Montreal City Crime Data Analysis and Visualization

In this notebook, I analyzed Montreal city crime data (2015 - 2017) with a goal of providing answers to the following main questions among other findings noted in the analysis:

- 1. What are the top 3 prevalent crimes or offenses committed in 2015, 2016 and 2017 in Montreal City?
- 2. Which neighborhoods recorded the highest crime incidents in 2015, 2016 and 2017 and what are the crime types in these neighborhoods?
- 3. Which neighborhood has the highest cases of murder crime in 2015, 2016 and 2017?
- 4. What time of the day did most crime incidents occur in 2015, 2016 and 2017?
- 5. Which top 5 police stations (PDQ) got the most crime complaints in 2015, 2016 and 2017?
- 6. Which are the top 3 PDQs that got least crime complaints in 2015, 2016 and 2017?

Data source and descriptions

The dataset used in this analysis contains the criminal acts or crimes registered by the Police Department of the City of Montreal (SPVM) and made available on Montreal Open Data Portal (http://donnees.ville.montreal.qc.ca/dataset/actes-criminels)).

Import packages and define settings

```
In [134]:
           import pandas as pd
             import numpy as np
             from dateutil.parser import parse
             import calendar
             import folium
             from folium import plugins
             from IPython.display import display html, HTML
             from geopy.geocoders import Nominatim
             import json
             # Plotly packages
             import plotly
             from plotly import tools
             import plotly.graph_objs as go
             from plotly.offline import init_notebook_mode, iplot
             # settings
             init_notebook_mode(connected=True)
             pd.set_option('display.max_colwidth', 130)
```

```
In [147]: ► #crimedata functions
              # define chart marker colors in two lists
              markercol = ['rgba(31, 119, 180, 0.5)', 'rgba(255, 127, 14, 0.5)',
                           'rgba(50, 171, 96, 0.5)', 'rgba(214, 39, 40, 0.5)',
                          'rgba(148, 103, 189, 0.5)', 'rgba(140, 86, 75, 0.5)']
              linecol = ['rgba(31, 119, 180, 1.0)', 'rgba(255, 127, 14, 1.0)',
                          'rgba(50, 171, 96, 1.0)', 'rgba(214, 39, 40, 1.0)',
                          'rgba(148, 103, 189, 1.0)', 'rgba(140, 86, 75, 1.0)']
              def map_data(mappings, x):
                  '''a function to map columns descriptions in French to English'''
                  for i, j in mappings:
                      if i == x:
                          return j
              def embed map(map):
                  '''a function to embed map in notebook '''
                  map.save(outfile="map.html")
                  return HTML('<iframe src="{i}" style="width: 100%; height: 510px; border: none"></iframe>
              def generate_map(df, yr):
                  '''a function for creating an interactive map'''
                  # exclude null location values
                  ref = df[(df['YEAR'] == yr) & (df['COORDS'] != (1.0, 1.0))].copy()
                  # create base map
                  crimemap = folium.Map(location=[ref['LAT'].mean(), ref['LON'].mean()], zoom start=11)
                  # create an instance of marker cluster for crimes in the dataset
                  crimes = plugins.MarkerCluster().add_to(crimemap)
                  # Loop through the dataset and add each crime point to the marker cluster
                  for lat, lon, category in zip(ref['LAT'], ref['LON'], ref['ADAPTED_CATEGORY']):
                      folium.Marker(location=[lat, lon], icon=None, popup=category).add_to(crimes)
                  return embed_map(crimemap)
              def extract address(col):
                  '''a function to extract addresses from latitudes and longitudes'''
                  coord = list(col)
                  slist = []
                  # Set a custom user_agent for the Nominatim geocoder
                  geolocator = Nominatim(user_agent="my-custom-application")
                  for i in coord:
                      jlist = []
                      location = geolocator.reverse(i, timeout=10)
                      # Convert Location to JSON string and then to a dictionary
                      json_string = json.dumps(location.raw)
                      dat = json.loads(json_string)
                      # Extract neighborhood address
                      for j in dat['address'].keys():
                          if j not in ['house_number', 'city', 'region', 'state', 'postcode', 'country', 'ce
                              jlist.append(dat['address'][j])
                      locstr = ", ".join(jlist)
                      slist.append(locstr)
                  return slist
```

```
def plotchart(chdata, chlayout, titlelist, yearlist, subtitlelist):
    '''a function to plot a chart with 1 row and 3 columns figure'''
   # define subplots
   fig = tools.make_subplots(rows=1, cols=3,
        subplot_titles=(["<b>{}</b>".format(i) for i in subtitlelist]),
        shared_yaxes=True,horizontal_spacing=(0.05),print_grid=False)
   # an empty list to hold chart data definitions for each plot
   trace_list = []
   for i in range(3):
        data = chdata['trace_data'][chdata['trace_data']['YEAR'] == yearlist[i]]
        tracex = go.Bar(x=data[chdata['x']], y=data[chdata['y']], name=titlelist[i], width=0.1
            text=data[chdata['y']], textposition='outside', hoverinfo='text',
            outsidetextfont=dict(size=10), cliponaxis=False,
            marker=dict(color=markercol[:3][i], line=dict(color=linecol[:3][i], width=1)))
        trace_list.append(tracex)
        # define each subplot order of selection
        m = np.array([1, 1, 1, 2, 1, 3]).reshape(3, 2)
        # append each subplot data definitions to the figure instance
        fig.append_trace(trace_list[i], m[i][0], m[i][1])
   # define layout settings
   for i in fig['layout']['annotations']:
        i['font'] = dict(size=12)
   for i in range(1, 4):
        fig['layout']['yaxis' + '{}'.format(i)].update(title=chlayout['yaxistitle'],
            titlefont=dict(size=12, color='rgb(107, 107, 107)'), showticklabels=False, showgri
        fig['layout']['xaxis' + '{}'.format(i)].update(titlefont=dict(size=11, color='rgb(107)
            tickfont=dict(size=11, color='rgb(107, 107, 107)'), tickangle=chlayout['tickangle
   # update layout settings
   fig['layout'].update(height=chlayout['height'], width=chlayout['width'], showlegend=False
        autosize=False, title=chlayout['title'], titlefont=dict(size=14),
        paper_bgcolor='rgba(245, 246, 249, 1)', plot_bgcolor='rgba(245, 246, 249, 1)')
   return iplot(fig)
```

