Crimes Section

July 9, 2020

1 Basic Statistics of the Data Set

This is part of the book Chapter 6

The statistics shown here intend to enlighten the characteristics of the data set used to contruct the ABM.

The data was provided by the PMMG, to whom we are very thankful.

The **programmer** is Eric Araujo.

The last update was in 29/9/2019

```
[1]: import pandas as pd
  import matplotlib.pyplot as plt
  plt.style.use('ggplot')
  import geopandas
  import descartes
  import numpy as np

import folium
  import seaborn as sns

%matplotlib inline
```

```
[2]: xls = pd.ExcelFile('data/Lavras.xls')
registers = pd.read_excel(xls, 'OCORRENCIAS')
```

2 Registers

```
[3]: registers.head(2)

[3]: Ano Fato Bairro \
0 2014 ANISIO ALVES DE ABREU
1 2014 ANISIO ALVES DE ABREU

Desc Longa Meio Utilizado \
0 AGRESSAO FISICA SEM EMPREGO DE INSTRUMENTOS
1 ARMAS DE FOGO
```

```
O BENS / VALORES DE ESTABELECIMENTO /PESSOA JURI...
     1 BENS / VALORES DE ESTABELECIMENTO /PESSOA JURI...
      Descrição Longa Local Imediato Descrição Subclasse Nat Principal \
     0
                 POSTO DE COMBUSTIVEL
     1
                 POSTO DE COMBUSTIVEL
                                                                  ROUBO
      Dia da Semana Fato Dia Numérico Fato
                                                      Faixa 1 Hora Fato
                                                                          Latitude \
             QUARTA-FEIRA
                                          8 De 03:00 as 03:59
     0
                                                                        -21.254166
             QUARTA-FEIRA
                                          29 De 00:00 as 00:59
                                                                        -21.254166
       Longitude Mês Numérico Fato Município
                                                       Número REDS \
     0 -44.999031
                                        LAVRAS 2014-000525047-001
                                   1
     1 -44.999031
                                        LAVRAS 2014-002172720-001
                                   1
       Tentado/Consumado Nat Principal UNIDADE_AREA
                                                      Unidade Área Militar \
     0
                               TENTADO
                                             08 BPM 54 CIA PM/8 BPM/6 RPM
                             CONSUMADO
                                             08 BPM 54 CIA PM/8 BPM/6 RPM
     1
       Município - Código
     0
                    313820
     1
                    313820
[4]: registers.columns
[4]: Index(['Ano Fato', 'Bairro', 'Desc Longa Meio Utilizado',
            'Desc Longa Subgrupo Complemento Nat', 'Descrição Longa Local Imediato',
            'Descrição Subclasse Nat Principal', 'Dia da Semana Fato',
            'Dia Numérico Fato', 'Faixa 1 Hora Fato', 'Latitude', 'Longitude',
            'Mês Numérico Fato', 'Município', 'Número REDS',
            'Tentado/Consumado Nat Principal', 'UNIDADE AREA',
            'Unidade Área Militar', 'Município - Código'],
           dtype='object')
[5]: # Renaming columns
     translate_cols = {'Ano Fato': 'year',
                       'Mês Numérico Fato': 'month',
                       'Dia Numérico Fato': 'day',
                       'Dia da Semana Fato': 'week_day',
                       'Bairro': 'neighborhood',
                       'Desc Longa Meio Utilizado': 'violence_type',
                       'Desc Longa Subgrupo Complemento Nat': 'type_object_robbed',
                       'Descrição Longa Local Imediato': 'location_description',
                       'Descrição Subclasse Nat Principal': 'crime_type',
                       'Faixa 1 Hora Fato': 'one_h_window',
```

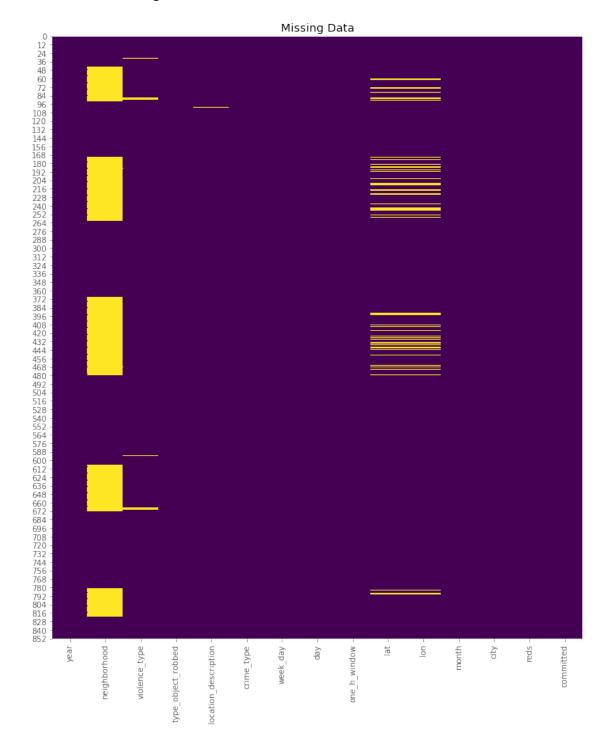
Desc Longa Subgrupo Complemento Nat \

```
'Latitude': 'lat',
                        'Longitude': 'lon',
                        'Município': 'city',
                        'Número REDS': 'reds',
                        'Tentado/Consumado Nat Principal': 'committed',
                        'UNIDADE_AREA': 'pm_area',
                        'Unidade Área Militar': 'pm_unity',
                        'Município - Código': 'city_code'
                       }
      registers.rename(columns=translate_cols, inplace=True)
 [6]: registers.columns
 [6]: Index(['year', 'neighborhood', 'violence_type', 'type_object_robbed',
             'location_description', 'crime_type', 'week_day', 'day', 'one_h_window',
             'lat', 'lon', 'month', 'city', 'reds', 'committed', 'pm_area',
             'pm_unity', 'city_code'],
            dtype='object')
 [7]: # Removing irrelevant columns
      registers = registers[['year', 'neighborhood', 'violence_type', _
      'location_description', 'crime_type', 'week_day', 'day',
             'one_h_window', 'lat', 'lon', 'month', 'city', 'reds', 'committed',
            ]]
      # Get years 2014-2018
      registers = registers[registers.year.isin([2014,2015,2016,2017,2018])]
 [8]: # Removing white spaces in one h window column
      registers.one_h_window = registers.one_h_window.str.rstrip()
     Checking for duplicates in the registers' tab
 [9]: print('Number of registers in the file: ', len(registers))
      print('Number of unique REDS - IDs: ', len(set(registers.reds)))
     Number of registers in the file: 862
     Number of unique REDS - IDs: 862
     Checking for missing data in the registers' tab
[10]: missing_values = ['INVÁLIDO', 'ESCOLARIDADE - IGNORADA', 'IGNORADA', L
      → 'PREENCHIMENTO OPCIONAL', 'INVALIDO',
                        'ESTADO CIVIL - IGNORADO', 'ESTADO CIVIL - NAO DECLARADO', L
      → 'PREENCHIMENTO OPCIONAL',
                        'GRAU DA LESAO - IGNORADO' ]
```

registers.replace(to_replace=missing_values, value=np.nan, inplace=True)

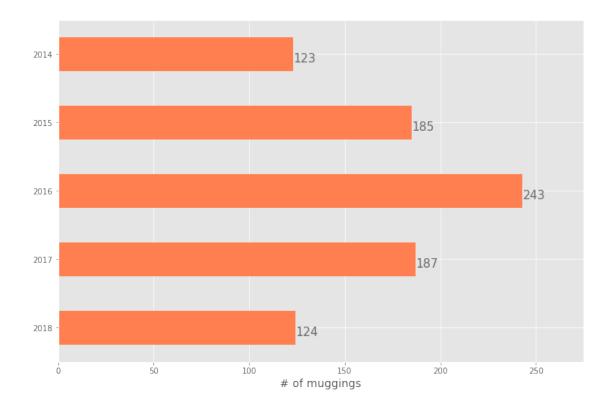
```
[11]: plt.figure(figsize=(12,14))
sns.heatmap(registers.isnull(), cbar = False, cmap = 'viridis')
plt.title('Missing Data')
```

[11]: Text(0.5, 1, 'Missing Data')



```
[12]: # Location description
      pd.value_counts(registers.location_description)[:10]
[12]: VIA DE ACESSO PUBLICA
                                                          429
      POSTO DE COMBUSTIVEL
                                                          118
      BAR / LANCHONETE / RESTAURANTE / SIMILAR
                                                           68
      OUTROS - ESTABELECIMENTOS COMERCIAIS / SERVICOS
                                                           34
      CASA
                                                           33
      MERCEARIA / SACOLAO / SUPERMERCADO
                                                           28
      CONFEITARIA / PADARIA / PANIFICADORA
                                                           26
     LOJA DIVERSA
                                                           17
      SITIO
                                                           15
      BOATE / CASA DE SHOW / SIMILAR
                                                           11
     Name: location_description, dtype: int64
[13]: 100*pd.value_counts(registers.location_description)[:10]/pd.
      →value_counts(registers.location_description).sum()
[13]: VIA DE ACESSO PUBLICA
                                                          49.883721
      POSTO DE COMBUSTIVEL
                                                          13.720930
      BAR / LANCHONETE / RESTAURANTE / SIMILAR
                                                           7.906977
      OUTROS - ESTABELECIMENTOS COMERCIAIS / SERVICOS
                                                           3.953488
                                                           3.837209
     MERCEARIA / SACOLAO / SUPERMERCADO
                                                           3.255814
      CONFEITARIA / PADARIA / PANIFICADORA
                                                           3.023256
     LOJA DIVERSA
                                                           1.976744
      SITIO
                                                           1.744186
      BOATE / CASA DE SHOW / SIMILAR
                                                           1.279070
      Name: location_description, dtype: float64
[14]: # Means used to commit the crime
      pd.value_counts(registers.violence_type)[:10]
[14]: ARMAS DE FOGO
                                                                        376
      INSTRUMENTO CONTUNDENTE / CORTANTE / PERFURANTE (ARMA BRANCA)
                                                                        184
      AGRESSAO FISICA SEM EMPREGO DE INSTRUMENTOS
                                                                         94
                                                                         87
      OUTROS MEIOS (DESCREVER EM CAMPO ESPECIFICO)
                                                                         83
      SIMULACRO DE ARMA DE FOGO
                                                                         27
      EMBOSCADA (INCLUI SAIDINHA DE BANCO)
                                                                          2
      Name: violence_type, dtype: int64
[15]: pd.value_counts(registers.violence_type)[:10]*100/pd.value_counts(registers.
       →violence_type)[:10].sum()
```

```
[15]: ARMAS DE FOGO
                                                                        44.079719
      INSTRUMENTO CONTUNDENTE / CORTANTE / PERFURANTE (ARMA BRANCA)
                                                                        21.570926
     AGRESSAO FISICA SEM EMPREGO DE INSTRUMENTOS
                                                                        11.019930
                                                                        10.199297
      OUTROS MEIOS (DESCREVER EM CAMPO ESPECIFICO)
                                                                        9.730363
     SIMULACRO DE ARMA DE FOGO
                                                                        3.165299
     EMBOSCADA (INCLUI SAIDINHA DE BANCO)
                                                                         0.234467
     Name: violence_type, dtype: float64
[16]: # Crimes per year
      ax = registers.year.value_counts().sort_index(ascending=True).plot(kind='barh',
                                                                      figsize=((12,8)),
                                                                      color='coral')
      ax.set_alpha(0.8)
      #plt.title('Robberies per year in Lauras (2014-2018)', fontsize=18)
      plt.xlabel('# of muggings', fontsize=14)
      plt.xlim(0,275)
      # create a list to collect the plt.patches data
      totals = []
      # find the values and append to list
      for i in ax.patches:
         totals.append(i.get_width())
          #print(totals)
      for i in ax.patches:
         # get_width pulls left or right; get_y pushes up or down
          ax.text(i.get_width()+.3, i.get_y()+.37, \
                  str(round((i.get_width()), 2)), fontsize=15,
      color='dimgrey')
      # invert for largest on top
      ax.invert_yaxis()
```



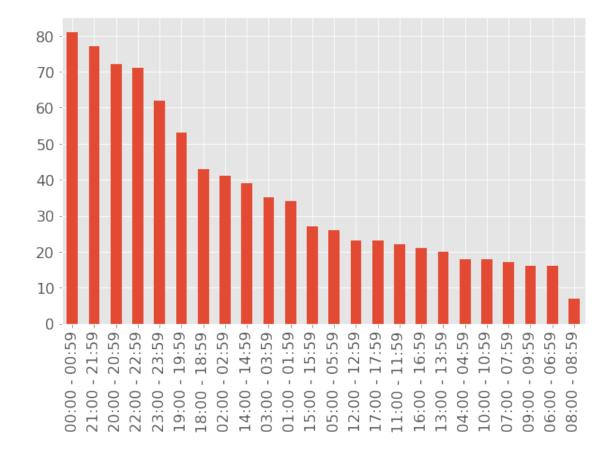
```
[17]: one_h_replacement_dict = {
                                    'De 00:00 as 00:59': '00:00 - 00:59',
                                    'De 01:00 as 01:59': '01:00 - 01:59',
                                    'De 02:00 as 02:59': '02:00 - 02:59',
                                    'De 03:00 as 03:59': '03:00 - 03:59',
                                    'De 04:00 as 04:59': '04:00 - 04:59',
                                    'De 05:00 as 05:59': '05:00 - 05:59',
                                    'De 06:00 as 06:59': '06:00 - 06:59',
                                    'De 07:00 as 07:59': '07:00 - 07:59',
                                    'De 08:00 as 08:59': '08:00 - 08:59',
                                    'De 09:00 as 09:59': '09:00 - 09:59',
                                    'De 10:00 as 10:59': '10:00 - 10:59',
                                    'De 11:00 as 11:59': '11:00 - 11:59',
                                    'De 12:00 as 12:59': '12:00 - 12:59',
                                    'De 13:00 as 13:59': '13:00 - 13:59',
                                    'De 14:00 as 14:59': '14:00 - 14:59',
                                    'De 15:00 as 15:59': '15:00 - 15:59',
                                    'De 16:00 as 16:59': '16:00 - 16:59',
                                    'De 17:00 as 17:59': '17:00 - 17:59',
                                    'De 18:00 as 18:59': '18:00 - 18:59',
                                    'De 19:00 as 19:59': '19:00 - 19:59',
                                    'De 20:00 as 20:59': '20:00 - 20:59',
                                    'De 21:00 as 21:59': '21:00 - 21:59',
                                    'De 22:00 as 22:59': '22:00 - 22:59',
```

```
'De 23:00 as 23:59': '23:00 - 23:59'}
```

[18]: registers.one_h_window.replace(one_h_replacement_dict, inplace=True)
#registers.one_h_window

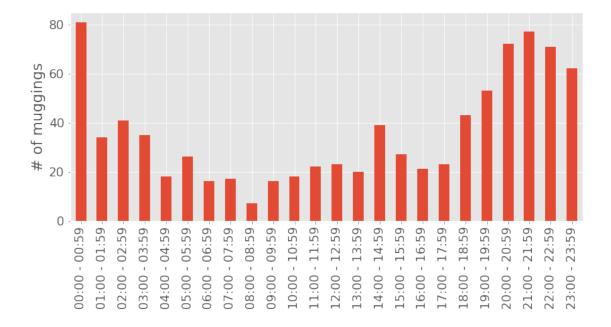
```
[41]: #plt.rcParams.update({'font.size': 16})
plt.tick_params(labelsize=16)
registers.one_h_window.value_counts().plot(kind='bar', figsize=((10,6)))
```

[41]: <matplotlib.axes._subplots.AxesSubplot at 0x12e454250>



```
'03:00 - 03:59',
                '04:00 - 04:59',
                '05:00 - 05:59'
day_times = ['06:00 - 06:59',
              '07:00 - 07:59',
              '08:00 - 08:59',
              '09:00 - 09:59',
              '10:00 - 10:59',
              '11:00 - 11:59',
              '12:00 - 12:59',
              '13:00 - 13:59',
              '14:00 - 14:59',
              '15:00 - 15:59',
              '16:00 - 16:59',
              '17:00 - 17:59'
            ]
ax = registers.one_h_window.value_counts().sort_index().plot(kind='bar',_
\rightarrowfigsize=((12,5)))
#ax.set_xlabel('Time of the day')
ax.set_ylabel('# of muggings')
```

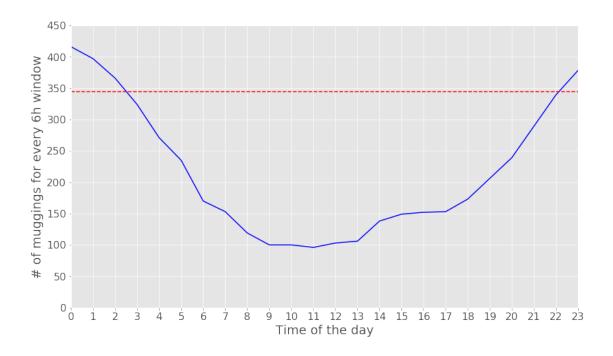
[28]: Text(0, 0.5, '# of muggings')



```
print('Total of crimes committed in night time: \t', len(registers[registers.
       →one_h_window.isin(night_times)]))
     Total of crimes committed in day time:
                                                        249
     Total of crimes committed in night time:
                                                        613
[30]: times_crimes = registers.one_h_window.value_counts().sort_index().
       →reset_index(drop=True)
[31]: times_crimes.mean()
[31]: 35.91666666666664
[32]: times_crimes.std()
[32]: 21.993905611736427
[33]: times_crimes.median()
[33]: 26.5
[34]: times_crimes_window = times_crimes.append(times_crimes)
      times_crimes_window = times_crimes_window.rolling(window=6).sum()[5:-19]
      times_crimes_window
[34]: 5
            235.0
      6
            170.0
      7
            153.0
            119.0
      8
      9
            100.0
      10
            100.0
             96.0
      11
      12
            103.0
      13
            106.0
      14
            138.0
      15
            149.0
      16
            152.0
      17
            153.0
      18
            173.0
      19
            206.0
      20
            239.0
      21
            289.0
      22
            339.0
      23
            378.0
            416.0
      0
      1
            397.0
      2
            366.0
```

```
324.0
      3
      4
            271.0
      Name: one_h_window, dtype: float64
[35]: tc_window_df = pd.DataFrame(times_crimes_window).sort_index()
[36]: # When the cummulative sum of crimes exceeds 40% of the total
      tc_window_df['hot_times'] = tc_window_df.one_h_window > len(set(registers.
       \rightarrowreds))*0.4
[37]: tc_window_df[tc_window_df.hot_times]
[37]:
          one_h_window hot_times
      0
                 416.0
                             True
      1
                 397.0
                             True
      2
                 366.0
                             True
      23
                 378.0
                             True
[38]: ax = tc_window_df.plot(figsize=((14,8)), color='b', legend=False)
      ax.set_ylim(0,450)
      ax.set_xlabel('Time of the day')
      ax.set_ylabel('# of muggings for every 6h window')
      ax.set_xticks(np.arange(0, 24, 1))
      # Line dividing 40% of the cases
      hline = len(set(registers.reds))*0.4
      ax.axhline(y=hline, color='r', linestyle='--')
```

[38]: <matplotlib.lines.Line2D at 0x12de553d0>



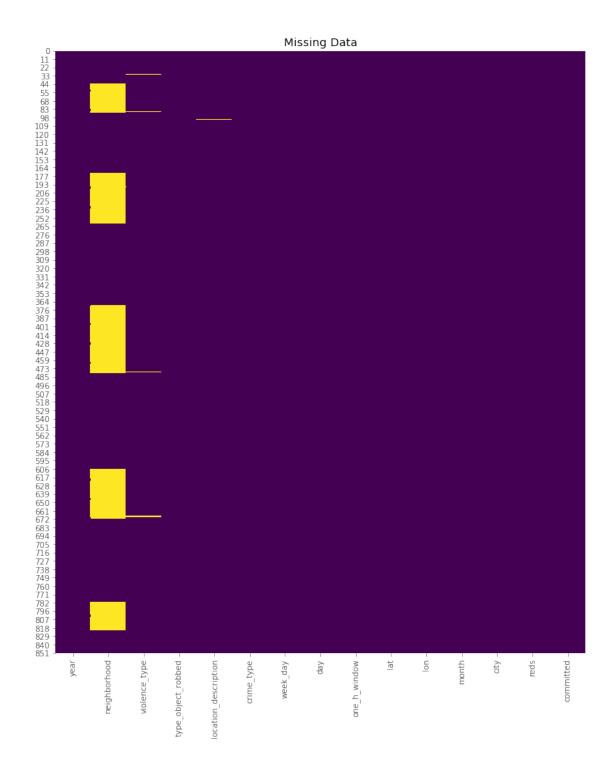
2.0.1 Cleaning the registers data

```
[29]: registers.dropna(axis=0, subset=['lat', 'lon'], inplace=True)
len(registers)

[29]: 803

[30]: plt.figure(figsize=(12,14))
    sns.heatmap(registers.isnull(), cbar = False, cmap = 'viridis')
    plt.title('Missing Data')

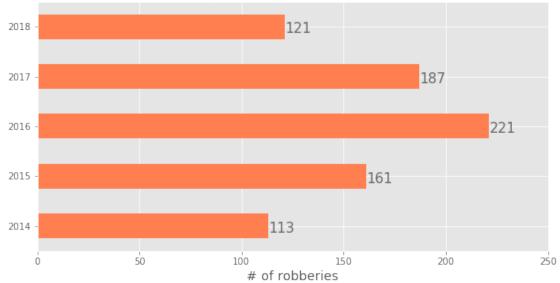
[30]: Text(0.5, 1, 'Missing Data')
```



```
plt.title('Robberies per year in Lavras (2014-2018)', fontsize=18)
plt.xlabel('# of robberies', fontsize=14)
plt.xlim(0,250)
# create a list to collect the plt.patches data
totals = []
# find the values and append to list
for i in ax.patches:
    totals.append(i.get_width())
    print(totals)
for i in ax.patches:
    # get_width pulls left or right; get_y pushes up or down
    ax.text(i.get_width()+.3, i.get_y()+.38, \
            str(round((i.get_width()), 2)), fontsize=15,
color='dimgrey')
# invert for largest on top
ax.invert_yaxis()
```

[121] [121, 187] [121, 187, 221] [121, 187, 221, 161] [121, 187, 221, 161, 113]

Robberies per year in Lavras (2014-2018)



```
[32]: # Location description
pd.value_counts(registers.location_description)[:10]
```

[32]:	VIA DE ACESSO PUBLICA	407
	POSTO DE COMBUSTIVEL	113
	BAR / LANCHONETE / RESTAURANTE / SIMILAR	62
	CASA	29
	OUTROS - ESTABELECIMENTOS COMERCIAIS / SERVICOS	29
	MERCEARIA / SACOLAO / SUPERMERCADO	27
	CONFEITARIA / PADARIA / PANIFICADORA	24
	LOJA DIVERSA	16
	SITIO	12
	BOATE / CASA DE SHOW / SIMILAR	11
	Name: location_description, dtype: int64	

[33]: # Means used to commit the crime pd.value_counts(registers.violence_type)[:10]

[33]: ARMAS DE FOGO

INSTRUMENTO CONTUNDENTE / CORTANTE / PERFURANTE (ARMA BRANCA)

AGRESSAO FISICA SEM EMPREGO DE INSTRUMENTOS

AMEACA

OUTROS MEIOS (DESCREVER EM CAMPO ESPECIFICO)

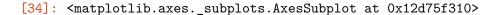
SIMULACRO DE ARMA DE FOGO

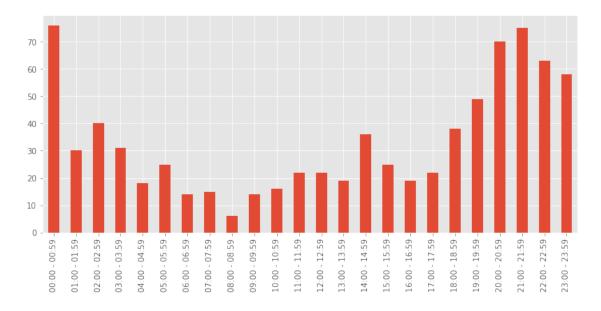
EMBOSCADA (INCLUI SAIDINHA DE BANCO)

Name: violence_type, dtype: int64

[34]: registers.one_h_window.value_counts().sort_index().plot(kind='bar', ⊔

→figsize=((12,5)))



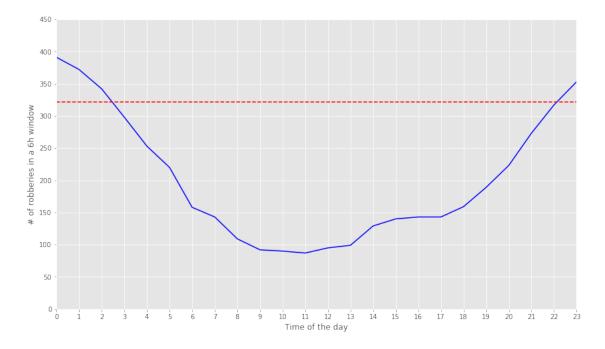


```
[35]: times_crimes = registers.one_h_window.value_counts().sort_index().
       →reset_index(drop=True)
      times_crimes_window = times_crimes.append(times_crimes)
      times_crimes_window = times_crimes_window.rolling(window=6).sum()[5:-19]
      times_crimes_window
[35]: 5
            220.0
      6
            158.0
      7
            143.0
      8
            109.0
             92.0
      9
      10
             90.0
      11
             87.0
             95.0
      12
      13
             99.0
            129.0
      14
      15
            140.0
      16
            143.0
      17
            143.0
      18
            159.0
      19
            189.0
      20
            223.0
      21
            273.0
      22
            317.0
      23
            353.0
      0
            391.0
      1
            372.0
      2
            342.0
      3
            298.0
            253.0
      Name: one_h_window, dtype: float64
[36]: tc_window_df = pd.DataFrame(times_crimes_window).sort_index()
      # When the cummulative sum of crimes exceeds 40% of the total
      tc_window_df['hot_times'] = tc_window_df.one_h_window > len(set(registers.
       \rightarrowreds))*0.4
[37]: tc_window_df[tc_window_df.hot_times]
[37]:
          one_h_window hot_times
                 391.0
                             True
                             True
      1
                 372.0
      2
                 342.0
                             True
      23
                 353.0
                             True
[38]: ax = tc_window_df.plot(figsize=((14,8)), color='b', legend=False)
      ax.set_ylim(0,450)
```

```
ax.set_xlabel('Time of the day')
ax.set_ylabel('# of robberies in a 6h window')
ax.set_xticks(np.arange(0, 24, 1))

# Line dividing 40% of the cases
hline = len(set(registers.reds))*0.4
ax.axhline(y=hline, color='r', linestyle='--')
```

[38]: <matplotlib.lines.Line2D at 0x12de4b850>



[]: