Los Angeles and New York crime dataset analysis

Course: BIG DATA TOOLS & TECHNO. II
Prof. Nizar Ali

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Project overview

This project focuses on analyzing and comparing crime rates in Los Angeles and New York City using extensive datasets provided by the respective police departments.

The aim is to integrate and process these datasets using various big data tools and technologies to uncover patterns and insights into crime dynamics in these two major cities.

Objectives

The primary objective of this project is to compare crime rates between Los Angeles and New York City by analyzing large datasets.

This analysis aims to identify patterns, trends, and insights into the crime dynamics in these two major cities.

Dataset and sources

LYPD DATA

Arrest Data from 2010 to 2019 -: https://catalog.data.gov/dataset/arrest-data-from-2010-to-2019

Crime Data from 2020 to Present -:https://catalog.data.gov/dataset/crime-data-from-2020-to-present

LAPD Calls for Service 2019-: https://catalog.data.gov/dataset/lapd-calls-for-service-2019

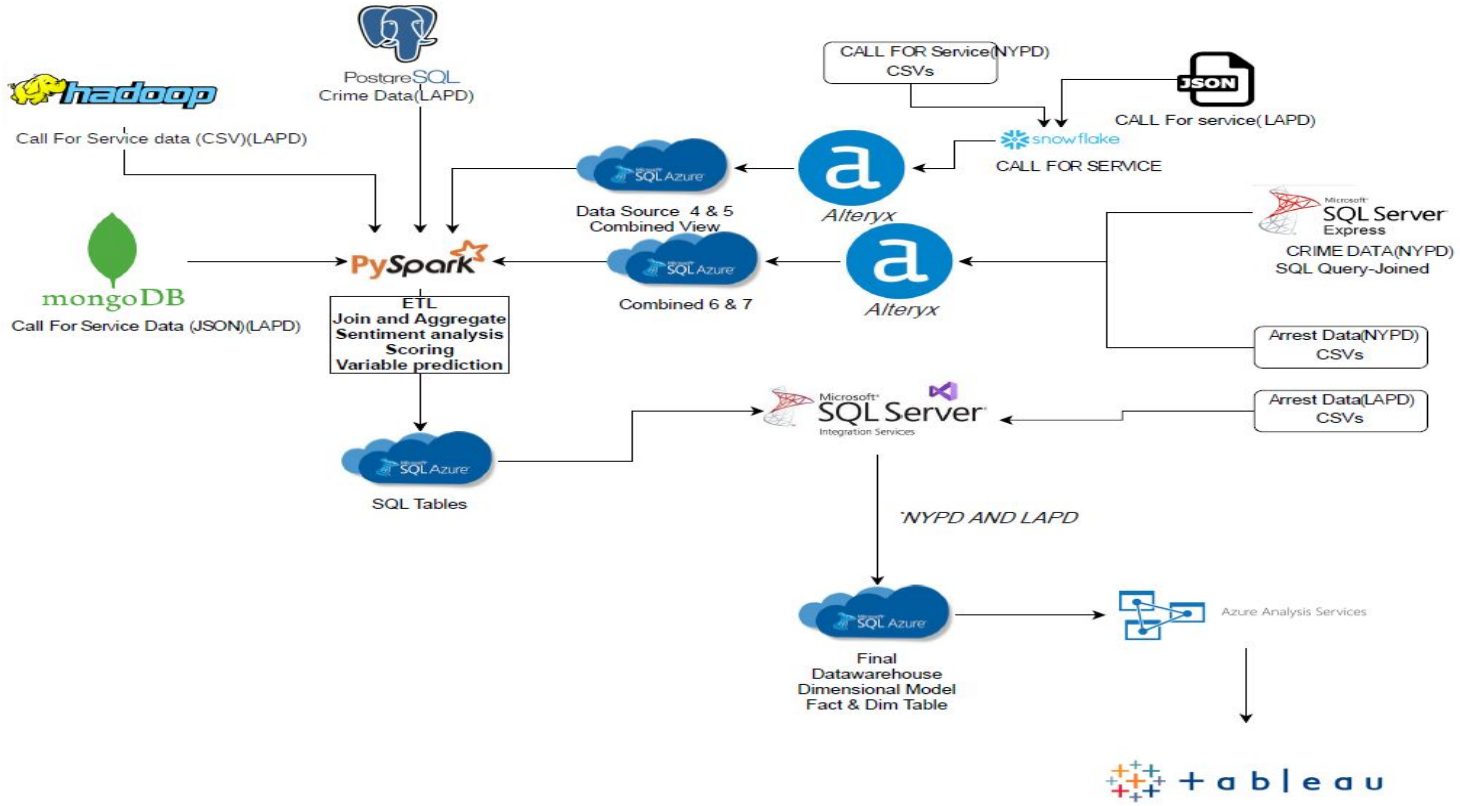
NEW YORK DATA

Arrest Data -: https://catalog.data.gov/dataset/nypd-arrests-data-historic

Call for Service Data -: https://catalog.data.gov/dataset/nypd-calls-for-service-historic

Crime Data -: https://catalog.data.gov/dataset/nypd-complaint-data-historic

Data flow chart



Data Preprocessing and transformation

```python
# Function to clean address
def clean_address(address):
    if address:
        return ' '.join(address.split())
    return address

clean_address_udf = udf(clean_address, StringType())
```

```python
# Apply the cleaning function to the Address and Cross Street columns
combined_arrest_data = combined_arrest_data.withColumn('Address', clean_address_udf(col('Address')))
combined_arrest_data = combined_arrest_data.withColumn('Cross Street', clean_address_udf(col('Cross Street')))
```

```python
# Drop specified columns
columns_to_drop = [
    'Descent Code', 'Charge Group Code', 'Charge Group Description',
    'Disposition Description', 'Booking Date', 'Booking Time',
    'Booking Location', 'Booking Location Code'
]
combined_arrest_data = combined_arrest_data.drop(*columns_to_drop)
```

```python
# Fill missing values in 'Charge Description, cross street'
combined_arrest_data = combined_arrest_data.withColumn('Charge Description', when(col('Charge Description').isNull(), 'Unknown').otherwise(col('Charge Description')))
```

```python
combined_arrest_data = combined_arrest_data.withColumn('Cross Street', when(col('Cross Street').isNull(), 'NAN').otherwise(col('Cross Street')))
# Fill missing values with an empty string
combined_arrest_data = combined_arrest_data.na.fill('NAN')
```

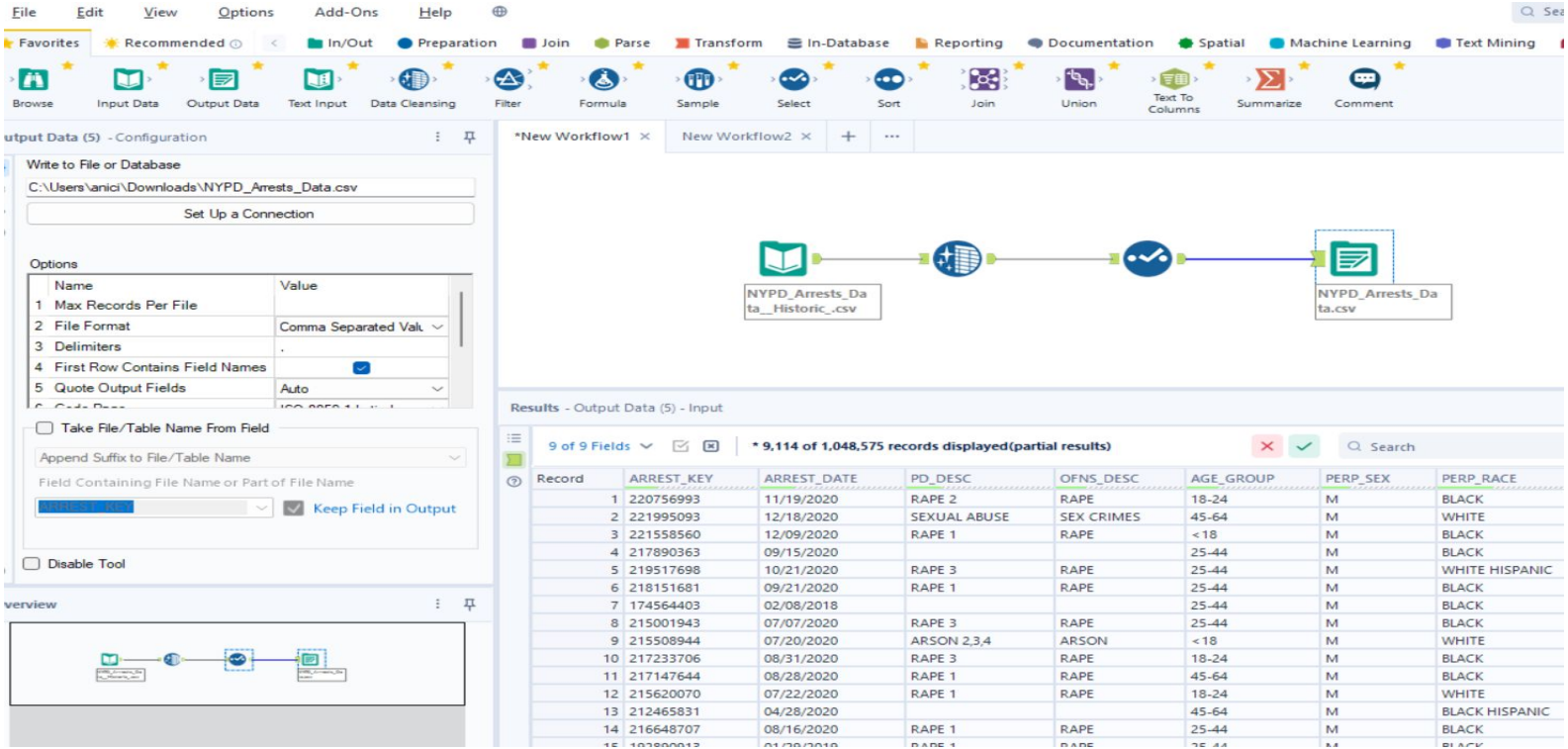
```python
# Show some results to verify final DataFrame
combined_arrest_data.show(5)
```

Data Preprocessing and transformation

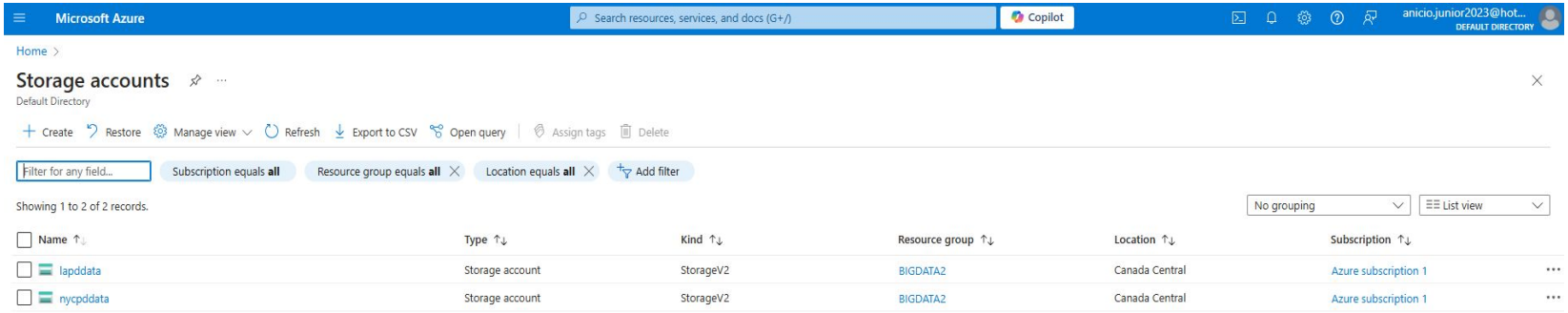
```
# Show some results to verify final DataFrame
combined_arrest_data.show(5)
```

```
+---------+-----------+-----------+----+-------+---------+-----------------+---+--------+---------------+---------+-----------------+-----------+-----------+-------+---------+--------------------+
|Report ID|Report Type|Arrest Date|Time|Area ID|Area Name|Reporting District|Age|Sex Code|Arrest Type Code|   Charge|Charge Description|    Address|Cross Street|    LAT|      LON|            Location|
+---------+-----------+-----------+----+-------+---------+-----------------+---+--------+---------------+---------+-----------------+-----------+-----------+-------+---------+--------------------+
|100629383|        RFC| 12/09/2010|1945|     06|Hollywood|             0647| 20|       M|              M|41.27CLAMC|DRINKING IN PUBLIC|  GORDON ST|  SUNSET BL|34.0981|-118.3196|POINT (-118.3196 ...|
|100629384|        RFC| 12/09/2010|1845|     06|Hollywood|             0648| 20|       M|              M|41.27CLAMC|DRINKING IN PUBLIC| CARLTON WY|  WESTERN AV|34.1005|-118.3093|POINT (-118.3093 ...|
|100629386|        RFC| 12/09/2010|1845|     06|Hollywood|             0648| 18|       M|              M|41.27CLAMC|DRINKING IN PUBLIC| CARLTON WY|  WESTERN AV|34.1005|-118.3093|POINT (-118.3093 ...|
|100629387|        RFC| 12/09/2010|2355|     06|Hollywood|             0665| 21|       F|              M|41.27CLAMC|DRINKING IN PUBLIC| LAS PALMAS|SANTA MONICA|34.0907|-118.3363|POINT (-118.3363 ...|
|100629388|        RFC| 12/09/2010|2335|     06|Hollywood|             0637| 19|       M|              M|41.27CLAMC|DRINKING IN PUBLIC|1700 VINE ST|           |34.1016|-118.3267|POINT (-118.3267 ...|
+---------+-----------+-----------+----+-------+---------+-----------------+---+--------+---------------+---------+-----------------+-----------+-----------+-------+---------+--------------------+
only showing top 5 rows
```

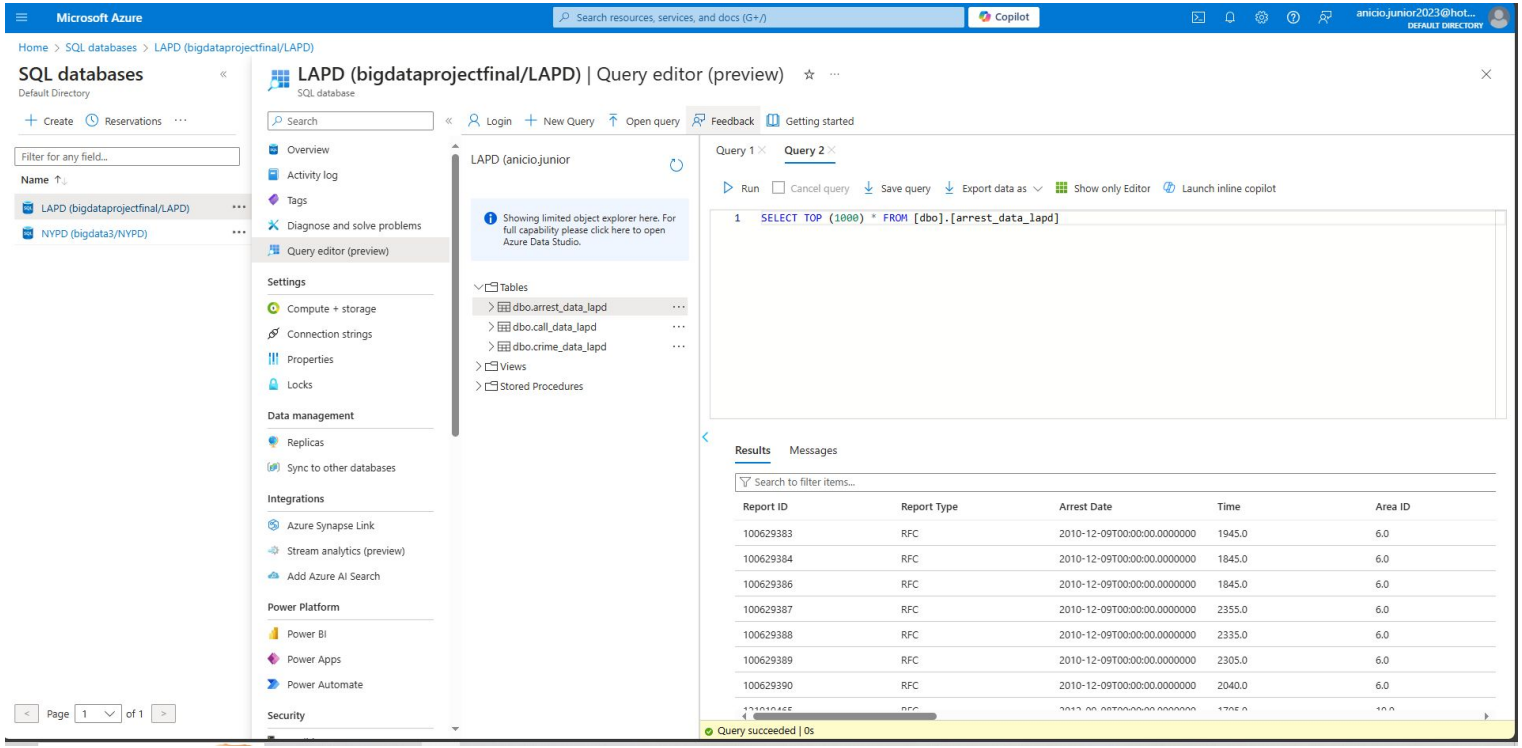
Data Preprocessing and transformation



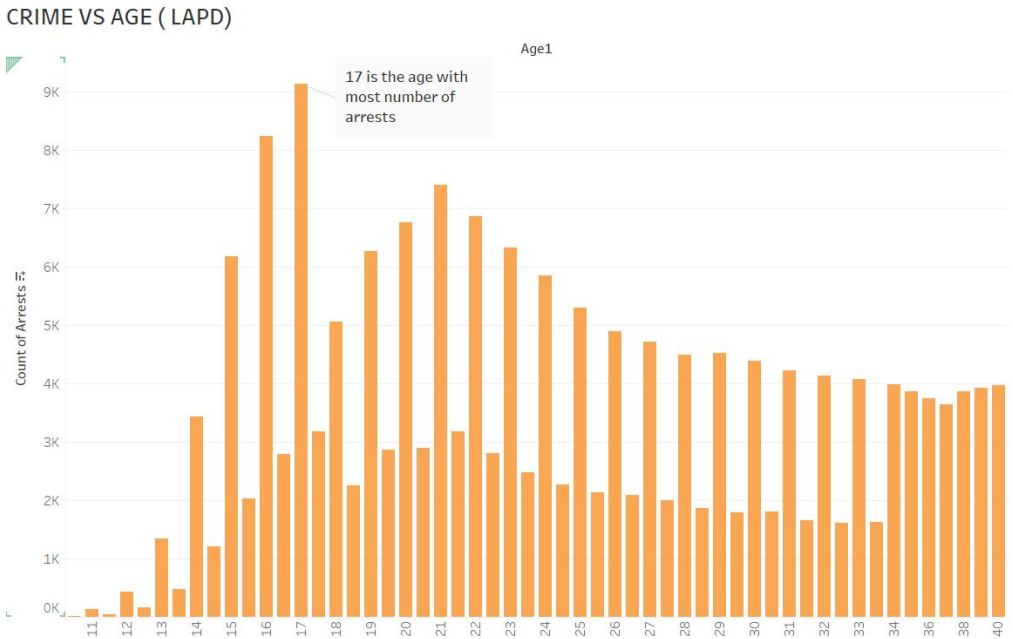
Azure connectivity

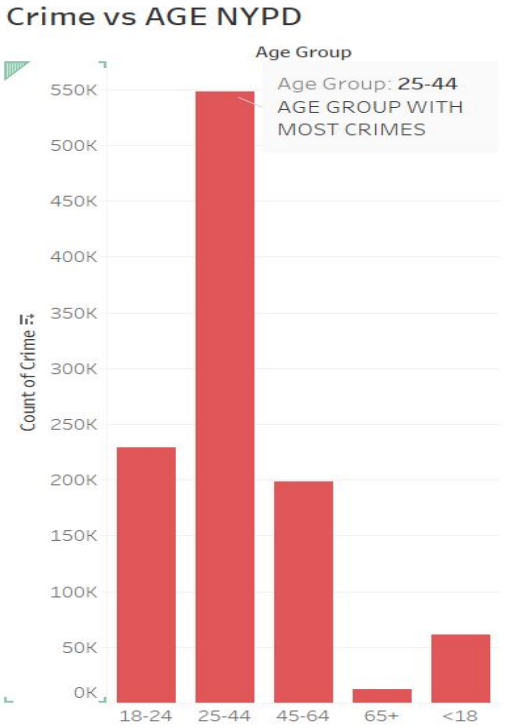


Azure connectivity



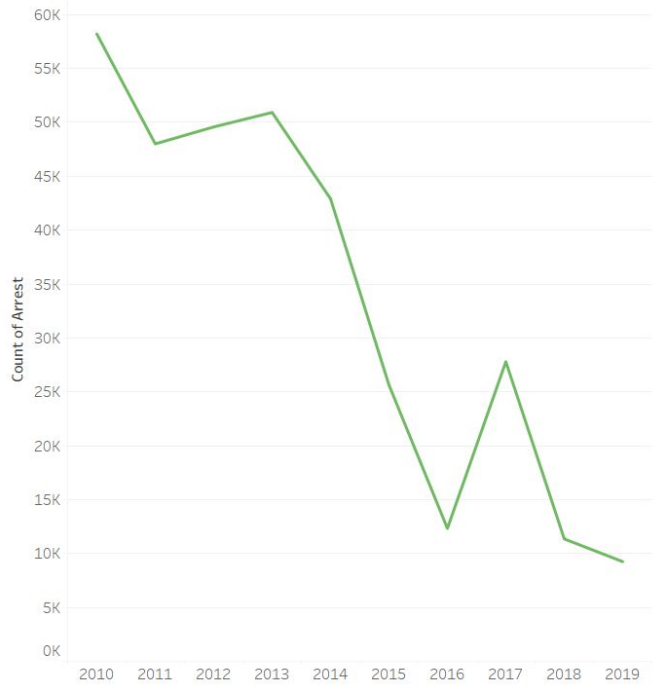
Results and graphs



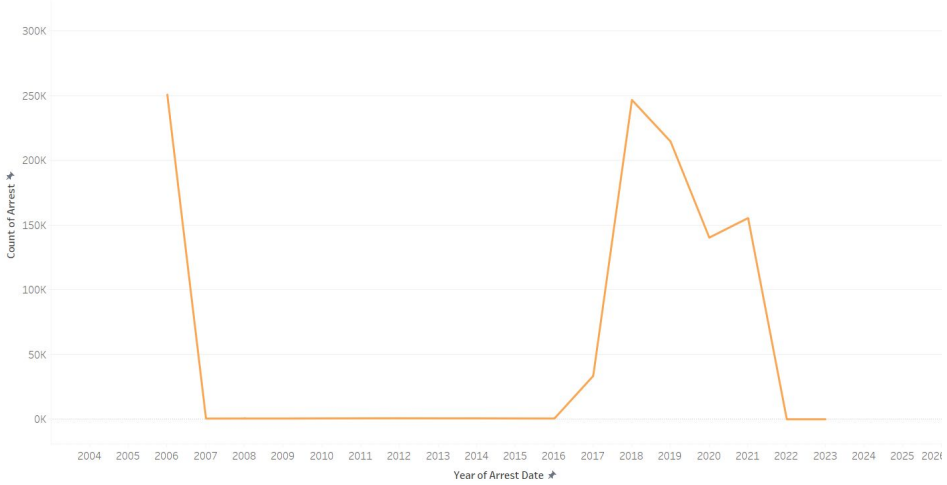


Results and graphs

Arrest VS YEAR (LAPD)

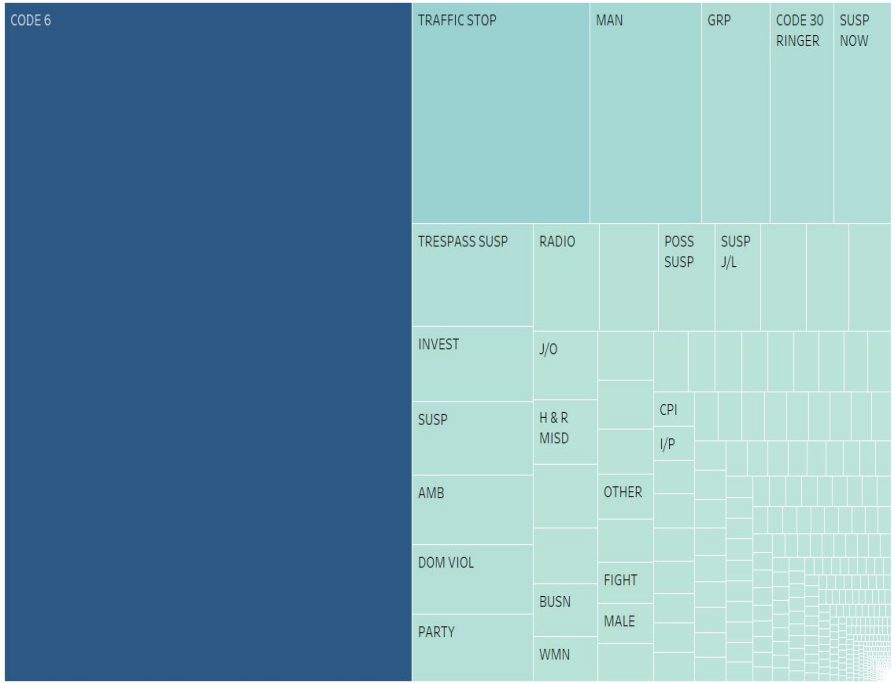


ARREST VS YEAR (NYPD)

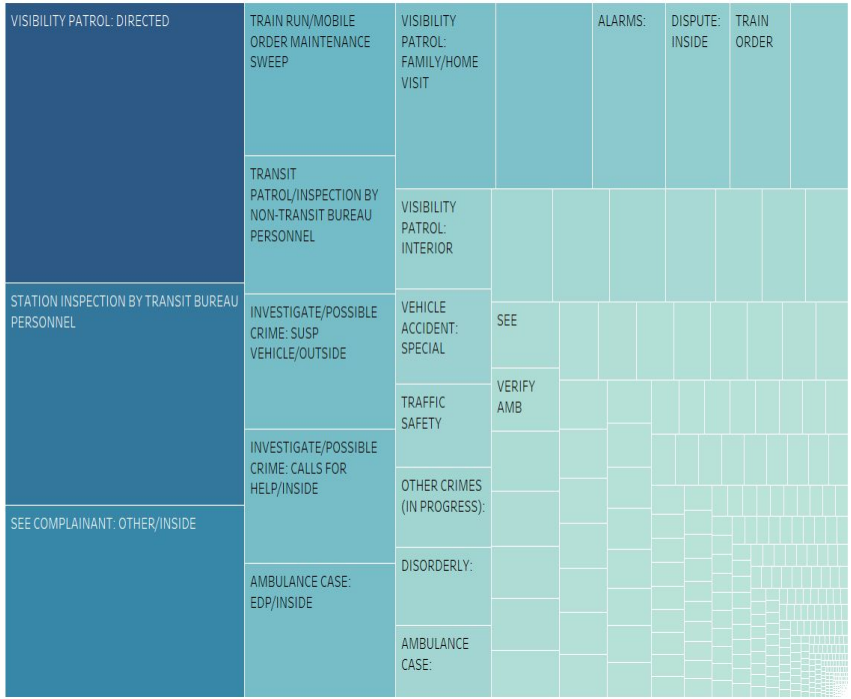


Results and graphs

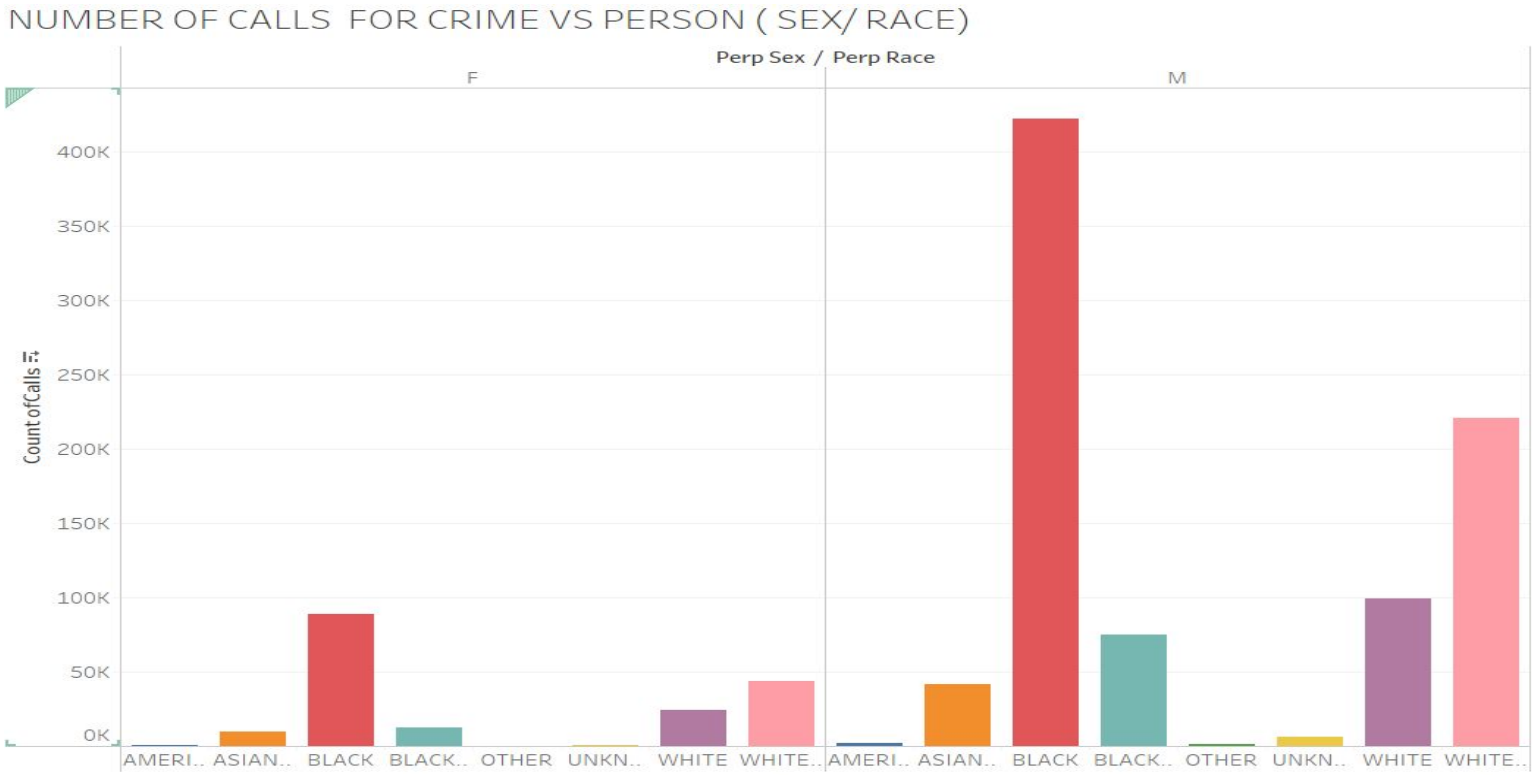
Most Calls For Service (LAPD)



Most CALLS for Service(NYPD)

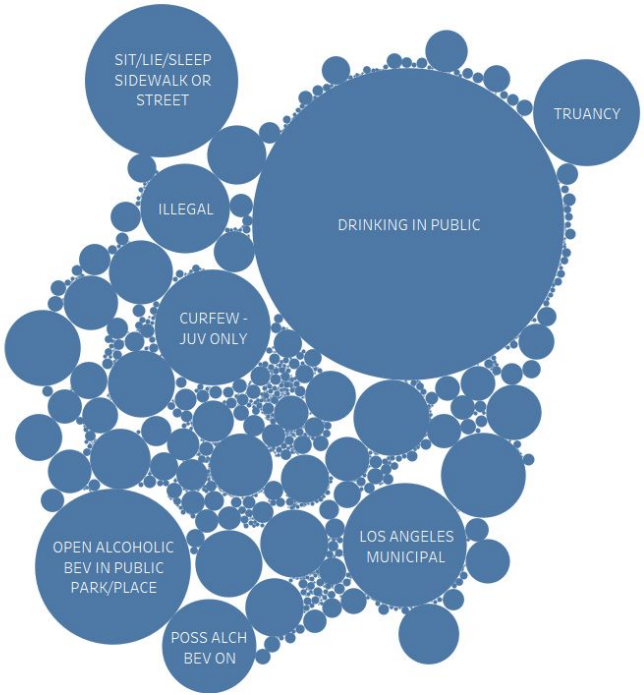


Results and graphs

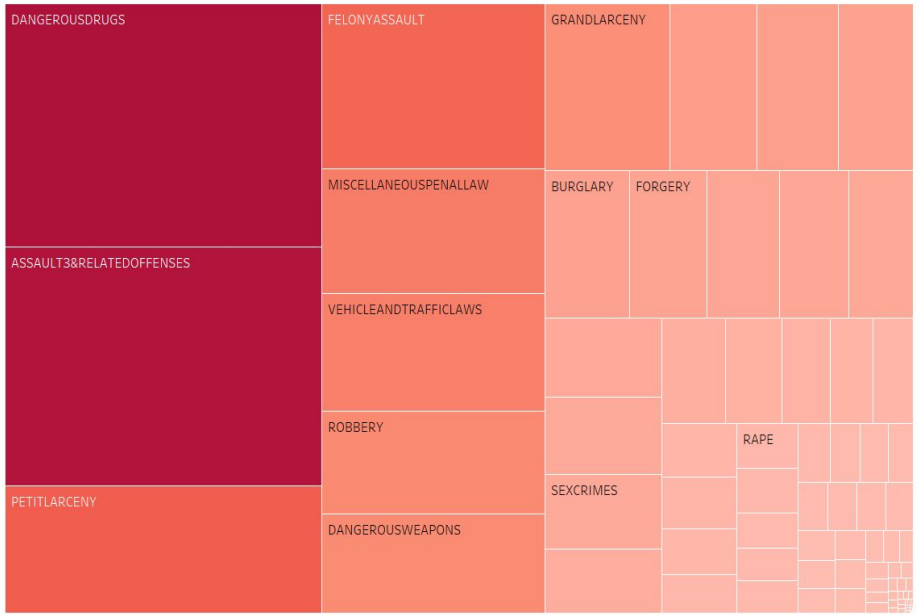


Results and graphs

MOST CRIMES (LAPD)



MOST CRIMES(NYPD)



Conclusion

Key Findings:

- **Crime vs. Age:**
  - **LAPD:** Peak arrests at age 17, smaller peak around ages 23-24.
  - **NYPD:** Highest crime count in the 25-44 age group.
- **Arrests vs. Year:**
  - **LAPD:** Overall declining trend from 2010 to 2019.
  - **NYPD:** Significant drop around 2012, followed by fluctuations.
- **Most Calls for Service:**
  - **LAPD:** 'Code 6' is the most frequent call.
  - **NYPD:** 'Visibility Patrol Directed' is the most common call.

Implications:

- Differences in age-related crime trends suggest targeted interventions.
- Yearly arrest trends indicate varying impacts of policies and socio-economic factors.
- Most common service calls reflect different policing priorities and strategies.

Conclusion:

- Understanding these patterns can help optimize law enforcement strategies and improve crime prevention efforts.
- Insights are valuable for developing targeted interventions in both cities.