

Project - 1

November 1, 2021

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import style
import seaborn as sns
```

```
[2]: service_311 = pd.read_csv("311_Service_Requests_from_2010_to_Present.csv")
```

```
/usr/local/lib/python3.7/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]: service_311.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]: service_311.shape
```

```
[4]: (300698, 53)
```

```
[5]: import datetime
```

```
[6]: from datetime import time
```

```
[9]: df = pd.read_csv("311_Service_Requests_from_2010_to_Present.csv", parse_dates = [
    ↳ ["Created Date", "Closed Date"])
```

```
[11]: df["Request_Closing_Time"] = df["Closed Date"] - df["Created Date"]
```

```
[12]: df["Request_Closing_Time"]
```

```
[12]: 0      00:55:15
      1      01:26:16
      2      04:51:31
      3      07:45:14
```

```

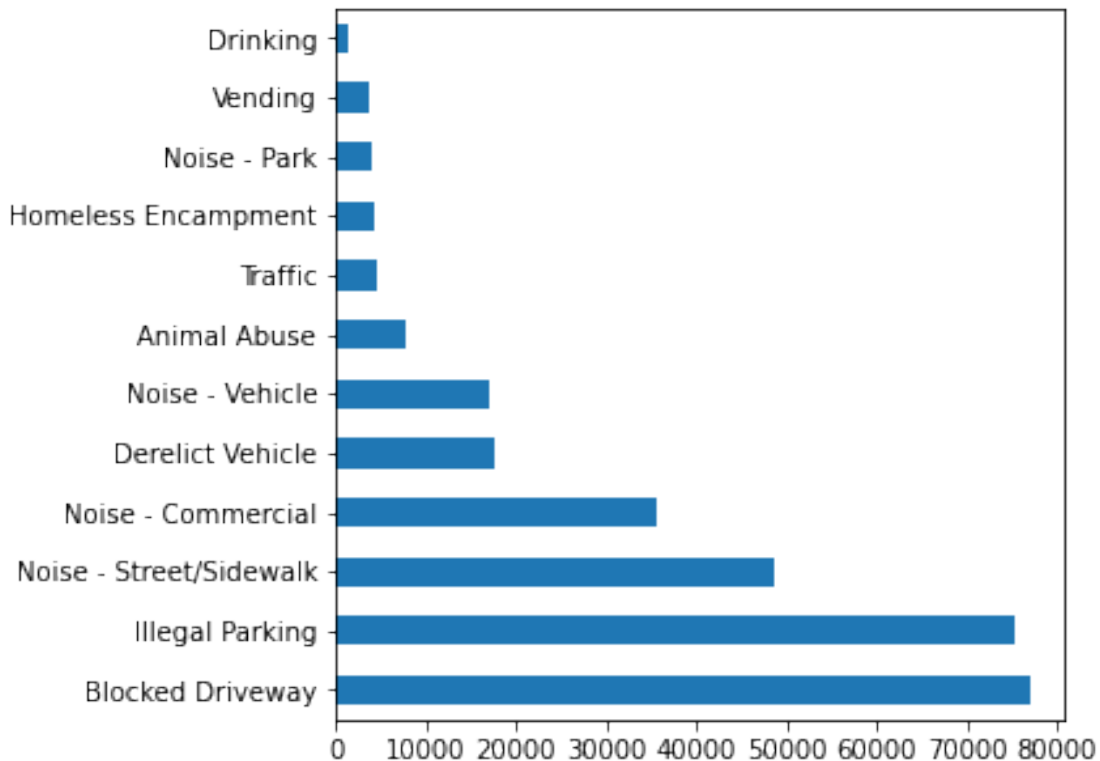
4          03:27:02
...
300693      NaT
300694    02:00:31
300695    03:07:17
300696    04:05:33
300697    04:08:49
Name: Request_Closing_Time, Length: 300698, dtype: timedelta64[ns]

```

```

[15]: service_311["Complaint Type"].value_counts().head(12).plot(kind = '
      ↪ 'barh',figsize = (5,5));

```



```

[16]: df["Request_Closing_Time"].describe()

```

```

[16]: count          298534
      mean    0 days 04:18:51.832782
      std     0 days 06:05:22.141833
      min           0 days 00:01:00
      25%         0 days 01:16:33
      50%    0 days 02:42:55.500000
      75%         0 days 05:21:00
      max        24 days 16:52:22

```

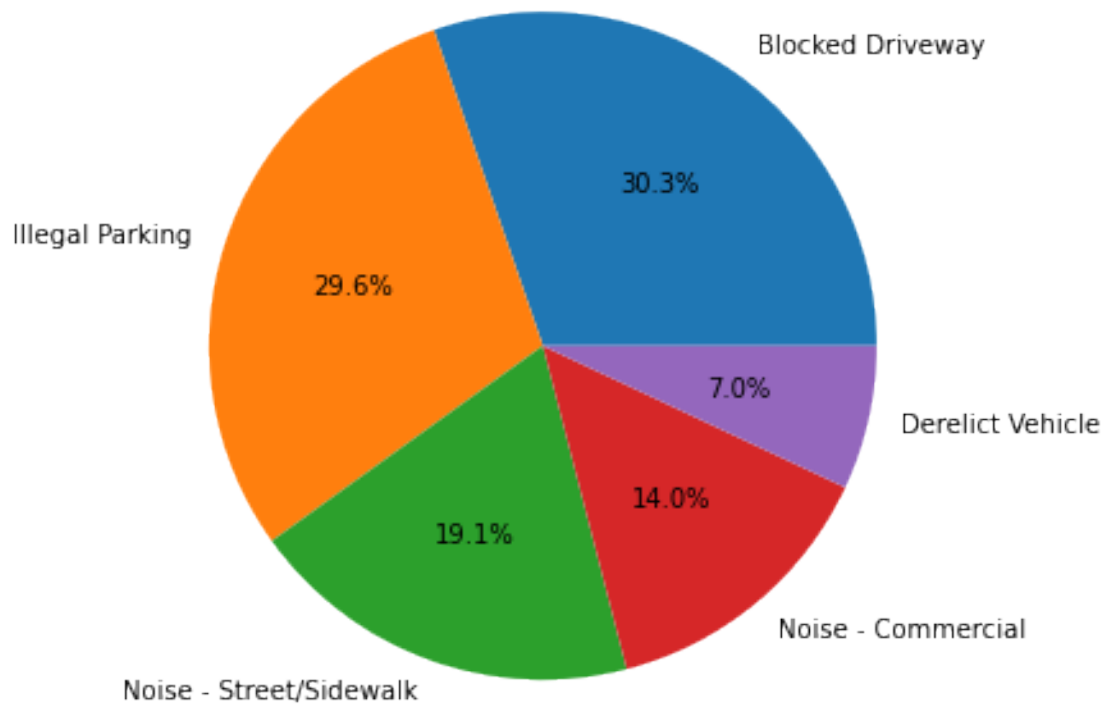
Name: Request_Closing_Time, dtype: object

```
[18]: major_complaints = service_311.dropna(subset = ["Complaint Type"])
major_complaints = service_311.groupby("Complaint Type")
sorted_complaint_type = major_complaints.size().sort_values(ascending = False)
sorted_complaint_type = sorted_complaint_type.to_frame("count").reset_index()
sorted_complaint_type
sorted_complaint_type.head(28)
```

```
[18]:
```

	Complaint Type	count
0	Blocked Driveway	77044
1	Illegal Parking	75361
2	Noise - Street/Sidewalk	48612
3	Noise - Commercial	35577
4	Derelect Vehicle	17718
5	Noise - Vehicle	17083
6	Animal Abuse	7778
7	Traffic	4498
8	Homeless Encampment	4416
9	Noise - Park	4042
10	Vending	3802
11	Drinking	1280
12	Noise - House of Worship	931
13	Posting Advertisement	650
14	Urinating in Public	592
15	Bike/Roller/Skate Chronic	427
16	Panhandling	307
17	Disorderly Youth	286
18	Illegal Fireworks	168
19	Graffiti	113
20	Agency Issues	6
21	Squeegee	4
22	Ferry Complaint	2
23	Animal in a Park	1

```
[19]: sorted_complaint_type = sorted_complaint_type.head()
plt.figure(figsize = (6,6))
plt.pie(sorted_complaint_type['count'],labels = _
↪sorted_complaint_type["Complaint Type"],autopct = "%1.1f%%")
plt.show()
```



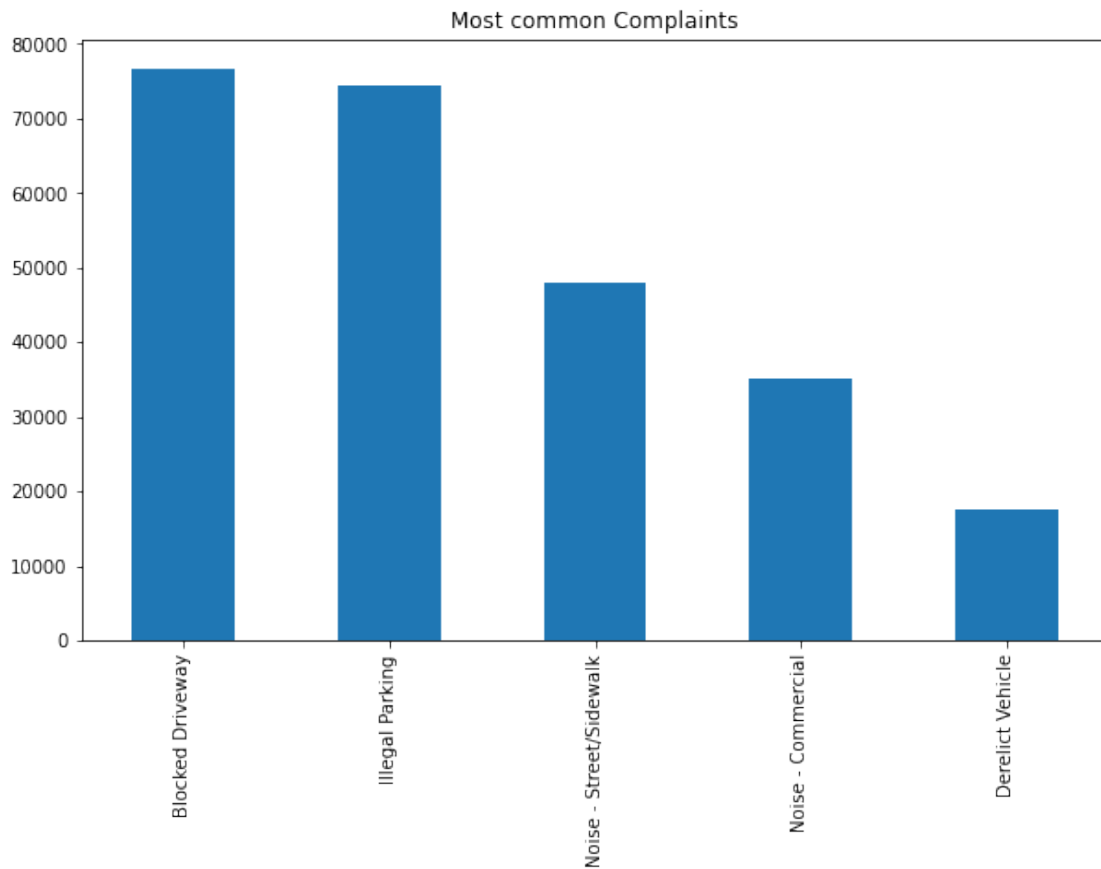
```
[20]: def prepareData(df):
        df['Resolution_Time'] = (df['Closed Date'] - df['Created Date']).dt.
        ↳total_seconds()
        df_clean=df[df['Resolution_Time'].notnull()]
        df_perfect = df_clean[df_clean['Closed Date'] >= df_clean['Created Date']]
        df_perfect['Day of Week'] = df_perfect['Created Date'].dt.dayofweek
        df_perfect['Day of Month'] = df_perfect['Created Date'].dt.day
        df_perfect['Month'] = df_perfect['Created Date'].dt.month
        df_perfect['Year'] = df_perfect['Created Date'].dt.year
        df_perfect=df_perfect[df_perfect.Borough!='Unspecified']
        return df_perfect
```

```
[21]: df_perfect = prepareData(df)
        df_perfect.shape
```

```
[21]: (298068, 59)
```

```
[22]: (df_perfect['Complaint Type'].value_counts()).head().plot(kind='bar',
        ↳figsize=(10,6), title = 'Most common Complaints')
```

```
[22]: <AxesSubplot:title={'center':'Most common Complaints'}>
```



```
[23]: df['Request_Closing_Time'].mean()
```

```
[23]: Timedelta('0 days 04:18:51.832782')
```

```
[24]: df_avg_res_time_city = df_perfect.groupby(['City', 'Complaint Type']).  
      ↪ Resolution_Time.mean()
```

```
[26]: df_avg_res_time = df_perfect.groupby('Complaint Type').Resolution_Time.mean().  
      ↪ sort_values(ascending=True)  
df_avg_res_time.head(28)
```

```
[26]: Complaint Type  
Posting Advertisement      7112.891975  
Illegal Fireworks         9940.101190  
Noise - Commercial       11291.632884  
Noise - House of Worship  11495.874058  
Noise - Park             12246.158157
```

Noise - Street/Sidewalk	12377.738882
Traffic	12415.252002
Disorderly Youth	12810.902098
Noise - Vehicle	12918.914430
Urinating in Public	13055.991554
Bike/Roller/Skate Chronic	13523.545024
Drinking	13879.309748
Vending	14449.060358
Squeegee	14564.250000
Homeless Encampment	15716.052536
Panhandling	15741.963934
Illegal Parking	16149.479466
Blocked Driveway	17057.298659
Animal Abuse	18768.513712
Graffiti	25744.504425
Derelect Vehicle	26445.913579

Name: Resolution_Time, dtype: float64

```
[27]: from scipy import stats
import statsmodels.api as sm
from statsmodels.formula.api import ols
```

```
[28]: complaintTypecity = pd.DataFrame({'count':service_311.groupby(['Complaint_Type', 'City']).size()}).reset_index()
complaintTypecity
```

```
[28]:
```

	Complaint Type	City	count
0	Animal Abuse	ARVERNE	38
1	Animal Abuse	ASTORIA	125
2	Animal Abuse	BAYSIDE	37
3	Animal Abuse	BELLEROSE	7
4	Animal Abuse	BREEZY POINT	2
..
759	Vending	STATEN ISLAND	25
760	Vending	SUNNYSIDE	15
761	Vending	WHITESTONE	1
762	Vending	WOODHAVEN	6
763	Vending	WOODSIDE	15

[764 rows x 3 columns]

```
[29]: service_311.groupby(['Borough', 'Complaint Type', 'Descriptor']).size()
```

```
[29]:
```

Borough	Complaint Type	Descriptor	
BRONX	Animal Abuse	Chained	132
		In Car	36
		Neglected	673

	No Shelter	71
	Other (complaint details)	311
	...	
Unspecified Noise - Vehicle	Engine Idling	11
	Posting Advertisement Vehicle	1
	Traffic Truck Route Violation	1
	Vending In Prohibited Area	2
	Unlicensed	5

Length: 288, dtype: int64

```
[30]: df_dis_youth = df_perfect[df_perfect['Complaint Type']=='Disorderly Youth']
df_dis_youth = df_dis_youth.loc[:,['Resolution_Time']]
df_dis_youth.head()
```

```
[30]: Resolution_Time
4670      713.0
9034     4605.0
12027    2345.0
12176    19415.0
17181     6849.0
```

```
[31]: df_noise_veh = df_perfect[df_perfect['Complaint Type']=='Noise - Vehicle']
df_noise_veh = df_noise_veh.loc[:,['Resolution_Time']]
df_noise_veh.head()
```

```
[31]: Resolution_Time
87      22949.0
156     7254.0
172    11319.0
221    10937.0
319     2615.0
```

```
[32]: df_type_res = df_perfect.loc[:, ['Complaint Type', 'Resolution_Time']]
df_type_res.head()
df_type_res.columns
```

```
[32]: Index(['Complaint Type', 'Resolution_Time'], dtype='object')
```

```
[33]: fvalue, pvalue = stats.f_oneway(df_dis_youth, df_noise_veh)
pvalue
```

```
[33]: array([0.91269878])
```

```
[34]: df_post_ad = df_perfect[df_perfect['Complaint Type']=='Posting Advertisement']
df_post_ad = df_post_ad.loc[:,['Resolution_Time']]
df_post_ad.head()
```



```
[34]: Resolution_Time
      39          7596.0
      42          7745.0
      46          7834.0
      49          8042.0
      51          8137.0
```

```
[35]: df_der_veh = df_perfect[df_perfect['Complaint Type']=='Derelict Vehicle']
      df_der_veh = df_der_veh.loc[:,['Resolution_Time']]
      df_der_veh.head()
```

```
[35]: Resolution_Time
      14          37763.0
      151         14221.0
      255          4913.0
      256         14879.0
      295          2712.0
```

```
[36]: fvalue, pvalue = stats.f_oneway(df_post_ad, df_der_veh)
      pvalue
```

```
[36]: array([7.28776953e-35])
```

```
[37]: df_perfect['Complaint_Type']=df_perfect['Complaint Type']
      df_type_res = df_perfect.loc[:, ['Complaint_Type','Resolution_Time']]
      model = ols('Resolution_Time ~ Complaint_Type', data=df_type_res).fit()
      anova_table = sm.stats.anova_lm(model, typ=2)
      anova_table
```

```
[37]:
```

	sum_sq	df	F	PR(>F)
Complaint_Type	3.784839e+12	20.0	410.258598	0.0
Residual	1.374816e+14	298047.0	NaN	NaN

```
[38]: df_city_type = pd.crosstab(df_perfect.City , df_perfect.Complaint_Type)
```

```
[39]: from scipy.stats import chi2_contingency
      from scipy.stats import chi2

      table = df_city_type

      stat, p, dof, expected = chi2_contingency(table)
      print('dof=%d' % dof)
      print(expected)

      prob = 0.95
      critical = chi2.ppf(prob, dof)
      print('probability=%.3f, critical=%.3f, stat=%.3f' % (prob, critical, stat))
```

```

if abs(stat) >= critical:
    print('Dependent (reject H0)')
else:
    print('Independent (fail to reject H0)')

alpha = 1.0 - prob
print('significance=%.3f, p=%.3f' % (alpha, p))
if p <= alpha:
    print('Dependent (reject H0)')
else:
    print('Independent (fail to reject H0)')

```

```

dof=1040
[[5.73350737e+00 3.11515400e-01 5.66574169e+01 ... 3.31741755e+00
 4.37007385e-01 2.80068584e+00]
 [1.64968644e+02 8.96314763e+00 1.63018841e+03 ... 9.54511504e+01
 1.25738943e+01 8.05833700e+01]
 [1.86599603e+01 1.01384103e+00 1.84394139e+02 ... 1.07966862e+01
 1.42226040e+00 9.11495938e+00]
 ...
 [6.41892211e+01 3.48755650e+00 6.34305536e+02 ... 3.71399974e+01
 4.89249632e+00 3.13549511e+01]
 [9.23615914e+01 5.01822989e+00 9.12699480e+02 ... 5.34405809e+01
 7.03979170e+00 4.51165029e+01]
 [3.12736765e+00 1.69917491e-01 3.09040456e+01 ... 1.80950048e+00
 2.38367665e-01 1.52764682e+00]]
probability=0.950, critical=1116.137, stat=110425.867
Dependent (reject H0)
significance=0.050, p=0.000
Dependent (reject H0)

```

[]:

[]: