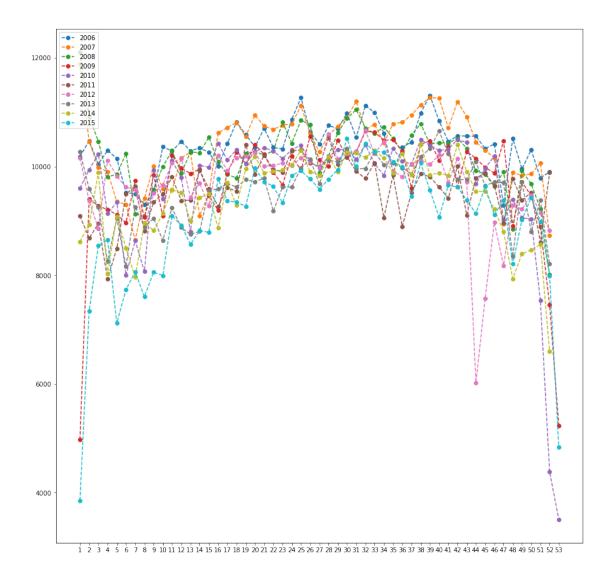


# 0.1.3 3. Number of crimes by year and by week

Code: *col5\_report\_year\_week.py* This code provides number of crimes by year and by week.

```
In [5]: fig = plt.figure(figsize=(15, 15))
    df = pd.read_table('col5_report_year_week.out',header=-1)
    df = df.sort([0,1], ascending=True)
    for i in df[0].unique():
        plt.plot(list(df[df[0]==i][1]),list(df[df[0]==i][2]),linestyle='--', marker='o',laber
    plt.legend(loc="upper left")
    plt.xticks(list(df[1].unique()))
    plt.show()
```

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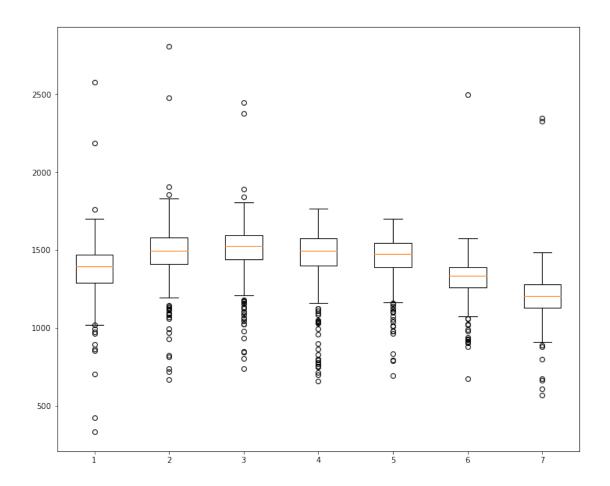
After plotting by week, it is clear that the week 44 in 2012 has an abnormal trend. It is probably because of hurricane sandy happened during that period.

# 0.1.4 4. Number of crimes by day of week

Code: *col5\_report\_year\_week\_day.py* This code provides number of crimes by year and weeknum in each day of the week.

```
In [6]: fig = plt.figure(figsize=(12, 10))
    df = pd.read_table('col5_report_year_week_day.out',header=-1)
    df = df.sort([0,1], ascending=True)
    dow = list()
    for i in (list(range(1,8))):
        dow.append(df[df[2]==i][3])
    plt.boxplot(dow)
    plt.show()
```

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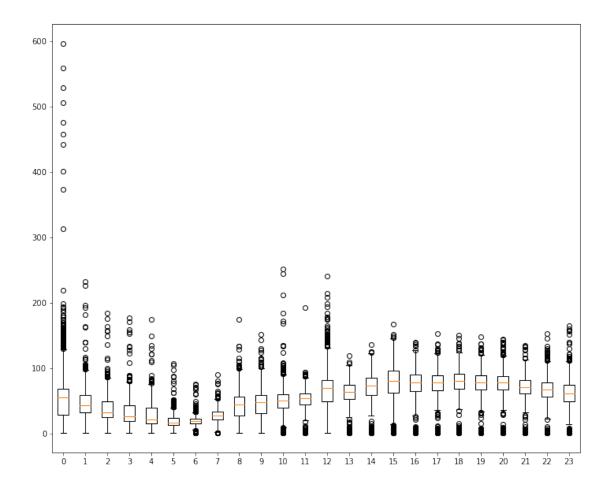
By looking at the boxplot, Tue-Fri is the peak and weekend has less compliants. Also, week-days have more outliars.

# 0.1.5 5. Number of compliants by hour

Code: col1\_2\_hour.py. This code is to check number of crimes by hour.

```
In [7]: fig = plt.figure(figsize=(12, 10))
    df = pd.read_table('col1_2_hour.out',header=-1)
    df = df.sort([0,1,2,3], ascending=True)
    dow = list()
    for i in (list(range(0,24))):
        dow.append(df[df[3]==i][4])
    plt.boxplot(dow)
    plt.xticks(list(range(1,25)),list(range(0,24)))
    plt.show()
```

/Users/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_main\_\_.py:3: FutureWarning: sort app.launch\_new\_instance()



Number of crimes is gradually increasing after 7 am and then start decreasing after midnight. However, there are some very high abnormal outliars happened between 12 - 1 am.

```
In [8]: dfzero = df[df[3]==0]
        dfoutliar = dfzero[dfzero[4]>300]
        dfoutliar
Out[8]:
                   0
                       1
                          2
                             3
                                   4
        17430
               2006
                      52
                          7
                             0
                                506
               2007
                                 596
        17454
                             0
                       1
                          1
               2008
                          2
                                 529
        26214
        34926
               2009
                       1
                             0
                                 373
        52374 2010
                                 475
                      53
                          5
        61134
               2011
                      52
                          6
                             0
                                 558
        69894
               2012
                      52
                          7
                             0
                                 442
```

1 2

69942 2013

```
78702 2014 1 3 0 458
87390 2015 1 4 0 313
```

The outliars are usually between 12 and 1 am on the new year. It makes sense since the NYE celebrations.

# 0.1.6 6. Number of crimes by length of compliant

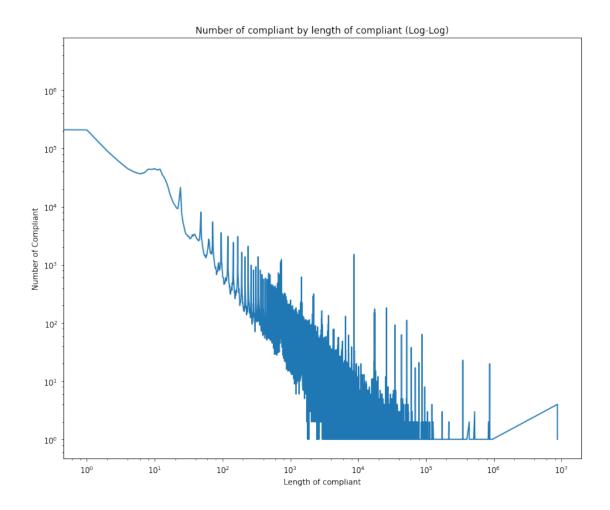
In [9]: fig = plt.figure(figsize=(12, 10))

Code: *length\_of\_compliant.py*. The code provides number of crimes by the length of complaint. When the crime has no to\_date and to\_time, the length will be counted as **0**.

```
df = pd.read_table('col1_2_3_4_length_of_compliant.out',header=-1)
df = df.sort([0], ascending=True)

plt.plot(list(df[0]),list(df[1]),linestyle='-')
plt.title("Number of compliant by length of compliant (Log-Log)")
plt.xlabel("Length of compliant")
plt.ylabel("Number of Compliant")
plt.xscale('log')
plt.yscale('log')
plt.show()
```

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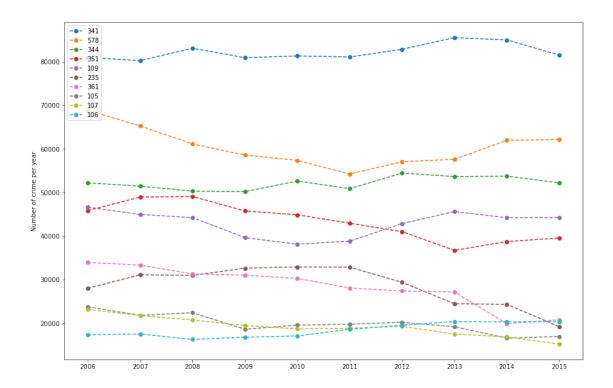
Overall,number of crimes decays over the length of the compliant after transforming to log-log.

# 0.1.7 7. Number of compliants by year and by crime type

Code: *col5\_6\_report\_kycd.py*. The code provides number of crimes by year and by KY\_CD, which stands for the crime type.

```
In [10]: fig = plt.figure(figsize=(15, 10))
    df = pd.read_table('col5_6_report_year_kycd.out',header=-1)
    df.columns = ["year","kycd","count"]
    #find the top 10 KY_CD
    top_kycd = df.groupby(["kycd"], as_index=False).sum().sort(["count"], ascending=False)[
    df = df.sort(["year"],ascending=True)
    for i in list(top_kycd):
        plt.plot(list(df[df["kycd"]==i]["year"]),list(df[df["kycd"]==i]["count"]),linestyle
    plt.legend(loc="upper left")
    plt.xticks(list(df["year"].unique()))
    plt.ylabel("Number of crime per year")
    plt.show()
```

/Users/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_main\_\_.py:5: FutureWarning: sort/Users/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_main\_\_.py:6: FutureWarning: sort/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_main\_\_.py:6: FutureWarning: sort/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_\_main\_\_.py:6: FutureWarning: sort/sunevan/anaconda/lib/python3.6/site-packages/ipykernel/\_\_\_\_.



From the list, the crime 341 is the most popluar and did not decrease over the year. It seems that crime 235 and 361 may contribute the year over year decrease as the trend continued going down after 2012.

# 0.1.8 8. Number of compliants by year and by borough

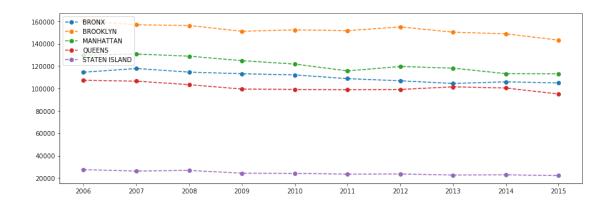
Code: *col5\_13\_report\_year\_borough.py*. The code provides number of crime by number of compliants, year, and borough.

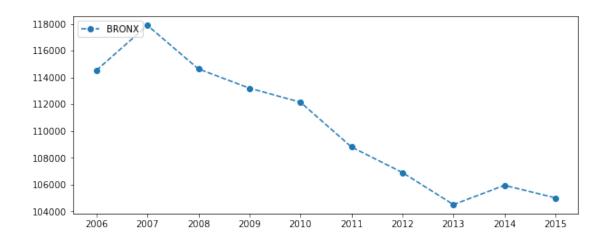
```
In [11]: fig = plt.figure(figsize=(15,5))
    df = pd.read_table('col5_13_report_year_borough.out',header=-1)
    df = df.sort([0,1], ascending=True)
    for i in df[1].unique():
        plt.plot(list(df[df[1]==i][0]),list(df[df[1]==i][2]),linestyle='--', marker='o',lab
    plt.legend(loc="upper left")
    plt.xticks(list(df[0].unique()))
    plt.show()
```

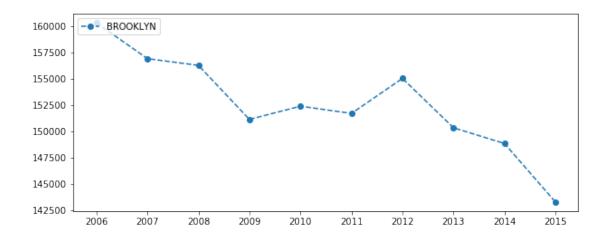
```
for i in df[1].unique():
    fig = plt.figure(figsize=(10,4))
```

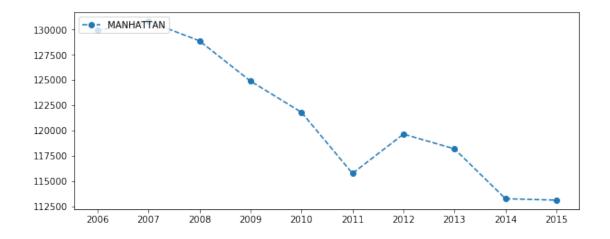
```
plt.plot(list(df[df[1]==i][0]),list(df[df[1]==i][2]),linestyle='--', marker='o',lab
plt.legend(loc="upper left")
plt.xticks(list(df[0].unique()))
plt.show()
```

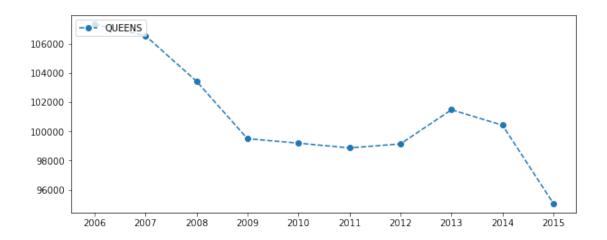
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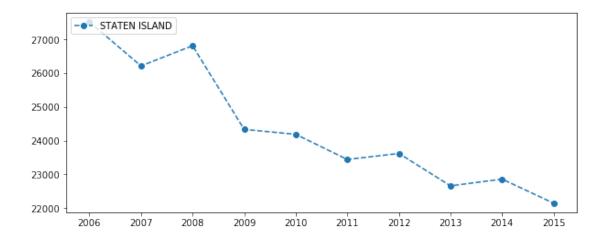












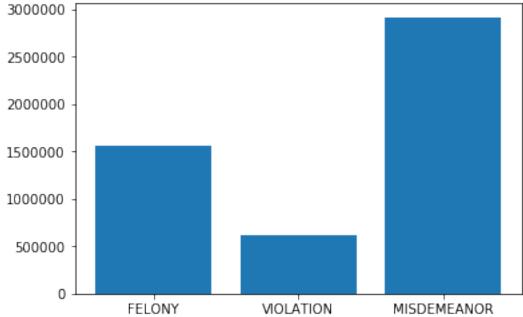
Looking at the stacked the line chart, the decreasing trend is not very clear. After plotting the trend borough by borough, it is more clear that broklyn contributes the most to the decrease, followed by manhattan.

# Data Trend 6-18

# April 17, 2017

#### 0.0.1 Column 11 - Offense Level

```
In [2]: data = pd.read_table('column11_data_quality.out', header = -1)
In [3]: stats = {}
    for level in ['FELONY', 'MISDEMEANOR', 'VIOLATION']:
        stats[level] = len(data[data[0]==level])
In [4]: plt.bar(range(len(stats)), stats.values(), align='center')
        plt.xticks(range(len(stats)), stats.keys())
        plt.show()
```



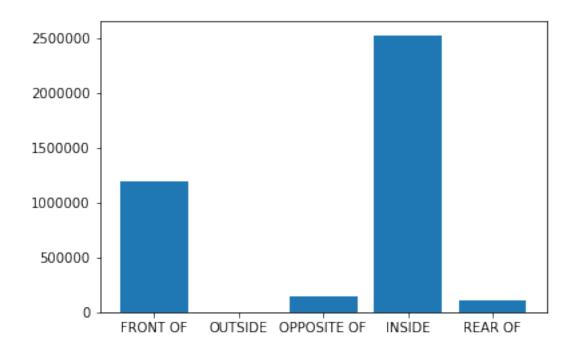
# 0.0.2 Column 13 - Borough Names

```
In [5]: data = pd.read_table('column13_data_quality.out', header = -1)
```

```
In [6]: stats = {}
       for boro in ['BRONX', 'BROOKLYN', 'MANHATTAN', 'QUEENS', 'STATEN ISLAND']:
           stats[boro] = len(data[data[0]==boro])
In [7]: plt.bar(range(len(stats)),stats.values(),align='center')
       plt.xticks(range(len(stats)), stats.keys(), rotation=17)
       plt.show()
      1600000
      1400000
      1200000
      1000000
        800000
        600000
        400000
        200000
             0
               STATEN ISLAND
                           MANHATTAN
                                        BROOKLYN
                                                     BRONX
                                                                QUEENS
```

# 0.0.3 Column 15 - Occurrence Location Description

```
In [8]: data = pd.read_table('column15_data_quality.out', header = -1)
In [9]: stats = {}
    for loc in ['INSIDE', 'OUTSIDE', 'OPPOSITE OF', 'FRONT OF', 'REAR OF']:
        stats[loc] = len(data[data[0]==loc])
In [10]: plt.bar(range(len(stats)), stats.values(), align='center')
        plt.xticks(range(len(stats)), stats.keys())
        plt.show()
```



# Heatmap of Crimes

# April 17, 2017

# 0.1 Heatmap of Crimes

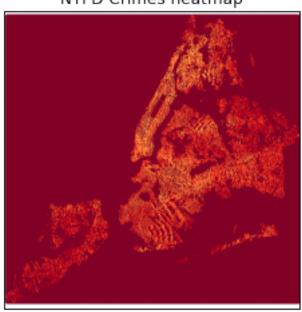
Now, we can plot a heatmap of crimes through history. From the heatmap we can find out whether there is any crime in an invalid location, for example, on the river.

#### 0.1.1 Map of NYC

```
In [2]: # You should have installed mpl_toolkits.basemap first. If not, try conda install basema
        # or use anaconda prompt 'conda install -c conda-forge basemap-data-hires=1.0.8.dev0'
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from matplotlib import cm
        from mpl_toolkits.basemap import Basemap
In [3]: df = pd.read_csv("/Users/xinyan/Downloads/NYPD_Complaint_Data_Historic.csv")
/Users/xinyan/.local/lib/python3.5/site-packages/IPython/core/interactiveshell.py:2717: DtypeWar
  interactivity=interactivity, compiler=compiler, result=result)
In [5]: """
        Heatmap of geolocated collisions in New York City area of a selected year.
        This map only plot data with location information, i.e. latitude and longitude.
        Data without location will be ignored.
        Baesd on tweets heatmap by Kelsey Jordahl, Enthought in Scipy 2013 geospatial tutorial.
        See more in github page: https://github.com/kjordahl/SciPy2013.
        def heatmap(df):
            The function turns input dataframe into heatmap. Input should contain feature
            'LATITUDE' and 'LONGITUDE', otherwise it will not be plotted.
            west, south, east, north = -74.26, 40.49, -73.70, 40.92 # NYC
            a = np.array(df['Latitude'].dropna())
```

# In [10]: heatmap(df)

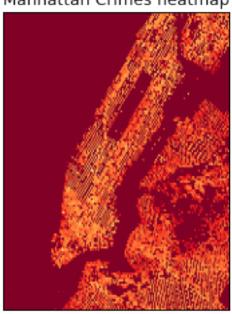




# 0.1.2 Map of 5 boroughs

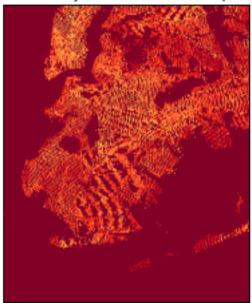
heatmap\_mht(df)

# Manhattan Crimes heatmap

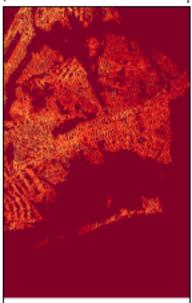


# heatmap\_bk(df)





# Queens Crimes heatmap



Staten Island Crimes heatmap

