## m-crime-da-notebook

September 8, 2023

# 1 Montreal Crime Data Analysis

#### 1.0.1 Step 1: Install + Import Necessary Libraries

```
[]: # !pip install pandas
# !pip install numpy
# !pip install matplotlib
# !pip install seaborn
[1]: import os
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
```

#### 1.0.2 Step 2: Reading + Exploring Data

```
[2]: data = pd.read_csv('mtl-crime-data.csv')
[3]:
     data.head()
[3]:
       Unnamed: 0
                               category
                                               date postal_code
                                                                     city \
     0
                   Motor vehicle theft
                                        2018-09-13
                                                        H1Z 1S9
                                                                 MONTREAL
                    Motor vehicle theft 2018-04-30
                                                        H1Z 1S9 MONTREAL
     1
                 1
                 2
     2
                          Home Invasion 2018-01-10
                                                        H1Z 2V6
                                                                 MONTREAL
     3
                 3
                               Mischief 2018-11-12
                                                        H1Z 2V6 MONTREAL
     4
                 4
                               Mischief 2018-08-15
                                                        H1Z 2V6 MONTREAL
      neighbourhood year
                            count
                                   longitude
                                              latitude
     O Saint-Michel
                      2018
                                1
                                     -73.626
                                                45.567
     1 Saint-Michel 2018
                                1
                                     -73.626
                                                45.567
     2 Saint-Michel 2018
                                1
                                     -73.629
                                                45.569
     3 Saint-Michel 2018
                                1
                                     -73.629
                                                45.569
     4 Saint-Michel 2018
                                1
                                     -73.629
                                                45.569
[4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 136642 entries, 0 to 136641
Data columns (total 10 columns):
    Column
                   Non-Null Count
                                    Dtype
    _____
                   -----
                                    ____
 0
    Unnamed: 0
                   136642 non-null int64
 1
    category
                   136642 non-null object
 2
    date
                   136642 non-null object
 3
    postal_code
                   136642 non-null object
 4
    city
                   136642 non-null object
 5
    neighbourhood 136642 non-null object
 6
                   136642 non-null int64
    year
 7
    count
                   136642 non-null int64
 8
                   136642 non-null float64
    longitude
    latitude
                   136642 non-null float64
dtypes: float64(2), int64(3), object(5)
memory usage: 10.4+ MB
```

#### 1.0.3 Step 3: Cleaning + Preprocessing

```
[5]: data.columns
```

**Step 3.1: Removing unwanted columns** We will drop the "Unnamed" column as it is just used for indexing purposes, and doesn't include any values.

[8]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 136642 entries, 0 to 136641
Data columns (total 9 columns):

| # | Column      | Non-Null Count  | Dtype  |
|---|-------------|-----------------|--------|
|   |             |                 |        |
| 0 | category    | 136642 non-null | object |
| 1 | date        | 136642 non-null | object |
| 2 | postal code | 136642 non-null | object |

```
3
    city
                   136642 non-null object
 4
    neighbourhood 136642 non-null object
 5
                   136642 non-null int64
    year
 6
    count
                   136642 non-null int64
 7
    longitude
                   136642 non-null float64
    latitude
                   136642 non-null float64
dtypes: float64(2), int64(2), object(5)
memory usage: 9.4+ MB
```

Step 3.2: Converting datatypes to required format "Date" column is currently in string format, and we need it to be in datetime format in order to perform analysis.

```
[9]: data['date'] = pd.to_datetime(data['date'])
[10]: pd.DataFrame(data.dtypes, columns=['Datatype']).rename_axis('Columns')
[10]:
                            Datatype
      Columns
      category
                              object
      date
                     datetime64[ns]
      postal_code
                              object
                              object
      city
      neighbourhood
                              object
      year
                               int64
      count
                               int64
      longitude
                             float64
      latitude
                             float64
     Step 3.3: Checking for null values
```

```
[11]: pd.DataFrame(data.isnull().sum(), columns=['Missing Values']).
       ⇔rename_axis('Feature')
```

```
[11]:
                       Missing Values
      Feature
      category
                                     0
                                     0
      date
      postal_code
                                     0
      city
                                     0
                                     0
      neighbourhood
      year
                                     0
      count
                                     0
      longitude
                                     0
      latitude
```

Dataset doesn't include null values.

#### 1.0.4 Step 4: Data Analysis + Visualization

Question 1: Which neighbourhoods have the highest number of reported crimes?

```
[12]: top_neighbourhood = pd.DataFrame(data['neighbourhood'].value_counts()).

⇔rename({"neighbourhood":"Case Reported"}, axis = 1).

⇔rename_axis("Neighbourhood").head(10)

top_neighbourhood.style.bar()
```

[12]: <pandas.io.formats.style.Styler at 0x2029baf6150>

```
top_neighbourhood = top_neighbourhood.sort_values(by='count', ascending=False).

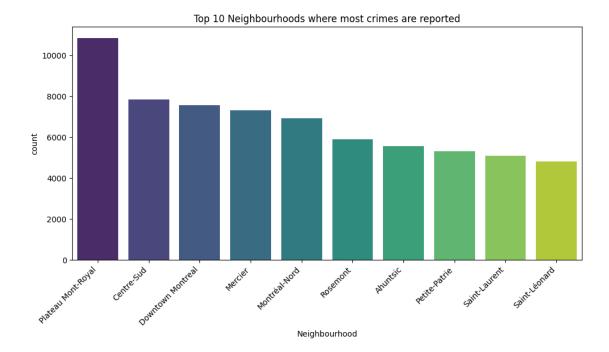
_head(10)

plt.figure(figsize=(10, 6))
sns.barplot(data=top_neighbourhood, x=top_neighbourhood.index, y='count',
_palette='viridis')

plt.xlabel('Neighbourhood')
plt.ylabel('count')
plt.title('Top 10 Neighbourhoods where most crimes are reported')

# Rotating x-axis labels for better visibility
plt.xticks(rotation=45, ha='right')

# Display the plot
plt.tight_layout()
plt.show()
```

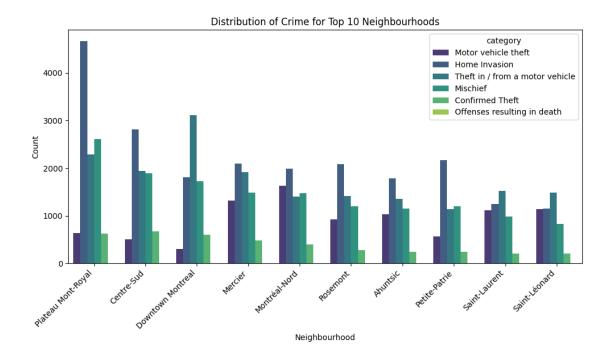


Analysis shows that the neighbourhoods with the highest number of reported crimes in Montreal between 2015 - 2021 are:

- 1. Plateau Mont-Royal (total of 10844 reported crimes)
- 2. Centre-Sud (total of 7849 reported crimes)
- 3. Downtown Montreal (total of 7565 reported crimes)
- 4. Mercier (total of 7302 reported crimes)
- 5. Montréal-Nord (total of 6913 reported crimes)
- 6. Rosemont (total of 5907 reported crimes)
- 7. Ahuntsic (total of 5575 reported crimes)
- 8. Petite-Patrie (total of 5327 reported crimes)
- 9. Saint-Laurent (total of 5094 reported crimes)
- 10. Saint-Léonard (total of 4822 reported crimes)

#### Question 2: What are the most frequent crimes by neighbourhood?

```
[16]: # Calculate the top 10 neighbourhoods with highest crime rate
      top_neighbourhoods = data['neighbourhood'].value_counts().head(10).index
      # Filter the data to include only the top 10 neighbourhoods
      data_top_neighbourhoods = data[data['neighbourhood'].isin(top_neighbourhoods)]
      # Get the order of neighborhoods by crime count in descending order
      neighbourhood_order = data_top_neighbourhoods.
       -groupby('neighbourhood')['category'].count().sort values(ascending=False).
       ⇔index
      # Create a count plot using Seaborn
      plt.figure(figsize=(10, 6))
      sns.countplot(data=data_top_neighbourhoods, x='neighbourhood', hue='category', u
       ⇒palette='viridis', order=neighbourhood_order)
      # Adding labels and title
      plt.xlabel('Neighbourhood')
      plt.ylabel('Count')
      plt.title('Distribution of Crime for Top 10 Neighbourhoods')
      # Slightly rotate the x-axis labels for readability
      plt.xticks(rotation=45, ha='right')
      # Display the plot
      plt.tight_layout()
      plt.show()
```



Home Invasions are the most reported type of crime in all neighbourhoods, except in **Downtown**, Saint-Laurent, and Saint Léonard; where Theft In/From a Vehicle has a higher rate.

Question 3: Number of reported crimes committed annualy in Montreal

| [17]: |      | count |
|-------|------|-------|
|       | Year |       |
|       | 2015 | 24222 |
|       | 2016 | 23407 |
|       | 2017 | 22611 |
|       | 2018 | 19419 |
|       | 2019 | 18650 |
|       | 2020 | 17786 |
|       | 2021 | 10547 |

By default, the value\_counts() method sorts values in descending order, so we can see that 2015 had the highest number of reported crimes (24,222 counts), while 2021 had the lowest number of reported crimes (10547).

## Question 4: What are the general trends in the reported crimes between 2015-2021

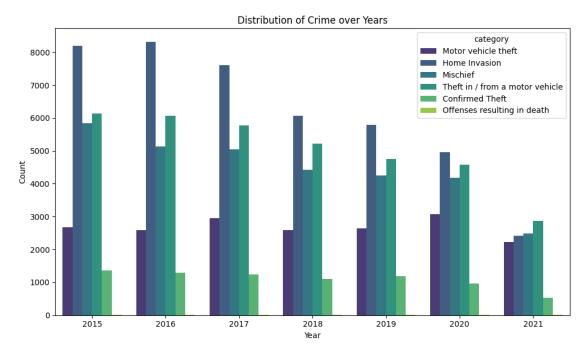
- What are the trends in the categories of reported crimes?
- Did crime rates increase or decrease?

#### 1. What are the trends in the categories of reported crimes?

```
[18]: plt.figure(figsize=(10, 6))
    sns.countplot(data=data, x='year', hue='category', palette='viridis')

# Adding labels and title
    plt.xlabel('Year')
    plt.ylabel('Count')
    plt.title('Distribution of Crime over Years')

# Display the plot
    plt.tight_layout()
    plt.show()
```



**Home Invasions** are the most reported crime from 2015-2020. In 2021 **Confirmed Theft** was the most reported crime.

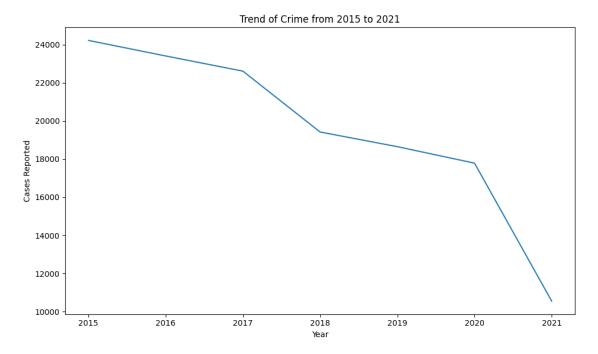
Theft In/From a Motor Vehicle is the second most reported crime in Montreal from 20215-2021, followed by Mischief.

## 2. Did crime rates increase or decrease?

```
sns.lineplot(data=year_wise_trend, x=year_wise_trend.index, y='Case Reported')

# Adding labels and title
plt.xlabel('Year')
plt.ylabel('Cases Reported')
plt.title('Trend of Crime from 2015 to 2021')

# Display the plot
plt.tight_layout()
plt.show()
```



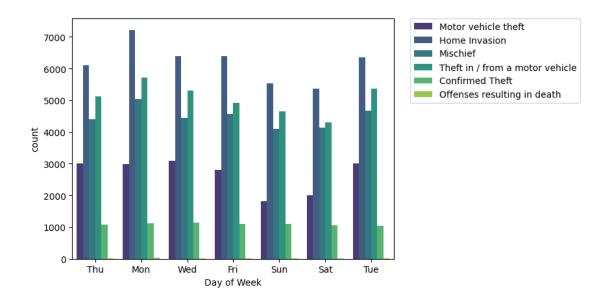
We can see that crime reports **decreased** from 2015-2021.

Question 5: On what days and months are a higher number of crimes reported? Let's explore the types of crime committed depending on the day of the week.

```
[20]: data['Month'] = data['date'].apply(lambda time: time.month)
   data['Day of Week'] = data['date'].apply(lambda time: time.dayofweek)

[21]: dmap = {0:'Mon',1:'Tue',2:'Wed',3:'Thu',4:'Fri',5:'Sat',6:'Sun'}
   data['Day of Week'] = data['Day of Week'].map(dmap)

[22]: sns.countplot(x='Day of Week',data=data,hue='category',palette='viridis')
   plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```



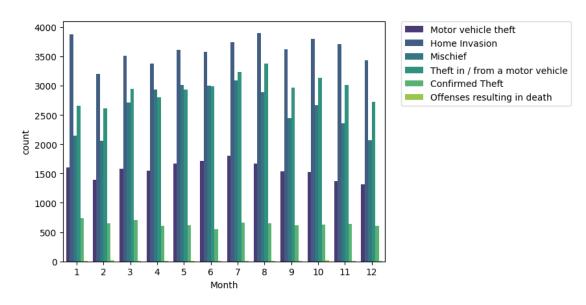
We can see that **Monday** has the highest rate of *Home Invasions*.

Saturday and Sunday have the lowest number of Motor Vehicle Thefts.

Let's explore the types of crime committed depending on the month of the year.

```
[23]: sns.countplot(x='Month', data=data, hue='category', palette='viridis') plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```

[23]: <matplotlib.legend.Legend at 0x202a3931d90>



January and August have the highest numbers of *Home Invasions*.

July and August have the highest numbers of Confirmed Thefts.

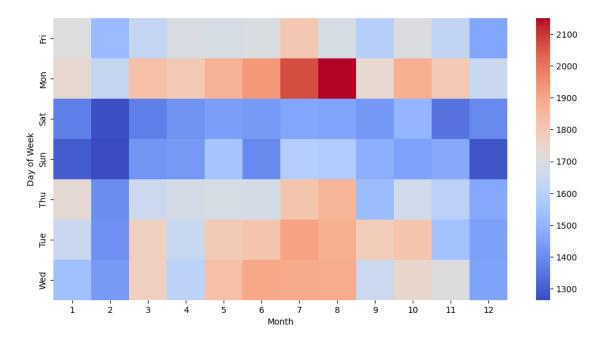
Let's explore the overall correlation between the time of year and the reported number of crimes between 2015-2021.

Which days/months have the highest number of total reported crimes?

```
[24]: data.head()
[24]:
                                     date postal_code
                                                            city neighbourhood year
                     category
         Motor vehicle theft 2018-09-13
                                              H1Z 1S9
                                                        MONTREAL
                                                                   Saint-Michel
                                                                                  2018
      0
      1
         Motor vehicle theft 2018-04-30
                                              H1Z 1S9
                                                        MONTREAL Saint-Michel
                                                                                  2018
      2
               Home Invasion 2018-01-10
                                              H1Z 2V6
                                                        MONTREAL
                                                                   Saint-Michel
                                                                                  2018
      3
                     Mischief 2018-11-12
                                              H1Z 2V6
                                                        MONTREAL
                                                                   Saint-Michel
                                                                                  2018
      4
                     Mischief 2018-08-15
                                              H1Z 2V6
                                                        MONTREAL Saint-Michel
                                                                                  2018
                 longitude
                            latitude
                                       Month Day of Week
         count
      0
                   -73.626
                               45.567
                                           9
              1
                                                      Thu
                   -73.626
                                           4
      1
              1
                              45.567
                                                      Mon
      2
              1
                   -73.629
                              45.569
                                           1
                                                      Wed
      3
                   -73.629
              1
                              45.569
                                           11
                                                      Mon
      4
              1
                   -73.629
                               45.569
                                           8
                                                      Wed
[25]: #correlation between days/months
      dayMonth = data.groupby(by=['Day of Week', 'Month']).count()['category'].

unstack()
      dayMonth.head()
[25]: Month
                      1
                            2
                                   3
                                                5
                                                      6
                                                            7
                                                                                10
                                         4
                                                                   8
                                                                         9
                                                                                      11
      Day of Week
      Fri
                    1706
                          1519
                                 1630
                                       1692
                                             1685
                                                    1694
                                                          1809
                                                                 1685
                                                                       1590
                                                                              1698
                                                                                    1624
      Mon
                    1742
                          1636
                                 1830
                                       1803
                                             1872
                                                    1931
                                                          2068
                                                                 2150
                                                                       1743
                                                                              1879
                                                                                    1802
                                                                       1425
      Sat
                    1372
                          1275
                                 1371
                                       1417
                                              1443
                                                    1428
                                                          1459
                                                                 1456
                                                                              1506
                                                                                    1345
      Sun
                    1299
                          1265
                                 1420
                                       1428
                                              1550
                                                    1394
                                                                 1578
                                                                       1487
                                                                              1450
                                                                                    1470
                                                          1587
                                       1679
                                              1685
      Thu
                    1734
                          1403
                                 1652
                                                    1674
                                                          1818
                                                                 1861
                                                                       1521
                                                                              1666
                                                                                    1607
      Month
                      12
      Day of Week
      Fri
                    1460
      Mon
                    1647
      Sat
                    1394
      Sun
                    1285
      Thu
                    1466
[26]: plt.figure(figsize=(12,6))
      sns.heatmap(dayMonth,cmap='coolwarm')
```

[26]: <Axes: xlabel='Month', ylabel='Day of Week'>



Overall, **Monday** has the *highest* number of reported crimes in the period between 2015-2021, followed by **Tuesday** and **Wednesday**.

We can also see that **June**, **July** and **August** have *higher* numbers of reported crimes.

#### 2015 - 2021 Crime Data Analysis Final Results

- Neighborhood Insights:
  - Plateau
  - Centre-Sud
  - Downtown
- Crime Trends:
  - Mondays have the highest number of reported crimes, followed by Tuesday and Wednesday.
  - August is the month with the most number of reported crimes, followed by July and June.
  - Home invasions are the most reported crime, followed by theft in/from a motor vehicle and mischief.
- Yearly Statistics:
  - -2015 had the highest crime rate (24,222 cases), and 2021 the lowest (10,547 cases).
  - Overall, crime reports decreased every year from 2015 to 2021.