#### **Customer Service Requests Analysis**

### In [3]:

```
# import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import style
```

#### In [4]:

```
data_311=pd.read_csv('311_Service_Requests_from_2010_to_Present.csv')
```

/usr/local/lib/python3.7/site-packages/IPython/core/interactiveshell.py:306
3: DtypeWarning: Columns (48,49) have mixed types.Specify dtype option on import or set low\_memory=False.

interactivity=interactivity, compiler=compiler, result=result)

# In [5]:

```
#View top 5 records
data_311.head()
```

### Out[5]:

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Typ
0	32310363	12/31/2015 11:59:45 PM	01-01- 16 0:55	NYPD	New York City Police Department	Noise - Street/Sidewalk	Loud Music/Party	Street/Sidewa
1	32309934	12/31/2015 11:59:44 PM	01-01- 16 1:26	NYPD	New York City Police Department	Blocked Driveway	No Access	Street/Sidewa
2	32309159	12/31/2015 11:59:29 PM	01-01- 16 4:51	NYPD	New York City Police Department	Blocked Driveway	No Access	Street/Sidewa
3	32305098	12/31/2015 11:57:46 PM	01-01- 16 7:43	NYPD	New York City Police Department	Illegal Parking	Commercial Overnight Parking	Street/Sidewa
4	32306529	12/31/2015 11:56:58 PM	01-01- 16 3:24	NYPD	New York City Police Department	Illegal Parking	Blocked Sidewalk	Street/Sidewa

5 rows × 53 columns

- 1. Understand the dataset:
- 2. 1 Identify the shape of the dataset

```
'Complaint Type', 'Descriptor', 'Location Type', 'Incident Zip',
    'Incident Address', 'Street Name', 'Cross Street 1', 'Cross Street
2',

'Intersection Street 1', 'Intersection Street 2', 'Address Type',
    'City', 'Landmark', 'Facility Type', 'Status', 'Due Date',
    'Resolution Description', 'Resolution Action Updated Date',
    'Community Board', 'Borough', 'X Coordinate (State Plane)',
    'Y Coordinate (State Plane)', 'Park Facility Name', 'Park Borough',
    'School Name', 'School Number', 'School Region', 'School Code',
    'School Phone Number', 'School Address', 'School City', 'School State',

'School Zip', 'School Not Found', 'School or Citywide Complaint',
    'Vehicle Type', 'Taxi Company Borough', 'Taxi Pick Up Location',
    'Bridge Highway Name', 'Bridge Highway Direction', 'Road Ramp',
    'Bridge Highway Segment', 'Garage Lot Name', 'Ferry Direction',
    'Ferry Terminal Name', 'Latitude', 'Longitude', 'Location'],
    dtype='object')
```

1. 2 Identify variables with null values

# In [9]:

# # check the null values in dataset data\_311.isnull().sum()

# Out[9]:

Unique Key	0
Created Date	0
Closed Date	2164
Agency	0
Agency Name	0
Complaint Type	0
Descriptor	5914
Location Type	131
Incident Zip	2615
Incident Address	44410
Street Name	44410
Cross Street 1	49279
Cross Street 2	49779
Intersection Street 1	256840
Intersection Street 2	257336
Address Type	2815
City	2614
Landmark	300349
Facility Type	2171
Status	0
Due Date	3
Resolution Description	9
Resolution Action Updated Date	2187
Community Board	2107
Borough	0
_	3540
<pre>X Coordinate (State Plane) Y Coordinate (State Plane)</pre>	3540
· · · · · · · · · · · · · · · · · · ·	_
Park Facility Name	0
Park Borough	0
School Name	0
School Number	0
School Region	1
School Code	1
School Phone Number	0
School Address	0
School City	0
School State	0
School Zip	1
School Not Found	0
School or Citywide Complaint	300698
Vehicle Type	300698
Taxi Company Borough	300698
Taxi Pick Up Location	300698
Bridge Highway Name	300455
Bridge Highway Direction	300455
Road Ramp	300485
Bridge Highway Segment	300485
Garage Lot Name	300698
Ferry Direction	300697
Ferry Terminal Name	300696
Latitude	3540
Longitude	3540
Location	3540
dtype: int64	

dtype: int64

- 1. Perform basic data exploratory analysis:
- 2. 1 Utilize missing value treatment
- 3. 2 Analyze the date column and remove the entries if it has an incorrect timeline
- 4. 3 Draw a frequency plot for city-wise complaints
- 5. 4 Draw scatter and hexbin plots for complaint concentration across Brooklyn