Load the data

Out[148]:

| | CATEGORIE | DATE | QUART | PDQ | X | Υ | LONGITUDE | LATITUDE |
|--------|-------------------------------------|----------------|-------|------|---------------|--------------|------------|-----------|
| 12763 | Introduction | 2015-01- 01 | soir | 27.0 | 293203.472992 | 5.045436e+06 | -73.648516 | 45.548740 |
| 41627 | Introduction | 2015-01- 01 | jour | 16.0 | 299612.000006 | 5.036898e+06 | -73.566352 | 45.471990 |
| 56091 | Vol dans / sur véhicule à moteur | 2015-01- 01 | jour | 15.0 | 298038.765999 | 5.034604e+06 | -73.586441 | 45.451332 |
| 1782 | Vols qualifiés | 2015-01- 01 | soir | 16.0 | 299173.405992 | 5.035151e+06 | -73.571941 | 45.456266 |
| 54900 | Vol dans / sur véhicule à moteur | 2015-01- 01 | nuit | 20.0 | NaN | NaN | NaN | NaN |
| | | | | | | | | |
| 225190 | Vol dans / sur véhicule à moteur | 2022-12- 31 | jour | 10.0 | 287477.098001 | 5.043502e+06 | -73.721784 | 45.531224 |
| 225191 | Vol de véhicule à moteur | 2022-12- 31 | jour | 31.0 | 295686.990996 | 5.045723e+06 | -73.616715 | 45.551361 |
| 225192 | Vol dans / sur véhicule à moteur | 2022-12- 31 | jour | 21.0 | 300335.540996 | 5.040250e+06 | -73.557129 | 45.502162 |
| 238211 | Vols qualifiés | 2022-12- 31 | nuit | 22.0 | 300687.888991 | 5.043679e+06 | -73.552649 | 45.533012 |
| 229328 | Méfait | 2022-12- 31 | jour | 48.0 | 301719.003998 | 5.048206e+06 | -73.539476 | 45.573752 |

242483 rows × 8 columns

In [149]:

display the first five records of the data
df.head()

Out[149]:

| | CATEGORIE | DATE | QUART | PDQ | x | Υ | LONGITUDE | LATITUDE |
|-------|-------------------------------------|----------------|-------|------|---------------|--------------|------------|-----------|
| 12763 | Introduction | 2015-01- 01 | soir | 27.0 | 293203.472992 | 5.045436e+06 | -73.648516 | 45.548740 |
| 41627 | Introduction | 2015-01- 01 | jour | 16.0 | 299612.000006 | 5.036898e+06 | -73.566352 | 45.471990 |
| 56091 | Vol dans / sur véhicule à moteur | 2015-01- 01 | jour | 15.0 | 298038.765999 | 5.034604e+06 | -73.586441 | 45.451332 |
| 1782 | Vols qualifiés | 2015-01- 01 | soir | 16.0 | 299173.405992 | 5.035151e+06 | -73.571941 | 45.456266 |
| 54900 | Vol dans / sur véhicule à moteur | 2015-01- 01 | nuit | 20.0 | NaN | NaN | NaN | NaN |

In [150]: # determine the number of records in the dataset
print('The dataset contains {0} rows and {1} columns.'.format(df.shape[0], df.shape[1]))

The dataset contains 242483 rows and 8 columns.

Crime categories mappings

Columns such as 'CATEGORIE' for crime types as well as 'QUART' field defining the time of the day when the crime event was reported are in French. However, for the purpose of this analysis, these will be summarized in English.

Define new columns and drop non-useful columns

The date field included in the dataset was parsed and separated into 'YEAR' and 'MONTH' columns. The PDQ column is converted to string and the alphabets 'PDQ' are appended to each value. Columns 'X' and 'Y' are not relevant to the analysis and hence dropped.

Define data scope for the analysis

For the purpose of this analysis, the scope of data will be restricted to period between 2015 and 2017.

```
In [154]: 
# extract only '2015 - 2017' records from the dataset
xdf = df[df['YEAR'] != '2018'].copy()
```

Generate crime map

To generate the map, the latitude and longitude fields in the dataset were first chained together and defined as a new column. This was then used to summarize and aggregate crime events in the same neighborhoods to enable easy mapping.

```
In [155]:  # chain latitudes and longitudes together as a new column
    xdf['COORDS'] = list(zip(xdf['LAT'], xdf['LON']))

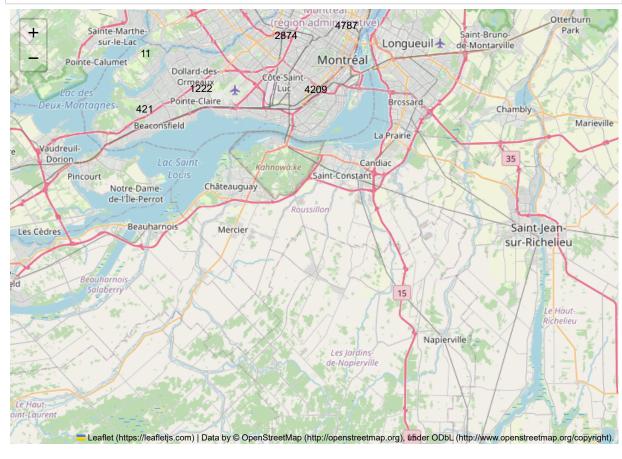
# summarize data
    aggcrime = xdf.groupby(['COORDS', 'ADAPTED_CATEGORY', 'YEAR']).agg({'ADAPTED_CATEGORY':'count}

# split 'COORDS' into two columns
    aggcrime['LAT'] = aggcrime['COORDS'].apply(lambda x: str(x).split(",")[0].replace("(", "")).ascaggcrime['LON'] = aggcrime['COORDS'].apply(lambda x: str(x).split(",")[1].replace(")", "")).ascaggcrime['LON'] = aggcrime['COORDS'].apply(lambda x: str(x).split(",")[1].replace(")", "")).ascaggcrime['COORDS'].apply(lambda x: str(x).split(",")[1].replace(")", "")).ascaggcrime['COORDS'].apply(")
```

In [156]:

a crime map of 2015 is shown here. Maps of 2016 and 2017 can be displayed by changing the ye generate_map(aggcrime, '2015')

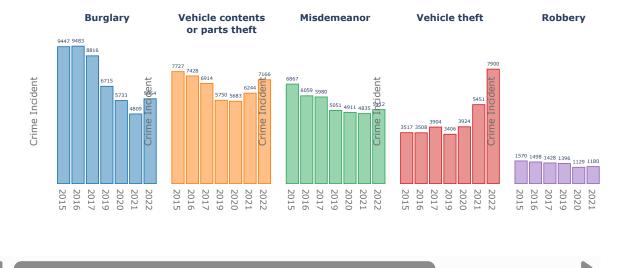
Out[156]:



Plot crime categories distribution

```
In [169]: | # create an empty list to hold chart data definitions for each plot
              trace list = []
              # define subplots
              fig = tools.make_subplots(rows=1, cols=6, subplot_titles=(["<b>{}</b>".format(i) for i in titl
                  shared_yaxes=True, horizontal_spacing=(0.02), print_grid=False)
              # a matrix for subplot selection order
              m = np.array([1, 1, 1, 2, 1, 3, 1, 4, 1, 5, 1, 6]).reshape(6, 2)
              # define chart data
              for i in range(6):
                  data = aggdf[aggdf['ADAPTED CATEGORY'] == titlelist[i]]
                  tracex = go.Bar(x=data['YEAR'], y=data['INCIDENT_COUNT'], text=data['INCIDENT_COUNT'],
                      textposition='outside', hoverinfo='text', outsidetextfont=dict(size=10),
                      cliponaxis=False, name='', showlegend=False, width=0.85,
                      marker=dict(color=markercol[i], line=dict(color=linecol[i], width=1)))
                  trace list.append(tracex)
                  # append each subplot data definitions to the figure instance
                  fig.append trace(trace list[i], m[i][0], m[i][1])
              # define layout settings
              for i in fig['layout']['annotations']:
                  i['font'] = dict(size=11)
                  i['y'] = 1.2
                  i['yanchor'] = 'top'
              fig['layout']['annotations'][1]['text'] = '<b>Vehicle contents<br>or parts theft</b>'
              for i in range(1, 7):
                  fig['layout']['yaxis' + '{}'.format(i)].update(title='Crime Incident',
                      titlefont=dict(size=11, color='rgb(107, 107, 107)'), tickfont=dict(size=10, color='rgb')
                      showticklabels=False, showgrid=True)
                  fig['layout']['xaxis' + '{}'.format(i)].update(titlefont=dict(size=11, color='rgb(107, 107))
                      tickfont=dict(size=10, color='rgb(107, 107, 107)'))
              # update layout settings
              fig['layout'].update(height=350, width=950, showlegend=False, autosize=False,
                  title="<b>Crime Events Distribution: 2015 - 2017</b>", titlefont=dict(size=14),
                  paper bgcolor='rgba(245, 246, 249, 1)', plot bgcolor='rgba(245, 246, 249, 1)')
              iplot(fig)
```

Crime Events Distribution: 2015 - 2017



Observation:

