

Hacktiv8 PTP Introduction to Data Science Projects 1 // Pt Plot

Title: Hacktiv8 PTP Introduction to Data Science Projects 1: Publication-Grade Plc

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Publication-grade Plot Introduction

The aim of this projects is to introduce you to data visualization with Python as corpossible. Using what you've learned; download the London Crime Dataset from Krof crime in major metropolitan areas, such as London, occurs in distinct patterns. criminal reports by month, LSOA borough, and major/minor category from Jan 200

This dataset contains:

- 1soa code: this represents a policing area
- borough: the london borough for which the statistic is related
- major_category: the major crime category
- minor_category: the minor crime category
- value: the count of the crime for that particular borough, in that particular mo
- year : the year of the summary statistic
- month: the month of the summary statistic

Formulate a question and derive a statistical hypothesis test to answer the questic that you're able to make decisions using data in a scientific manner. And the impo Examples of questions can be:

- What is the change in the number of crime incidents from 2011 to 2016?
- What were the top 3 crimes per borough in 2016?

Please make sure that you have completed the session for this course, namely Ac part of this Program.

Note: You can take a look at Project Rubric below:

Criteria

Meet Expectations

Criteria	Meet Expectations
Area Plot	Mengimplementasikan Area Plot Menggunakan Matplotlib Dengan [Kegunaan Plot/Grafik
Histogram	Mengimplementasikan Histogram Menggunakan Matplotlib Dengan Kegunaan Plot/Grafik.
Bar Chart	Mengimplementasikan Bar Chart Menggunakan Matplotlib Dengan Kegunaan Plot/Grafik.
Pie Chart	Mengimplementasikan Pie Chart Menggunakan Matplotlib Dengan I Kegunaan Plot/Grafik.
Box Plot	Mengimplementasikan Box Plot Menggunakan Matplotlib Dengan D Kegunaan Plot/Grafik.
Scatter Plot	Mengimplementasikan Scatter Plot Menggunakan Matplotlib Denga Kegunaan Plot/Grafik.
Word Clouds	Mengimplementasikan Word Clouds Menggunakan Wordclouds Librar Dengan Kegunaan Plot/Grafik.
Folium Maps	Mengimplementasikan London Maps Menggunakan Folium.
Preprocessing	Student Melakukan Preproses Dataset Sebelum Menerapkan Visualisasi.
Apakah Kode Berjalan Tanpa Ada Eror?	Seluruh Kode Berfungsi Dan Dibuat Dengan Benar.
Area Plot	Menarik Informasi/Kesimpulan Berdasarkan Area Plot Yang Telah Student I
Histogram	Menarik Informasi/Kesimpulan Berdasarkan Histogram Yang Telah Student
Bar Chart	Menarik Informasi/Kesimpulan Berdasarkan Bar Chart Yang Telah Student
Pie Chart	Menarik Informasi/Kesimpulan Berdasarkan Pie Chart Yang Telah Student I
Box Plot	Menarik Informasi/Kesimpulan Berdasarkan Box Plot Yang Telah Student B
Scatter Plot	Menarik Informasi/Kesimpulan Berdasarkan Scatter Plot Yang Telah Studer
Overall Analysis	Menarik Informasi/Kesimpulan Dari Keseluruhan Plot Yang Dapat Menjawa

Exploring Datasets with pandas

pandas is an essential data analysis toolkit for Python. From their website (http://p

pandas is a Python package providing fast, flexible, and expressive data s make working with "relational" or "labeled" data both easy and intuitive. It ε fundamental high-level building block for doing practical, **real world** data ε

The course heavily relies on *pandas* for data wrangling, analysis, and visualizatior some time and familizare yourself with the *pandas* API Reference: http://pandas.pydata.org/pandas-docs/stable/api.html).

The first thing we'll do is import two key data analysis modules: pandas and Numi

```
In [1]:
           1
              import numpy as np
              import pandas as pd
In [2]:
              df = pd.read csv('london crime by lsoa.csv')
           2
             print ('Data read into a pandas dataframe!')
          Data read into a pandas dataframe!
In [3]:
           1
              # Let's view the top 5 rows of the dataset using the head() function
           2
              df.head()
            Isoa code
                         borough
                                         major_category
                                                               minor_category
                                                                               value year month
         0 E01001116
                                                         Burglary in Other Buildings
                                                                                      2016
                       Croydon
                                  Burglary
                                                                                           11
         1 E01001646
                      Greenwich
                                  Violence Against the Person Other violence
                                                                               0
                                                                                      2016 11
            E01000677
                                  Violence Against the Person Other violence
                      Bromley
                                                                               0
                                                                                      2015 5
            E01003774
                       Redbridge
                                  Burglary
                                                         Burglary in Other Buildings 0
                                                                                      2016 3
           E01004563
                      Wandsworth Robbery
                                                         Personal Property
                                                                                      2008 6
In [4]:
              # We can also veiw the bottom 5 rows of the dataset using the tail()
              df.tail()
                  Isoa_code borough
                                              major_category
                                                                      minor_category value
                                                                                            year
         13490599 E01000504
                             Brent
                                       Criminal Damage
                                                              Criminal Damage To Dwelling 0
                                                                                            2015
         13490600 E01002504
                             Hillingdon
                                       Robbery
                                                              Personal Property
                                                                                      1
                                                                                            2015
                                                              Burglary in a Dwelling
         13490601 E01004165
                             Sutton
                                       Burglary
                                                                                      0
                                                                                            2011
         13490602 E01001134
                             Croydon
                                       Robbery
                                                              Business Property
                                                                                      0
                                                                                            2011
         13490603 E01003413
                                       Violence Against the Person Wounding/GBH
                             Merton
                                                                                            2015
          When analyzing a dataset, it's always a good idea to start by getting basic information
          can do this by using the info() method.
In [5]:
              print(df.info())
              print(df.describe())
              print('minor_category ',df.minor_category.unique())
              print('major_category ',df.major_category.unique())
              print('borough ',df.borough.unique())
```

value

False

```
7 print("To check if any colun has null values")
8 print(df.isnull().any())
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13490604 entries, 0 to 13490603
Data columns (total 7 columns):
   Column
                   Dtype
---
                    ____
   lsoa code
                  object
 0
 1 borough
                   object
 2
   major_category object
 3 minor category object
 4
    value
                    int64
 5
   year
                   int64
 6
    month
                    int64
dtypes: int64(3), object(4)
memory usage: 720.5+ MB
None
                          year
             value
count 1.349060e+07 1.349060e+07 1.349060e+07
mean 4.779444e-01 2.012000e+03 6.500000e+00
std 1.771513e+00 2.581989e+00 3.452053e+00
min 0.000000e+00 2.008000e+03 1.000000e+00
25% 0.000000e+00 2.010000e+03 3.750000e+00
50%
    0.000000e+00 2.012000e+03 6.500000e+00
    1.000000e+00 2.014000e+03 9.250000e+00
75%
      3.090000e+02 2.016000e+03 1.200000e+01
minor category ['Burglary in Other Buildings' 'Other violence' 'Personal Property'
 'Other Theft' 'Offensive Weapon' 'Criminal Damage To Other Building'
 'Theft/Taking of Pedal Cycle' 'Motor Vehicle Interference & Tampering'
 'Theft/Taking Of Motor Vehicle' 'Wounding/GBH' 'Other Theft Person'
 'Common Assault' 'Theft From Shops' 'Possession Of Drugs' 'Harassment'
 'Handling Stolen Goods' 'Criminal Damage To Dwelling'
 'Burglary in a Dwelling' 'Criminal Damage To Motor Vehicle'
 'Other Criminal Damage' 'Counted per Victim' 'Going Equipped'
 'Other Fraud & Forgery' 'Assault with Injury' 'Drug Trafficking'
 'Other Drugs' 'Business Property' 'Other Notifiable' 'Other Sexual'
 'Theft From Motor Vehicle' 'Rape' 'Murder']
major_category ['Burglary' 'Violence Against the Person' 'Robbery' 'Theft and Handling'
 'Criminal Damage' 'Drugs' 'Fraud or Forgery' 'Other Notifiable Offences'
 'Sexual Offences']
borough ['Croydon' 'Greenwich' 'Bromley' 'Redbridge' 'Wandsworth' 'Ealing'
 'Hounslow' 'Newham' 'Sutton' 'Haringey' 'Lambeth' 'Richmond upon Thames'
 'Hillingdon' 'Havering' 'Barking and Dagenham' 'Kingston upon Thames'
 'Westminster' 'Hackney' 'Enfield' 'Harrow' 'Lewisham' 'Brent' 'Southwark'
 'Barnet' 'Waltham Forest' 'Camden' 'Bexley' 'Kensington and Chelsea'
 'Islington' 'Tower Hamlets' 'Hammersmith and Fulham' 'Merton'
 'City of London']
To check if any colun has null values
lsoa code
                 False
borough
                 False
major_category False
minor_category
                 False
```

```
year
                            False
           month
                            False
           dtype: bool
           To get the list of column headers we can call upon the dataframe's .columns para
In [6]:
              df.columns.values
           array(['lsoa_code', 'borough', 'major_category', 'minor_category',
                  'value', 'year', 'month'], dtype=object)
           Similarly, to get the list of indicies we use the .index parameter.
In [7]:
               df.index.values
           array([
                                  1,
                                           2, ..., 13490601, 13490602, 13490603])
            Rename column
In [8]:
               df.rename(columns={'borough':'District'}, inplace=True)
            2
               df.head()
             Isoa_code
                           District
                                          major_category
                                                                minor_category value year month
          0 E01001116
                                                          Burglary in Other Buildings 0
                        Croydon
                                                                                       2016 11
                                   Burglary
          1 E01001646
                        Greenwich
                                   Violence Against the Person Other violence
                                                                                       2016 11
            E01000677
                        Bromley
                                   Violence Against the Person Other violence
                                                                                       2015 5
            E01003774
                        Redbridge
                                   Burglary
                                                          Burglary in Other Buildings 0
                                                                                       2016 3
            E01004563 Wandsworth Robbery
                                                          Personal Property
                                                                                0
                                                                                       2008 6
           To view the dimensions of the dataframe, we use the .shape parameter.
In [9]:
               print(df.shape)
           (13490604, 7)
           Let's make one dataset that contains value 1 in value features.
In [10]:
               criminal = df[df['value'] == 1]
In [11]:
               df1 = df.copy()
```

```
2 df1.drop(['lsoa_code','minor_category'], axis=1, inplace=True)
```

~	1 0	-
~	αt	- 1
J	uт	_1

	District	major_category	value	year	month
0	Croydon	Burglary	0	2016	11
1	Greenwich	Violence Against the Person	0	2016	11
2	Bromley	Violence Against the Person	0	2015	5
3	Redbridge	Burglary	0	2016	3
4	Wandsworth	Robbery	0	2008	6
13490599	Brent	Criminal Damage	0	2015	2
13490600	Hillingdon	Robbery	1	2015	6
13490601	Sutton	Burglary	0	2011	2
13490602	Croydon	Robbery	0	2011	5
13490603	Merton	Violence Against the Person	0	2015	6

 $13490604 \text{ rows} \times 5 \text{ columns}$

```
In [12]:
           1
             drugs = df1[(df1['major_category'] == 'Drugs') & (df1['year'] == 201
              print(drugs.value.sum())
           38914
In [13]:
              df sum = df1.groupby(['year', 'District']).size().reset index(name='c-
           2
             print(df_sum)
              print(df sum.columns)
               year
                               District count_per_year
             2008 Barking and Dagenham
           0
                                                34560
             2008
                                 Barnet
                                               63648
           2
               2008
                                 Bexley
                                                42852
           3
               2008
                                               54516
                                  Brent
           4
               2008
                                                58212
                                Bromley
               . . .
           . .
          292 2016
                                 Sutton
                                               35832
          293 2016
                                                45792
                          Tower Hamlets
          294 2016
                         Waltham Forest
                                                45144
          295 2016
                             Wandsworth
                                                55404
          296 2016
                                                40740
                            Westminster
          [297 rows x 3 columns]
          Index(['year', 'District', 'count_per_year'], dtype='object')
```

table = df1.pivot_table(values='value', index=['year'],columns=['maj

major_category	Burglary	Criminal Damage	Drugs	Fraud or Forgery	Other Notifiable Offences	Robbery	Se) Offen
year							
2008	88092	91872	68804	5325	10112	29627	1273
2009	90619	85565	60549	0	10644	29568	0
2010	86826	77897	58674	0	10768	32341	0
2011	93315	70914	57550	0	10264	36679	0
2012	93392	62158	51776	0	10675	35260	0
2013	87222	56206	50278	0	10811	29337	0
2014	76053	59279	44435	0	13037	22150	0
2015	70489	62976	39785	0	14229	21383	0
2016	68285	64071	38914	0	15809	22528	0

Visualizing Data using Matplotlib Matplotlib: Standard Python Visualization Library

The primary plotting library we will explore in the course is <u>Matplotlib (http://matplc</u> website:

Matplotlib is a Python 2D plotting library which produces publication quality hardcopy formats and interactive environments across platforms. Matplotli scripts, the Python and IPython shell, the jupyter notebook, web applicatio graphical user interface toolkits.

If you are aspiring to create impactful visualization with python, Matplotlib is an es disposal.

Matplotlib.Pyplot

One of the core aspects of Matplotlib is ${\tt matplotlib.pyplot}$.

Let's start by importing Matplotlib and Matplotlib.pyplot as follows:

```
1 # we are using the inline backend
2 %matplotlib inline
3
4 import matplotlib as mpl
5 import matplotlib.pyplot as plt
```

In [16]:

mpl.style.use(['ggplot']) # optional: for ggplot-like style

Area Pots (Series/Dataframe)

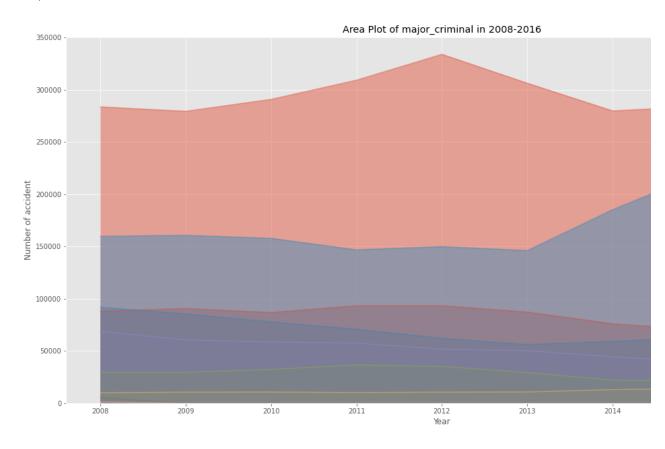
What is a line plot and why use it?

An Area chart or area plot is a type of plot which displays information as a series connected by straight line segments. It is a basic type of chart common in many finance a continuous data set. These are best suited for trend-based visualizations c

Questions:

1. what most major_criminal in 2008-2016?

```
In [17]:
          1
            # Write your function below
            table.plot(kind='area',
          2
                        alpha=0.45,
          3
                          stacked=False,
          4
          5
                          figsize=(20, 10), # pass a tuple (x, y) size
          6
          7
            # Graded-Funtion Begin (~1 Lines)
          8
         9
            # Graded-Funtion End
        10
            plt.title('Area Plot of major_criminal in 2008-2016') # add a title
        11
            plt.ylabel('Number of accident') # add y-label
        12
            plt.xlabel('Year') # add x-label
        13
        14
        15
            plt.show()
```



Insight:

Based on graph, Thef and Handling is most major criminal happend during 2008-2

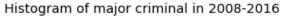
Histogram

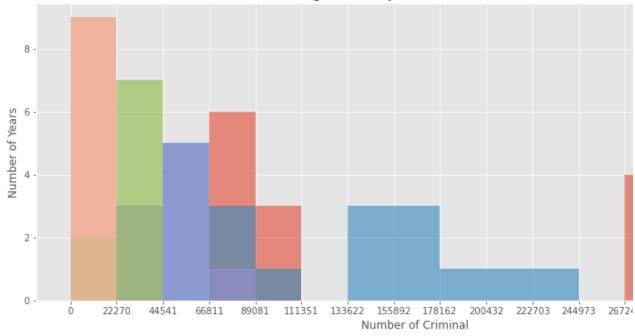
A histogram is a way of representing the frequency distribution of numeric dataset the x-axis into bins, assigns each data point in our dataset to a bin, and then coun that have been assigned to each bin. So the y-axis is the frequency or the number that we can change the bin size and usually one needs to tweak it so that the distribution of numeric dataset to a bin, and then countributed in the countributed in the property of the property of

Question:

1. Frequency major case criminal in London (Make your own questions)

```
In [18]:
          1
             # Write your function below
          2
             count, bin edges = np.histogram(table, 15)
             table.plot(kind ='hist',
          3
                       figsize=(15, 6),
          4
                       bins=15,
          5
                       alpha=0.6,
          6
          7
                       xticks=bin edges,
          8
          9
             # Graded-Funtion Begin (~2 Lines)
         10
             # Graded-Funtion End
         11
         12
            plt.title('Histogram of major criminal in 2008-2016') # add a title
         13
            plt.ylabel('Number of Years') # add y-label
         14
             plt.xlabel('Number of Criminal ') # add x-label
         15
         16
         17
            plt.show()
```





Insight: Most frequency cases in london between 2008-2016 is Thef and Handling

Bar Charts (Dataframe)

A bar plot is a way of representing data where the *length* of the bars represents th feature/variable. Bar graphs usually represent numerical and categorical variables

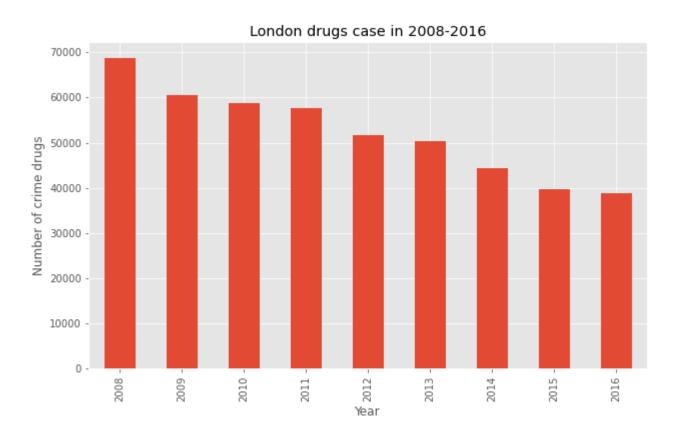
To create a bar plot, we can pass one of two arguments via kind parameter in pl

- kind=bar creates a vertical bar plot
- kind=barh creates a horizontal bar plot

Question:

1. Yearly drug case in London from 2008-2016?

```
In [19]:
            # Write your function below
         1
         2
            table bar = table['Drugs']
            table_bar.plot(kind='bar', figsize=(10, 6))
          3
            # Graded-Funtion Begin (~1 Lines)
         5
            # Graded-Funtion End
         6
         7
         8
            plt.xlabel('Year') # add to x-label to the plot
         9
            plt.ylabel('Number of crime drugs') # add y-label to the plot
        10
            plt.title('London drugs case in 2008-2016') # add title to the plot
         11
         12
            plt.show()
```



Insight:

Drug case in London is decrasing in 2008-2016

Pie Charts

A pie chart is a circualr graphic that displays numeric proportions by dividing a c slices. You are most likely already familiar with pie charts as it is widely used in bu

create pie charts in Matplotlib by passing in the kind=pie keyword.

Question:

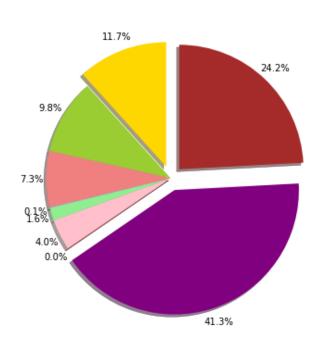
(Make your own questions)

```
In [20]:
           1
              table pie = table.transpose()
              table pie['total'] = table.sum()
           2
           3
             table pie
           4
                                 2008
                                        2009
                                               2010
                                                      2011
                                                             2012
                                                                    2013
                                                                           2014
                                                                                  2015
                                                                                         2010
                           year
                 major_category
         Burglary
                                       90619
                                                     93315
                                                                   87222
                                                                                 70489
                                88092
                                              86826
                                                            93392
                                                                          76053
                                                                                        68285
         Criminal Damage
                                91872
                                       85565
                                              77897
                                                     70914
                                                            62158
                                                                   56206
                                                                          59279
                                                                                 62976
                                                                                        64071
         Drugs
                                68804
                                       60549
                                                     57550
                                                                   50278
                                                                          44435
                                                                                39785
                                              58674
                                                            51776
                                                                                        38914
         Fraud or Forgery
                                5325
                                       0
                                              0
                                                     0
                                                            0
                                                                   0
                                                                          0
                                                                                        0
                                                                                 0
         Other Notifiable Offences
                                10112
                                      10644
                                              10768
                                                     10264
                                                            10675
                                                                   10811
                                                                          13037
                                                                                 14229
                                                                                        15809
         Robbery
                                29627
                                       29568
                                              32341
                                                     36679
                                                            35260
                                                                   29337
                                                                          22150
                                                                                 21383
                                                                                        22528
         Sexual Offences
                                1273
                                                     0
                                                                   0
                                       0
                                              0
                                                            0
                                                                          0
                                                                                 0
                                                                                        0
         Theft and Handling
                                283692 279492 290924
                                                     309292 334054 306372 279880 284022 29413
         Violence Against the Person 159844 160777 157894 146901 150014 146181 185349 218740 23238
In [23]:
              # Write your function below
              # ratio for each continent with which to offset each wedge.
              colors_list = ['gold', 'yellowgreen', 'lightcoral', 'lightskyblue',
              explode list = [0.1, 0, 0, 0, 0, 0, 0, 0.1, 0.1]
              # Graded-Funtion Begin (~8 Lines)
              table_pie['total'].plot(kind='pie',
                                       figsize=(15, 6),
                                       autopct='%1.1f%%',
                                       startangle=90,
                                       shadow=True,
                                                               # turn off labels on pie
                                       labels=None,
                                       colors=colors list, # add custom colors
                                       # the ratio between the center of each pie sli
                                       pctdistance=1.12,
```

```
17
                          explode=explode list # 'explode'
18
   # Graded-Funtion End
19
20
21
   # scale the title up by 12% to match pctdistance
   plt.title('Major criminal case in London (2008-2016)', y=1.12)
23
   plt.axis('equal')
24
25
26
  # add legend
   plt.legend(labels=table pie.index, loc='upper left')
28
29 plt.show()
```

Major criminal case in London (2008-2016)





Insight:

Theft and Handling is most major criminal case in London during 2008-2016, with

Box Plots

A box plot is a way of statistically representing the distribution of the data through

- Minimun: Smallest number in the dataset.
- First quartile: Middle number between the minimum and the median.

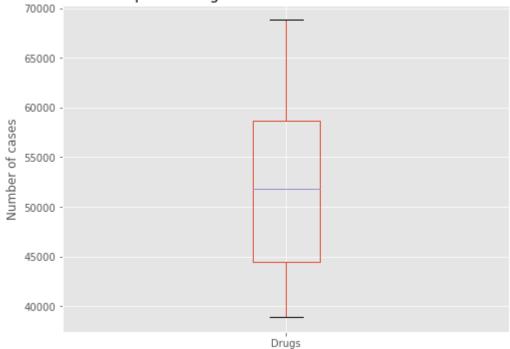
- Second quartile (Median): Middle number of the (sorted) dataset.
- Third quartile: Middle number between median and maximum.
- Maximum: Highest number in the dataset.

Question:

1. Describe drug case in London from 2008-2016? (Make your own questions)

```
In [24]:
          1
             # Write your function below
            table bar.plot(kind='box', figsize=(8, 6))
            # Graded-Funtion Begin (~1 Lines)
          3
          4
          5
            # Graded-Funtion End
          6
          7
            plt.title('Box plot of drugs case in London from 1980 - 2013')
            plt.ylabel('Number of cases')
          9
         10
            plt.show()
```





Insight: Drugs max cases is around 70000 cases (Make your own Insight)

Scatter Plots

A scatter plot (2D) is a useful method of comparing variables against each oth

line plots in that they both map independent and dependent variables on a 2D connected together by a line in a line plot, they are not connected in a scatter plot considered to express a trend. With further analysis using tools like regression, we this relationship and use it to predict trends outside the dataset.

Question:

1. lets compare Drugs and Robbery

(Make your own questions)

```
In [68]:
1    table_scatter = table[['Drugs','Robbery']]
2    table_scatter = table_scatter.reset_index()
3    table_scatter

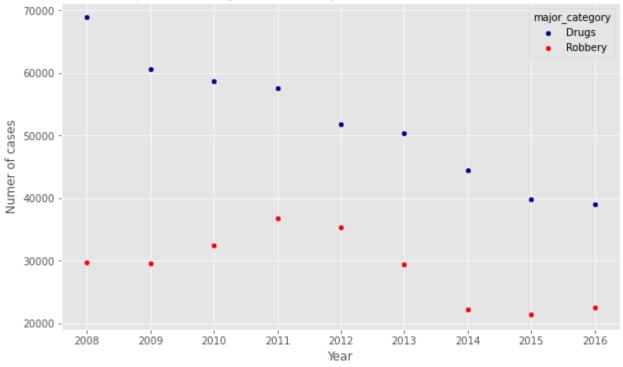
major_category year Drugs Robbery

0     2008 68804 29627
```

major_category	year	Drugs	Robbery
0	2008	68804	29627
1	2009	60549	29568
2	2010	58674	32341
3	2011	57550	36679
4	2012	51776	35260
5	2013	50278	29337
6	2014	44435	22150
7	2015	39785	21383
8	2016	38914	22528

```
In [84]:
          1
            # Write your function below
          2
            # Graded-Funtion Begin (~1 Lines)
          3
            ax1 = table scatter.plot(kind='scatter', x='year', y='Drugs', figsize
          4
            ax2 = table_scatter.plot(kind='scatter', x='year', y='Robbery', figs
          5
          6
          7
            # Graded-Funtion End
          8
          9
            plt.title('Comparation Drugs and Robbery cases in London from 2008-2
            plt.xlabel('Year')
         10
            plt.ylabel('Numer of cases')
         11
            plt.show()
         12
```

Comparation Drugs and Robbery cases in London from 2008-2016



Word Clouds

word clouds (also known as text clouds or tag clouds) work in a simple way: the na source of textual data (such as a speech, blog post, or database), the bigger and cloud.

```
# !conda install -c conda-forge wordcloud --yes
          4 # !pip install wordcloud
          5
          6 # import package and its set of stopwords
          7 from wordcloud import WordCloud, STOPWORDS
          8
          9 print ('Wordcloud is installed and imported!')
         Wordcloud is installed and imported!
In [28]:
             stopwords = set(STOPWORDS)
In [29]:
            # table minor = df[['minor category']]
            source_dataset = ' '.join(df.major_category)
In [30]:
            # instantiate a word cloud object
            your_wordcloud = WordCloud(
          2
          3
                 background_color='white',
                 max words=2000,
          4
          5
                 stopwords=stopwords
          6
          7
            # generate the word cloud
            your wordcloud.generate(source dataset)
          <wordcloud.wordcloud.WordCloud at 0x7fc1e4e41850>
```

```
# Write your function below

# Graded-Funtion Begin (~1 Lines)

plt.imshow(your_wordcloud, interpolation='bilinear')

# Graded-Funtion End

plt.axis('off')

plt.show()

| Drugs Theft | Drugs Craman | Decemp Violence | Person | Decemp Vio
```

```
Drugs Theft Handling Violence Person Notifiable Offences Notifiabl
```

Folium

Folium is a powerful Python library that helps you create several types of Leaflet r results are interactive makes this library very useful for dashboard building.

From the official Folium documentation page:

Folium builds on the data wrangling strengths of the Python ecosystem an of the Leaflet.js library. Manipulate your data in Python, then visualize it in Folium.

Folium makes it easy to visualize data that's been manipulated in Python c map. It enables both the binding of data to a map for choropleth visualizati Vincent/Vega visualizations as markers on the map.

The library has a number of built-in tilesets from OpenStreetMap, Mapbox, supports custom tilesets with Mapbox or Cloudmade API keys. Folium sup and TopoJSON overlays, as well as the binding of data to those overlays to with color-brewer color schemes.

```
#!conda install -c conda-forge folium=0.5.0 --yes
import folium

print('Folium installed and imported!')
Folium installed and imported!
```



In [35]:

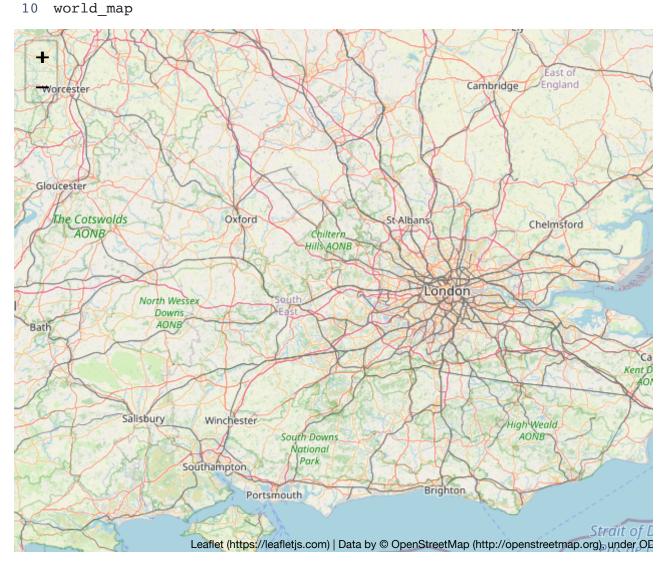
define the world map centered around London with a low zoom level
world_map = folium.Map(location=[51.509865, -0.118092], zoom_start=8

```
# Write your function below

Graded-Funtion Begin (~1 Lines)

# Graded-Funtion End

# display world map
world_map
```



Thanks For Completing This Labs!