## Exploratory Data Analysis + Data Pre-Processing & Data Cleaning

```
In []: # imports
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

# read in data
raw_data = pd.read_csv('../data/raw/Bos_crime_2023.csv')
In []: print(raw_data.describe())
print(raw_data.info())
```

```
_id OFFENSE_CODE OFFENSE_CODE_GROUP
                                                           SHOOTING
count 81133.000000 81133.000000
                                                  0.0
                                                      81133.000000
      40567.000000
                      2340.595861
                                                  NaN
                                                           0.008073
mean
std
       23421.224032
                      1175.333353
                                                  NaN
                                                           0.089488
min
           1.000000
                       111.000000
                                                  NaN
                                                           0.000000
25%
       20284.000000
                      1106.000000
                                                  NaN
                                                           0.000000
50%
      40567.000000
                      2907.000000
                                                  NaN
                                                           0.000000
75%
       60850.000000
                      3201.000000
                                                  NaN
                                                           0.000000
max
       81133.000000
                      3831.000000
                                                  NaN
                                                           1.000000
               YEAR
                            MONTH
                                           HOUR UCR_PART
                                                                    Lat \
count 81133.000000 81133.000000 81133.000000
                                                      0.0 7.528700e+04
mean
        2023.044261
                         6.361481
                                      12.486411
                                                      NaN 4.232312e+01
std
           0.205675
                         3.530441
                                       6.564979
                                                      NaN 1.576223e-01
min
        2023.000000
                         1.000000
                                       0.000000
                                                      NaN 1.327335e-07
25%
        2023.000000
                         3.000000
                                       8.000000
                                                      NaN 4.229755e+01
50%
        2023.000000
                         6.000000
                                      13.000000
                                                      NaN 4.232866e+01
75%
       2023.000000
                         9.000000
                                      18.000000
                                                      NaN 4.234906e+01
       2024.000000
                        12.000000
                                      23.000000
                                                      NaN 4.239504e+01
max
               Long
count 7.528700e+04
mean -7.108282e+01
      2.610060e-01
std
min
      -7.120251e+01
25%
     -7.109943e+01
50%
     -7.107775e+01
      -7.106090e+01
75%
      5.249645e-08
max
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 81133 entries, 0 to 81132
Data columns (total 18 columns):
    Column
                          Non-Null Count Dtype
#
---
    -----
                          _____
0
     id
                          81133 non-null
                                          int64
1
    INCIDENT NUMBER
                          81133 non-null object
 2
    OFFENSE CODE
                          81133 non-null int64
3
    OFFENSE_CODE_GROUP
                          0 non-null
                                          float64
4
     OFFENSE DESCRIPTION
                          81133 non-null object
5
    DISTRICT
                          80929 non-null object
 6
     REPORTING AREA
                          81133 non-null object
7
    SHOOTING
                          81133 non-null
                                          int64
8
     OCCURRED ON DATE
                          81133 non-null
                                          object
9
     YEAR
                          81133 non-null
                                          int64
10 MONTH
                          81133 non-null int64
11 DAY_OF_WEEK
                          81133 non-null
                                          object
12 HOUR
                          81133 non-null int64
13 UCR PART
                          0 non-null
                                          float64
14 STREET
                          81133 non-null object
15
    Lat
                          75287 non-null
                                          float64
16 Long
                          75287 non-null
                                          float64
 17 Location
                          75287 non-null
                                          object
dtypes: float64(4), int64(6), object(8)
memory usage: 11.1+ MB
None
```

In [ ]: # print all types of OFFENSE\_DESCRIPTION
print(raw\_data['OFFENSE\_DESCRIPTION'].unique())

```
['FRAUD - IMPERSONATION' 'VANDALISM' 'INVESTIGATE PERSON'
 'M/V - LEAVING SCENE - PROPERTY DAMAGE'
 'LARCENY THEFT FROM MV - NON-ACCESSORY' 'M/V ACCIDENT - PERSONAL INJURY'
 'THREATS TO DO BODILY HARM' 'SICK ASSIST - DRUG RELATED ILLNESS'
 'FRAUD - WELFARE' 'M/V ACCIDENT - OTHER' 'SICK ASSIST' 'VERBAL DISPUTE'
 'INVESTIGATE PROPERTY' 'ASSAULT - SIMPLE' 'BURGLARY - RESIDENTIAL'
 'TOWED MOTOR VEHICLE' 'ASSAULT - AGGRAVATED'
 'PROPERTY - LOST THEN LOCATED' 'HARASSMENT/ CRIMINAL HARASSMENT'
 'PROPERTY - LOST/ MISSING' 'OPERATING UNDER THE INFLUENCE (OUI) ALCOHOL'
 'VAL - OPERATING W/O AUTHORIZATION LAWFUL' 'LICENSE PREMISE VIOLATION'
 'SICK/INJURED/MEDICAL - PERSON' 'M/V ACCIDENT - PROPERTY DAMAGE'
 'LARCENY THEFT FROM BUILDING' 'AUTO THEFT'
 'M/V ACCIDENT - INVOLVING PEDESTRIAN - INJURY'
 'MURDER, NON-NEGLIGENT MANSLAUGHTER' 'PROPERTY - FOUND'
 'LIQUOR/ALCOHOL - DRINKING IN PUBLIC' 'VAL - VIOLATION OF AUTO LAW'
 'ROBBERY' 'LARCENY THEFT OF MV PARTS & ACCESSORIES' 'MISSING PERSON'
 'FIRE REPORT' 'FRAUD - CREDIT CARD / ATM FRAUD' 'LARCENY ALL OTHERS'
 'FIREARM/WEAPON - FOUND OR CONFISCATED'
 'DANGEROUS OR HAZARDOUS CONDITION' 'LARCENY SHOPLIFTING'
 'VIOLATION - CITY ORDINANCE' 'BALLISTICS EVIDENCE/FOUND'
 'DRUGS - POSSESSION/ SALE/ MANUFACTURING/ USE' 'TRESPASSING'
 'MISSING PERSON - LOCATED'
 'STOLEN PROPERTY - BUYING / RECEIVING / POSSESSING'
 'SERVICE TO OTHER AGENCY' 'FORGERY / COUNTERFEITING' 'M/V PLATES - LOST'
 'SUDDEN DEATH' 'DEATH INVESTIGATION' 'LARCENY THEFT OF BICYCLE'
 'SICK/INJURED/MEDICAL - POLICE' 'M/V ACCIDENT - POLICE VEHICLE'
 'VAL - OPERATING AFTER REV/SUSP.' 'PROPERTY - STOLEN THEN RECOVERED'
 'WEAPON VIOLATION - CARRY/ POSSESSING/ SALE/ TRAFFICKING/ OTHER'
 'ANIMAL INCIDENTS (DOG BITES, LOST DOG, ETC)'
 'M/V - LEAVING SCENE - PERSONAL INJURY'
 'CHILD REQUIRING ASSISTANCE (FOMERLY CHINS)'
 'PRISONER - SUICIDE / SUICIDE ATTEMPT' 'FRAUD - FALSE PRETENSE / SCHEME'
 'EVADING FARE' 'BURGLARY - COMMERICAL'
 'MISSING PERSON - NOT REPORTED - LOCATED'
 'M/V ACCIDENT - INVOLVING BICYCLE - NO INJURY'
 'M/V ACCIDENT - OTHER CITY VEHICLE'
 'WARRANT ARREST - OUTSIDE OF BOSTON WARRANT'
 'RECOVERED - MV RECOVERED IN BOSTON (STOLEN OUTSIDE BOSTON)'
 'FRAUD - WIRE' 'LANDLORD - TENANT' 'EXTORTION OR BLACKMAIL'
 'M/V ACCIDENT - INVOLVING BICYCLE - INJURY' 'SEARCH WARRANT'
 'DISTURBING THE PEACE/ DISORDERLY CONDUCT/ GATHERING CAUSING ANNOYANCE/ NOISY PA
R'
 'PROPERTY - ACCIDENTAL DAMAGE' 'NOISY PARTY/RADIO-NO ARREST' 'GRAFFITI'
 'LARCENY PICK-POCKET' 'OTHER OFFENSE'
 'DRUGS - POSSESSION OF DRUG PARAPHANALIA'
 'WARRANT ARREST - BOSTON WARRANT (MUST BE SUPPLEMENTAL)'
 'FUGITIVE FROM JUSTICE' 'HARBOR INCIDENT / VIOLATION'
 'INJURY BICYCLE NO M/V INVOLVED'
 'M/V ACCIDENT - INVOLVING PEDESTRIAN - NO INJURY'
 'FIRE REPORT/ALARM - FALSE' 'AUTO THEFT - MOTORCYCLE / SCOOTER' 'ARSON'
 'SUICIDE / SUICIDE ATTEMPT' 'EMBEZZLEMENT'
 'AUTO THEFT - LEASED/RENTED VEHICLE'
 'RECOVERED - MV RECOVERED IN BOSTON (STOLEN IN BOSTON) MUST BE SUPPLEMENTAL'
 'BOMB THREAT' 'TRUANCY / RUNAWAY' 'MANSLAUGHTER - VEHICLE - NEGLIGENCE'
 'AIRCRAFT INCIDENTS' 'AFFRAY' 'DRUNKENNESS' 'INTIMIDATING WITNESS'
 'LARCENY PURSE SNATCH - NO FORCE' 'LIQUOR LAW VIOLATION'
 'BREAKING AND ENTERING (B&E) MOTOR VEHICLE' 'ANIMAL ABUSE'
 'FIREARM/WEAPON - LOST' 'MURDER, NON-NEGLIGIENT MANSLAUGHTER'
 'LARCENY THEFT FROM COIN-OP MACHINE'
 'KIDNAPPING/CUSTODIAL KIDNAPPING/ ABDUCTION'
```

```
'PROTECTIVE CUSTODY / SAFEKEEPING'
'FIREARM/WEAPON - ACCIDENTAL INJURY / DEATH' 'PROSTITUTION - SOLICITING'
'EXPLOSIVES - POSSESSION OR USE' 'OBSCENE PHONE CALLS' 'PROSTITUTION'
'BREAKING AND ENTERING (B&E) MOTOR VEHICLE (NO PROPERTY STOLEN)'
'OPERATING UNDER THE INFLUENCE (OUI) DRUGS'
'POSSESSION OF BURGLARIOUS TOOLS']
```

```
In []: # create a list of all crime descriptions and their counts
    crime_list = raw_data['OFFENSE_DESCRIPTION'].value_counts().reset_index().values

# print crime_list
print(crime_list)

# save crime_list to csv
crime_df = pd.DataFrame(crime_list)
crime_df.to_csv('../data/processed/crime_list.csv', index=False)

print('Done!')
```

[['INVESTIGATE PERSON', 8741], ['SICK ASSIST', 6869], ['M/V - LEAVING SCENE - PRO PERTY DAMAGE', 4759], ['INVESTIGATE PROPERTY', 3610], ['TOWED MOTOR VEHICLE', 342 4], ['ASSAULT - SIMPLE', 3210], ['VANDALISM', 3132], ['LARCENY SHOPLIFTING', 298 8], ['PROPERTY - LOST/ MISSING', 2519], ['LARCENY THEFT FROM BUILDING', 2164], ['LARCENY THEFT FROM MV - NON-ACCESSORY', 2026], ['M/V ACCIDENT - PROPERTY DAMAG E', 1910], ['THREATS TO DO BODILY HARM', 1864], ['VERBAL DISPUTE', 1854], ['DRUGS - POSSESSION/ SALE/ MANUFACTURING/ USE', 1810], ['LARCENY ALL OTHERS', 1713], ['M/V ACCIDENT - OTHER', 1648], ['ASSAULT - AGGRAVATED', 1585], ['MISSING PERSON - LOCATED', 1283], ['HARASSMENT/ CRIMINAL HARASSMENT', 1273], ['FRAUD - FALSE PRE TENSE / SCHEME', 1262], ['SICK/INJURED/MEDICAL - PERSON', 1236], ['VAL - VIOLATIO N OF AUTO LAW', 1138], ['AUTO THEFT', 1121], ['M/V ACCIDENT - PERSONAL INJURY', 1 041], ['PROPERTY - FOUND', 985], ['SICK ASSIST - DRUG RELATED ILLNESS', 812], ['R OBBERY', 794], ['LARCENY THEFT OF BICYCLE', 679], ['BURGLARY - RESIDENTIAL', 67 8], ['M/V ACCIDENT - INVOLVING PEDESTRIAN - INJURY', 572], ['FRAUD - CREDIT CARD / ATM FRAUD', 542], ['WARRANT ARREST - OUTSIDE OF BOSTON WARRANT', 497], ['DEATH INVESTIGATION', 447], ['FORGERY / COUNTERFEITING', 444], ['FIRE REPORT', 439], ['SUDDEN DEATH', 437], ['LANDLORD - TENANT', 407], ['TRESPASSING', 400], ['VAL -OPERATING AFTER REV/SUSP.', 397], ['LICENSE PREMISE VIOLATION', 396], ['SICK/INJU RED/MEDICAL - POLICE', 385], ['MISSING PERSON', 376], ['MISSING PERSON - NOT REPO RTED - LOCATED', 367], ['LARCENY THEFT OF MV PARTS & ACCESSORIES', 340], ['SERVIC E TO OTHER AGENCY', 333], ['M/V - LEAVING SCENE - PERSONAL INJURY', 332], ['BURGL ARY - COMMERICAL', 323], ['WEAPON VIOLATION - CARRY/ POSSESSING/ SALE/ TRAFFICKIN G/ OTHER', 307], ['FRAUD - IMPERSONATION', 297], ['RECOVERED - MV RECOVERED IN BO STON (STOLEN OUTSIDE BOSTON)', 292], ['BALLISTICS EVIDENCE/FOUND', 276], ['FRAUD - WIRE', 266], ['M/V ACCIDENT - POLICE VEHICLE', 261], ['PROPERTY - ACCIDENTAL DA MAGE', 261], ['M/V ACCIDENT - OTHER CITY VEHICLE', 259], ['ANIMAL INCIDENTS (DOG BITES, LOST DOG, ETC)', 230], ['M/V ACCIDENT - INVOLVING BICYCLE - INJURY', 226], ['AUTO THEFT - MOTORCYCLE / SCOOTER', 222], ['FIREARM/WEAPON - FOUND OR CONFISCAT ED', 193], ['STOLEN PROPERTY - BUYING / RECEIVING / POSSESSING', 174], ['M/V PLAT ES - LOST', 151], ['DISTURBING THE PEACE/ DISORDERLY CONDUCT/ GATHERING CAUSING A NNOYANCE/ NOISY PAR', 148], ['SEARCH WARRANT', 121], ['M/V ACCIDENT - INVOLVING P EDESTRIAN - NO INJURY', 118], ['VIOLATION - CITY ORDINANCE', 111], ['M/V ACCIDENT - INVOLVING BICYCLE - NO INJURY', 103], ['OPERATING UNDER THE INFLUENCE (OUI) ALC OHOL', 95], ['EXTORTION OR BLACKMAIL', 95], ['FRAUD - WELFARE', 82], ['WARRANT AR REST - BOSTON WARRANT (MUST BE SUPPLEMENTAL)', 79], ['LIQUOR/ALCOHOL - DRINKING I N PUBLIC', 75], ['LIQUOR LAW VIOLATION', 71], ['AUTO THEFT - LEASED/RENTED VEHICL E', 67], ['LARCENY PICK-POCKET', 63], ['NOISY PARTY/RADIO-NO ARREST', 63], ['HARB OR INCIDENT / VIOLATION', 63], ['VAL - OPERATING W/O AUTHORIZATION LAWFUL', 57], ['FUGITIVE FROM JUSTICE', 46], ['GRAFFITI', 46], ['EMBEZZLEMENT', 46], ['DANGEROU S OR HAZARDOUS CONDITION', 44], ['PROPERTY - LOST THEN LOCATED', 43], ['FIRE REPO RT/ALARM - FALSE', 43], ['AFFRAY', 40], ['PROPERTY - STOLEN THEN RECOVERED', 34], ['SUICIDE / SUICIDE ATTEMPT', 32], ['OTHER OFFENSE', 31], ['MURDER, NON-NEGLIGENT MANSLAUGHTER', 30], ['BOMB THREAT', 30], ['AIRCRAFT INCIDENTS', 26], ['RECOVERED - MV RECOVERED IN BOSTON (STOLEN IN BOSTON) MUST BE SUPPLEMENTAL', 25], ['EVADING FARE', 25], ['INJURY BICYCLE NO M/V INVOLVED', 23], ['ARSON', 20], ['DRUGS - POSS ESSION OF DRUG PARAPHANALIA', 19], ['LARCENY PURSE SNATCH - NO FORCE', 17], ['BRE AKING AND ENTERING (B&E) MOTOR VEHICLE', 16], ['TRUANCY / RUNAWAY', 15], ['INTIMI DATING WITNESS', 14], ['CHILD REQUIRING ASSISTANCE (FOMERLY CHINS)', 10], ['PRISO NER - SUICIDE / SUICIDE ATTEMPT', 10], ['DRUNKENNESS', 8], ['FIREARM/WEAPON - LOS T', 7], ['ANIMAL ABUSE', 7], ['LARCENY THEFT FROM COIN-OP MACHINE', 6], ['KIDNAPP ING/CUSTODIAL KIDNAPPING/ ABDUCTION', 6], ['BREAKING AND ENTERING (B&E) MOTOR VEH ICLE (NO PROPERTY STOLEN)', 4], ['OPERATING UNDER THE INFLUENCE (OUI) DRUGS', 4], ['MANSLAUGHTER - VEHICLE - NEGLIGENCE', 3], ['PROSTITUTION - SOLICITING', 3], ['O BSCENE PHONE CALLS', 3], ['PROTECTIVE CUSTODY / SAFEKEEPING', 2], ['FIREARM/WEAPO N - ACCIDENTAL INJURY / DEATH', 2], ['EXPLOSIVES - POSSESSION OR USE', 2], ['MURD ER, NON-NEGLIGIENT MANSLAUGHTER', 1], ['PROSTITUTION', 1], ['POSSESSION OF BURGLA RIOUS TOOLS', 1]]

```
In []: # draw a bar chart of the top 10 crimes using different colors for each crime

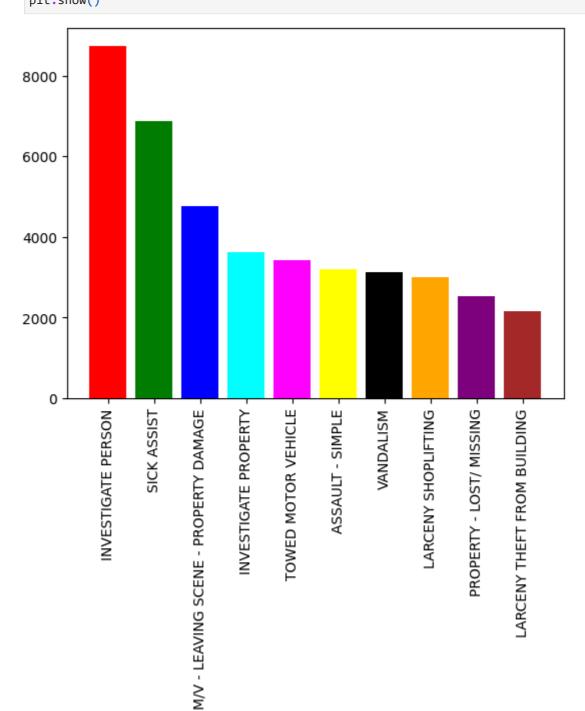
# create a list of the top 10 crimes
top_10_crimes = raw_data['OFFENSE_DESCRIPTION'].value_counts().head(10).reset_in

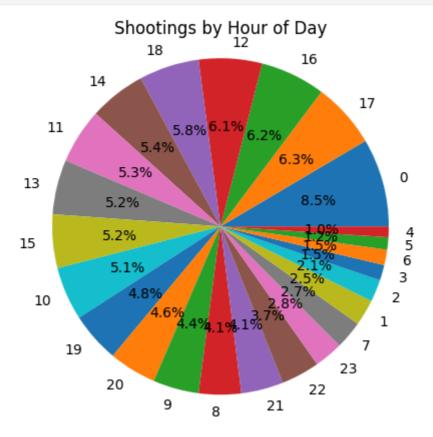
# draw a bar chart of the top 10 crimes
plt.bar(range(len(top_10_crimes)), [val[1] for val in top_10_crimes], align='cen
plt.xticks(range(len(top_10_crimes)), [val[0] for val in top_10_crimes])
plt.xticks(rotation=90)

# save bar chart to file

plt.savefig('../reports/figures/top_10_crimes.png') # need to call this before p

plt.show()
```





```
In []: # drop Offense Code group column
    raw_data.drop('OFFENSE_CODE_GROUP', axis=1, inplace=True)

# drio UCR_PART column
    raw_data.drop('UCR_PART', axis=1, inplace=True)

# drop Lat and Long and Location columns
    raw_data.drop('Lat', axis=1, inplace=True)
    raw_data.drop('Long', axis=1, inplace=True)
    raw_data.drop('Location', axis=1, inplace=True)

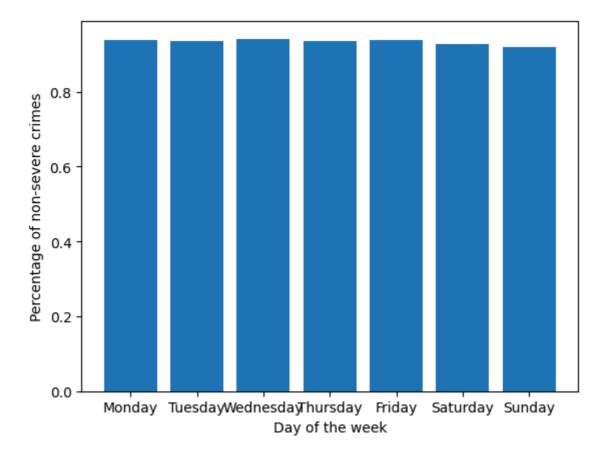
# drop incident number column
    raw_data.drop('INCIDENT_NUMBER', axis=1, inplace=True)

print(raw_data.shape)

# never run this repeatedLy!!!!!!!!!!!
```

```
In [ ]: # drop rows with null values]
        raw_data.dropna(inplace=True)
        raw_data.reset_index(drop=True, inplace=True)
        print(raw_data.shape)
       (80929, 12)
In [ ]: # save cleaned data to csv
        raw_data.to_csv('../data/processed/cleaned_data.csv', index=False)
        print('Done!')
       Done!
In [ ]: # Feature Engineering
        data = pd.read_csv('../data/processed/cleaned_data.csv')
        # add a new column set default to 0 for all rows
        data['Severe_crimes'] = 0
        # print the number of 0s in the new column
        print(data['Severe_crimes'].value_counts())
        # set the value of Sevre_crimes to 1 if the crime involves shooting
        data.loc[data['SHOOTING'] == 1, 'Severe_crimes'] = 1
        # set the value of Sevre_crimes to 1 if the crime description contains the follo
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('ASSAULT', case=False), 'Sever
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('MURDER', case=False), 'Severe
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('ARSON', case=False), 'Severe_
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('KIDNAPPING', case=False), 'Se
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('MANSLAUGHTER', case=False), '
        data.loc[data['OFFENSE_DESCRIPTION'].str.contains('BREAKING', case=False), 'Seve
       Severe crimes
            80929
       Name: count, dtype: int64
In [ ]: # save cleaned data to csv
        data.to_csv('../data/processed/cleaned_data.csv', index=False)
        print('Done!')
       Done!
In [ ]: # read in the data
        data = pd.read_csv('../data/processed/Cleaned_Data.csv')
In [ ]: # print the number of 0s in the new column
        print(data['Severe_crimes'].value_counts())
       Severe_crimes
           75556
             5373
       Name: count, dtype: int64
In [ ]: # use the 2024 data as testing data
        test_data = data[data["YEAR"] == 2024]
        # use train test split to split the data into training and validation data
```

```
train_data, val_data = train_test_split(data[data["YEAR"] != 2024], test_size=0.
        # save the data
        train_data.to_csv("../data/processed/train_data.csv", index=False)
        val_data.to_csv("../data/processed/val_data.csv", index=False)
        test_data.to_csv(".../data/processed/test_data.csv", index=False)
In [ ]: # describe the data
        print("Training Data")
        print(train_data.describe())
      Training Data
                      _id OFFENSE_CODE
                                           SHOOTING
                                                        YEAR
                                                                    MONTH \
      count 61877.000000 61877.000000 61877.000000 61877.0 61877.000000
             39011.358615 2338.513115
                                           0.008307
                                                      2023.0
                                                                 6.614283
      mean
                                                         0.0
      std
             22557.935929 1175.798607
                                            0.090763
                                                                 3.418810
                                           0.000000 2023.0
                 2.000000 111.000000
                                                                1.000000
      min
           19365.000000 1106.000000
      25%
                                           0.000000 2023.0
                                                                4.000000
      50% 39015.000000 2907.000000
                                           0.000000 2023.0
                                                                7.000000
                                           0.000000 2023.0 10.000000
      75% 58555.000000 3201.000000
             81133.000000 3831.000000
                                           1.000000 2023.0 12.000000
      max
                     HOUR Severe_crimes
      count 61877.000000 61877.000000
      mean
               12.500558
                               0.066632
      std
                              0.249386
                6.569869
               0.000000
                              0.000000
      min
      25%
                8.000000
                              0.000000
      50%
                13.000000
                              0.000000
      75%
                18.000000
                              0.000000
      max
                23.000000
                               1.000000
In [ ]:
In [ ]: # do a percentage of non-severe crimes by day of the week
        percentage_non_severe_crimes = []
        for day in ["Monday", "Tuesday", "Wednesday", "Thursday", "Friday", "Saturday",
            percentage_non_severe_crimes.append(train_data[train_data["DAY_OF_WEEK"] ==
        plt.bar(["Monday", "Tuesday", "Wednesday", "Thursday", "Friday", "Saturday", "Su
        plt.xlabel("Day of the week")
        plt.ylabel("Percentage of non-severe crimes")
        # save the plot
        plt.savefig("../reports/figures/percentage_non_severe_crimes_by_day_of_week.png"
        plt.show()
```



## Feature Engineering & Feature Selection

```
In [ ]: from sklearn.preprocessing import LabelEncoder
        # read in train, test, and val data
        train = pd.read_csv('../data/processed/train_data.csv')
        test = pd.read_csv('../data/processed/test_data.csv')
        val = pd.read_csv('../data/processed/val_data.csv')
In [ ]: # remove id column
        train = train.drop(columns=['_id'])
        # find duplicate rows
        print(train.duplicated().sum())
       173
In [ ]: # remove duplicate rows
        train = train.drop_duplicates()
        # find duplicate rows
        print(train.duplicated().sum())
       0
In [ ]: # remove columns that are not needed
        # remove REPORTING AREA, SHOOTING
        train = train.drop(columns=['REPORTING_AREA', 'SHOOTING'])
        train.head()
```

```
OFFENSE CODE OFFENSE DESCRIPTION DISTRICT OCCURRED ON DATE
Out[ ]:
                                 PROPERTY - LOST/
                                                                       2023-11-28
        0
                     3201
                                                         A1
                                                                                    2023
                                                                      00:00:00+00
                                         MISSING
                                                                       2023-05-30
                     1402
                                      VANDALISM
                                                                                    2023
        1
                                                         Α1
                                                                      00:09:00+00
                                        FORGERY /
                                                                       2023-06-15
        2
                     1001
                                                        E13
                                                                                    2023
                                                                      09:21:00+00
                                 COUNTERFEITING
                                                                       2023-01-19
                                      VANDALISM
        3
                     1402
                                                          E5
                                                                                    2023
                                                                       10:31:00+00
                                                                       2023-10-16
                     1831
                                       SICK ASSIST
                                                                                    2023
        4
                                                          B3
                                                                       10:01:00+00
In [ ]:
        # remove YEAR since its all 2023
        train = train.drop(columns=['YEAR'])
In [ ]:
        # change OCCURRED_ON_DATE to datetime
        train['OCCURRED_ON_DATE'] = pd.to_datetime(train['OCCURRED_ON_DATE'])
In [ ]: # change day of week to numbers monday = 0, sunday = 6
        train['DAY_OF_WEEK'] = train['OCCURRED_ON_DATE'].dt.dayofweek
        train.head()
        # remove year from OCCURRED_ON_DATE
        train['OCCURRED_ON_DATE'] = train['OCCURRED_ON_DATE'].dt.strftime('%m-%d')
       # Remove spaces in OFFENSE_DESCRIPTION
In [ ]:
        train['OFFENSE_DESCRIPTION'] = train['OFFENSE_DESCRIPTION'].str.replace(' ', '')
        remove = ['-', '(', ')', '/']
        for r in remove:
            train['OFFENSE_DESCRIPTION'] = train['OFFENSE_DESCRIPTION'].str.replace(r,
In [ ]: # do the same for test and val data
        test = test.drop(columns=['REPORTING_AREA', 'SHOOTING', 'YEAR'])
        val = val.drop(columns=['REPORTING AREA', 'SHOOTING', 'YEAR'])
        # change OCCURRED ON DATE to datetime
        test['OCCURRED_ON_DATE'] = pd.to_datetime(test['OCCURRED_ON_DATE'])
        val['OCCURRED_ON_DATE'] = pd.to_datetime(val['OCCURRED_ON_DATE'])
        # change day of week to numbers monday = 0, sunday = 6
        test['DAY OF WEEK'] = test['OCCURRED ON DATE'].dt.dayofweek
        val['DAY_OF_WEEK'] = val['OCCURRED_ON_DATE'].dt.dayofweek
        # remove year from OCCURRED_ON_DATE
        test['OCCURRED_ON_DATE'] = test['OCCURRED_ON_DATE'].dt.strftime('%m-%d')
        val['OCCURRED_ON_DATE'] = val['OCCURRED_ON_DATE'].dt.strftime('%m-%d')
        # Remove spaces in OFFENSE DESCRIPTION
        test['OFFENSE_DESCRIPTION'] = test['OFFENSE_DESCRIPTION'].str.replace(' ', '')
        val['OFFENSE_DESCRIPTION'] = val['OFFENSE_DESCRIPTION'].str.replace(' ', '')
        for r in remove:
```

```
test['OFFENSE_DESCRIPTION'] = test['OFFENSE_DESCRIPTION'].str.replace(r, '')
            val['OFFENSE_DESCRIPTION'] = val['OFFENSE_DESCRIPTION'].str.replace(r, '')
        # check the data types of each column
        train.dtypes
Out[]: OFFENSE_CODE
                                int64
        OFFENSE_DESCRIPTION
                               object
        DISTRICT
                               object
        OCCURRED_ON_DATE
                               object
        MONTH
                                int64
                                int32
        DAY_OF_WEEK
        HOUR
                                int64
        STREET
                               object
        Severe_crimes
                               int64
        dtype: object
In [ ]: # remove street column
        train = train.drop(columns=['STREET'])
        test = test.drop(columns=['STREET'])
        val = val.drop(columns=['STREET'])
In [ ]: # encode all non-numeric columns
        # reset the label encoder
        le_description = LabelEncoder()
        le_district = LabelEncoder()
        # fit and transform the label encoder
        train['OFFENSE_DESCRIPTION'] = le_description.fit_transform(train['OFFENSE_DESCR
        train['DISTRICT'] = le_district.fit_transform(train['DISTRICT'])
In [ ]: train.head()
Out[ ]:
           OFFENSE CODE OFFENSE DESCRIPTION DISTRICT OCCURRED ON DATE MON'
        0
                     3201
                                               87
                                                           0
                                                                            11-28
        1
                     1402
                                              110
                                                                            05-30
        2
                                               32
                                                           9
                     1001
                                                                            06-15
        3
                     1402
                                              110
                                                          11
                                                                            01-19
                                                                            10-16
        4
                     1831
                                               96
                                                           4
In [ ]: # do the same for test and val data
        # add new label if the value is unseen in the training data
        test['OFFENSE_DESCRIPTION'] = test['OFFENSE_DESCRIPTION'].map(lambda s: '<unknow</pre>
        val['OFFENSE_DESCRIPTION'] = val['OFFENSE_DESCRIPTION'].map(lambda s: '<unknown>
        # add new label if the value is unseen in the training data
        test['DISTRICT'] = test['DISTRICT'].map(lambda s: '<unknown>' if s not in le dis
        val['DISTRICT'] = val['DISTRICT'].map(lambda s: '<unknown>' if s not in le_distr
        # add <unknown> to the classes
        le_description.classes_ = np.append(le_description.classes_, '<unknown>')
        le_district.classes_ = np.append(le_district.classes_, '<unknown>')
```

```
# transform the label encoder
        test['OFFENSE_DESCRIPTION'] = le_description.transform(test['OFFENSE_DESCRIPTION
        val['OFFENSE_DESCRIPTION'] = le_description.transform(val['OFFENSE_DESCRIPTION']
        test['DISTRICT'] = le_district.transform(test['DISTRICT'])
        val['DISTRICT'] = le_district.transform(val['DISTRICT'])
        # check the data types of each column
        train.dtypes
Out[]: OFFENSE_CODE
                                int64
        OFFENSE_DESCRIPTION int32
        DISTRICT
                              int32
        OCCURRED_ON_DATE object
        MONTH
                              int64
                              int32
        DAY_OF_WEEK
        HOUR
                              int64
                              int64
        Severe_crimes
        dtype: object
In [ ]: # save the processed data
        train.to_csv('.../data/processed/train_data_processed.csv', index=False)
        test.to_csv('../data/processed/test_data_processed.csv', index=False)
        val.to_csv('../data/processed/val_data_processed.csv', index=False)
In [ ]: # add synthetic data to the training data
        # add 1000 rows of synthetic data
        synthetic_data = train.sample(n=1000, replace=True)
        train = pd.concat([train, synthetic_data])
        # show number of unique values in each column
        train.nunique()
Out[]: OFFENSE_CODE
                               115
        OFFENSE DESCRIPTION
                               116
        DISTRICT
                               14
        OCCURRED_ON_DATE
                             365
        MONTH
                               12
        DAY_OF_WEEK
                                7
        HOUR
                                24
        Severe_crimes
        dtype: int64
In [ ]: # count the number of 1 and 0 in the Severe crimes column
        train['Severe crimes'].value counts()
Out[]: Severe_crimes
           58527
             4177
        Name: count, dtype: int64
```

## Machine Learning Model Selection

```
In [ ]: # imports
   import pandas as pd
   import numpy as np
```

```
import joblib
        from sklearn.preprocessing import LabelEncoder
        from sklearn.ensemble import RandomForestClassifier
In [ ]: # load in train, test, val data
        test = pd.read_csv('../data/processed/test_data_processed.csv')
        train = pd.read_csv('../data/processed/train_data_processed.csv')
        val = pd.read_csv('../data/processed/val_data_processed.csv')
        le_datetime = joblib.load('../models/datetime_encoder.pkl')
        # encode datetime
        test['OCCURRED_ON_DATE'] = le_datetime.transform(test['OCCURRED_ON_DATE'])
        train['OCCURRED_ON_DATE'] = le_datetime.transform(train['OCCURRED_ON_DATE'])
        val['OCCURRED_ON_DATE'] = le_datetime.transform(val['OCCURRED_ON_DATE'])
In [ ]: # remove _id column
        val = val.drop('_id', axis=1)
        test = test.drop('_id', axis=1)
In [ ]: # drop DISTRICT column
        val = val.drop('DISTRICT', axis=1)
        test = test.drop('DISTRICT', axis=1)
        train = train.drop('DISTRICT', axis=1)
In [ ]: # define X and y
        test_target = test['Severe_crimes']
        test_features = test.drop('Severe_crimes', axis=1)
        val_target = val['Severe_crimes']
        val_features = val.drop('Severe_crimes', axis=1)
        train_target = train['Severe_crimes']
        train_features = train.drop('Severe_crimes', axis=1)
In [ ]: # train model
        rf = RandomForestClassifier(n_estimators=1000, max_depth=10, random_state=42)
        rf.fit(train_features, train_target)
Out[ ]: ▼
                                   RandomForestClassifier
        RandomForestClassifier(max depth=10, n estimators=1000, random state=4
        2)
In [ ]: # save model
        joblib.dump(rf, '../models/rf_model_week11.pkl')
        print('model saved')
       model saved
In [ ]: # read in other encoder
        # Load other encoders
        le_district = joblib.load('../models/le_district.pkl')
        le_description = joblib.load('../models/le_description.pkl')
```

```
# save train, test, val data with pickle
joblib.dump(train_features, '../data/processed/train_features.pkl')
joblib.dump(train_target, '../data/processed/train_target.pkl')

joblib.dump(val_features, '../data/processed/val_features.pkl')

joblib.dump(val_target, '../data/processed/val_target.pkl')

joblib.dump(test_features, '../data/processed/test_features.pkl')

joblib.dump(test_target, '../data/processed/test_target.pkl')

print('data saved')
```

data saved

```
In []: # read in other encoder
# load other encoders
le_district = joblib.load('../models/le_district.pkl')
le_description = joblib.load('../models/le_description.pkl')

# save train, test, val data with pickle
joblib.dump(train_features, '../data/processed/train_features.pkl')
joblib.dump(train_target, '../data/processed/train_target.pkl')

joblib.dump(val_features, '../data/processed/val_features.pkl')
joblib.dump(test_features, '../data/processed/val_target.pkl')

joblib.dump(test_features, '../data/processed/test_features.pkl')
joblib.dump(test_target, '../data/processed/test_target.pkl')

print('data saved')
```

data saved