# In [10]:

#As we seen Closed Date is important column and have many missing values
data\_311['Closed Date'].isnull()]

# Out[10]:

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Lo
416	32305700	12/31/2015 02:16:04 PM	NaN	NYPD	New York City Police Department	Illegal Parking	Posted Parking Sign Violation	Str
611	32309308	12/31/2015 09:58:06 AM	NaN	NYPD	New York City Police Department	Noise - Street/Sidewalk	Loud Music/Party	Str
1648	32303348	12/30/2015 05:13:42 AM	NaN	NYPD	New York City Police Department	Illegal Parking	Commercial Overnight Parking	Str
1816	32294519	12/29/2015 10:44:50 PM	NaN	NYPD	New York City Police Department	Derelict Vehicle	With License Plate	Str
1965	32296487	12/29/2015 07:09:13 PM	NaN	NYPD	New York City Police Department	Derelict Vehicle	With License Plate	Str
300273	30287350	03/29/2015 02:40:19 PM	NaN	NYPD	New York City Police Department	Blocked Driveway	No Access	Str
300492	30284963	03/29/2015 08:50:15 AM	NaN	NYPD	New York City Police Department	Vending	Unlicensed	Str
300496	30285492	03/29/2015 08:44:13 AM	NaN	NYPD	New York City Police Department	Vending	Unlicensed	Str
300620	30282717	03/29/2015 01:55:35 AM	NaN	NYPD	New York City Police Department	Noise - Commercial	Loud Music/Party	Club/Ba
300693	30281872	03/29/2015 12:33:41 AM	NaN	NYPD	New York City Police Department	Noise - Commercial	Loud Music/Party	Club/Ba
2164 rows × 53 columns								

### In [11]:

```
data_311[data_311['Unique Key']==32305700]
```

# Out[11]:

	Unique Key	Created Date	Closed Date	Agency	Agency Name	Complaint Type	Descriptor	Location Type
416	32305700	12/31/2015 02:16:04 PM	NaN	NYPD	New York City Police Department	Illegal Parking	Posted Parking Sign Violation	Street/Sidewalk
1 row	ıs × 53 colu	ımns						
4								

# In [16]:

```
data_311['Created Date'] = data_311['Created Date'].astype('datetime64[ns]')
data_311['Closed Date'] = data_311['Closed Date'].astype('datetime64[ns]')
data_311[['Created Date','Closed Date']]
```

# Out[16]:

	Created Date	Closed Date
0	2015-12-31 23:59:45	2016-01-01 00:55:00
1	2015-12-31 23:59:44	2016-01-01 01:26:00
2	2015-12-31 23:59:29	2016-01-01 04:51:00
3	2015-12-31 23:57:46	2016-01-01 07:43:00
4	2015-12-31 23:56:58	2016-01-01 03:24:00
300693	2015-03-29 00:33:41	NaT
300694	2015-03-29 00:33:28	2015-03-29 02:33:59
300695	2015-03-29 00:33:03	2015-03-29 03:40:20
300696	2015-03-29 00:33:02	2015-03-29 04:38:35
300697	2015-03-29 00:33:01	2015-03-29 04:41:50

300698 rows × 2 columns

```
In [20]:
```

```
data_311['Request_Closing_Time']=data_311['Closed Date']-data_311['Created Date']
data_311['Request_Closing_Time']
Out[20]:
0
         0 days 00:55:15
1
         0 days 01:26:16
2
         0 days 04:51:31
3
         0 days 07:45:14
4
         0 days 03:27:02
300693
                     NaT
300694
         0 days 02:00:31
300695
         0 days 03:07:17
         0 days 04:05:33
300696
300697
         0 days 04:08:49
Name: Request_Closing_Time, Length: 300698, dtype: timedelta64[ns]
In [24]:
#Num of Complaints closed after Due Date
data_311['Due Date'] = data_311['Due Date'].astype('datetime64[ns]')
over_time = data_311[data_311['Due Date']<data_311['Closed Date']]['Unique Key'].count()</pre>
less_time = data_311[data_311['Due Date']>data_311['Closed Date']]['Unique Key'].count()
same_time = data_311[data_311['Due Date']==data_311['Closed Date']]['Unique Key'].count()
not_closed = data_311['Closed Date'].isnull().sum()
times = np.array([over_time,less_time,same_time,not_closed])
names = np.array(['over_time','less_time','same_time','not_closed'])
times = pd.Series(times)
```

#### Out[24]:

dataframe

	time_taken	time
0	over_time	38580
1	less_time	259892
2	same_time	61
3	not_closed	2164

names = pd.Series(names)

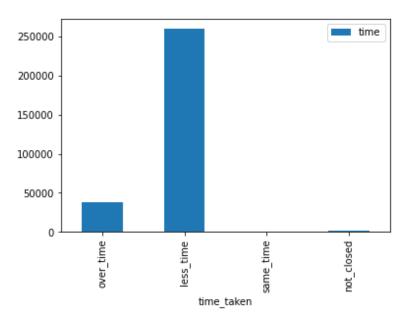
dataframe = pd.DataFrame({'time\_taken':names, 'time':times})

# In [26]:

```
dataframe.plot(kind='bar',x='time_taken')
```

# Out[26]:

<AxesSubplot:xlabel='time\_taken'>

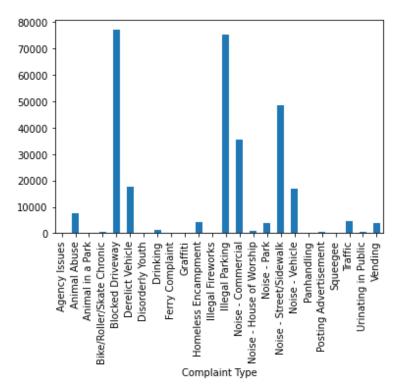


# In [28]:

```
data_311.groupby('Complaint Type')['Unique Key'].count().plot(kind='bar')
```

### Out[28]:

# <AxesSubplot:xlabel='Complaint Type'>

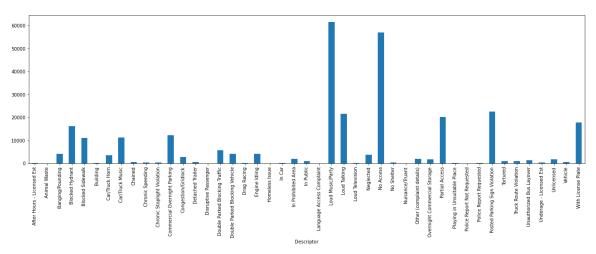


### In [30]:

```
data_311.groupby('Descriptor')['Unique Key'].count().plot(kind='bar',figsize=[22,6])
```

### Out[30]:

# <AxesSubplot:xlabel='Descriptor'>



# In [31]:

```
data_311.groupby('Status')['Status'].agg('count')
```

# Out[31]:

Status

Assigned 786 Closed 298471 Draft 2 Open 1439

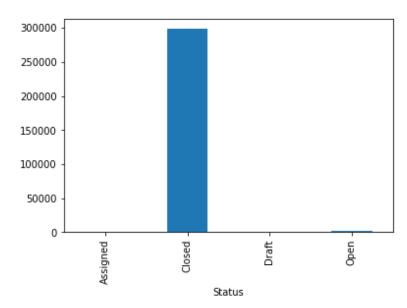
Name: Status, dtype: int64

# In [32]:

```
data_311.groupby('Status')['Status'].agg('count').plot(kind='bar')
```

# Out[32]:

# <AxesSubplot:xlabel='Status'>



#### In [33]:

```
data_311.groupby('Location Type')['Unique Key'].count()
```

#### Out[33]:

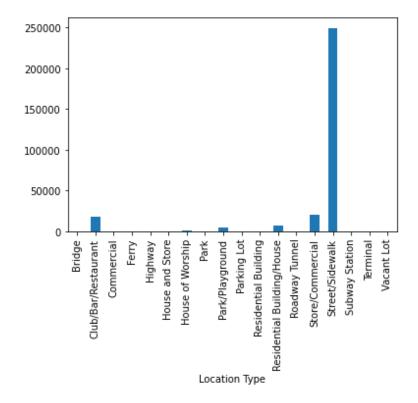
Location Type Bridge 2 Club/Bar/Restaurant 17360 Commercial 62 Ferry 1 Highway 215 House and Store 93 929 House of Worship Park 1 Park/Playground 4773 Parking Lot 117 Residential Building 227 Residential Building/House 6960 Roadway Tunnel 35 Store/Commercial 20381 249299 Street/Sidewalk Subway Station 34 Terminal 1 Vacant Lot 77 Name: Unique Key, dtype: int64

# In [34]:

```
data_311.groupby('Location Type')['Unique Key'].count().plot(kind='bar')
```

## Out[34]:

<AxesSubplot:xlabel='Location Type'>



```
In [40]:
```

```
#convert Request_Closing_Time to mins
data_311['Request_Closing_Time'] = data_311['Request_Closing_Time'].astype('timedelta64
[m]')
```

#### In [41]:

```
data 311['Request Closing Time']
Out[41]:
0
            55.0
1
            86.0
2
          291.0
3
          465.0
4
          207.0
300693
             NaN
300694
          120.0
300695
          187.0
300696
          245.0
300697
          248.0
```

#### In [42]:

```
data_311.groupby('Location Type')['Request_Closing_Time'].agg('mean').sort_values()
```

### Out[42]:

```
Location Type
Subway Station
                                 141.970588
Club/Bar/Restaurant
                                 185.769954
House of Worship
                                 191.523193
Store/Commercial
                                 197.783035
Park/Playground
                                 206.836876
Highway
                                 223.074766
Bridge
                                 229.000000
Roadway Tunnel
                                 266.085714
Street/Sidewalk
                                 268.211355
Residential Building
                                 288.775330
House and Store
                                 300.462366
Residential Building/House
                                 309.203222
Parking Lot
                                 319.863248
Commercial
                                 320.193548
Vacant Lot
                                 448.103896
Park
                               20210.000000
Ferry
                                        NaN
Terminal
                                        NaN
Name: Request_Closing_Time, dtype: float64
```

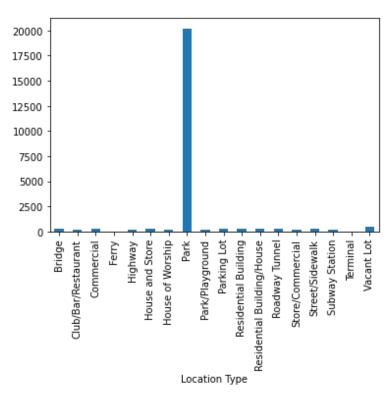
Name: Request\_Closing\_Time, Length: 300698, dtype: float64

# In [43]:

data\_311.groupby('Location Type')['Request\_Closing\_Time'].agg('mean').plot(kind='bar')

# Out[43]:

<AxesSubplot:xlabel='Location Type'>

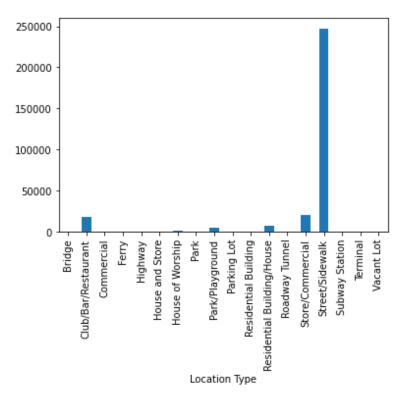


# In [45]:

data\_311.groupby('Location Type')['Request\_Closing\_Time'].agg('count').plot(kind='bar')

# Out[45]:

<AxesSubplot:xlabel='Location Type'>



# In [46]:

```
data_311.groupby('Complaint Type')['Request_Closing_Time'].agg('mean')
```

# Out[46]:

Complaint Type	
Agency Issues	315.333333
Animal Abuse	312.489830
Animal in a Park	20210.000000
Bike/Roller/Skate Chronic	225.693396
Blocked Driveway	284.148041
Derelict Vehicle	441.542756
Disorderly Youth	213.167832
Drinking	231.391373
Ferry Complaint	NaN
Graffiti	428.752212
Homeless Encampment	261.637908
Illegal Fireworks	165.470238
Illegal Parking	269.762894
Noise - Commercial	188.524272
Noise - House of Worship	191.287406
Noise - Park	204.343859
Noise - Street/Sidewalk	206.415509
Noise - Vehicle	215.038220
Panhandling	262.072131
Posting Advertisement	118.251543
Squeegee	242.500000
Traffic	206.614991
Urinating in Public	217.302365
Vending	240.524901
<pre>Name: Request_Closing_Time,</pre>	dtype: float64

#### In [47]:

```
df = data_311[['Complaint Type', 'Request_Closing_Time']].sort_values(by='Complaint Type'
).reset_index().drop('index',axis=1)
df.columns = ['Complaint_Type','Request_Closing_Time']
df
```

### Out[47]:

Complaint_Type	Request_Closing_Tim	ıe
----------------	---------------------	----

0	Agency Issues	67.0
1	Agency Issues	623.0
2	Agency Issues	162.0
3	Agency Issues	452.0
4	Agency Issues	177.0
300693	Vending	17.0
300694	Vending	19.0
300695	Vending	137.0
300696	Vending	68.0
300697	Vending	128.0

300698 rows × 2 columns

### In [48]:

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

lm = ols('Request_Closing_Time ~ Complaint_Type',data=df).fit()
table = sm.stats.anova_lm(lm)
print(table)
```

```
PR(>F)
                      df
                                sum_sq
                                             mean_sq
                                                               F
Complaint_Type
                    23.0
                          1.455041e+09
                                        6.326263e+07
                                                      491.829393
                                                                      0.0
Residual
                298511.0 3.839663e+10
                                        1.286272e+05
                                                             NaN
                                                                      NaN
```

Ho = Complaint Type and Location Type are not related.

H1 = Complaint Type and Location Type are related

### In [50]:

```
df1 = data_311[pd.notnull(data_311['Location Type'])]
df2 = df1[['Complaint Type', 'Location Type']].sort_values(by='Complaint Type').reset_ind
ex().drop('index',axis=1)
df2
```

### Out[50]:

	Complaint Type	Location Type
0	Animal Abuse	Residential Building/House
1	Animal Abuse	Residential Building/House
2	Animal Abuse	Residential Building/House
3	Animal Abuse	Residential Building/House
4	Animal Abuse	Residential Building/House
300562	Vending	Street/Sidewalk
300563	Vending	Store/Commercial
300564	Vending	Street/Sidewalk
300565	Vending	Residential Building/House
300566	Vending	Street/Sidewalk

300567 rows × 2 columns

### In [51]:

```
data_crosstab = pd.crosstab(df2['Complaint Type'], df2['Location Type'], margins = False)
from scipy.stats import chi2_contingency
g,p,dof,expctd = chi2_contingency(data_crosstab)
g,p
```

### Out[51]:

(1638407.580569627, 0.0)

### In [ ]: