DV0101EN-2-3-1-Pie-Charts-Box-Plots-Scatter-Plots-and-Bubble-Plots-py-v2.0

March 17, 2019

Pie Charts, Box Plots, Scatter Plots, and Bubble Plots

0.1 Introduction

In this lab session, we continue exploring the Matplotlib library. More specificatly, we will learn how to create pie charts, box plots, scatter plots, and bubble charts.

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1 Exploring Datasets with pandas and Matplotlib

Toolkits: The course heavily relies on *pandas* and **Numpy** for data wrangling, analysis, and visualization. The primary plotting library we will explore in the course is Matplotlib.

Dataset: Immigration to Canada from 1980 to 2013 - International migration flows to and from selected countries - The 2015 revision from United Nation's website.

The dataset contains annual data on the flows of international migrants as recorded by the countries of destination. The data presents both inflows and outflows according to the place of birth, citizenship or place of previous / next residence both for foreigners and nationals. In this lab, we will focus on the Canadian Immigration data.

2 Downloading and Prepping Data

Import primary modules.

```
In [1]: import numpy as np # useful for many scientific computing in Python
import pandas as pd # primary data structure library
```

Let's download and import our primary Canadian Immigration dataset using *pandas* read_excel() method. Normally, before we can do that, we would need to download a module which *pandas* requires to read in excel files. This module is **xlrd**. For your convenience, we have pre-installed this module, so you would not have to worry about that. Otherwise, you would need to run the following line of code to install the **xlrd** module:

```
!conda install -c anaconda xlrd --yes
```

Download the dataset and read it into a pandas dataframe.

Data downloaded and read into a dataframe!

Let's take a look at the first five items in our dataset.

```
In [3]: df_can.head()
```

```
Out [3]:
                           Coverage
                                              OdName
                                                       AREA AreaName
                                                                        REG
                  Туре
           Immigrants Foreigners
                                         Afghanistan
                                                        935
                                                                 Asia 5501
        0
           Immigrants
                        Foreigners
                                             Albania
                                                        908
                                                                        925
        1
                                                              Europe
                        Foreigners
           Immigrants
                                             Algeria
                                                        903
                                                              Africa
                                                                        912
        3
           Immigrants
                        Foreigners
                                     American Samoa
                                                        909
                                                             Oceania
                                                                        957
           Immigrants
                        Foreigners
                                             Andorra
                                                        908
                                                                        925
                                                              Europe
                    RegName
                              DEV
                                               DevName
                                                         1980
                                                                     2004
                                                                            2005
                                                                                  2006
        0
                                   Developing regions
                                                                     2978
                                                                            3436
                                                                                  3009
              Southern Asia
                              902
                                                           16
                                                               . . .
        1
           Southern Europe
                              901
                                    Developed regions
                                                                     1450
                                                                           1223
                                                                                   856
                                                            1
                                                               . . .
        2
           Northern Africa
                                   Developing regions
                                                                            3626
                                                                                  4807
                              902
                                                           80
                                                               . . .
                                                                     3616
                                   Developing regions
        3
                  Polynesia
                                                            0
                                                                        0
                                                                               0
                              902
                                                                                     1
                                                                . . .
           Southern Europe
                              901
                                    Developed regions
                                                                        0
                                                                               0
                                                                                     1
            2007
                  2008
                        2009
                               2010
                                      2011
                                            2012
                                                   2013
           2652
                  2111
                        1746
                               1758
                                      2203
                                            2635
                                                   2004
        0
            702
                   560
                                561
                                       539
        1
                         716
                                             620
                                                    603
        2
           3623
                  4005
                        5393
                               4752
                                     4325
                                            3774
                                                   4331
        3
                     0
                            0
                                  0
               0
                                         0
                                               0
                                                      0
        4
               1
                     0
                            0
                                  0
                                         0
                                               1
                                                      1
```

[5 rows x 43 columns]

Let's find out how many entries there are in our dataset.

Clean up data. We will make some modifications to the original dataset to make it easier to create our visualizations. Refer to *Introduction to Matplotlib and Line Plots* and *Area Plots, Histograms, and Bar Plots* for a detailed description of this preprocessing.

3 Visualizing Data using Matplotlib

Import Matplotlib.

```
In [6]: %matplotlib inline
    import matplotlib as mpl
    import matplotlib.pyplot as plt
    mpl.style.use('ggplot') # optional: for ggplot-like style
    # check for latest version of Matplotlib
    print('Matplotlib version: ', mpl.__version__) # >= 2.0.0
Matplotlib version: 3.0.2
```

4 Pie Charts

A pie chart is a circualr graphic that displays numeric proportions by dividing a circle (or pie) into proportional slices. You are most likely already familiar with pie charts as it is widely used in business and media. We can create pie charts in Matplotlib by passing in the kind=pie keyword.

Let's use a pie chart to explore the proportion (percentage) of new immigrants grouped by continents for the entire time period from 1980 to 2013.

Step 1: Gather data.

We will use *pandas* groupby method to summarize the immigration data by Continent. The general process of groupby involves the following steps:

1. **Split:** Splitting the data into groups based on some criteria.

In [7]: # group countries by continents and apply sum() function

df_continents = df_can.groupby('Continent', axis=0).sum()

- 2. **Apply:** Applying a function to each group independently: .sum() .count() .mean() .std() .aggregate() .apply() .etc..
- 3. **Combine:** Combining the results into a data structure.

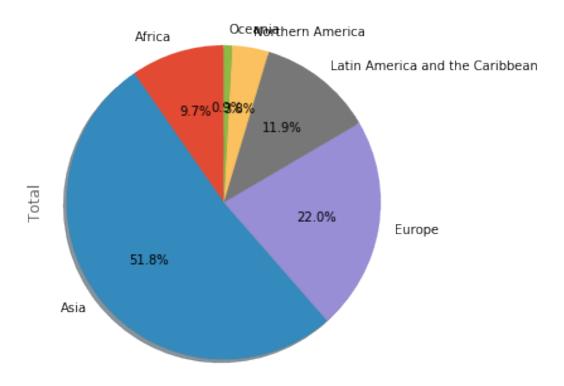
```
# note: the output of the groupby method is a `groupby' object.
        # we can not use it further until we apply a function (eq .sum())
        print(type(df_can.groupby('Continent', axis=0)))
        df_continents.head()
<class 'pandas.core.groupby.generic.DataFrameGroupBy'>
Out[7]:
                                           1980
                                                   1981
                                                          1982
                                                                 1983
                                                                        1984
                                                                                1985
        Continent
                                                   4363
                                                                         2639
        Africa
                                           3951
                                                          3819
                                                                 2671
                                                                                2650
        Asia
                                          31025
                                                  34314
                                                         30214
                                                                24696
                                                                       27274
                                                                               23850
                                          39760
                                                  44802 42720
                                                                24638
                                                                       22287
        Europe
                                                                               20844
        Latin America and the Caribbean
                                          13081
                                                  15215 16769
                                                                15427 13678
                                                                               15171
        Northern America
                                                  10030
                                           9378
                                                          9074
                                                                 7100
                                                                         6661
                                                                                6543
                                           1986
                                                   1987
                                                          1988
                                                                 1989
                                                                               2005
        Continent
        Africa
                                           3782
                                                   7494
                                                          7552
                                                                 9894
                                                                        . . .
                                                                              27523
        Asia
                                          28739
                                                  43203
                                                         47454
                                                                60256
                                                                        . . .
                                                                             159253
                                          24370
                                                  46698
                                                         54726
                                                                60893
        Europe
                                                                              35955
        Latin America and the Caribbean
                                          21179
                                                  28471
                                                         21924
                                                                25060
                                                                              24747
                                                   7705
        Northern America
                                           7074
                                                          6469
                                                                 6790
                                                                               8394
                                             2006
                                                     2007
                                                             2008
                                                                      2009
                                                                              2010
        Continent
        Africa
                                           29188
                                                    28284
                                                            29890
                                                                    34534
                                                                             40892
        Asia
                                          149054 133459 139894 141434 163845
        Europe
                                           33053
                                                    33495
                                                            34692
                                                                    35078
                                                                             33425
```

Latin America and the Caribbean Northern America	24676 9613	26011 9463	26547 10190	26867 8995	28818 8142
Continent	2011	2012	2013	Total	
Africa	35441	38083	38543	618948	
Asia	146894	152218	155075	3317794	
Europe	26778	29177	28691	1410947	
Latin America and the Caribbean	27856	27173	24950	765148	
Northern America	7677	7892	8503	241142	

[5 rows x 35 columns]

Step 2: Plot the data. We will pass in kind = 'pie' keyword, along with the following additional parameters: - autopct - is a string or function used to label the wedges with their numeric value. The label will be placed inside the wedge. If it is a format string, the label will be fmt%pct. - startangle - rotates the start of the pie chart by angle degrees counterclockwise from the x-axis. - shadow - Draws a shadow beneath the pie (to give a 3D feel).

Immigration to Canada by Continent [1980 - 2013]



The above visual is not very clear, the numbers and text overlap in some instances. Let's make a few modifications to improve the visuals:

- Remove the text labels on the pie chart by passing in legend and add it as a seperate legend using plt.legend().
- Push out the percentages to sit just outside the pie chart by passing in pctdistance parameter.
- Pass in a custom set of colors for continents by passing in colors parameter.
- **Explode** the pie chart to emphasize the lowest three continents (Africa, North America, and Latin America and Carribbean) by pasing in explode parameter.

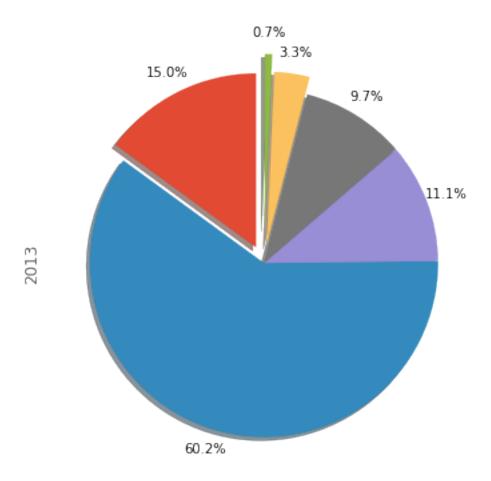
```
pctdistance=1.12, # the ratio between the center of each
                                    colors=colors_list, # add custom colors
                                    explode=explode_list # 'explode' lowest 3 continents
    # scale the title up by 12% to match pctdistance
   plt.title('Immigration to Canada by Continent [1980 - 2013]', y=1.12)
   plt.axis('equal')
    # add legend
   plt.legend(labels=df_continents.index, loc='upper left')
   plt.show()
                         Immigration to Canada by Continent [1980 - 2013]
                                          0.9%3.8%
     Africa
     Asia
     Europe
                                                   11.9%
     Latin America and the Caribbean
     Northern America
     Oceania
ota
                                                         22.0%
```

Question: Using a pie chart, explore the proportion (percentage) of new immigrants grouped by continents in the year 2013.

Note: You might need to play with the explore values in order to fix any overlapping slice values.

)

Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x7fc51cc40be0>



5 Box Plots

A box plot is a way of statistically representing the *distribution* of the data through five main dimensions:

- 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.

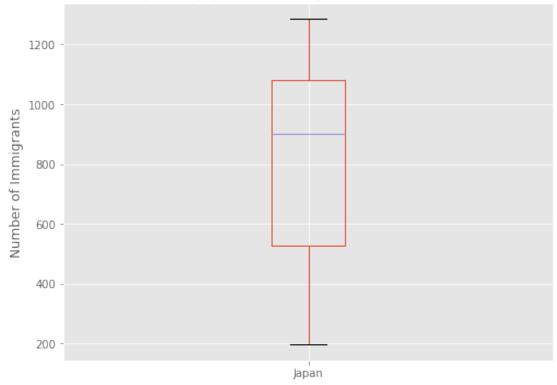
To make a box plot, we can use kind=box in plot method invoked on a *pandas* series or dataframe.

Let's plot the box plot for the Japanese immigrants between 1980 - 2013.

Step 1: Get the dataset. Even though we are extracting the data for just one country, we will obtain it as a dataframe. This will help us with calling the dataframe.describe() method to view the percentiles.

Step 2: Plot by passing in kind='box'.

Box plot of Japanese Immigrants from 1980 - 2013



We can immediately make a few key observations from the plot above: 1. The minimum number of immigrants is around 200 (min), maximum number is around 1300 (max), and median number of immigrants is around 900 (median). 2. 25% of the years for period 1980 - 2013 had an annual immigrant count of ~500 or fewer (First quartile). 2. 75% of the years for period 1980 - 2013 had an annual immigrant count of ~1100 or fewer (Third quartile).

We can view the actual numbers by calling the describe() method on the dataframe.

```
In [13]: df_japan.describe()
Out[13]: Country
                         Japan
         count
                     34.000000
                    814.911765
         mean
                    337.219771
         std
         min
                    198.000000
         25%
                    529.000000
         50%
                    902.000000
         75%
                   1079.000000
                   1284.000000
         max
```

One of the key benefits of box plots is comparing the distribution of multiple datasets. In one of the previous labs, we observed that China and India had very similar immigration trends. Let's analyize these two countries further using box plots.

Question: Compare the distribution of the number of new immigrants from India and China for the period 1980 - 2013.

Step 1: Get the dataset for China and India and call the dataframe **df_CI**.

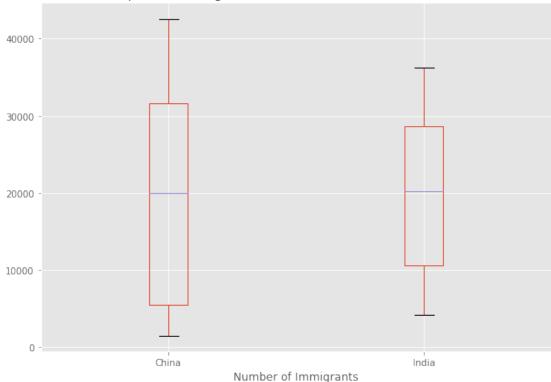
```
In [14]: ### type your answer here
         df_CI= df_can.loc[['China', 'India'], years].transpose()
         df_CI.head()
Out[14]: Country
                  China
                          India
         1980
                    5123
                           8880
         1981
                    6682
                           8670
         1982
                    3308
                           8147
         1983
                    1863
                           7338
         1984
                    1527
                           5704
```

Let's view the percentages associated with both countries using the describe() method.

```
Out[15]: Country
                         China
                                        India
         count
                     34.000000
                                   34.000000
                  19410.647059 20350.117647
         mean
                  13568.230790
                                10007.342579
         std
         min
                   1527.000000
                                 4211.000000
         25%
                   5512.750000 10637.750000
         50%
                  19945.000000 20235.000000
                  31568.500000 28699.500000
         75%
                  42584.000000 36210.000000
         max
```

Step 2: Plot data.

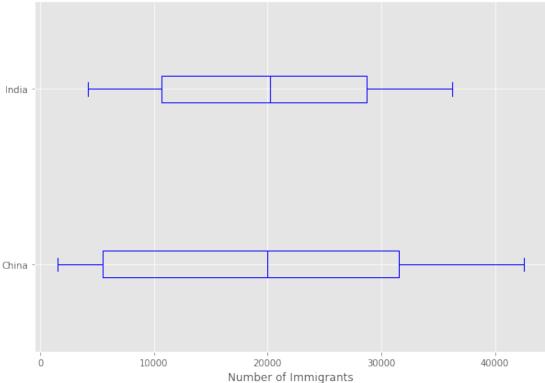




We can observe that, while both countries have around the same median immigrant population (~20,000), China's immigrant population range is more spread out than India's. The maximum population from India for any year (36,210) is around 15% lower than the maximum population from China (42,584).

If you prefer to create horizontal box plots, you can pass the vert parameter in the **plot** function and assign it to *False*. You can also specify a different color in case you are not a big fan of the default red color.





Subplots

Often times we might want to plot multiple plots within the same figure. For example, we might want to perform a side by side comparison of the box plot with the line plot of China and India's immigration.

To visualize multiple plots together, we can create a figure (overall canvas) and divide it into subplots, each containing a plot. With **subplots**, we usually work with the **artist layer** instead of the **scripting layer**.

Typical syntax is:

```
fig = plt.figure() # create figure
ax = fig.add_subplot(nrows, ncols, plot_number) # create subplots
```

Where - nrows and ncols are used to notionally split the figure into (nrows * ncols) sub-axes, - plot_number is used to identify the particular subplot that this function is to create within the notional grid. plot_number starts at 1, increments across rows first and has a maximum of nrows * ncols as shown below.

We can then specify which subplot to place each plot by passing in the ax parameter in plot() method as follows:

```
In [18]: fig = plt.figure() # create figure

ax0 = fig.add_subplot(1, 2, 1) # add subplot 1 (1 row, 2 columns, first plot)
ax1 = fig.add_subplot(1, 2, 2) # add subplot 2 (1 row, 2 columns, second plot). See tip
```

```
# Subplot 1: Box plot

df_CI.plot(kind='box', color='blue', vert=False, figsize=(20, 6), ax=ax0) # add to subp

ax0.set_title('Box Plots of Immigrants from China and India (1980 - 2013)')

ax0.set_xlabel('Number of Immigrants')

ax0.set_ylabel('Countries')

# Subplot 2: Line plot

df_CI.plot(kind='line', figsize=(20, 6), ax=ax1) # add to subplot 2

ax1.set_title ('Line Plots of Immigrants from China and India (1980 - 2013)')

ax1.set_ylabel('Number of Immigrants')

ax1.set_xlabel('Years')

plt.show()

Box Plots of immigrants from China and India (1980 - 2013)

Line Plots of immigrants from China and India (1980 - 2013)

Line Plots of immigrants from China and India (1980 - 2013)
```

** * Tip regarding subplot convention **

In the case when nrows, ncols, and plot_number are all less than 10, a convenience exists such that the a 3 digit number can be given instead, where the hundreds represent nrows, the tens represent ncols and the units represent plot_number. For instance,

```
subplot(211) == subplot(2, 1, 1)
```

produces a subaxes in a figure which represents the top plot (i.e. the first) in a 2 rows by 1 column notional grid (no grid actually exists, but conceptually this is how the returned subplot has been positioned).

Let's try something a little more advanced.

Previously we identified the top 15 countries based on total immigration from 1980 - 2013.

Question: Create a box plot to visualize the distribution of the top 15 countries (based on total immigration) grouped by the *decades* 1980s, 1990s, and 2000s.

Step 1: Get the dataset. Get the top 15 countries based on Total immigrant population. Name the dataframe **df_top15**.

Out[19]:		Continent \
out[19].	Country	continent /
	India	Asia
	China	Asia
	United Kingdom of Great Britain and Northern Ir	Europe
	Philippines	Asia
	Pakistan	Asia
	United States of America	Northern America
	Iran (Islamic Republic of)	Asia
	Sri Lanka	Asia
	Republic of Korea	Asia
	Poland	Europe
	Lebanon	Asia
	France	Europe
	Jamaica	Latin America and the Caribbean
	Viet Nam	Asia
	Romania	Europe
		•
		Region \
	Country	
	India	Southern Asia
	China Hill Company of the Company of	Eastern Asia
	United Kingdom of Great Britain and Northern Ir	Northern Europe
	Philippines	South-Eastern Asia
	Pakistan	Southern Asia
	United States of America	Northern America
	Iran (Islamic Republic of)	Southern Asia
	Sri Lanka	Southern Asia
	Republic of Korea	Eastern Asia
	Poland	Eastern Europe
	Lebanon	Western Asia
	France	Western Europe
	Jamaica	Caribbean
	Viet Nam	South-Eastern Asia
	Romania	Eastern Europe
		DevName 1980 \
	Country	
	India	Developing regions 8880
	China	Developing regions 5123
	United Kingdom of Great Britain and Northern Ir	Developed regions 22045
	Philippines	Developing regions 6051
	Pakistan	Developing regions 978
	United States of America	Developed regions 9378
	Iran (Islamic Republic of)	Developing regions 1172
	Sri Lanka	Developing regions 185
	Republic of Korea	Developing regions 1011
	Poland	Developed regions 863

Lebanon France Jamaica Viet Nam Romania	Develo Develo Develo	oping re oped re ping re ping re	gions gions gions	1409 1729 3198 1191 375
	1981	1982	1983	\
Country India China United Kingdom of Great Britain and Northern Ir Philippines Pakistan United States of America Iran (Islamic Republic of) Sri Lanka Republic of Korea Poland Lebanon France Jamaica	8670 6682 24796 5921 972 10030 1429 371 1456 2930 1119 2027 2634	8147 3308 20620 5249 1201 9074 1822 290 1572 5881 1159 2219 2661	7338 1863 10015 4562 900 7100 1592 197 1081 4546 789 1490 2455	
Viet Nam	1829 438	2162 583	3404 543	
Romania	438 1984	583 1985	543 1986 .	\
Country India China United Kingdom of Great Britain and Northern Ir Philippines Pakistan United States of America Iran (Islamic Republic of) Sri Lanka Republic of Korea Poland Lebanon France Jamaica Viet Nam Romania	5704 1527 10170 3801 668 6661 1977 1086 847 3588 1253 1169 2508 7583 524	1816 9564 3150 514 6543 1648 845 962 2819 1683 1177 2938	7150 . 1960 . 9470 . 4166 . 691 . 7074 . 1794 . 1838 . 1208 . 4808 . 2576 . 1298 . 4649 . 2741 . 656 .	
Country India China United Kingdom of Great Britain and Northern Ir Philippines	36210 42584 7258 18139	33848 33518 7140 18400	28742 27642 8216 19837	·

Pakistan	14314	13127	10124	
United States of America	8394	9613		
Iran (Islamic Republic of)	5837			
Sri Lanka	4930	4714	4123	
Republic of Korea	5832		5920	
Poland	1405			
Lebanon	3709			
France	4429			
Jamaica	1945		2141	
Viet Nam	1852			
Romania	5048			
nomania	3040	4400	3034	
	2008	2009	2010	\
Country India	28261	29456	34235	
China	30037			
United Kingdom of Great Britain and Northern Ir	8979	8876	8724	
Philippines	24887		38617	
Pakistan	8994	7217	6811	
United States of America	10190		8142	
	6475		7477	
Iran (Islamic Republic of)				
Sri Lanka	4756			
Republic of Korea	7294		5537	
Poland	1267		795	
Lebanon	3566			
France	4532		4646	
Jamaica	2334		2321	
Viet Nam	1784		1942	
Romania	2837	2076	1922	
Country	2011	2012	2013	\
India	27509	30933	33087	
China	28502	33024	34129	
United Kingdom of Great Britain and Northern Ir	6204	6195	5827	
Philippines	36765	34315	29544	
Pakistan	7468	11227	12603	
United States of America	7676	7891	8501	
Iran (Islamic Republic of)	7479	7534	11291	
Sri Lanka	3309	3338	2394	
Republic of Korea	4588	5316	4509	
Poland	720	779	852	
Lebanon	3072	1614		
			2172	
France	4080	6280	5623	
Jamaica	2059	2182	2479	
Viet Nam	1723	1731	2112	
Romania	1776	1588	1512	

```
Total
Country
India
                                                       691904
China
                                                       659962
United Kingdom of Great Britain and Northern Ir...
                                                       551500
Philippines
                                                       511391
Pakistan
                                                       241600
United States of America
                                                       241122
Iran (Islamic Republic of)
                                                       175923
Sri Lanka
                                                       148358
Republic of Korea
                                                       142581
Poland
                                                       139241
Lebanon
                                                       115359
France
                                                       109091
Jamaica
                                                       106431
Viet Nam
                                                        97146
Romania
                                                        93585
[15 rows x 38 columns]
```

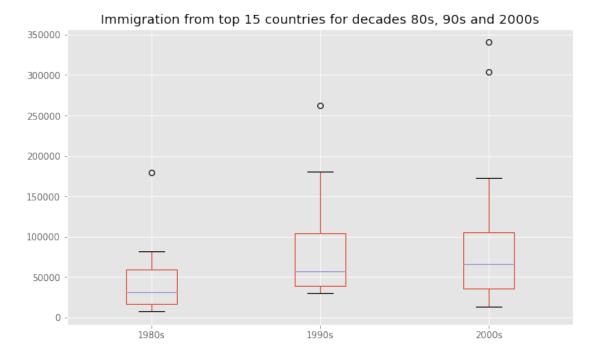
Step 2: Create a new dataframe which contains the aggregate for each decade. One way to do that: 1. Create a list of all years in decades 80's, 90's, and 00's. 2. Slice the original dataframe df_can to create a series for each decade and sum across all years for each country. 3. Merge the three series into a new data frame. Call your dataframe **new_df**.

```
In [20]: ### type your answer here
         years_80s = list(map(str, range(1980, 1990)))
         years_90s = list(map(str, range(1990, 2000)))
         years_00s = list(map(str, range(2000, 2010)))
         # slice the original dataframe df_can to create a series for each decade
         df_80s = df_top15.loc[:, years_80s].sum(axis=1)
         df_90s = df_top15.loc[:, years_90s].sum(axis=1)
         df_00s = df_top15.loc[:, years_00s].sum(axis=1)
         # merge the three series into a new data frame
         new_df = pd.DataFrame({'1980s': df_80s, '1990s': df_90s, '2000s':df_00s})
         # display dataframe
         new_df.head()
Out [20]:
                                                               1980s
                                                                       1990s
                                                                               2000s
         Country
         India
                                                               82154
                                                                      180395
                                                                              303591
         China
                                                               32003
                                                                      161528
                                                                              340385
         United Kingdom of Great Britain and Northern Ir...
                                                              179171
                                                                      261966
                                                                               83413
         Philippines
                                                               60764
                                                                      138482
                                                                              172904
         Pakistan
                                                               10591
                                                                       65302 127598
```

Let's learn more about the statistics associated with the dataframe using the describe() method.

Out[21]:		1980s	1990s	2000s
	count	15.000000	15.000000	15.000000
	mean	44418.333333	85594.666667	97471.533333
	std	44190.676455	68237.560246	100583.204205
	min	7613.000000	30028.000000	13629.000000
	25%	16698.000000	39259.000000	36101.500000
	50%	30638.000000	56915.000000	65794.000000
	75%	59183.000000	104451.500000	105505.500000
	max	179171.000000	261966.000000	340385.000000

Step 3: Plot the box plots.



Note how the box plot differs from the summary table created. The box plot scans the data and identifies the outliers. In order to be an outlier, the data value must be: * larger than Q3 by at least 1.5 times the interquartile range (IQR), or, * smaller than Q1 by at least 1.5 times the IQR.

Let's look at decade 2000s as an example: * Q1 (25%) = 36,101.5 * Q3 (75%) = 105,505.5 * IQR = Q3 - Q1 = 69,404

Using the definition of outlier, any value that is greater than Q3 by 1.5 times IQR will be flagged as outlier.

Outlier > 105,505.5 + (1.5 * 69,404) Outlier > 209,611.5

China and India are both considered as outliers since their population for the decade exceeds 209,611.5.

The box plot is an advanced visualization tool, and there are many options and customizations that exceed the scope of this lab. Please refer to Matplotlib documentation on box plots for more information.

6 Scatter Plots

A scatter plot (2D) is a useful method of comparing variables against each other. Scatter plots look similar to line plots in that they both map independent and dependent variables on a 2D graph. While the datapoints are connected together by a line in a line plot, they are not connected in a scatter plot. The data in a scatter plot is considered to express a trend. With further analysis using tools like regression, we can mathematically calculate this relationship and use it to predict trends outside the dataset.

Let's start by exploring the following:

Using a scatter plot, let's visualize the trend of total immigrantion to Canada (all countries combined) for the years 1980 - 2013.

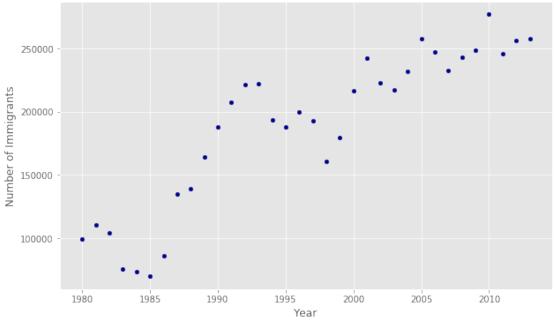
Step 1: Get the dataset. Since we are expecting to use the relationship betewen years and total population, we will convert years to int type.

```
In [24]: # we can use the sum() method to get the total population per year
         df_tot = pd.DataFrame(df_can[years].sum(axis=0))
         # change the years to type int (useful for regression later on)
         df_tot.index = map(int, df_tot.index)
         # reset the index to put in back in as a column in the df_tot dataframe
         df_tot.reset_index(inplace = True)
         # rename columns
         df_tot.columns = ['year', 'total']
         # view the final dataframe
        df_tot.head()
Out [24]:
           year
                   total
        0 1980
                   99137
         1 1981 110563
         2 1982 104271
         3 1983 75550
         4 1984 73417
```

Step 2: Plot the data. In Matplotlib, we can create a scatter plot set by passing in kind='scatter' as plot argument. We will also need to pass in x and y keywords to specify the columns that go on the x- and the y-axis.

```
In [25]: df_tot.plot(kind='scatter', x='year', y='total', figsize=(10, 6), color='darkblue')
    plt.title('Total Immigration to Canada from 1980 - 2013')
    plt.xlabel('Year')
    plt.ylabel('Number of Immigrants')
    plt.show()
```





Notice how the scatter plot does not connect the datapoints together. We can clearly observe an upward trend in the data: as the years go by, the total number of immigrants increases. We can mathematically analyze this upward trend using a regression line (line of best fit).

So let's try to plot a linear line of best fit, and use it to predict the number of immigrants in 2015.

Step 1: Get the equation of line of best fit. We will use **Numpy**'s polyfit() method by passing in the following: - x: x-coordinates of the data. - y: y-coordinates of the data. - deg: Degree of fitting polynomial. 1 = linear, 2 = quadratic, and so on.

```
Out[26]: array([ 5.56709228e+03, -1.09261952e+07])
```

The output is an array with the polynomial coefficients, highest powers first. Since we are plotting a linear regression y= a*x + b, our output has 2 elements [5.56709228e+03, -1.09261952e+07] with the slope in position 0 and intercept in position 1. Step 2: Plot the regression line on the scatter plot.

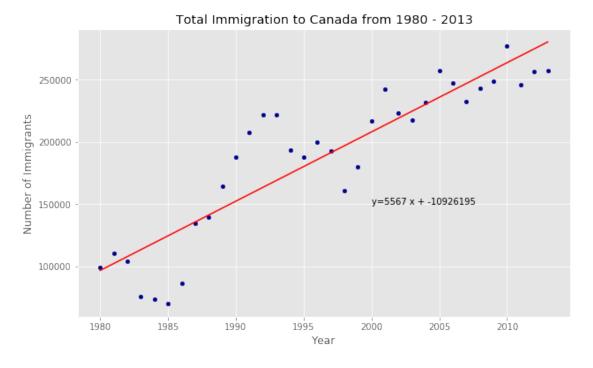
```
In [27]: df_tot.plot(kind='scatter', x='year', y='total', figsize=(10, 6), color='darkblue')

plt.title('Total Immigration to Canada from 1980 - 2013')
plt.xlabel('Year')
plt.ylabel('Number of Immigrants')

# plot line of best fit
plt.plot(x, fit[0] * x + fit[1], color='red') # recall that x is the Years
plt.annotate('y={0:.0f} x + {1:.0f}'.format(fit[0], fit[1]), xy=(2000, 150000))

plt.show()

# print out the line of best fit
'No. Immigrants = {0:.0f} * Year + {1:.0f}'.format(fit[0], fit[1])
```



```
Out[27]: 'No. Immigrants = 5567 * Year + -10926195'
```

Using the equation of line of best fit, we can estimate the number of immigrants in 2015:

```
No. Immigrants = 5567 * Year - 10926195
No. Immigrants = 5567 * 2015 - 10926195
No. Immigrants = 291,310
```

When compared to the actuals from Citizenship and Immigration Canada's (CIC) 2016 Annual Report, we see that Canada accepted 271,845 immigrants in 2015. Our estimated value of 291,310 is within 7% of the actual number, which is pretty good considering our original data came from United Nations (and might differ slightly from CIC data).

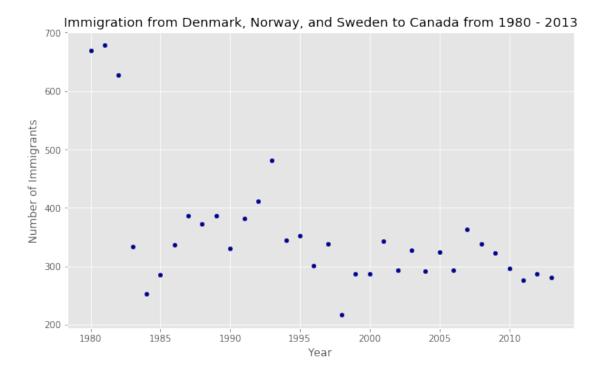
As a side note, we can observe that immigration took a dip around 1993 - 1997. Further analysis into the topic revealed that in 1993 Canada introcuded Bill C-86 which introduced revisions to the refugee determination system, mostly restrictive. Further amendments to the Immigration Regulations cancelled the sponsorship required for "assisted relatives" and reduced the points awarded to them, making it more difficult for family members (other than nuclear family) to immigrate to Canada. These restrictive measures had a direct impact on the immigration numbers for the next several years.

Question: Create a scatter plot of the total immigration from Denmark, Norway, and Sweden to Canada from 1980 to 2013?

Step 1: Get the data: 1. Create a dataframe the consists of the numbers associated with Denmark, Norway, and Sweden only. Name it **df_countries**. 2. Sum the immigration numbers across all three countries for each year and turn the result into a dataframe. Name this new dataframe **df_total**. 3. Reset the index in place. 4. Rename the columns to **year** and **total**. 5. Display the resulting dataframe.

```
In [47]: # create df_countries dataframe
         df_countries = df_can.loc[['Denmark', 'Norway', 'Sweden'], years].transpose()
          # create df_total by summing across three countries for each year
         df_total = pd.DataFrame(df_countries.sum(axis=1))
         # change the years to type int (useful for regression later on)
         df_total.index = map(int, df_total.index)
         # reset the index to put in back in as a column in the df_tot dataframe
         df_total.reset_index(inplace = True)
         # rename columns
         df_total.columns = ['year', 'total']
         # view the final dataframe
         df_total.head()
Out [47]:
            year
                 total
         0 1980
                    669
         1 1981
                    678
         2 1982
                    627
         3 1983
                    333
         4 1984
                    252
```

Step 2: Generate the scatter plot by plotting the total versus year in **df_total**.



In []:

7 Bubble Plots

A bubble plot is a variation of the scatter plot that displays three dimensions of data (x, y, z). The datapoints are replaced with bubbles, and the size of the bubble is determined by the third variable 'z', also known as the weight. In maplotlib, we can pass in an array or scalar to the keyword s to plot(), that contains the weight of each point.

Let's start by analyzing the effect of Argentina's great depression.

Argentina suffered a great depression from 1998 - 2002, which caused widespread unemployment, riots, the fall of the government, and a default on the country's foreign debt. In terms of income, over 50% of Argentines were poor, and seven out of ten Argentine children were poor at the depth of the crisis in 2002.

Let's analyze the effect of this crisis, and compare Argentina's immigration to that of it's neighbour Brazil. Let's do that using a bubble plot of immigration from Brazil and Argentina for the years 1980 - 2013. We will set the weights for the bubble as the *normalized* value of the population for each year.

Step 1: Get the data for Brazil and Argentina. Like in the previous example, we will convert the Years to type int and bring it in the dataframe.

```
In [32]: df_can_t = df_can[years].transpose() # transposed dataframe
         # cast the Years (the index) to type int
         df_can_t.index = map(int, df_can_t.index)
         # let's label the index. This will automatically be the column name when we reset the a
         df_can_t.index.name = 'Year'
          # reset index to bring the Year in as a column
         df_can_t.reset_index(inplace=True)
          # view the changes
         df_can_t.head()
Out[32]: Country Year Afghanistan
                                       Albania
                                                 Algeria American Samoa
                                                                            Andorra
                   1980
                                   16
                                              1
                                                       80
                                                                          0
                                                                                   0
         1
                   1981
                                   39
                                              0
                                                       67
                                                                          1
                                                                                   0
                                                                                            3
         2
                   1982
                                   39
                                              0
                                                                          0
                                                                                   0
                                                                                            6
                                                       71
         3
                   1983
                                   47
                                              0
                                                       69
                                                                          0
                                                                                   0
                                                                                            6
         4
                   1984
                                   71
                                              0
                                                       63
                                                                          0
                                                                                   0
                                                                                            4
         Country Antigua and Barbuda
                                         Argentina
                                                      Armenia
         0
                                                 368
                                                                . . .
         1
                                       0
                                                 426
                                                                . . .
         2
                                       0
                                                 626
                                                             0
                                                                . . .
         3
                                       0
                                                 241
                                                             0
                                                                . . .
         4
                                      42
                                                 237
                                                             0
                                                                . . .
         Country United States of America Uruguay
                                                         Uzbekistan
                                                                      Vanuatu
         0
                                         9378
                                                    128
                                                                   0
                                                                             0
         1
                                        10030
                                                                   0
                                                    132
                                                                             0
         2
                                         9074
                                                    146
                                                                   0
                                                                             0
         3
                                         7100
                                                                   0
                                                                             0
                                                    105
         4
                                         6661
                                                     90
                                                                   0
                                                                             0
         Country Venezuela (Bolivarian Republic of)
                                                          Viet Nam Western Sahara
                                                                                       Yemen
                                                                                   0
                                                                                           1
         0
                                                     103
                                                               1191
                                                                                           2
         1
                                                                                   0
                                                     117
                                                               1829
         2
                                                     174
                                                               2162
                                                                                   0
                                                                                           1
         3
                                                     124
                                                               3404
                                                                                   0
                                                                                           6
         4
                                                               7583
                                                                                           0
                                                     142
                                                                                   0
```

Country	Zambia	Zimbabwe
0	11	72
1	17	114
2	11	102
3	7	44
4	16	32

[5 rows x 196 columns]

Step 2: Create the normalized weights.

There are several methods of normalizations in statistics, each with its own use. In this case, we will use feature scaling to bring all values into the range [0,1]. The general formula is:

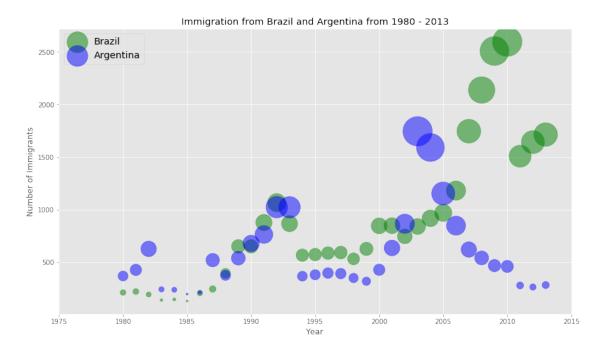
where X is an original value, X^{T} is the normalized value. The formula sets the max value in the dataset to 1, and sets the min value to 0. The rest of the datapoints are scaled to a value between 0-1 accordingly.

Step 3: Plot the data. - To plot two different scatter plots in one plot, we can include the axes one plot into the other by passing it via the ax parameter. - We will also pass in the weights using the s parameter. Given that the normalized weights are between 0-1, they won't be visible on the plot. Therefore we will: - multiply weights by 2000 to scale it up on the graph, and, - add 10 to compensate for the min value (which has a 0 weight and therefore scale with x2000).

```
In [34]: # Brazil
         ax0 = df_can_t.plot(kind='scatter',
                             x='Year',
                              y='Brazil',
                              figsize=(14, 8),
                                                          # transparency
                              alpha=0.5,
                              color='green',
                              s=norm_brazil * 2000 + 10, # pass in weights
                              xlim=(1975, 2015)
         # Argentina
         ax1 = df_can_t.plot(kind='scatter',
                              x='Year',
                              y='Argentina',
                              alpha=0.5,
                              color="blue",
                              s=norm_argentina * 2000 + 10,
                              ax = ax0
                             )
```

```
ax0.set_ylabel('Number of Immigrants')
ax0.set_title('Immigration from Brazil and Argentina from 1980 - 2013')
ax0.legend(['Brazil', 'Argentina'], loc='upper left', fontsize='x-large')
```

Out[34]: <matplotlib.legend.Legend at 0x7fc51c3e47b8>



The size of the bubble corresponds to the magnitude of immigrating population for that year, compared to the 1980 - 2013 data. The larger the bubble, the more immigrants in that year.

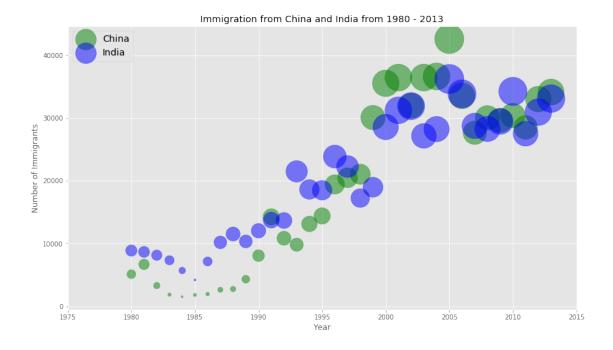
From the plot above, we can see a corresponding increase in immigration from Argentina during the 1998 - 2002 great depression. We can also observe a similar spike around 1985 to 1993. In fact, Argentina had suffered a great depression from 1974 - 1990, just before the onset of 1998 - 2002 great depression.

On a similar note, Brazil suffered the *Samba Effect* where the Brazilian real (currency) dropped nearly 35% in 1999. There was a fear of a South American financial crisis as many South American countries were heavily dependent on industrial exports from Brazil. The Brazilian government subsequently adopted an austerity program, and the economy slowly recovered over the years, culminating in a surge in 2010. The immigration data reflect these events.

Question: Previously in this lab, we created box plots to compare immigration from China and India to Canada. Create bubble plots of immigration from China and India to visualize any differences with time from 1980 to 2013. You can use **df_can_t** that we defined and used in the previous example.

Step 1: Normalize the data pertaining to China and India.

```
# normalize India data
         norm_india = (df_can_t['India'] - df_can_t['India'].min()) / (df_can_t['India'].max() -
   Step 2: Generate the bubble plots.
In [37]: ### type your answer here
         # China
         ax0 = df_can_t.plot(kind='scatter',
                             x='Year',
                             y='China',
                             figsize=(14, 8),
                             alpha=0.5,
                                                         # transparency
                             color='green',
                             s=norm_china * 2000 + 10, # pass in weights
                             xlim=(1975, 2015)
         ax1 = df_can_t.plot(kind='scatter',
                             x='Year',
                             y='India',
                             alpha=0.5,
                             color="blue",
                             s=norm\_india * 2000 + 10,
                             ax = ax0
         ax0.set_ylabel('Number of Immigrants')
         ax0.set_title('Immigration from China and India from 1980 - 2013')
         ax0.legend(['China', 'India'], loc='upper left', fontsize='x-large')
Out[37]: <matplotlib.legend.Legend at 0x7fc51c976400>
```



7.0.1 Thank you for completing this lab!

This notebook was created by Jay Rajasekharan with contributions from Ehsan M. Kermani, and Slobodan Markovic.

This notebook was recently revamped by Alex Aklson. I hope you found this lab session interesting. Feel free to contact me if you have any questions!

This notebook is part of a course on **Coursera** called *Data Visualization with Python*. If you accessed this notebook outside the course, you can take this course online by clicking here.

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