

There is also a first project to work on, based on the 311 service request data from Open data buffalo. Most of the variables are categorical in nature, the categories of major interest are probably

-Status, subject, reason, type, object type, council district, police district, neighborhood

-There is also a couple of date variables

It looks like status, subject, reason, type and object type might be interesting categorical variables. Which of these can be described by a limited number of categories? Which are not described easily as a limited number of outcomes

Figure out which of the outcomes fit into a limited number of categories, and then produce some graphs explaining how they are related to council district, police district, neighborhood.

Do different districts and/or neighborhoods tend to different numbers of 311 calls?

Do different months have different number of 311 reports? What about different years?

This is a two week long project. Show your results visually, make sure everything is labeled clearly in your ipynb file. Write a clear discussion of each finding and be sure the sections of the notebook are all clearly labeled.

Set up a repository for this project on GitHUB, save your ipynb file there, the pdf version of the ipynb and also a ReadMe file describing the project

Turn in the pdf of your notebook and also a link to the repository on GitHub.

Upload the pdf version of the ipynb file.

What I am really looking for in this project is:

- a.) Some interesting insights into 311 calls- search for at least one
- b.) A clear visual demonstration of that insight
- c.) Documentation of the use of GitHub

Import Data, set up libraries and other options

In []:

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

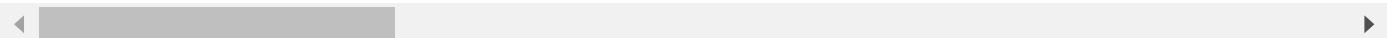
pd.set_option("display.max_columns", 200)
```

```
In [ ]: infile = "311_Service_Requests_20240221.csv"
df311 = pd.read_csv(infile)
df311.head()
```

```
<ipython-input-463-683123bb5431>:2: DtypeWarning: Columns (0) have mixed types. Specify dtype option on import or set low_memory=False.
df311 = pd.read_csv(infile)
```

Out[]:

	Case Reference	Open Date	Closed Date	Status	Subject	Reason	Type	Object Type	Address Number
0	514722	07/19/2010 12:06:00 PM	07/19/2010 02:15:00 PM	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	NaN F
1	1000802971	02/12/2018 06:47:00 AM	02/12/2018 10:33:00 AM	Closed	Buffalo Police Department	Police	Police Issue (Req_Serv)	Street	NaN I
2	541218	01/11/2011 09:12:00 AM	01/26/2011 10:01:00 AM	Closed	Utilities	National Grid	Streetlights (Req_Serv)	Street	NaN
3	1000708750	07/10/2017 11:36:00 AM	07/10/2017 02:27:00 PM	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	NaN
4	1000892866	08/18/2018 03:44:00 PM	08/20/2018 10:00:00 PM	Closed	Dept of Public Works	Forestry	Tree Planting Request (Req_Serv)	Property	NaN



Understand the data I am dealing with

```
In [ ]: df311.shape
```

```
Out[ ]: (1112730, 34)
```

Thats beefy

```
In [ ]: df311.columns
```

```
Out[ ]: Index(['Case Reference', 'Open Date', 'Closed Date', 'Status', 'Subject',
       'Reason', 'Type', 'Object Type', 'Address Number', 'Address Line 1',
       'Address Line 2', 'City', 'State', 'Zipcode', 'Property ID', 'Location',
       'Latitude', 'Longitude', 'Council District', 'Council District 2011',
       'Police District', 'Census Tract', 'Census Block Group', 'Census Block',
       'Neighborhood', 'X Coordinate', 'Y Coordinate', '2010 Census Tract',
       '2010 Census Block Group', '2010 Census Block', 'TRACTCE20',
       'GEOID20_tract', 'GEOID20_blockgroup', 'GEOID20_block'],
      dtype='object')
```

```
In [ ]: df311.dtypes
```

```
Out[ ]:   Case Reference      object
          Open Date        object
          Closed Date      object
          Status           object
          Subject          object
          Reason           object
          Type             object
          Object Type      object
          Address Number    object
          Address Line 1    object
          Address Line 2    object
          City              object
          State             object
          Zipcode          object
          Property ID       object
          Location          object
          Latitude          float64
          Longitude         float64
          Council District  object
          Council District 2011  object
          Police District   object
          Census Tract      object
          Census Block Group object
          Census Block       object
          Neighborhood       object
          X Coordinate      float64
          Y Coordinate      float64
          2010 Census Tract  object
          2010 Census Block Group object
          2010 Census Block  object
          TRACTCE20          object
          GEOID20_tract       object
          GEOID20_blockgroup  object
          GEOID20_block       object
          dtype: object
```

In []: df311.describe()

	Latitude	Longitude	X Coordinate	Y Coordinate
count	1.019924e+06	1.019924e+06	5.677520e+05	5.677520e+05
mean	4.299835e+01	-7.899725e+01	-7.135284e+06	4.525047e+06
std	1.296937e-02	2.080884e-02	3.614622e+06	1.711193e+06
min	4.283199e+01	-7.900000e+01	-8.784200e+06	-8.776799e+06
25%	4.300000e+01	-7.900000e+01	-8.779050e+06	5.290542e+06
50%	4.300000e+01	-7.900000e+01	-8.775132e+06	5.297801e+06
75%	4.300000e+01	-7.900000e+01	-8.772702e+06	5.301150e+06
max	4.300000e+01	-7.879899e+01	5.302890e+06	5.306748e+06

Prepare my dat for Analysis with plots

I am going to keep Open Date to reference Dates and the categorical variables : Status, subject, reason, type, object typethen compare their relationship with council district, police district, neighborhood.

In []: df311.columns

```
Out[ ]: Index(['Case Reference', 'Open Date', 'Closed Date', 'Status', 'Subject',
       'Reason', 'Type', 'Object Type', 'Address Number', 'Address Line 1',
       'Address Line 2', 'City', 'State', 'Zipcode', 'Property ID', 'Location',
       'Latitude', 'Longitude', 'Council District', 'Council District 2011',
       'Police District', 'Census Tract', 'Census Block Group', 'Census Block',
       'Neighborhood', 'X Coordinate', 'Y Coordinate', '2010 Census Tract',
       '2010 Census Block Group', '2010 Census Block', 'TRACTCE20',
       'GEOID20_tract', 'GEOID20_blockgroup', 'GEOID20_block'],
      dtype='object')
```

1. Get rid of 'Case Reference' I want to understand a general trend
2. Keep open date to reference Dates remove closed dates
3. Status subject reason type and object type are useful for understanding a general trend as they describe the nature of the call
4. I would like to understand the realtions of these variables to council district, police district, neighborhood

In []: working311 = df311[['Open Date', 'Status', 'Subject', 'Reason', 'Type', 'Object Type', 'Council District', 'Police District', 'Neighborhood']]
working311.head()

Out[]:

	Open Date	Status	Subject	Reason	Type	Object Type	Council District	Police District	Neighborhood
0	07/19/2010 12:06:00 PM	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
1	02/12/2018 06:47:00 AM	Closed	Buffalo Police Department	Police	Police Issue (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
2	01/11/2011 09:12:00 AM	Closed	Utilities	National Grid	Streetlights (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
3	07/10/2017 11:36:00 AM	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
4	08/18/2018 03:44:00 PM	Closed	Dept of Public Works	Forestry	Tree Planting Request (Req_Serv)	Property	UNKNOWN	UNKNOWN	UNKNOWN

In []: working311.dtypes

```
Out[ ]: Open Date      object
         Status       object
         Subject      object
         Reason       object
         Type         object
         Object Type   object
         Council District object
         Police District object
         Neighborhood  object
         dtype: object
```

Lets figure out a way to change the data type of the Date to be a Date type in python.

They look like Datetime so import the libaries moduels. And convert it to date time.

In []: `from datetime import datetime`

```
working311["Open Date"] = pd.to_datetime(working311["Open Date"])
```

```
<ipython-input-471-f6140a70ffc2>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
working311["Open Date"] = pd.to_datetime(working311["Open Date"])
```

In []: working311.dtypes

```
Out[ ]: Open Date      datetime64[ns]
         Status       object
         Subject      object
         Reason       object
         Type         object
         Object Type   object
         Council District object
         Police District object
         Neighborhood  object
         dtype: object
```

In []: working311.shape

```
Out[ ]: (1112730, 9)
```

In []: working311.isna().sum()

```
Out[ ]: Open Date      0
         Status       0
         Subject      0
         Reason       0
         Type         0
         Object Type   0
         Council District 1797
         Police District 0
         Neighborhood  0
         dtype: int64
```

This is such a small amount compared to 1112730 I am just going to drop all the rows where Council District is null.

```
In [ ]: working311=working311.loc[~working311["Council District"].isna()]

In [ ]: working311.shape

Out[ ]: (1110933, 9)
```

Which variables have limited catagories?

```
In [ ]: working311.columns

Out[ ]: Index(['Open Date', 'Status', 'Subject', 'Reason', 'Type', 'Object Type',
   'Council District', 'Police District', 'Neighborhood'],
   dtype='object')

In [ ]: len(working311.columns)

Out[ ]: 9

In [ ]: working311["Council District"].unique()

Out[ ]: array(['UNKNOWN', 'LOVEJOY', 'FILLMORE', 'SOUTH', 'NORTH', 'NIAGARA',
   'DELAWARE', 'UNIVERSITY', 'MASTEN', 'ELLICOTT'], dtype=object)

In [ ]: len(working311["Council District"].unique())

Out[ ]: 10
```

10 catagories, I am going to keep the "Unknown" catagory to account for all the unidentified values.

```
In [ ]: working311["Open Date"].unique()

Out[ ]: array(['2010-07-19T12:06:00.000000000', '2018-02-12T06:47:00.000000000',
   '2011-01-11T09:12:00.000000000', ..., '2024-02-20T09:22:00.000000000', '2024-02-20T09:07:00.000000000',
   '2024-02-20T08:36:00.000000000'], dtype='datetime64[ns]')

In [ ]: len(working311["Open Date"].unique())

Out[ ]: 839254
```

This is not a catagorical variable as it continually increases as time goes on but I just wanted to observe

```
In [ ]: working311["Status"].unique()

Out[ ]: array(['Closed', 'Open'], dtype=object)

In [ ]: len(working311["Status"].unique())
```

Out[]: 2

In []: working311["Subject"].unique()

```
Out[ ]: array(['Dept of Public Works', 'Buffalo Police Department', 'Utilities',
   'DPIS', 'Assessment & Taxation', 'Dept of Parking',
   'Buffalo Municipal Housing Authority', 'Buffalo Fire Department',
   'City Clerk', 'Office of the Mayor',
   'Office of Strategic Planning', 'Dept of Law',
   'Community Services & Rec. Program.', 'New Americans',
   'Human Resources', 'Management Information Systems', 'Test',
   'Knowledge Management'], dtype=object)
```

In []: len(working311["Subject"].unique())

Out[]: 18

This one is fairly small I will see how my graphs looks.

In []: working311["Reason"].unique()

```
Out[ ]: array(['Streets', 'Police', 'National Grid', 'Forestry', 'Rodent Control',
   'Sanitation', 'Housing', 'Buffalo Sewer Authority',
   'Animal Shelter', 'City Parks', 'Taxation',
   'Buffalo Water Authority', 'Engineering - Street Repairs',
   'Parking Violations Bureau', 'BMHA', 'BFD', 'Licenses',
   'City Clerk Issue', 'Engineering - Traffic',
   'Citizen Services - Quick Response Teams',
   'Citizens Services - Clean City',
   'Citizen Services - Save Our Streets', 'Real Estate',
   'Moving Violations', 'Administration', 'Streets/Sanitation',
   'Citizen Services - Graffiti', 'Telecommunications',
   'Freedom of Information', 'Adjudication - Ordinance Violation',
   'Assessment & Taxation', 'Harbor Master', 'Personnel',
   'Community Based Orgs', 'Buildings Division', 'Immigration', 'HR',
   'OSP', 'COB APP Issues', 'Youth Bureau',
   'Citizen Services - Thrive Program', 'Rodent_Pest Control',
   'Citizen Services - Good Neighbor', 'Assessment', 'Fair Housing',
   'Citizen Services - Tele-Engagement', 'Test',
   'Citizen Services - Weed & Seed', 'Knowledge',
   'Citizens Services - BreakOut & Love Buffalo', 'ADA'], dtype=object)
```

In []: len(working311["Reason"].unique())

Out[]: 51

This one is even larger but for a data set of a 1 million rows we can consider this to be a limited number of categories.

In []: working311["Type"].unique()

```
Out[ ]: array(['Sweeper (Req_Serv)', 'Police Issue (Req_Serv)',  

   'Streetlights (Req_Serv)', 'Tree Planting Request (Req_Serv)',  

   'Rodents (Req_Serv)', 'Excess Trash (Req_Serv)',  

   'Totes Replace (Req_Serv)', 'Tree Removal (Req_SERV)',  

   'Housing Violations (Req_Serv)', 'Street Flooding (Req_Serv)',  

   'Dead Animal Removal (Req_Serv)', 'Pest (Req_Serv)',  

   'Totes Combo (Req_Serv)', 'Animals (Req_Serv)',  

   'Snow Removal Inspection (Req_Serv)', 'Sewer (Req_Serv)',  

   'City Parks (Req_Serv)', 'Recycling Missed Pick Up (Req_Serv)',  

   'User Fee (Req_Serv)', 'Taxation Issue (Req_Serv)',  

   'Tree Trimming Request (Req_Serv)', 'Bulk Trash (Req_Serv)',  

   'Water (Req_Serv)', 'Street Snow Plowing (Req_Serv)',  

   'Olmsted Parks (Req_Serv)', 'Garbage Missed Pick Up (Req_Serv)',  

   'PW Vacant Lot (Req_Serv)', 'Pot Hole (Req_Serv)',  

   'PVB Single Meter (Req_Serv)', 'Tree Other (Req_Serv)',  

   'Recycling Tote Deliver (Req_Serv)',  

   'Pools and Splashpads (Req_Serv)', 'BMHA Issue (Req_Serv)',  

   'PVB Pay Station (Req_Serv)', 'Paving (Req_Serv)',  

   'Quality of Life Issue (Req_Serv)', 'Fire (Req_Serv)',  

   'Lead Paint Inspection (Req_Serv)', 'Totes Deliver (Req_Serv)',  

   'Electronic Waste (Req_Serv)', 'Boarding request (Req_Serv)',  

   'Electrical Issue (Req_Serv)', 'Fridge (Req_Serv)',  

   'Trash Ordinance Violation (Req_Serv)', 'Stump Removal (Req_Serv)',  

   'Rental Registration (Req_Serv)', 'Street Salting (Req_Serv)',  

   'Cave In (Req_Serv)', 'Christmas Tree (Req_Serv)',  

   'Dog License Issue (Req_Serv)',  

   'Street Snow Plowing Issue (Req_Serv)',  

   'Recycling Bin Pickup (Req_Serv)', 'Totes Audit (Req_Serv)',  

   'Totes Pickup (Req_Serv)', 'Fallen Tree Blocking Row (Req_Serv)',  

   'Pick and Pay (Req_Serv)', 'City Clerk Issue (Req_Serv)',  

   'Graffiti Private Property (Req_Serv)', 'Water Issue (Req_Serv)',  

   'Recycling Tote Combo (Req_Serv)',  

   'Leaves / Lawn Debris (Req_Serv)',  

   'Totes Abandon Pickup (Req_Serv)', 'Snow-Bank Removal (Req_Serv)',  

   'Illegal Dumping PrivateProperty (Req_Serv)',  

   'Recycling Tote Replace (Req_Serv)', 'Open311 Housing',  

   'Sign Maintenance (Req_Serv)', 'Sign Hazards (Req_Serv)',  

   'Water_Billing_Meter (Req_Serv)', 'Sidewalks (Req_Serv)',  

   'Parking Issues (Req_Serv)',  

   'Missed Pickup 2_Piece Large Trash (Req_Serv)',  

   'Illegal Dumping Street (Req_Serv)', 'QRT Snow Removal (Req_Serv)',  

   'Other Hole in Road (Req_Serv)', 'QRT Other Issue (Req_Serv)',  

   'Great American Clean-Up (Req_Serv)',  

   'Save Our Streets Program (Req_Serv)',  

   'Pavement Marking Lines (Req_Serv)',  

   'OSP Illegal Dumping (Req_Serv)',  

   'Buffalo Traffic Violations (Req_Serv)',  

   'Abandoned Vehicles (Req_Serv)', 'Illegal Dumping Curb (Req_Serv)',  

   'Recycling Tote Pickup (Req_Serv)', 'Bridge Issue (Req_Serv)',  

   'Recycling Tote Abandon Pickup (Req_Serv)',  

   'Tree Trimming Quality Issue (Req_Serv)',  

   'Basement Flooding (Req_Serv)',  

   'Damaged Street Light Pole (Req_Serv)', 'Curbs (Req_Serv)',  

   'Recycling - Escalated Questions (Req_Serv)',  

   'Rain Barrels (Req_Serv)', 'Fallen Tree Inspection (Req_Serv)',  

   'Fair Housing Issue (Req_Serv)', 'Inrem Real Estate (Req_Serv)',  

   'Water Tested (Req_Serv)', 'Tree Removal Quality Issue (Req_Serv)',  

   'PW Neighborhood Traffic Calming (Req_Serv)',  

   'Right of Way Issue (Req_Serv)',  

   'Damage from Street Worker (Req_Serv)', 'City Property (Req_Serv)'])
```

'Curb - Metal Protruding (Req_Serv)',
'Abandoned Vehicle Inspection (Req_Serv)',
'Private Property (Req_Serv)', 'Telecommunications (Req_Serv)',
'Basketball Hoop in RoW (Req_Serv)',
'PW Ongoing Construction (Req_Serv)',
'FOIL Records City Clerk (Req_Serv)', 'Illegal Dumping (Req_Serv)',
'Neighborhood CleanUp (Req_Serv)',
'Signal Timing Issue City (Req_Serv)',
'Assessment Issue (Req_Serv)',
'Other Adjudication Issue (Req_Serv)',
'Vehicle Blocking Snow Plow (Req_Serv)',
'Signal Other Issue (Req_Serv)', 'BFD Snow on Hydrant (Req_Serv)',
'Stump Removal Quality Issue (Req_Serv)', 'PVB Mobile App Ticket',
'Fire Hydrant Issue (Req_Serv)',
'Signal Out or Flashing (Req_Serv)',
'City Park Tree Issue (Req_Serv)',
'Tree Removal per Inspector (Req_Serv)', 'PVB Mobile App Issue',
'Waterfront Issues (Req_Serv)',
'Totes Report of Removed Broken Tote (Req_Serv)',
'Damage from Snow Removal (Req_Serv)',
'Engineering Operational (Req_Serv)',
'Park Garbage Pickup (Req_Serv)', 'Sneakers Hanging (Req_Serv)',
'Parks City (Req_Serv)', 'CBO (Req_Serv)',
'PW Missing Manhole Cover (Req_Serv)',
'FOIL Records Board of Ethics (Req_Serv)',
'Building Maintenance (Req_Serv)', 'Bike Rack (Req_Serv)',
'Obscene PW Engineering (Req_Serv)',
'FOIL Records Assessment & Taxation (Req_Serv)',
'NYS New Americans (Req_Serv)',
'Compensation & Benefits (Req_Serv)', 'Other (Req_Serv)',
'OSP Other Issue (Req_Serv)',
'QRT Illegal Dumping on Viaduct (Req_Serv)',
'Signal Issue DoT (Req_Serv)', 'Obscene City Property (Req_Serv)',
'Snow Removal Bus Stop/Shelter (Req_Serv)',
'FOIL Records EDPIS (Req_Serv)',
'CityHall_CityCourt Maintenance (Req_Serv)',
'Obscene Private Property (Req_Serv)',
'Special Event Totes (Req_Serv)',
'Tree Planting Quality Issue (Req_Serv)', 'PW Traffic (Req_Serv)',
'PW Engineering (Req_Serv)', 'Snow Removal (Req_Serv)',
'FOIL Records AdmFinance_Policy_UrbanAff (Req_Serv)',
'Recreation Center (Req_Serv)',
'FOIL Records Real Estate (Req_Serv)', 'Open311 Graffiti',
'FOIL Records Police Dept (Req_Serv)',
'Obscene PW Traffic (Req_Serv)', 'Small Cell (Req_Serv)',
'FOIL Records Law Dept (Req_Serv)',
'FOIL Records Division of Purchase (Req_Serv)',
'Tree Removal Challenge (Req_Serv)',
'Open311 APP Issues (Req_Serv)', 'Obscene Other (Req_Serv)',
'Youth Bureau Issue (Req_Serv)',
'FOIL Records Adjudication (Req_Serv)',
'FOIL Records Fire Dept (Req_Serv)',
'Thrive Work Development (Req_Serv)',
'FOIL Records Planning Board (Req_Serv)',
'FOIL Records Human Resources (Req_Serv)',
'BFD Fire Prevention (Req_Serv)',
'PVB School-Zone Camera Citations (Req_Serv)',
'Chicken Coop (Req_Serv)',
'FOIL Records Citizens Services (Req_Serv)',
'FOIL Records Traffic Violations (Req_Serv)',

```
'Good Neighbor (Req_Serv)', 'Obscene Parks City (Req_Serv)',  

'2020 Reassessment', 'FairHsg Other (Req_Serv)',  

'Tele-Engagement Citizen Services (Req_Serv)',  

'FOIL Records Compensation and Benefits (Req_Serv)',  

'FOIL Records Parking (Req_Serv)',  

'Tele-Engagement Board of Block Clubs (Req_Serv)',  

'FOIL Records Community Svrs & RecProg (Req_Serv)',  

'FOIL Records Civil Service (Req_Serv)', 'Pest_Private Trap',  

'FOIL Records Strategic Planning (Req_Serv)',  

'Tele-Engagement Community Services (Req_Serv)',  

'Obscene Schools (Req_Serv)',  

'FOIL Records Audit Control (Req_Serv)',  

'Parks Olmsted (Req_Serv)', 'Security Deposit (Req_Serv)',  

'Tele-Engagement Strategic Planning (Req_Serv)',  

'ReOpen Buffalo (Req_Serv)', 'Erie Basin Marina (Req_Serv)',  

'Tele-Engagement BETC (Req_Serv)',  

'Obscene Parks Olmsted (Req_Serv)',  

'FOIL Records Office of the Mayor (Req_Serv)',  

'Tele-Engagement Police (Req_Serv)',  

'FOIL Records Preservation Board (Req_Serv)', 'Test2',  

'Parking Meter Issue (Req_Serv)', 'QRT Vacant Lot (Req_Serv)',  

'Test_before', 'Pavement Markings Other (Req_Serv)',  

'Ordinance Violation (Req_Serv)',  

'Recycling Bin Delivery (Req_Serv)',  

'Parks Usage Survey (Req_Serv)', 'Block Club CleanUp (Req_Serv)',  

'Housing Violations (Prov_Info)', 'Obscene Thruway (Req_Serv)',  

'Weed & Seed Program (Req_Serv)', 'Schools (Req_Serv)',  

'Escalated Question', 'Neighborhood Watch Signs (Req_Serv)',  

'Thruway (Req_Serv)', 'RalphWilson Park Construction (Req_Serv)',  

'Scrap_It CurbSide (Req_Serv)',  

'QRT Illegal Dumping on Vacant Lot (Req_Serv)',  

'BreakOut & LoveBuffalo (Req_Serv)', 'ADA-Other (Req_Serv)',  

'ADA-PW Sidewalks (Req_Serv')], dtype=object)
```

In []: len(working311["Type"].unique())

Out[]: 223

Fairly large amount of categories, however still limited when we look at this number next to the number of rows.

reference:'Police District', 'Neighborhood'

In []: working311["Police District"].unique()

Out[]: array(['UNKNOWN', 'District D', 'District A', 'District E', 'District B',
'District C'], dtype=object)

In []: len(working311["Police District"].unique())

Out[]: 6

In []: working311["Neighborhood"].unique()

```
Out[ ]: array(['UNKNOWN', 'Riverside', 'Hopkins-Tiffet', 'South Park', 'Kenfield',
   'Black Rock', 'Schiller Park', 'Grant-Amherst', 'Elmwood Bryant',
   'Lovejoy', 'Fillmore-Leroy', 'First Ward', 'Elmwood Bidwell',
   'Central Park', 'Kaisertown', 'Fruit Belt', 'Broadway Fillmore',
   'Parkside', 'North Park', 'MLK Park', 'Genesee-Moselle',
   'West Side', 'Ellicott', 'Kensington-Bailey', 'Allentown',
   'West Hertel', 'Delavan Grider', 'Upper West Side',
   'Seneca-Cazenovia', 'Hamlin Park', 'Seneca Babcock',
   'Lower West Side', 'Masten Park', 'Pratt-Willert', 'Central',
   'University Heights'], dtype=object)
```

```
In [ ]: len(working311["Neighborhood"].unique())
```

```
Out[ ]: 36
```

```
In [ ]: len(working311["Object Type"].unique())
```

```
Out[ ]: 5
```

```
In [ ]: working311["Object Type"].unique()
```

```
Out[ ]: array(['Street', 'Property', 'Individual', 'Unknown', 'Organisation'],
   dtype=object)
```

```
In [ ]: #Lets add a separate year and month column
working311["Year"] = pd.DatetimeIndex(working311["Open Date"]).year
working311["Month"] = pd.DatetimeIndex(working311["Open Date"]).month
```

```
In [ ]: working311.shape
```

```
Out[ ]: (1110933, 11)
```

produce some graphs explaining how they are related to council district, police district, neighborhood.

```
In [ ]: working311.head()
```

Out[]:	Open Date	Status	Subject	Reason	Type	Object Type	Council District	Police District	Neighborhood
0	2010-07-19 12:06:00	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
1	2018-02-12 06:47:00	Closed	Buffalo Police Department	Police	Police Issue (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
2	2011-01-11 09:12:00	Closed	Utilities	National Grid	Streetlights (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
3	2017-07-10 11:36:00	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
4	2018-08-18 15:44:00	Closed	Dept of Public Works	Forestry	Tree Planting Request (Req_Serv)	Property	UNKNOWN	UNKNOWN	UNKNOWN



Lets organize and remind myself what I should be comparing again.

```
Index(['Open Date', 'Status', 'Subject', 'Reason', 'Type', 'Object Type', 'Council District', 'Police District', 'Neighborhood'], dtype='object')
```

```
'Status' 2, 'Subject' 18, 'Reason' 51, 'Type' 223, 'Object Type' 5'
```

```
'Council District' 10, 'Police District' 6, 'Neighborhood' 36
```

Lets Just make sure they are

```
In [ ]: Compare_1 = working311.groupby("Police District").Type.value_counts(normalize=False)
```

```
In [ ]: Compare_1 = pd.DataFrame(Compare_1)
Compare_1
```

Out[]:

Police District	Type	Type
District A	Housing Violations (Req_Serv)	26054
	Totes Replace (Req_Serv)	21202
	Garbage Missed Pick Up (Req_Serv)	8530
	Pot Hole (Req_Serv)	8124
	Recycling Missed Pick Up (Req_Serv)	6054
...
UNKNOWN	Tele-Engagement Citizen Services (Req_Serv)	1
	Tele-Engagement Community Services (Req_Serv)	1
	Tele-Engagement Police (Req_Serv)	1
	Test2	1
	Weed & Seed Program (Req_Serv)	1

1085 rows × 1 columns

In []: Compare_1.loc['District A', :].sum()

Out[]: Type 176476
dtype: int64

This would be hard to describe with a limited number of categories. It has a length of 1085 with 223 types and 6 police districts. Dealing with each category visually would be messy.

In []: Compare_2 = working311.groupby("Police District").Reason.value_counts(normalize=False)

In []: Compare_2 = pd.DataFrame(Compare_2)
Compare_2

Out[]:

Police District	Reason	
District A	Sanitation	55622
	Housing	29475
	Engineering - Street Repairs	17437
	Streets	15248
	Forestry	9511
...
UNKNOWN	Knowledge	4
	Citizens Services - BreakOut & Love Buffalo	2
	Citizen Services - Weed & Seed	1
	Fair Housing	1
	Test	1

266 rows × 1 columns

I may try to graph this and see what transpires. But I may just do away with it because it doesn't seem as though the outcomes fit easily into a limited number of category for Reasons.

```
In [ ]: Compare_3 = working311.groupby("Police District").Status.value_counts(normalize=False)
```

```
In [ ]: Compare_3=pd.DataFrame(Compare_3)
```

```
Compare_3
```

Out[]:

		Status	
Police District		Status	
District A	Closed	175801	
	Open	675	
District B	Closed	149414	
	Open	698	
District C	Closed	198554	
	Open	786	
District D	Closed	259651	
	Open	1047	
District E	Closed	230324	
	Open	823	
UNKNOWN	Closed	91945	
	Open	1215	

In []: Compare_4 = working311.groupby("Police District")['Object Type'].value_counts(normalize=True)

In []: Compare_4=pd.DataFrame(Compare_4)
Compare_4

Out[]:

		Object Type
Police District		Object Type
District A	Property	168230
	Street	8246
District B	Property	150111
	Street	1
District C	Property	199338
	Street	2
District D	Property	260698
District E	Property	231147
UNKNOWN	Street	63145
	Property	17933
	Unknown	6838
	Individual	5092
	Organisation	152

```
In [ ]: Compare_5 = working311.groupby("Police District").Type.value_counts(normalize=False)
```

```
In [ ]: Compare_5 =pd.DataFrame(Compare_5)
Compare_5
```

Out[]:

Police District	Type	
District A	Housing Violations (Req_Serv)	26054
	Totes Replace (Req_Serv)	21202
	Garbage Missed Pick Up (Req_Serv)	8530
	Pot Hole (Req_Serv)	8124
	Recycling Missed Pick Up (Req_Serv)	6054
...		...
UNKNOWN	Tele-Engagement Citizen Services (Req_Serv)	1
	Tele-Engagement Community Services (Req_Serv)	1
	Tele-Engagement Police (Req_Serv)	1
	Test2	1
	Weed & Seed Program (Req_Serv)	1

1085 rows × 1 columns

```
In [ ]: Compare_A2 = working311.groupby("Neighborhood").Reason.value_counts(normalize=False)
```

```
In [ ]: Compare_A2 =pd.DataFrame(Compare_A2)
Compare_A2
```

Out[]:

Neighborhood	Reason	
Allentown	Sanitation	5706
	Housing	2901
	Engineering - Street Repairs	1853
	Parking Violations Bureau	1550
	Streets	1420

West Side	Community Based Orgs	3
	Citizen Services - Good Neighbor	2
	Telecommunications	2
	Fair Housing	1
	OSP	1

1362 rows × 1 columns

In []: Compare_A1 = working311.groupby("Neighborhood").Type.value_counts(normalize=False)

In []: Compare_A1 = pd.DataFrame(Compare_A1)
Compare_A1

Out[]:

Neighborhood	Type	
Allentown	Housing Violations (Req_Serv)	1905
	Totes Replace (Req_Serv)	1674
	Parking Issues (Req_Serv)	1348
	Pot Hole (Req_Serv)	1039
	Garbage Missed Pick Up (Req_Serv)	934

West Side	PVB School-Zone Camera Citations (Req_Serv)	1
	Pavement Markings Other (Req_Serv)	1
	Schools (Req_Serv)	1
	Special Event Totes (Req_Serv)	1
	Tree Removal Challenge (Req_Serv)	1

5450 rows × 1 columns

```
In [ ]: Compare_A5 = working311.groupby("Neighborhood").Type.value_counts(normalize=False)
```

```
In [ ]: Compare_A5 = pd.DataFrame(Compare_A5)
Compare_A5
```

Out[]:

Neighborhood	Type	
Allentown	Housing Violations (Req_Serv)	1905
	Totes Replace (Req_Serv)	1674
	Parking Issues (Req_Serv)	1348
	Pot Hole (Req_Serv)	1039
	Garbage Missed Pick Up (Req_Serv)	934
...		...
West Side	PVB School-Zone Camera Citations (Req_Serv)	1
	Pavement Markings Other (Req_Serv)	1
	Schools (Req_Serv)	1
	Special Event Totes (Req_Serv)	1
	Tree Removal Challenge (Req_Serv)	1

5450 rows × 1 columns

```
In [ ]: Compare_A4 = working311.groupby("Neighborhood")['Object Type'].value_counts(normalize=False)
```

```
In [ ]: Compare_A4 = pd.DataFrame(Compare_A4)
Compare_A4
```

Out[]:

Object Type

Neighborhood	Object Type	
Allentown	Property	17515
Black Rock	Property	15937
Broadway Fillmore	Property	63542
	Street	1
Central	Property	16922
Central Park	Property	18995
Delavan Grider	Property	25309
Ellicott	Property	9125
Elmwood Bidwell	Property	47801
Elmwood Bryant	Property	30056
Fillmore-Leroy	Property	26772
First Ward	Property	10696
Fruit Belt	Property	13365
Genesee-Moselle	Property	45475
Grant-Amherst	Property	17429
Hamlin Park	Property	22920
Hopkins-Tifft	Property	29318
	Street	8245
Kaisertown	Property	24106
Kenfield	Property	26241
Kensington-Bailey	Property	52308
Lovejoy	Property	39716
	Street	1
Lower West Side	Property	15962
	Street	1
MLK Park	Property	20450
Masten Park	Property	34905
North Park	Property	54754
Parkside	Property	25742
Pratt-Willert	Property	11889
Riverside	Property	37894
Schiller Park	Property	39938
Seneca Babcock	Property	15118

Object Type

Neighborhood	Object Type	
Seneca-Cazenovia	Property	30175
South Park	Property	48729
	Street	1
UNKNOWN	Street	63145
	Property	16727
	Unknown	6838
	Individual	5092
	Organisation	152
University Heights	Property	38077
Upper West Side	Property	36746
West Hertel	Property	12497
West Side	Property	34306

```
In [ ]: Compare_A3 = working311.groupby("Neighborhood").Status.value_counts(normalize=False)
```

```
In [ ]: Compare_A3 = pd.DataFrame(Compare_A3)
Compare_A3
```

Out[]:

		Status
Neighborhood		
Allentown	Closed	17426
	Open	89
Black Rock	Closed	15864
	Open	73
Broadway Fillmore	Closed	63309
...
Upper West Side	Open	145
West Hertel	Closed	12451
	Open	46
West Side	Closed	34160
	Open	146

72 rows × 1 columns

```
In [ ]: Compare_B2 = working311.groupby("Council District").Reason.value_counts(normalize=False)
```

```
In [ ]: Compare_B2 = pd.DataFrame(Compare_B2)
Compare_B2
```

Out[]:

Council District	Reason	
DELAWARE	Sanitation	5346
	Streets	2517
	Housing	2202
	Engineering - Street Repairs	1967
	Parking Violations Bureau	988
...		...
UNKNOWN	Knowledge	5
	ADA	2
	Citizens Services - BreakOut & Love Buffalo	2
	Immigration	2
	Test	1

352 rows × 1 columns

```
In [ ]: Compare_B1 = working311.groupby("Council District").Type.value_counts(normalize=False)
```

```
In [ ]: Compare_B1 = pd.DataFrame(Compare_B1)
```

```
In [ ]: Compare_B1
```

Out[]:

Council District	Type	
DELAWARE	Totes Replace (Req_Serv)	1420
	Garbage Missed Pick Up (Req_Serv)	1264
	Housing Violations (Req_Serv)	1228
	Street Snow Plowing (Req_Serv)	1124
	Pot Hole (Req_Serv)	1108
...
UNKNOWN	Tele-Engagement Citizen Services (Req_Serv)	1
	Tele-Engagement Community Services (Req_Serv)	1
	Tele-Engagement Police (Req_Serv)	1
	Test2	1
	Thruway (Req_Serv)	1

1442 rows × 1 columns

In []: Compare_B5 = working311.groupby("Council District").Type.value_counts(normalize=False)

In []: Compare_B5 = pd.DataFrame(Compare_B5)

In []: Compare_B5

Council District	Type	
DELAWARE	Totes Replace (Req_Serv)	1420
	Garbage Missed Pick Up (Req_Serv)	1264
	Housing Violations (Req_Serv)	1228
	Street Snow Plowing (Req_Serv)	1124
	Pot Hole (Req_Serv)	1108
...
UNKNOWN	Tele-Engagement Citizen Services (Req_Serv)	1
	Tele-Engagement Community Services (Req_Serv)	1
	Tele-Engagement Police (Req_Serv)	1
	Test2	1
	Thruway (Req_Serv)	1

1442 rows × 1 columns

```
In [ ]: Compare_B4 = working311.groupby("Council District")['Object Type'].value_counts(normalize=False)
```

```
In [ ]: Compare_B4 = pd.DataFrame(Compare_B4)
```

```
In [ ]: Compare_B4
```

```
Out[ ]:          Object Type
```

Council District	Object Type	
DELAWARE	Property	17946
ELLICOTT	Property	21165
FILLMORE	Property	24398
LOVEJOY	Property	25279
	Street	7343
MASTEN	Property	23363
NIAGARA	Property	20330
NORTH	Property	16527
SOUTH	Property	17246
UNIVERSITY	Property	18240
UNKNOWN	Property	842963
	Street	64051
	Unknown	6838
	Individual	5092
	Organisation	152

```
In [ ]: Compare_B3 = working311.groupby("Council District").Status.value_counts(normalize=False)
```

```
In [ ]: Compare_B3 = pd.DataFrame(Compare_B3)
```

```
In [ ]: Compare_B3
```

Out[]:

Status

Council District	Status	
DELAWARE	Closed	17697
	Open	249
ELLICOTT	Closed	20748
	Open	417
FILLMORE	Closed	23897
	Open	501
LOVEJOY	Closed	32219
	Open	403
MASTEN	Closed	23060
	Open	303
NIAGARA	Closed	20019
	Open	311
NORTH	Closed	16172
	Open	355
SOUTH	Closed	17012
	Open	234
UNIVERSITY	Closed	17953
	Open	287
UNKNOWN	Closed	916912
	Open	2184

Graph all these relationships except the ones containing Types

In []: `Compare_B3.drop(labels="UNKNOWN", axis=0)`

Out[]:

Status

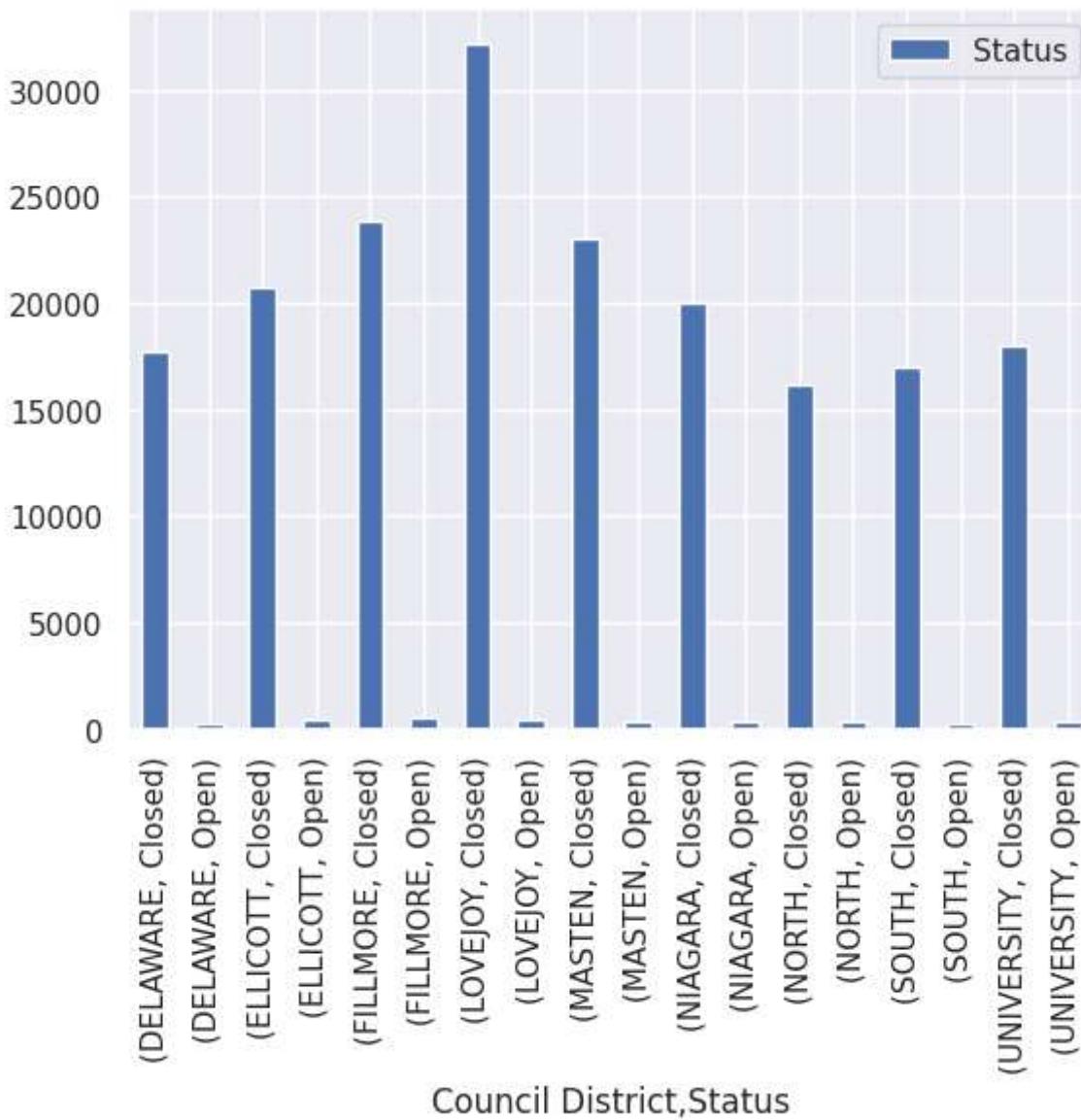
Council District	Status	
DELAWARE	Closed	17697
	Open	249
ELLICOTT	Closed	20748
	Open	417
FILLMORE	Closed	23897
	Open	501
LOVEJOY	Closed	32219
	Open	403
MASTEN	Closed	23060
	Open	303
NIAGARA	Closed	20019
	Open	311
NORTH	Closed	16172
	Open	355
SOUTH	Closed	17012
	Open	234
UNIVERSITY	Closed	17953
	Open	287

Graph the council district, police district, neighborhood vs status

In []: Compare_B3 = Compare_B3.drop(labels="UNKNOWN", axis=0)

In []: Compare_B3.plot(kind = 'bar')

Out[]: <Axes: xlabel='Council District,Status'>



I think that's pretty, there seems to be very few cases that are open most have been closed. Among the closed cases the distribution is roughly uniform. Suggesting the 311 services are optimized to treat all neighbourhoods evenly. I dropped the "Unknown" neighbourhood so we can see the impact of neighbourhood and whether or not cases are open or closed. Instead of having 916,000 unknown neighbourhood where we can't observe anything as they aren't identified.

In []: Compare_3

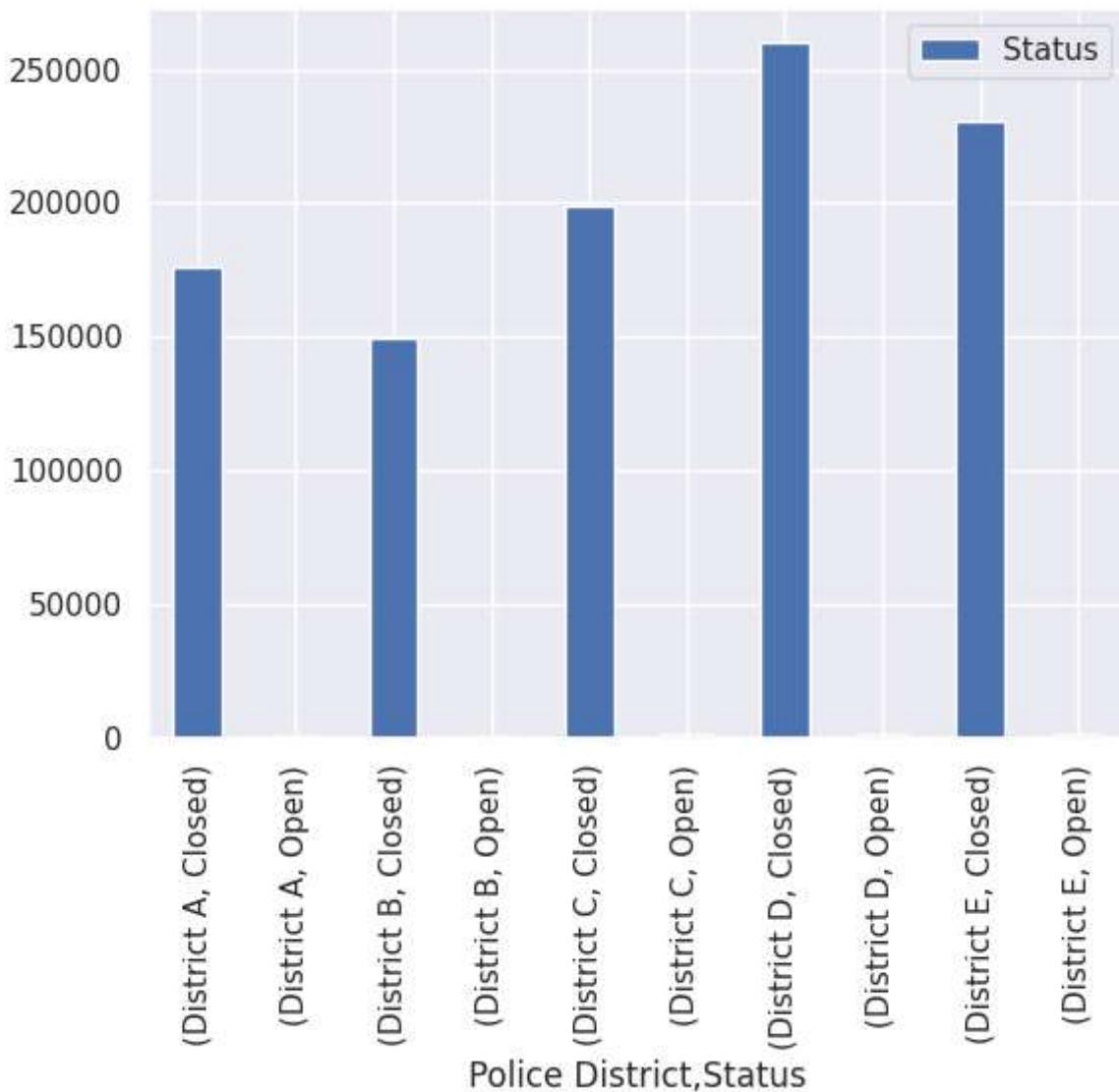
Out[]:

Police District	Status
District A	Closed 175801
	Open 675
District B	Closed 149414
	Open 698
District C	Closed 198554
	Open 786
District D	Closed 259651
	Open 1047
District E	Closed 230324
	Open 823
UNKNOWN	Closed 91945
	Open 1215

In []: Compare_3= Compare_3.drop(labels="UNKNOWN", axis=0)

In []: Compare_3.plot(kind = 'bar')

Out[]: <Axes: xlabel='Police District,Status'>



District D seems to have more cases than others but once again police district and council district both seem to have roughly an evenly distributed amounts of cases and closed cases.

```
In [ ]: # Lets do the same for object Type and police district  
Compare_4= Compare_4.drop(labels="UNKNOWN", axis=0)
```

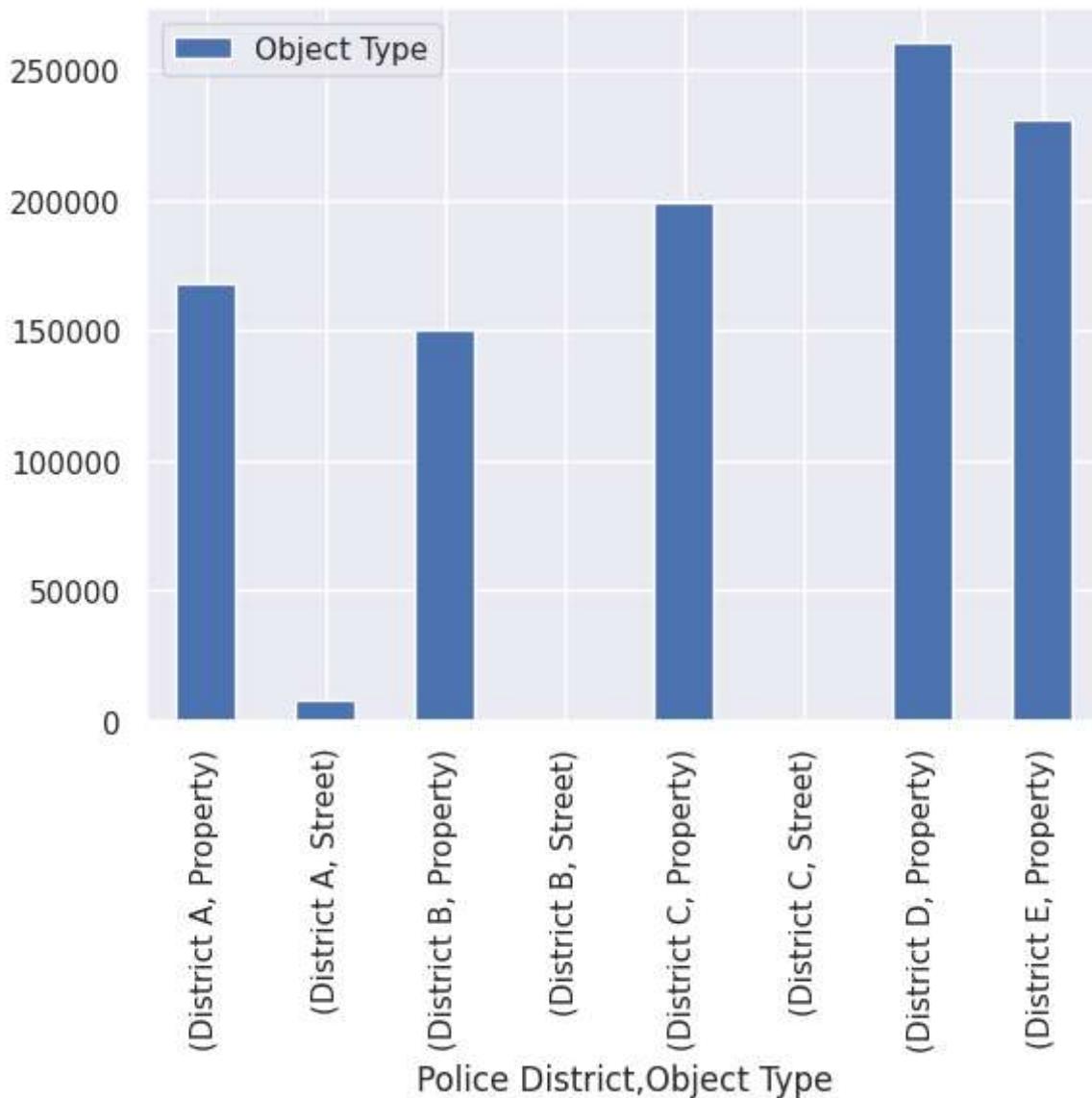
```
In [ ]: Compare_4
```

Out[]:

		Object Type
Police District	Object Type	
District A	Property	168230
	Street	8246
District B	Property	150111
	Street	1
District C	Property	199338
	Street	2
District D	Property	260698
District E	Property	231147

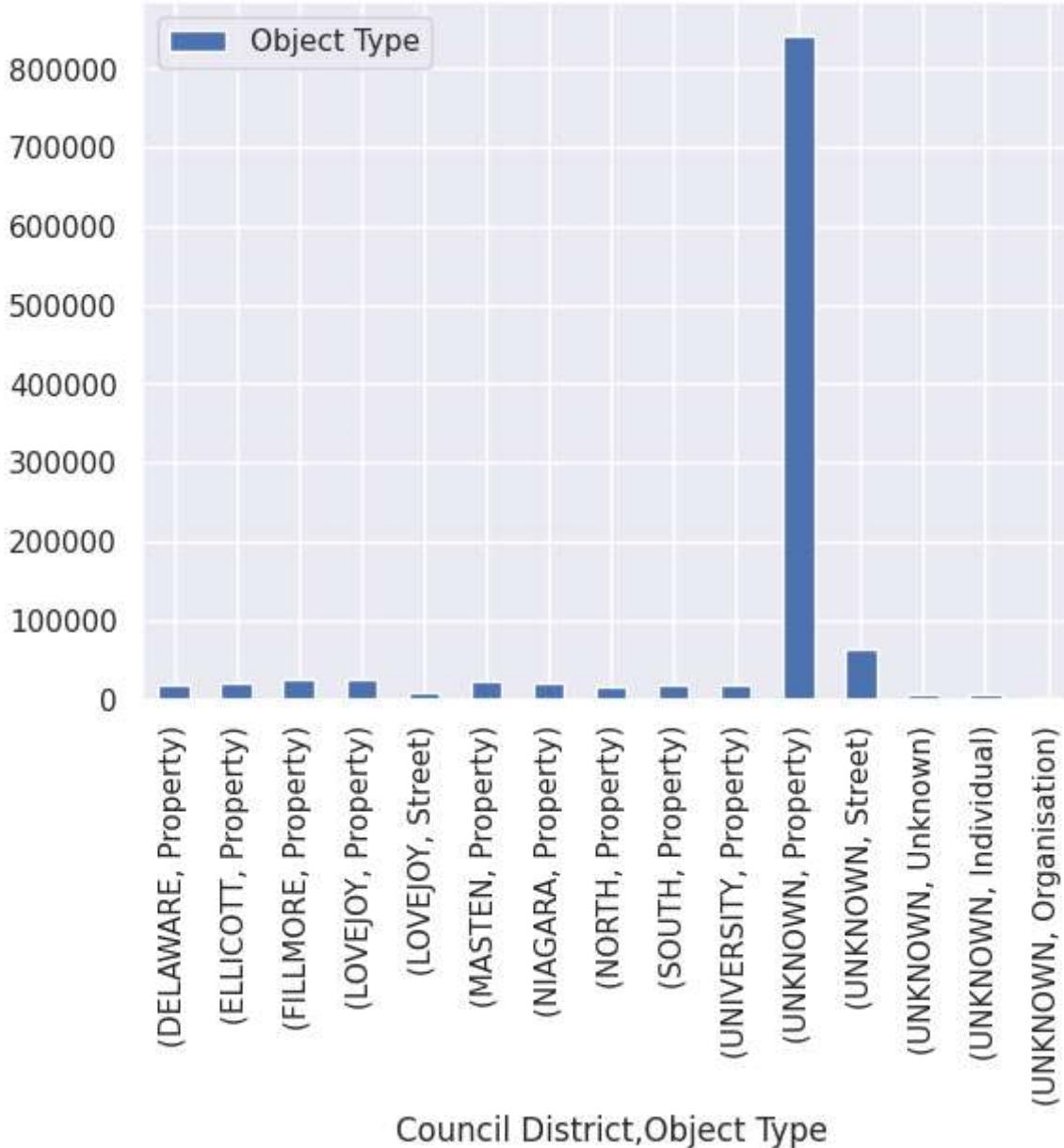
In []: Compare_4.plot(kind = 'bar')

Out[]: <Axes: xlabel='Police District, Object Type'>



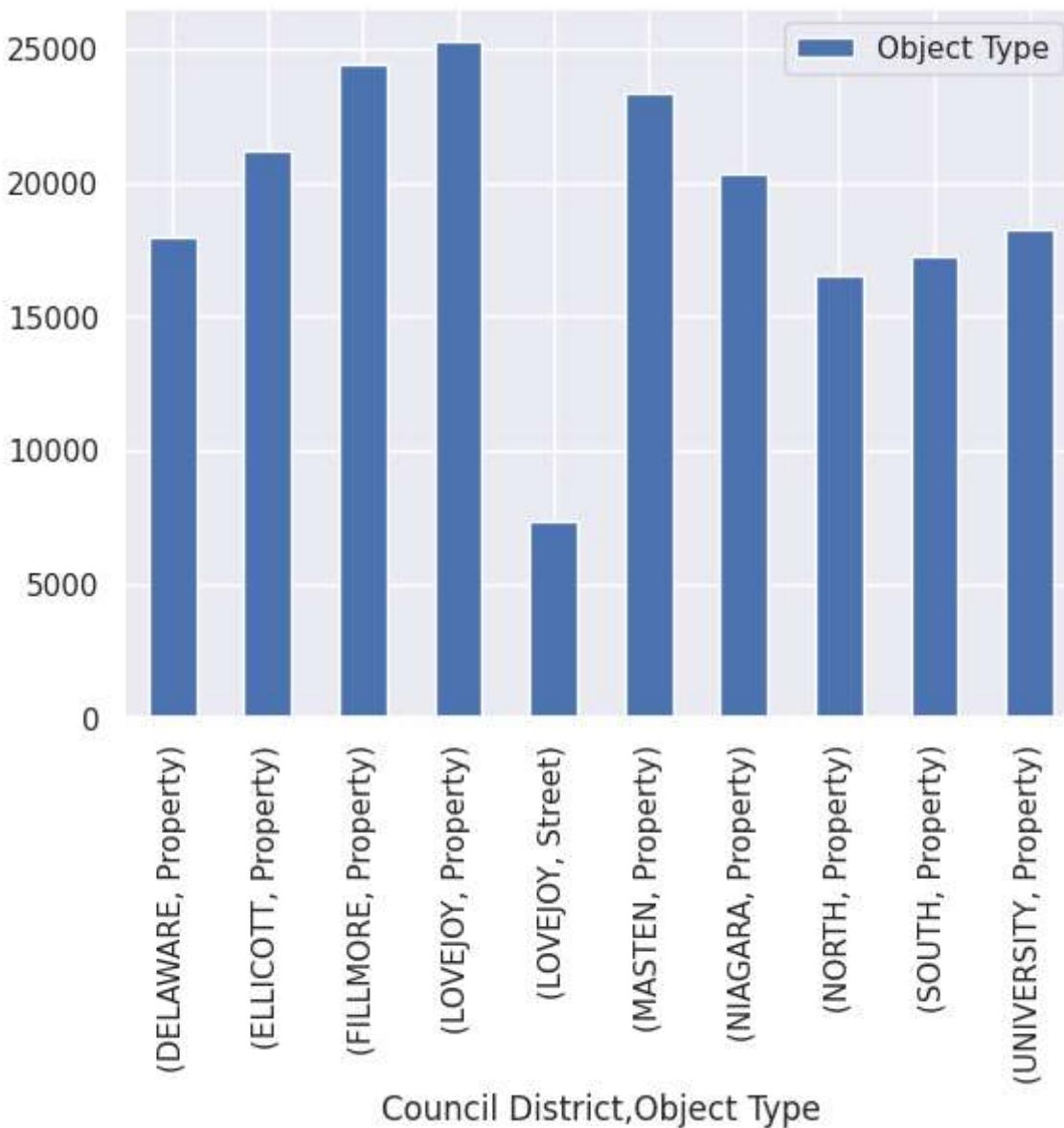
property D and E seems to have higher object type of Property, while only district A has any significant object

```
In [ ]: Compare_B4.plot(kind = 'bar')
Out[ ]: <Axes: xlabel='Council District, Object Type'>
```



Wanted to show why I am getting rid of it.

```
In [ ]: Compare_B4 = Compare_B4.drop(labels="UNKNOWN", axis=0)
In [ ]: Compare_B4.plot(kind = 'bar')
Out[ ]: <Axes: xlabel='Council District, Object Type'>
```

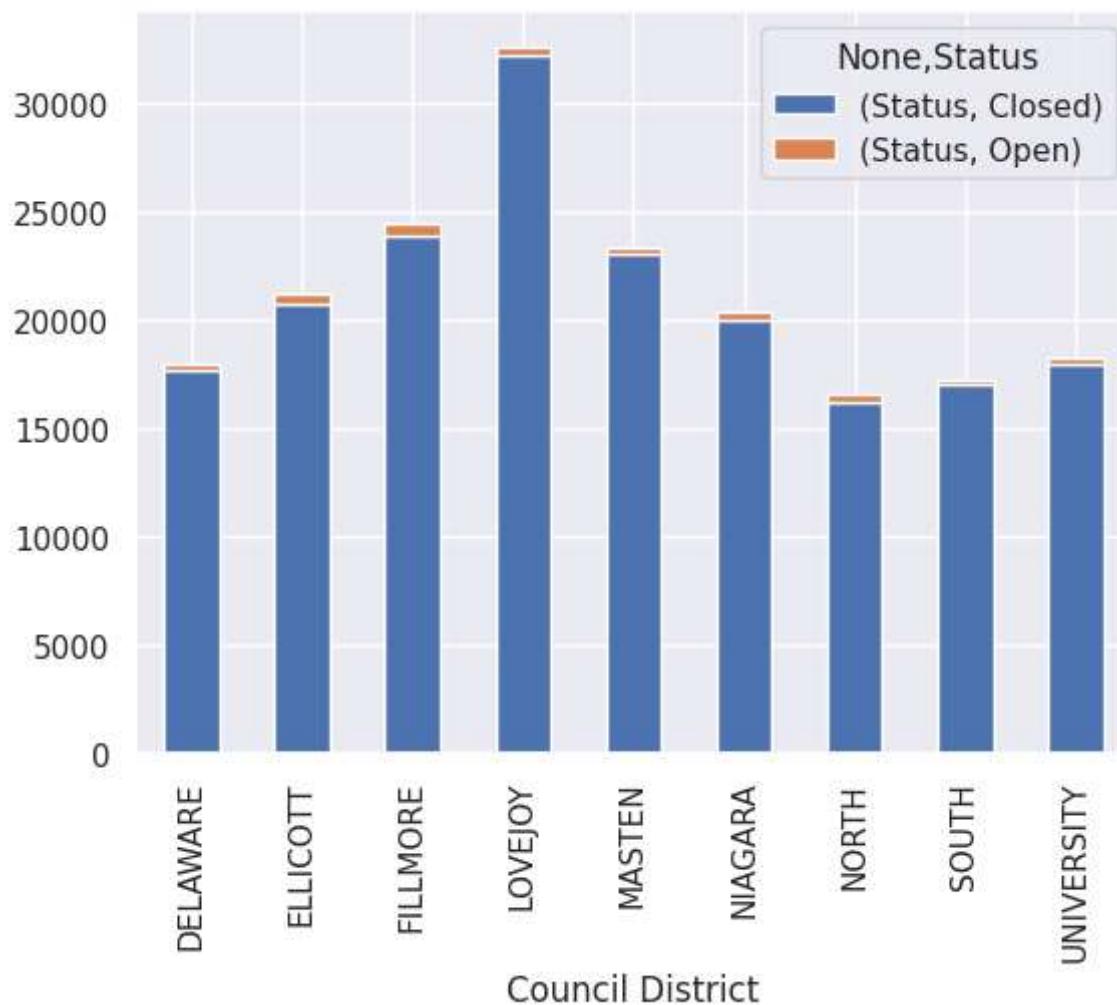


Overall There seems to be a uniform distribution among the property object type of all the council district while only lovejoy has the street object type.

Do different districts and/or neighborhoods tend to different numbers of 311 calls?

We can simply stack all the neighbor counts for each neighborhood to figure this out.

```
In [ ]: Compare_B3.unstack().plot(kind = 'bar', stacked= True)
Out[ ]: <Axes: xlabel='Council District'>
```



In []: Compare_B3

Out[]:

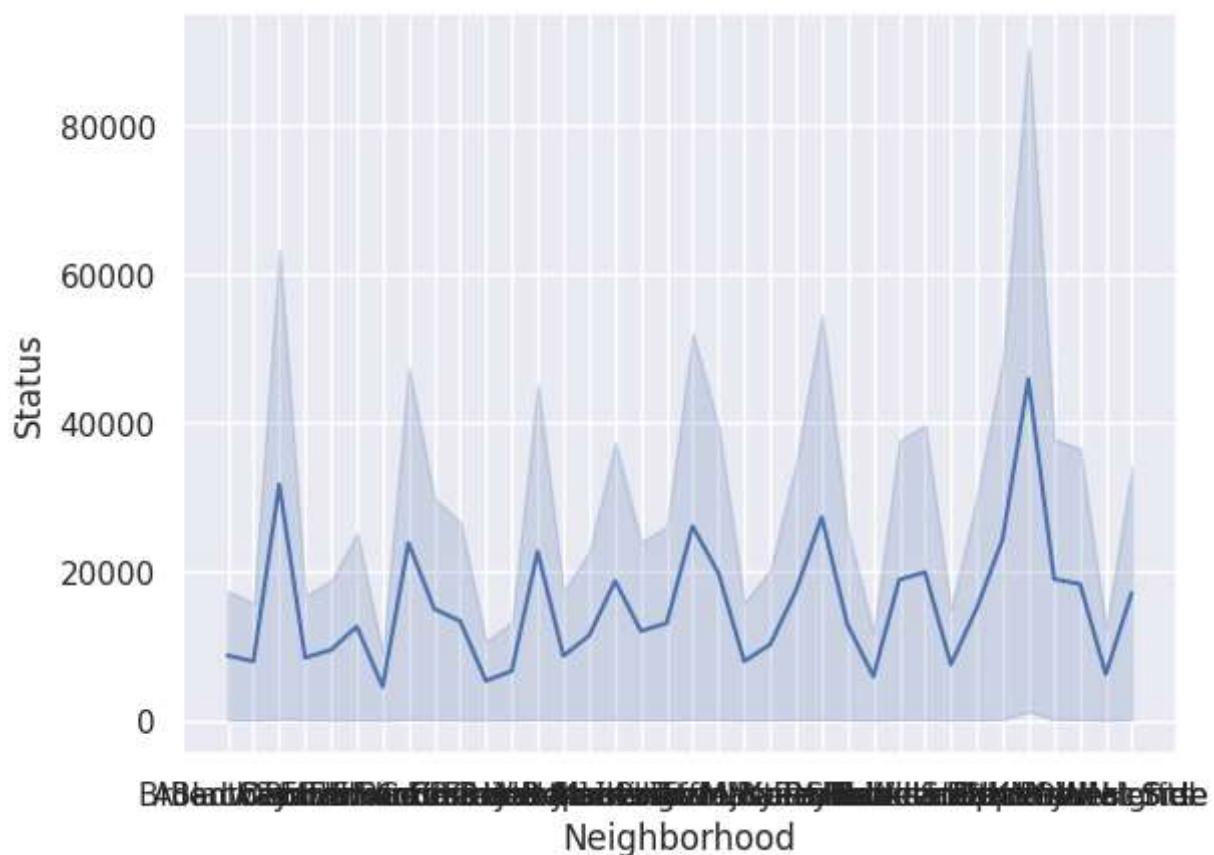
Status

Council District	Status	
DELAWARE	Closed	17697
	Open	249
ELLICOTT	Closed	20748
	Open	417
FILLMORE	Closed	23897
	Open	501
LOVEJOY	Closed	32219
	Open	403
MASTEN	Closed	23060
	Open	303
NIAGARA	Closed	20019
	Open	311
NORTH	Closed	16172
	Open	355
SOUTH	Closed	17012
	Open	234
UNIVERSITY	Closed	17953
	Open	287

In []: `Compare_B3.columns.values.tolist()`Out[]: `['Status']`

As seen here different neighborhood does indeed have different number of calls. In the above cell we can also see where or not the case is closed or open.

```
In [ ]: sns.set_theme(style="darkgrid")
ax = sns.lineplot(x="Neighborhood", y="Status", data=Compare_A3)
```



This may be a weird way of looking at it with a lineplot however as you can see there are diffrent peaks of varios diffrent levels, so difffrent neighbourhoods do tend to have diffrent amounts of 311 calls

I tried using other forms of plotting but this seems to be the best.

In []: Compare_A3

Out[]:

Status		
Neighborhood	Status	
Allentown	Closed	17426
	Open	89
Black Rock	Closed	15864
	Open	73
Broadway Fillmore	Closed	63309

Upper West Side	Open	145
	Closed	12451
West Hertel	Open	46
	Closed	34160
West Side	Open	146

72 rows × 1 columns

Do different months have different number of 311 reports? What about different years?

In []: working311.head()

	Open Date	Status	Subject	Reason	Type	Object Type	Council District	Police District	Neighborhood
0	2010-07-19 12:06:00	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
1	2018-02-12 06:47:00	Closed	Buffalo Police Department	Police	Police Issue (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
2	2011-01-11 09:12:00	Closed	Utilities	National Grid	Streetlights (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
3	2017-07-10 11:36:00	Closed	Dept of Public Works	Streets	Sweeper (Req_Serv)	Street	UNKNOWN	UNKNOWN	UNKNOWN
4	2018-08-18 15:44:00	Closed	Dept of Public Works	Forestry	Tree Planting Request (Req_Serv)	Property	UNKNOWN	UNKNOWN	UNKNOWN

```
In [ ]: yearGroup = working311.groupby('Year').Neighborhood.value_counts()
```

```
In [ ]: yearGroup = pd.DataFrame(yearGroup)
```

```
In [ ]: yearGroup
```

Out[]:

Year	Neighborhood	
2008	UNKNOWN	3463
	Broadway Fillmore	913
	Genesee-Moselle	718
	West Side	660
	Elmwood Bidwell	646

2024	Pratt-Willert	134
	First Ward	121
	West Hertel	95
	Ellicott	91
	UNKNOWN	1

612 rows × 1 columns

```
In [ ]: yearGroup.index.get_level_values(0).unique()
```

```
Out[ ]: Int64Index([2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018,
                   2019, 2020, 2021, 2022, 2023, 2024],
                   dtype='int64', name='Year')
```

```
In [ ]: total_value_sum = yearGroup.groupby(level='Year')['Neighborhood'].sum()
```

```
In [ ]: total_value_sum = pd.DataFrame(total_value_sum)
```

```
In [ ]: total_value_sum
```

Out[]:

Neighborhood

Year	Neighborhood
2008	17174
2009	60760
2010	54283
2011	62294
2012	57418
2013	60409
2014	73020
2015	83276
2016	71291
2017	71520
2018	78471
2019	76820
2020	73870
2021	77732
2022	99531
2023	82924
2024	10140

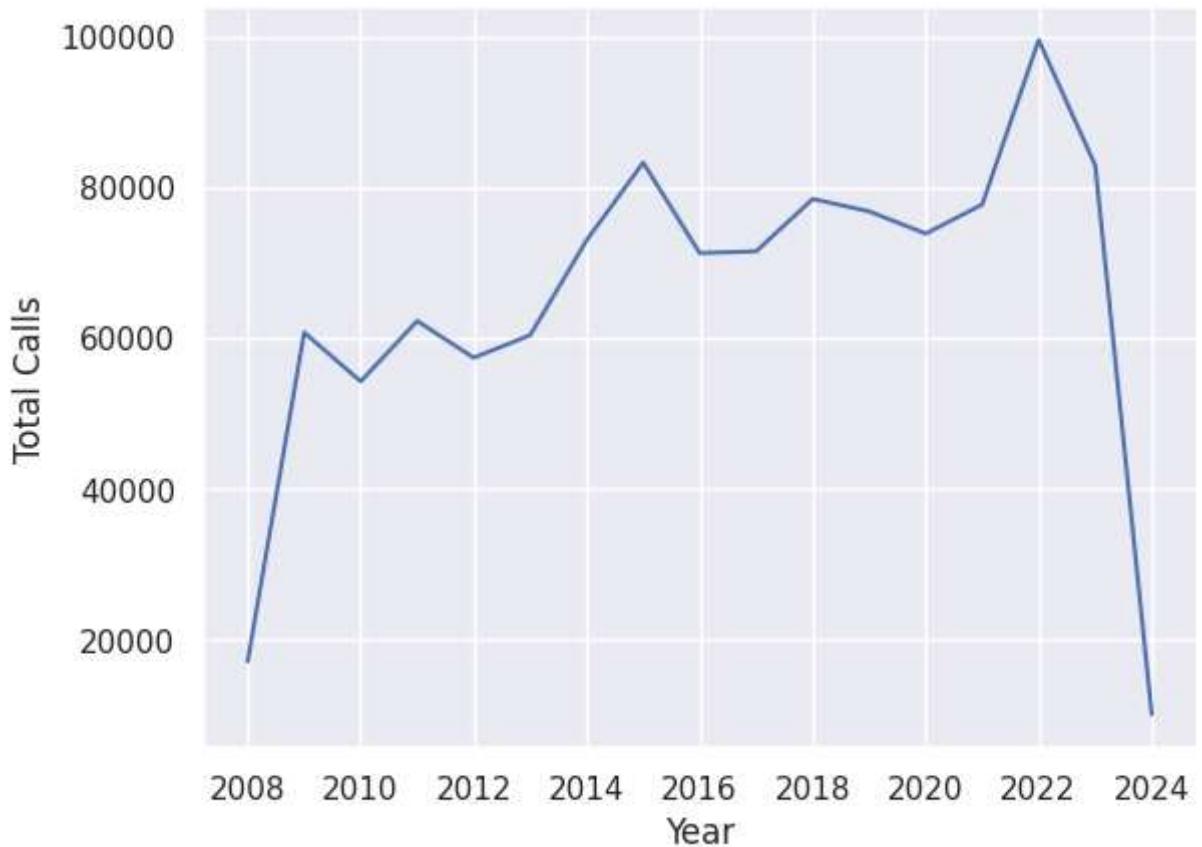
In []:

```
sns.set_theme(style="darkgrid")

ax = sns.lineplot(x="Year", y="Neighborhood", data=total_value_sum)
ax.set(xlabel='Year', ylabel='Total Calls')
```

Out[]:

```
[Text(0.5, 0, 'Year'), Text(0, 0.5, 'Total Calls')]
```



The number decreases significantly at the end, that just means it's not a complete year from tableau experience.

Lets do the same for the months

```
In [ ]: MonthGroup = working311.groupby('Month').Neighborhood.value_counts()
```

```
In [ ]: MonthGroup = pd.DataFrame(yearGroup)
```

```
In [ ]: total_value_sum2 = MonthGroup.groupby(level='Month')[ 'Neighborhood'].sum()
```

```
In [ ]: total_value_sum2 = pd.DataFrame(total_value_sum2)
```

```
In [ ]: total_value_sum2
```

Out[]:

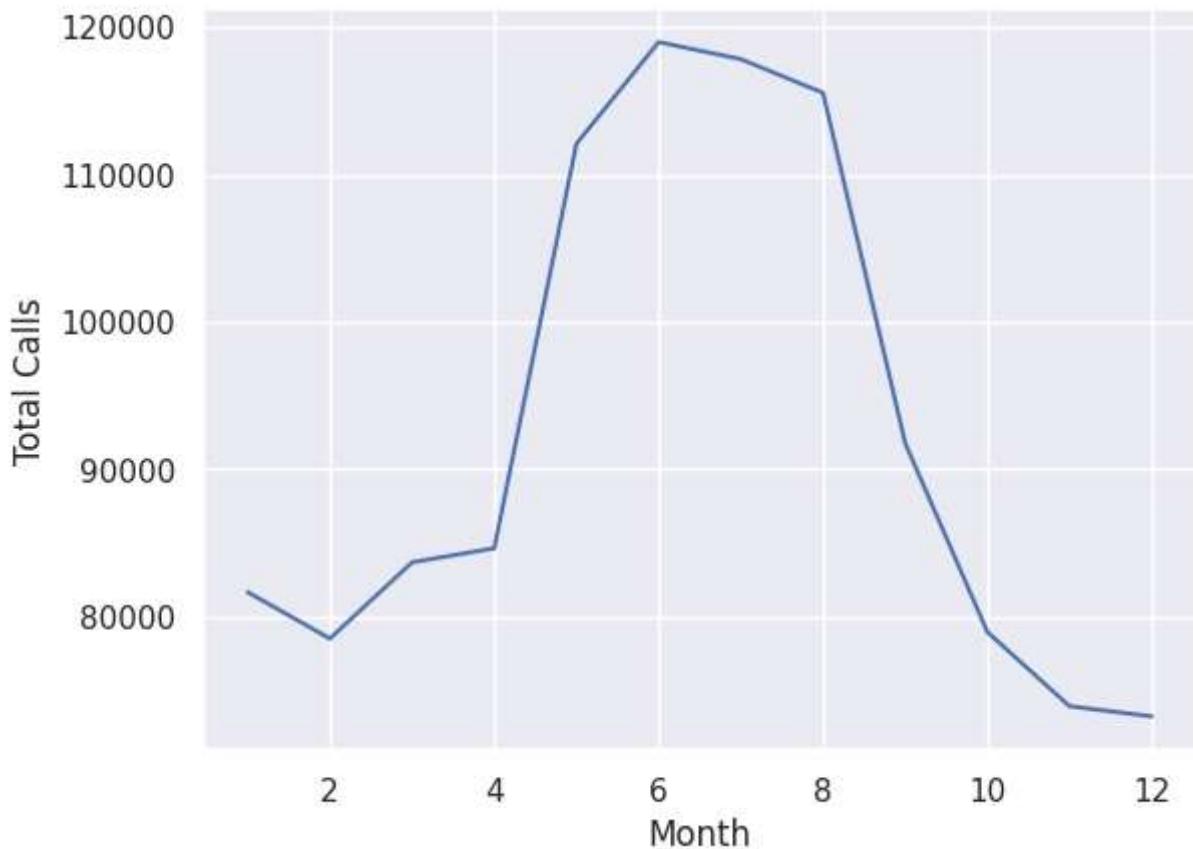
Neighborhood

Month	
1	81670
2	78519
3	83694
4	84645
5	112100
6	118997
7	117839
8	115553
9	91784
10	78965
11	73927
12	73240

In []: sns.set_theme(style="darkgrid")

ax = sns.lineplot(x='Month', y="Neighborhood", data=total_value_sum2)
ax.set(xlabel='Month', ylabel='Total Calls')

Out[]: [Text(0.5, 0, 'Month'), Text(0, 0.5, 'Total Calls')]



Looks like a bell shaped curve almost identical to a normal distribution shares certain properties of it , such as most of the calls are clustured in the middle. However it's not really the mean so it's inappopriate to say it's normal distribution but I thought that it's very interesting that this graph shares the same properties.

In []: