IMPORTING THE DATA

```
import pandas as pd
infile = "/content/311_Service_Requests_20240217.csv"
df = pd.read_csv(infile)
```

```
    ns (0) have mixed types. Specify dtype option on import or set low_memory=False.
```

```
df.columns
```

```
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')
```

```
df.shape
```

```
(118105, 34)
```

```
df.head()
```

	Case Reference	Open Date	Closed Date	Status	Subject	Reason	Type	Object Type
0	1001988123	02/13/2024 10:38:00 AM	02/14/2024 02:49:00 PM	Closed	Dept of Public Works	Engineering - Street Repairs	Pot Hole (Req_Serv)	Property
1	1001988270	02/13/2024 03:19:00 PM	02/14/2024 01:05:00 PM	Closed	Dept of Public Works	Rodent_Pest Control	Rodents (Req_Serv)	Property
2	1001988178	02/13/2024 12:15:00 PM	02/14/2024 01:04:00 PM	Closed	Dept of Public Works	Sanitation	Totes Replace (Req_Serv)	Property
3	1001988102	02/13/2024 10:06:00 AM	02/13/2024 12:54:00 PM	Closed	Dept of Parking	Parking Violations Bureau	Parking Issues (Req_Serv)	Property
4	1001988088	02/13/2024 09:54:00 AM	02/13/2024 02:17:00 PM	Closed	Dept of Public Works	Sanitation	Trash Ordinance Violation (Req_Serv)	Property

5 rows × 34 columns

LOOKING FOR WHICH VARIABLES HAVE LIMITED OUTCOMES

```
variables = ['Status', 'Subject', 'Reason', 'Type', 'Object Type']


for variable in variables:
    print(f"Unique values for '{variable}':")
    print(df[variable].value_counts())
    print()
```

```
Unique values for 'Status':
Closed    116389
Open        1716
Name: Status, dtype: int64

Unique values for 'Subject':
Dept of Public Works                  78048
DPIS                                  18629
Dept of Parking                        7141
Utilities                              6549
Buffalo Police Department              5262
Office of the Mayor                     741
Buffalo Municipal Housing Authority     467
Dept of Law                             425
Office of Strategic Planning            401
```

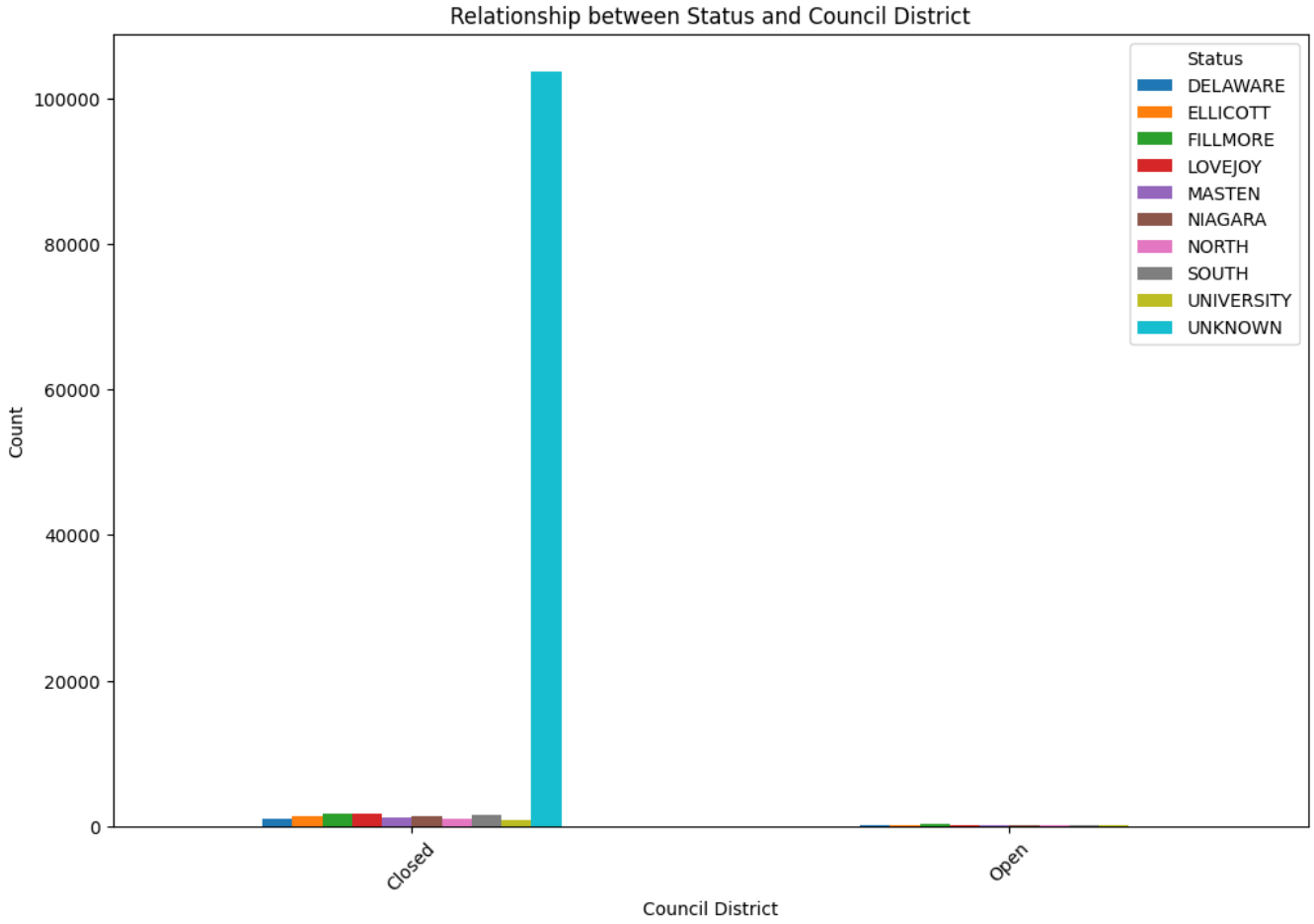
```
Assessment & Taxation                  200
Buffalo Fire Department                155
City Clerk                              70
Community Services & Rec. Program.      10
Human Resources                          4
New Americans                            2
Management Information Systems           1
Name: Subject, dtype: int64

Unique values for 'Reason':
Sanitation                             37998
Housing                                18629
Engineering - Street Repairs           12543
Streets                                 9662
Parking Violations Bureau                6986
Rodent Control                          5586
Police                                  5262
Forestry                                4999
Engineering - Traffic                   3655
National Grid                           3492
Animal Shelter                          2323
Buffalo Sewer Authority                 1872
Buffalo Water Authority                 1185
City Parks                               530
BMHA                                     467
Citizen Services - Quick Response Teams  311
Streets/Sanitation                       291
Rodent_Pest Control                      286
Real Estate                              260
Freedom of Information                   235
Adjudication - Ordinance Violation       190
Taxation                                 168
Citizen Services - Graffiti              162
Moving Violations                        155
BFD                                      155
Citizen Services - Save Our Streets      148
Personnel                                122
Citizens Services - Clean City           119
Administration                           114
City Clerk Issue                          43
Buildings Division                        34
Assessment & Taxation                     30
Licenses                                  27
```

Based on this, it looks like Object Type and Status can be defined by a limited amount of categories. The rest of the variables that were looked at have larger amounts of categories, with Type having the most as it does not even list them all as it has 180.

PRODUCING GRAPHS

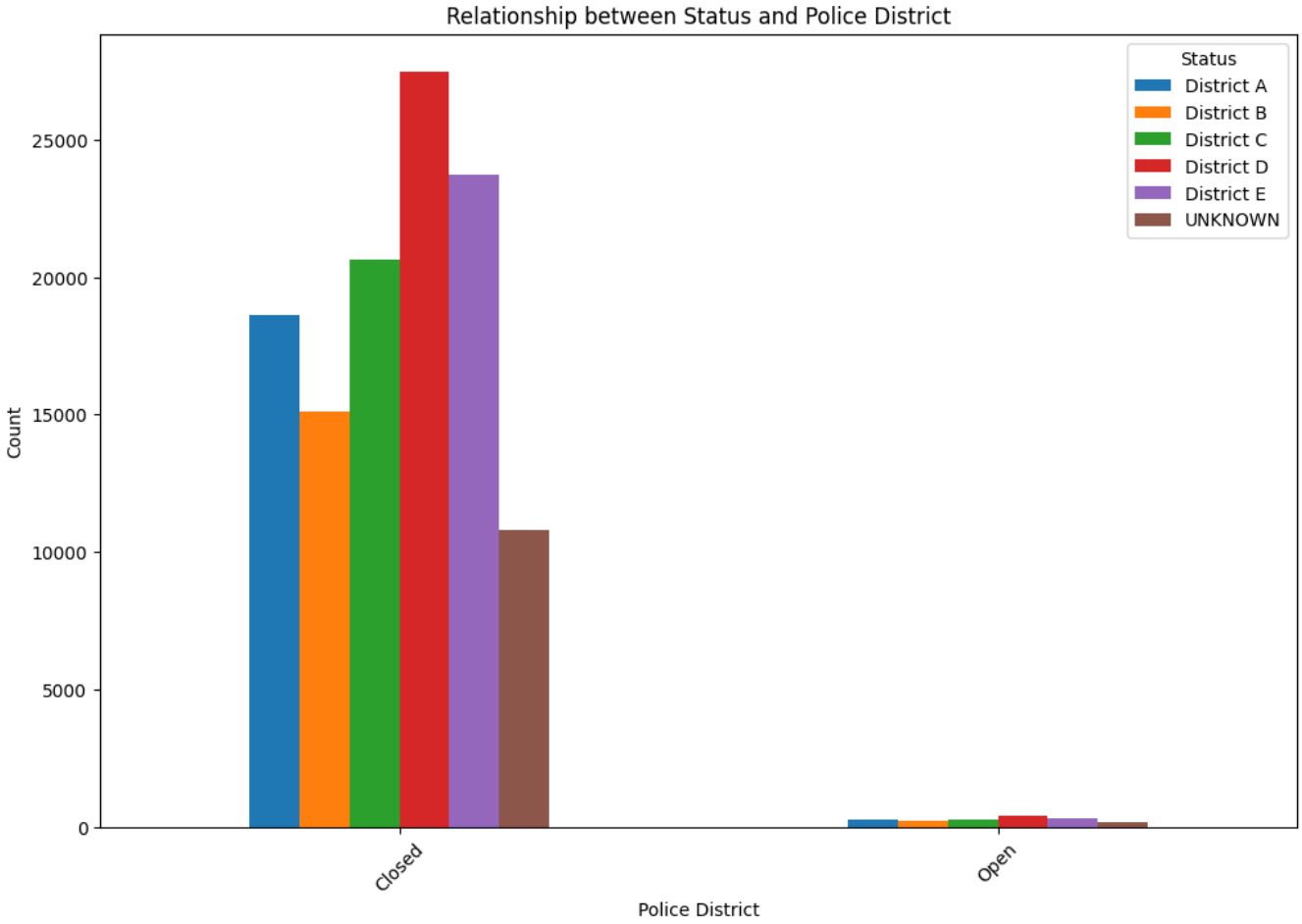
```
import matplotlib.pyplot as plt
```

```
grouped_data = df.groupby(['Status', 'Council District']).size().unstack(fill_value=
grouped_data.plot(kind='bar', figsize=(12, 8))
plt.title('Relationship between Status and Council District')
plt.xlabel('Council District')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.legend(title='Status')
plt.show()
```



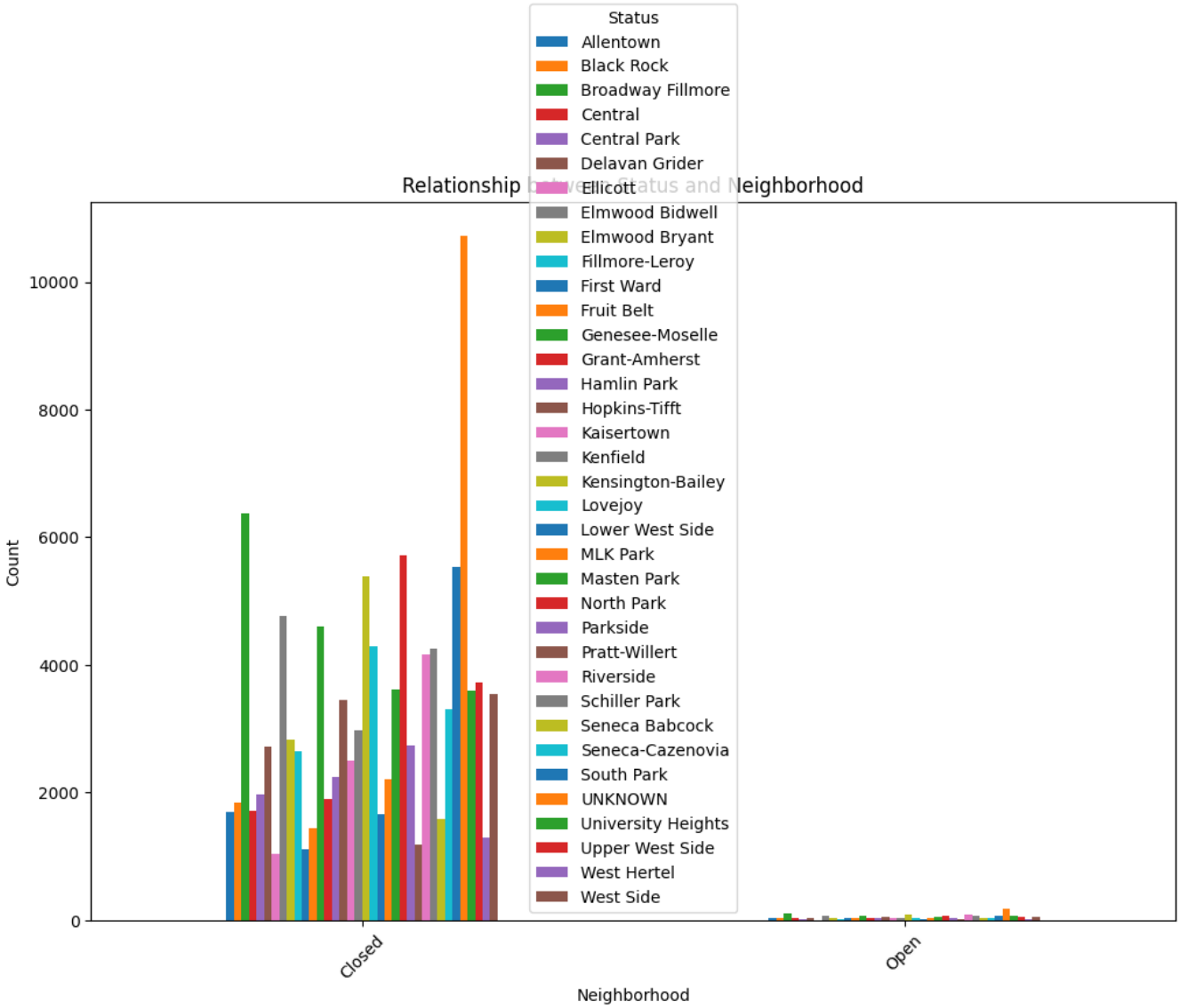
With the severe difference in amount of closed versus open observations it is not a surprise to see much more information on the closed side. The overwhelming majority of closed observations fall under the UNKNOWN council district.

```
grouped_data1 = df.groupby(['Status', 'Police District']).size().unstack(fill_value=
grouped_data1.plot(kind='bar', figsize=(12, 8))
plt.title('Relationship between Status and Police District')
plt.xlabel('Police District')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.legend(title='Status')
plt.show()
```



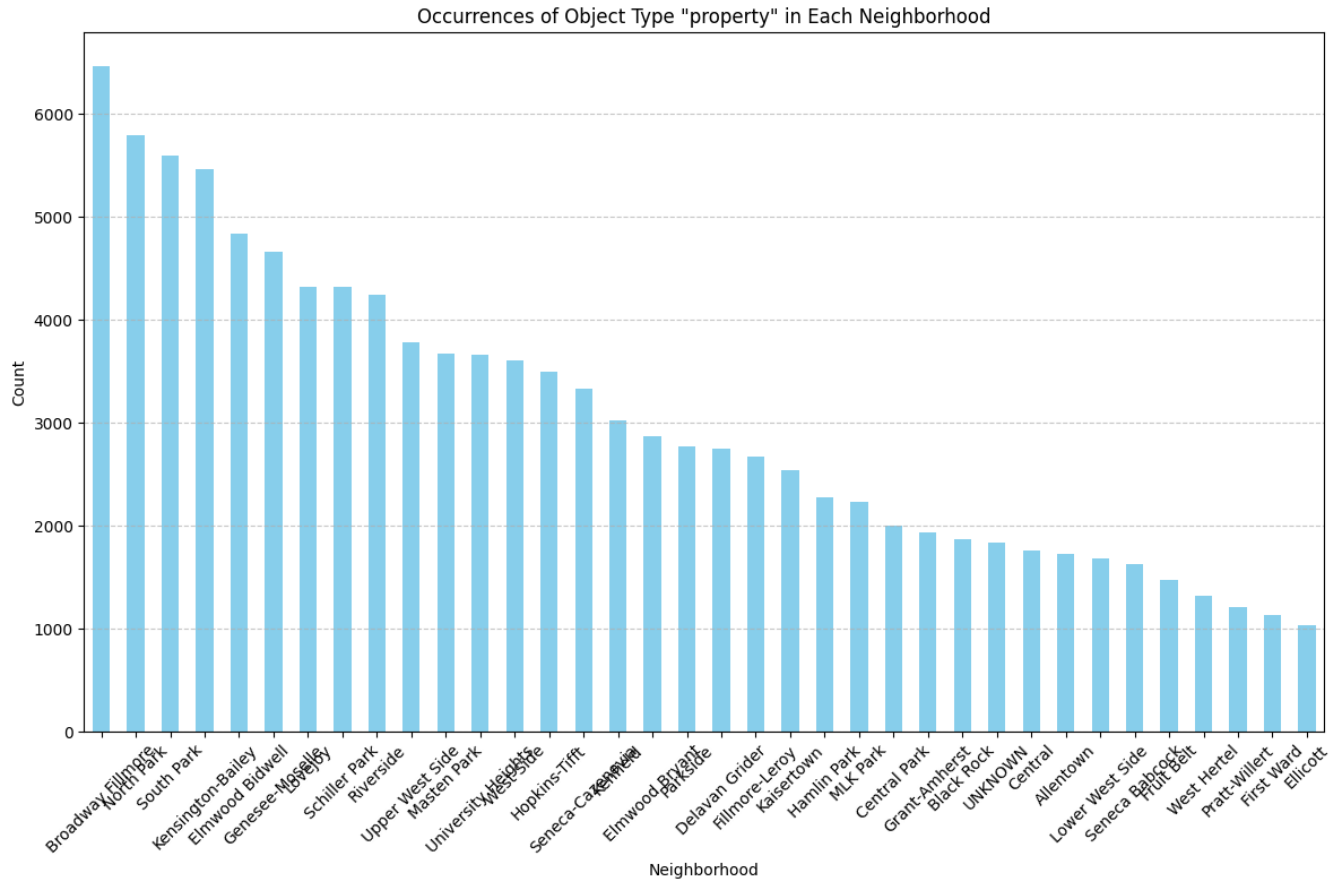
This shows much different results than the council district as there are far less unknowns. District D has the highest number of closed and open status'. District E has the second most for both closed and open.

```
grouped_data2 = df.groupby(['Status', 'Neighborhood']).size().unstack(fill_value=0)
grouped_data2.plot(kind='bar', figsize=(12, 8))
plt.title('Relationship between Status and Neighborhood')
plt.xlabel('Neighborhood')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.legend(title='Status')
plt.show()
```



While this is slightly difficult to read with the amount of neighborhoods that there are, we can still clearly see that UNKNOWN has the highest smount of closed and open status'. It seems to be a clear relationship between the number of closed and open observations as the neighborhoods with high number of closed observations also have high number of open observations.

```
property_data = df[df['Object Type'] == 'Property']
property_counts = property_data['Neighborhood'].value_counts()
plt.figure(figsize=(12, 8))
property_counts.plot(kind='bar', color='skyblue')
plt.title('Occurrences of Object Type "property" in Each Neighborhood')
plt.xlabel('Neighborhood')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```



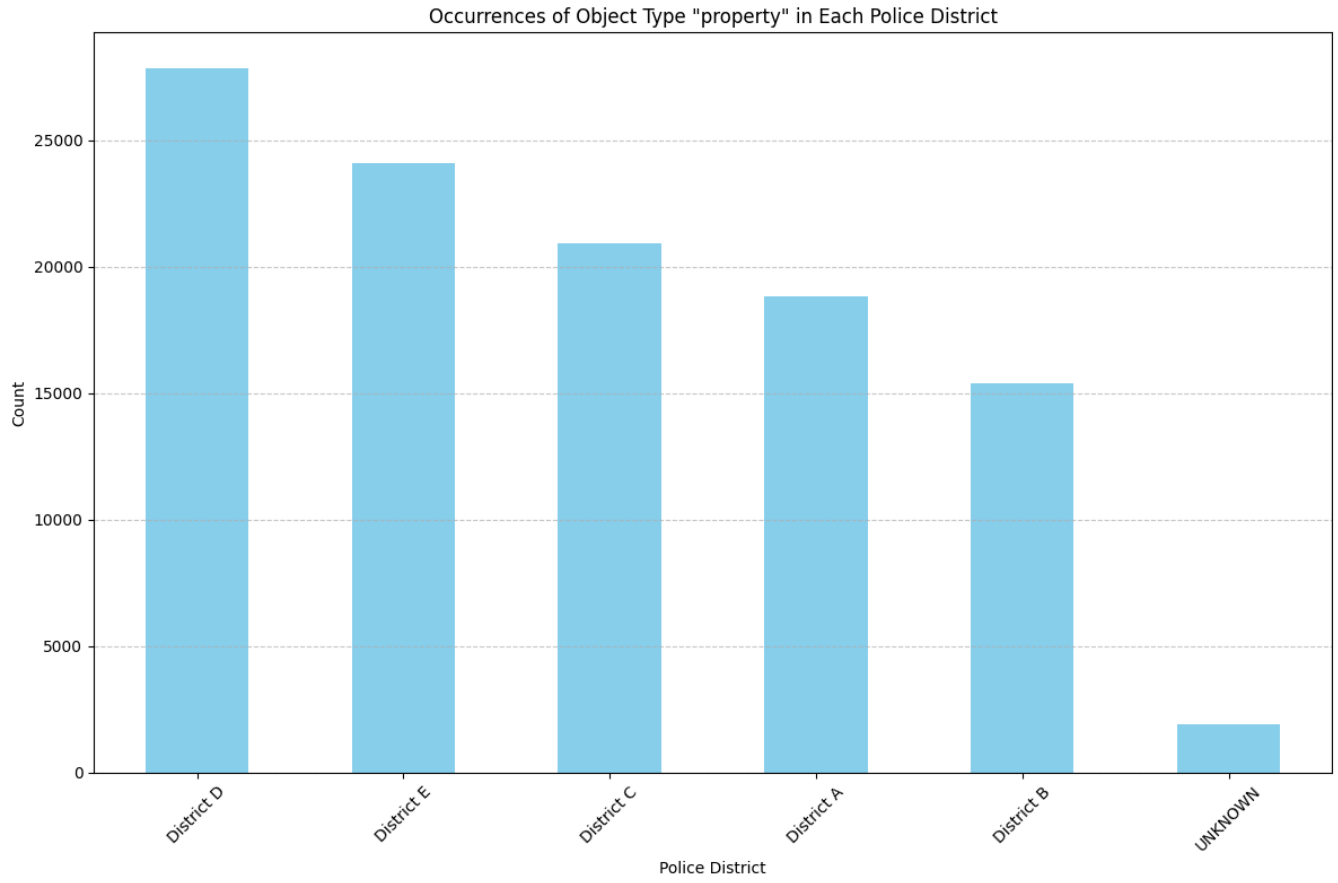

```
property_counts
```

Broadway Fillmore	6469
North Park	5794
South Park	5600
Kensington-Bailey	5469
Elmwood Bidwell	4842
Genesee-Moselle	4668
Lovejoy	4325
Schiller Park	4323
Riverside	4249
Upper West Side	3784
Masten Park	3669
University Heights	3667
West Side	3606
Hopkins-Tifft	3492
Seneca-Cazenovia	3336
Kenfield	3019
Elmwood Bryant	2869
Parkside	2776
Delavan Grider	2747
Fillmore-Leroy	2675
Kaisertown	2537
Hamlin Park	2279
MLK Park	2228
Central Park	1999
Grant-Amherst	1937
Black Rock	1872

```
UNKNOWN                1835
Central                1756
Allentown              1725
Lower West Side        1685
Seneca Babcock         1627
Fruit Belt             1475
West Hertel            1318
Pratt-Willert          1213
First Ward             1134
Ellicott               1037
Name: Neighborhood, dtype: int64
```

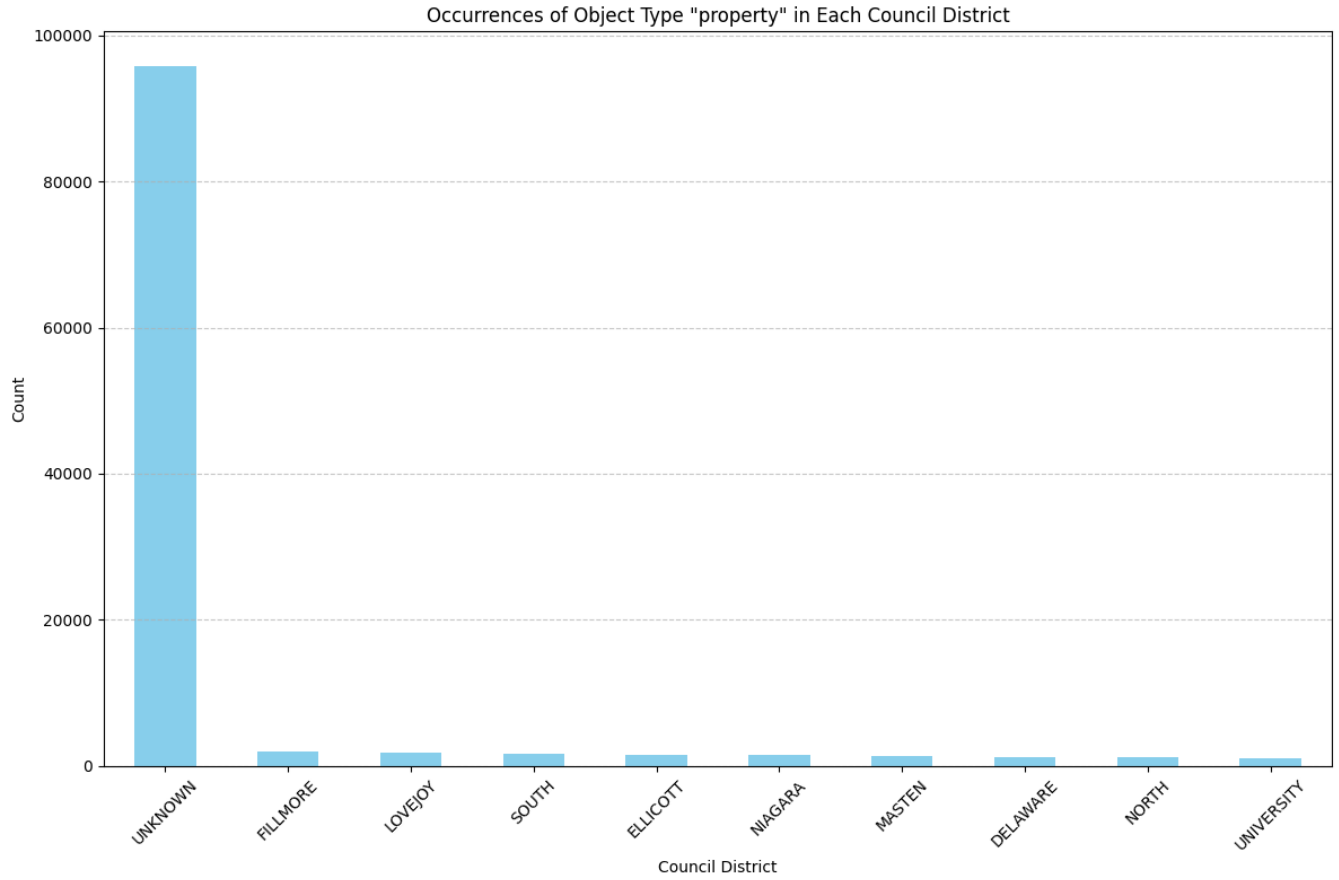
Here we can see that Broadway Fillmore has the most observations where the object type if property. Ellicott has the least.

```
property_counts = property_data['Police District'].value_counts()
plt.figure(figsize=(12, 8))
property_counts.plot(kind='bar', color='skyblue')
plt.title('Occurrences of Object Type "property" in Each Police District')
plt.xlabel('Police District')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```



Here we see that District D has the highest number of properties within it. UNKNOWN has the least amount.

```
property_counts = property_data['Council District'].value_counts()
plt.figure(figsize=(12, 8))
property_counts.plot(kind='bar', color='skyblue')
plt.title('Occurrences of Object Type "property" in Each Council District')
plt.xlabel('Council District')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```

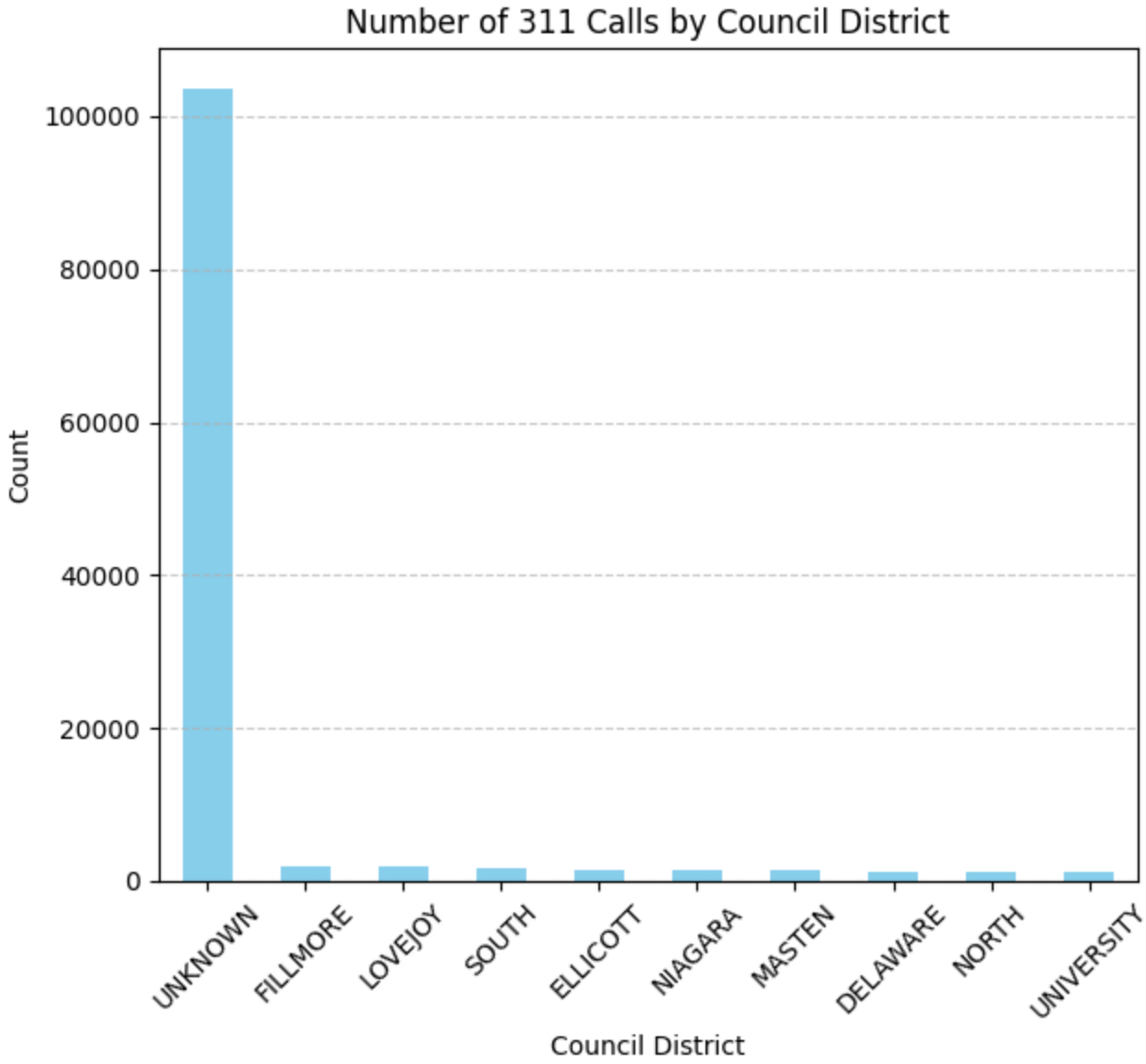


Similarly to how status performed with council district, UNKNOWN has the overwhelming amount of observations while the other council districts have rather similar numbers

NUMBER OF CALLS GRAPHS

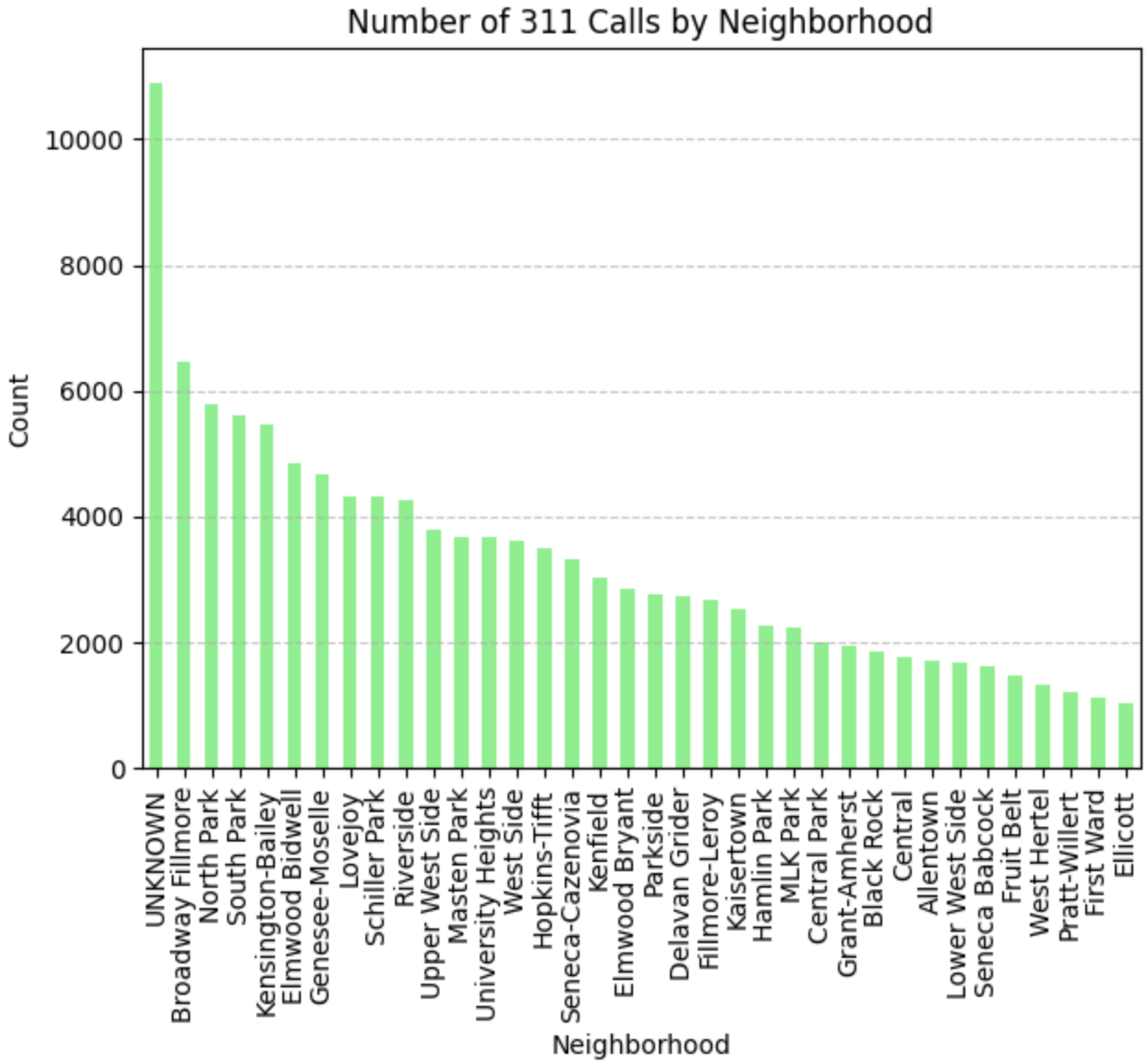
```
district_counts = df['Council District'].value_counts()
neighborhood_counts = df['Neighborhood'].value_counts()
```

```
plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 1)
district_counts.plot(kind='bar', color='skyblue')
plt.title('Number of 311 Calls by Council District')
plt.xlabel('Council District')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)

plt.tight_layout()
plt.show()
```



It is again evident that UNKNOWN has the highest number of 311 calls when looking at council district. Fillmore has the second highest.

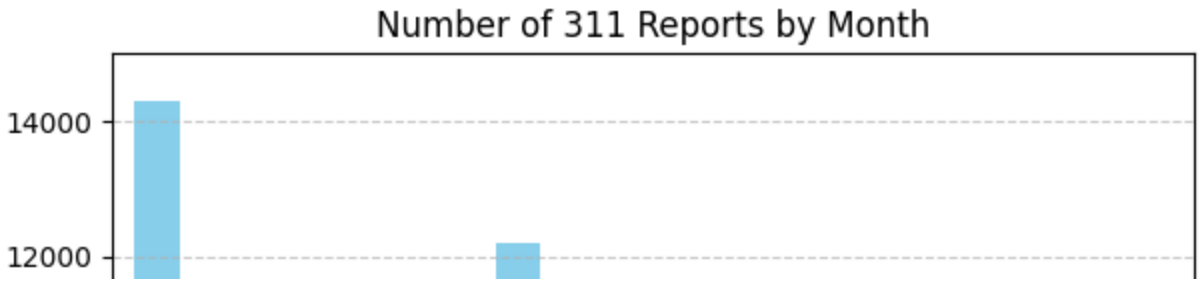
```
plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 2)
neighborhood_counts.plot(kind='bar', color='lightgreen')
plt.title('Number of 311 Calls by Neighborhood')
plt.xlabel('Neighborhood')
plt.ylabel('Count')
plt.xticks(rotation=90)
plt.grid(axis='y', linestyle='--', alpha=0.7)

plt.tight_layout()
plt.show()
```



Similarly to council district, UNKNOWN has the highest number of 311 calls when looking at it by neighborhood. 2nd is Broadway-Fillmore.

NUMBER OF CALLS BY MONTH AND YEAR

```
df['date'] = pd.to_datetime(df['Open Date'])


df['month'] = df['date'].dt.month
df['year'] = df['date'].dt.year


monthly_counts = df['month'].value_counts().sort_index()
yearly_counts = df['year'].value_counts().sort_index()


plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 1)
monthly_counts.plot(kind='bar', color='skyblue')
plt.title('Number of 311 Reports by Month')
plt.xlabel('Month')
plt.ylabel('Count')
plt.xticks(rotation=0)
plt.grid(axis='y', linestyle='--', alpha=0.7)


plt.tight_layout()
plt.show()
```



Here we can see January had the highest number of reported 311 cases. This is followed by the months surrounding and including summer with May, June, July and August all being grouped together much higher than any other month. November has the lowest amount of reported 311 cases.


```
plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 2)
yearly_counts.plot(kind='bar', color='lightgreen')
plt.title('Number of 311 Reports by Year')
plt.xlabel('Year')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)

plt.tight_layout()
plt.show()
```