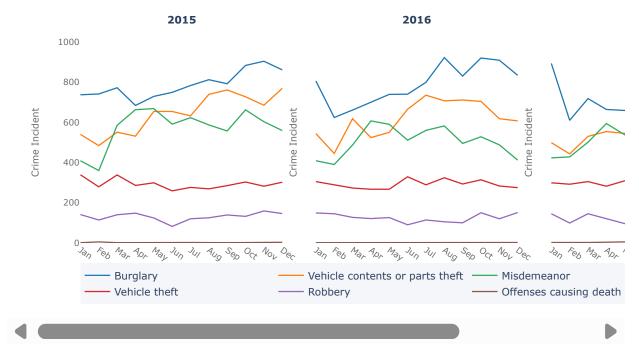
• Besides vehicle theft that increased by 9.4% between 2016 and 2017, all other crime types showed a downward trend in the 3-years period reviewed.

Plot crime incidents trends

```
In [159]: # summarize data by year, month and crime categories
mon_trend = xdf.groupby(['YEAR', 'MONTH', 'ADAPTED_CATEGORY']).agg({'ADAPTED_CATEGORY': 'count
# sort month column and change months in figures to names
mon_trend['MONTH'] = mon_trend['MONTH'].sort_values().apply(lambda x: calendar.month_abbr[x])
```

```
fig = tools.make_subplots(rows=1, cols=3, subplot_titles=(["<b>{}</b>".format(i) for i in year
                 shared_yaxes=True, horizontal_spacing=(0.05), print_grid=False)
             # a matrix for subplot selection order
             m = np.array([1, 1, 1, 2, 1, 3]).reshape(3, 2)
             # define marker colors
             scattcol = ['rgb(31, 119, 180)', 'rgb(255, 127, 14)', 'rgb(50, 171, 96)',
                         'rgb(214, 39, 40)', 'rgb(148, 103, 189)', 'rgb(140, 86, 75)']
             # define chart data
             for i in range(3):
                 trace_list = []
                 data = mon trend[mon trend['YEAR'] == yearlist[i]]
                 for j in range(6):
                     tracex = go.Scatter(x=data[data['ADAPTED CATEGORY'] == titlelist[j]]['MONTH'],
                         y=data[data['ADAPTED_CATEGORY'] == titlelist[j]]['INCIDENT_COUNT'], mode='lines',
                        marker=dict(color=scattcol[j]), line=dict(width=1.5), showlegend=False, name=title
                     if i == 0: # show legend for only the first subplot.
                        tracex.showlegend = True
                     trace_list.append(tracex)
                     # append each subplot data definitions to the figure instance
                     fig.append_trace(trace_list[j], m[i][0], m[i][1])
             # define layout settings
             for i in fig['layout']['annotations']:
                 i['font'] = dict(size=12)
                 i['y'] = 1.07
             fig['layout']['legend'] = dict(orientation="h")
             for i in range(1, 4):
                 fig['layout']['yaxis' + '{}'.format(i)].update(title='Crime Incident',
                     titlefont=dict(size=11, color='rgb(107, 107, 107)'), tickfont=dict(size=10, color='rgb(107, 107)')
                     range=[0, 1000], showgrid=True)
                 fig['layout']['xaxis' + '{}'.format(i)].update(titlefont=dict(size=11, color='rgb(107, 107))
                     tickfont=dict(size=10, color='rgb(107, 107, 107)'), tickangle=35, showgrid=False)
             # update layout settings
             fig['layout'].update(height=420, width=950, showlegend=True, autosize=False,
                 title="<b>Crime Incidents Trends: 2015 - 2017</b>",
                 iplot(fig)
```

Crime Incidents Trends: 2015 - 2017



Observations:

- 1. Apart from from few exceptions noted in certain months, almost all the crime categories exhibited the same trend patterns over the 3-years period.
- 2. Burglary recorded a significant decrease (31%) from 917 in January to 629 in February in 2017.

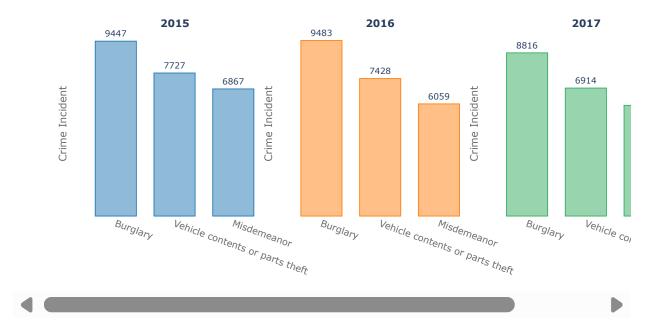
What are the top 3 prevalent crimes or offenses committed in 2015, 2016 and 2017 in Montreal City?

```
In [170]: # summarize data
tot =[aggdf[aggdf['YEAR']==year].sort_values('INCIDENT_COUNT', ascending=False)[:3] for year :
    prev = pd.concat(tot, ignore_index=True)

# display result in crosstab
display(pd.crosstab(index=prev['YEAR'], columns=prev['ADAPTED_CATEGORY'], values=prev['INCIDEN
# define chart data and plot
data = prev
chdata = {'trace_data':data, 'x':'ADAPTED_CATEGORY', 'y':'INCIDENT_COUNT'}
chlayout= {'height':400, 'width':850, 'title':"<b>Top 3 crime types committed in 2015, 2016 ar
plotchart(chdata, chlayout, titlelist, yearlist, subtitlelist=yearlist)
```

| Vehicle theft | Vehicle contents or parts theft | Misdemeanor | Burglary | ADAPTED_CATEGORY | |
|---------------|---------------------------------|-------------|----------|------------------|--|
| | | | | YEAR | |
| NaN | 7727.0 | 6867.0 | 9447.0 | 2015 | |
| NaN | 7428.0 | 6059.0 | 9483.0 | 2016 | |
| NaN | 6914.0 | 5980.0 | 8816.0 | 2017 | |
| NaN | 5750.0 | 5051.0 | 6715.0 | 2019 | |
| NaN | 5683.0 | 4911.0 | 5733.0 | 2020 | |
| 5451.0 | 6244.0 | 4835.0 | NaN | 2021 | |
| 7900 0 | 7166.0 | NaN | 5854 0 | 2022 | |

Top 3 crime types committed in 2015, 2016 and 2017



Observation:

• Burglary, Vehicle contents or parts theft and Misdemeanor are the top three prevalent crimes in the 3-years period.

Which neighborhoods recorded the highest crime incidents in 2015, 2016 and 2017 and what are the crime types in these neighborhoods?

```
In [162]: # exclude records with no location information
top_neighb = xdf[xdf['LAT']!=1.000000]

# summarize data
top = top_neighb.groupby(['YEAR', 'COORDS']).agg({'ADAPTED_CATEGORY': 'count'}).rename(columns)

tot =[top[top['YEAR']==year].sort_values('CRIME_INCIDENT', ascending=False).iloc[:1] for year
topdf = pd.concat(tot, ignore_index=True)

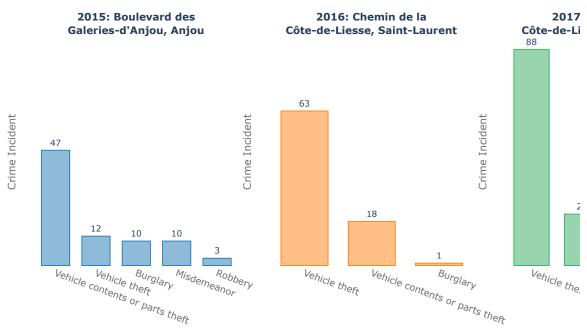
# extract neighborhood addresses as a new column using 'extract_address' function.
topdf['NEIGHBORHOOD'] = extract_address(topdf['COORDS'])

# define list to format subplot headers
klist = ["<b>2015: Boulevard des<br/>br>Galeries-d'Anjou, Anjou</br>
klist = ["<b>2016: Chemin de la<br>
br>Côte-de-Liesse, Saint-Laurent</br>
'<br/>'<b>2017: Chemin de la<br/>br>Côte-de-Liesse, Saint-Laurent</br>
```

| | YEAR | CRIME_INCIDENT | NEIGHBORHOOD |
|---|------|----------------|--------------------------------------------------------------------------------------------------------------------------------|
| 0 | 2015 | 82 | Boulevard des Galeries-d'Anjou, Anjou, Agglomération de Montréal, CA-QC |
| 1 | 2016 | 82 | Chemin de la Côte-de-Liesse, Saint-Laurent, Agglomération de Montréal, CA-QC |
| 2 | 2017 | 118 | Chemin de la Côte-de-Liesse, Saint-Laurent, Agglomération de Montréal, CA-QC |
| 3 | 2019 | 53 | Boulevard des Galeries-d'Anjou, Anjou, Agglomération de Montréal, CA-QC |
| 4 | 2020 | 68 | Boulevard Robert-Bourassa et rue Sainte-Catherine, Boulevard Robert-Bourassa, Quartier des Spectacles, Ville-Marie, Agglomérat |
| 5 | 2021 | 81 | Voie de Service Nord, Pointe-Claire, Agglomération de Montréal, CA-QC |
| 6 | 2022 | 142 | Voie de Service Nord, Pointe-Claire, Agglomération de Montréal, CA-QC |

Crime types in the neighborhoods with highest crime incidents in 2015, 2016 and 201

0



- 1. Boulevard des Galeries-d'Anjou, Anjou and Chemin de la Côte-de-Liesse, Saint-Laurent are neighborhoods with the highest number of crimes in 2015, 2016 and 2017.
- 2. Vehicle theft and vehicle contents or parts theft were the main common crime types in Boulevard des Galeries-d'Anjou, Anjou and Chemin de la Côte-de-Liesse, Saint-Laurent neighborhoods within the period reviewed.
- 3. Vehicle theft was particularly prominent in Chemin de la Côte-de-Liesse, Saint-Laurent in 2016 and in 2017. The neighborhood has about 40% increase in vehicle theft from 2016 to 2017. Vehicle contents or parts theft recorded 22%

Which neighborhood has the highest cases of murder crime in 2015, 2016 and 2017?

```
In [164]: # extract and summarize data
ddf = top_neighb[top_neighb['ADAPTED_CATEGORY']=="Offenses causing death"]
dtop = ddf.groupby(['YEAR', 'COORDS']).agg({'COORDS':'count'}).rename(columns={'COORDS':'DEATH}

# sort and extract neighborhood addresses as a new column using 'extract_address' function.
topdf = dtop.sort_values('DEATH_INCIDENT', ascending=False).reset_index(drop=True).iloc[:1]
topdf['NEIGHBORHOOD'] = extract_address(topdf['COORDS'])

# display result
display(topdf)
```

| | YEAR | COORDS D | EATH_INCIDENT | NEIGHBORHOOD |
|---|------|----------------------------|---------------|-----------------------------------------------------------------------------------------------------------------------------|
| 0 | 2016 | (45.511716, -73.562202) | 2 | Pavillon Sainte-Catherine, Rue Sainte-Catherine Est, Quartier des Spectacles, Ville-Marie, Agglomération de Montréal, CA-QC |

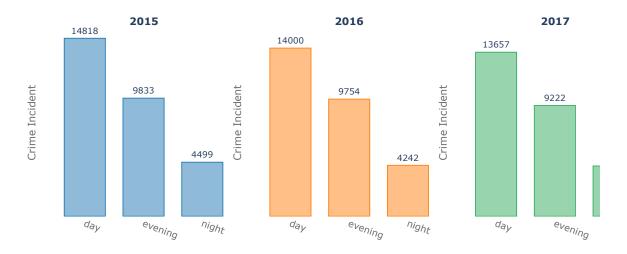
Observation:

• Two cases of murder occurred in 2016 at Pavillon Sainte-Catherine, Rue Sainte-Catherine Est, Quartier des Spectacles, Centre-Ville, Ville-Marie. These are the highest murder cases within the 3-years period.

What time of the day did most crime incidents occur in 2015, 2016 and 2017?

| QUART | day | evening | night | |
|-------|-------|---------|-------|--|
| YEAR | | | | |
| 2015 | 14818 | 9833 | 4499 | |
| 2016 | 14000 | 9754 | 4242 | |
| 2017 | 13657 | 9222 | 4184 | |
| 2019 | 11311 | 7510 | 3517 | |
| 2020 | 11141 | 6713 | 3548 | |
| 2021 | 12045 | 6807 | 3693 | |
| 2022 | 14417 | 8081 | 4994 | |

Crime incidents during the days, evenings and nights in 2015, 2016 and 2017



• Crimes committed during the days (jour) in each year were about twice the total crimes registered in the evenings (soir) and in the nights (nuit) for the same year. However daytime crimes recorded 5.4% decrease from 2015 to

Which top 5 police stations (PDQ) got the most crime complaints in 2015, 2016 and 2017?

Top 5 police stations (PDQ) with the highest crime complaints registered in 2015, 201

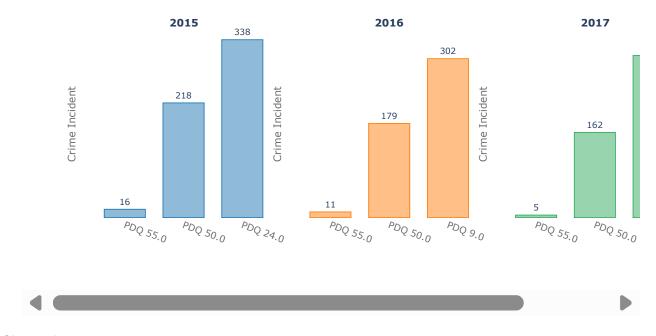


Observation:

 PDQ38 and PDQ7 are the prominent among the top stations that registered most crime incidents in 2015, 2016 and 2017.

Which are the top 3 PDQs that got least crime complaints in 2015, 2016 and 2017?

Top 5 police stations (PDQ) with the lowest crime complaints registered in 2015, 2016



Observation:

 PDQ0 and PDQ55 recorded the least crime cases among the 3 top stations with low crime registrations in 2015, 2016 and 2017.

Conclusion

- Burglary, Vehicle contents or parts theft and Misdemeanor are the three most prevalent crimes in the 3-years period.
- Besides vehicle theft that increased by 9.4% between 2016 and 2017, all other crime types showed a downward trend in the 3-years period reviewed.
- Offenses causing death are the least crime incidents noted during the 3-years period.
- Burglary recorded a significant decrease (31%) from 917 in January to 629 in February in 2017.
- Boulevard des Galeries-d'Anjou, Anjou and Chemin de la Côte-de-Liesse, Saint-Laurent are the neighborhoods with the highest number of crimes in 2015, 2016 and 2017. Vehicle theft and vehicle contents or parts theft are the

main common crime types noted in these neighborhoods.

- Vehicle theft was particularly prominent in Chemin de la Côte-de-Liesse, Saint-Laurent in 2016 and in 2017. The
 neighbourhood has about 40% increase in vehicle theft from 2016 to 2017. Vehicle contents or parts theft recorded
 22% increase within the same period.
- Two cases of murder occurred in 2016 at Pavillon Sainte-Catherine, Rue Sainte-Catherine Est, Quartier des Spectacles, Centre-Ville, Ville-Marie. These are the highest murder cases within the 3-years period based on the dataset.
- Crimes committed during the days (jour) in each year were about twice the total crimes registered in the evenings (soir) and in the nights (nuit) for the same year. However daytime crimes recorded 5.4% decrease from 2015 to 2016 and 2.2% decrease between 2016 and 2017.
- PDQ38 and PDQ7 are the prominent among the top stations that registered most crime incidents in 2015, 2016 and 2017.
- PDQ0 and PDQ55 recorded the least crime cases among the 3 top stations with low crime registrations in 2015, 2016 and 2017.