Handout for Session 8

1. Brief Overview of Data Analytics Using Pandas

```
Loading data
[1]: import pandas as pd
    base='https://raw.githubusercontent.com/chendaniely/pandas_for_everyone/master/data/'
     filename='gapminder.tsv'
    data=pd.read_csv(base+filename,sep='\t')
    data.head()
       country continent
                               lifeExp
                         year
                                                    gdpPercap
                                              pop
                                         8425333
0 Afghanistan
                   Asia 1952
                                 28.801
                                                  779.445314
1 Afghanistan
                   Asia 1957
                                 30.332
                                         9240934 820.853030
2 Afghanistan
                   Asia 1962
                                 3197 10267083 853.100710
3 Afghanistan
                   Asia 1967
                                 34.020 11537966 836.197138
4 Afghanistan
                   Asia 1972
                                36.088 13079460 739.981106
Obtaining Basic Information
[2]: data.shape
(1704, 6)
[3]: data.describe()
                                                  gdpPercap
             year
                       lifeExp
                                         pop
      1704.00000 1704.000000
                               1.704000e+03
                                                1704.000000
```

```
count
      1979.50000
                     59.474439
                                2.960121e+07
                                                7215.327081
mean
std
         17.26533
                     12.917107
                                1.061579e+08
                                                9857.454543
      1952.00000
                     23.599000 6.001100e+04
                                                 241.165877
min
25%
      1965.75000
                     48.198000
                                2.793664e+06
                                                1202.060309
50%
      1979.50000
                     60.712500 7.023596e+06
                                                3531.846989
                                                9325.462346
75%
      1993.25000
                     70.845500 1.958522e+07
      2007.00000
                     82.603000 1.318683e+09 113523.132900
max
```

```
[4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1704 entries, 0 to 1703
Data columns (total 6 columns):
country
             1704 non-null object
             1704 non-null object
continent
             1704 non-null int64
year
             1704 non-null float64
lifeExp
             1704 non-null int64
pop
gdpPercap
             1704 non-null float64
dtypes: float64(2), int64(2), object(2)
memory usage: 80.0+ KB
```

Manipulating data

```
[5]: data['gdp']=data['pop']*data['gdpPercap']/1e9
    data['pop']/=1e6
    data['gdpPercap']/=1e3
    data=data.set_index('year')
    data.head(3)
          country continent lifeExp
                                           pop gdpPercap
                                                                gdp
year
                             28.801
                                      8.425333
                                                 0.779445 6.567086
1952 Afghanistan
                      Asia
1957
     Afghanistan
                      Asia
                             30.332
                                      9.240934
                                                 0.820853 7.585449
1962 Afghanistan
                      Asia
                             31.997 10.267083
                                                 0.853101 8.758856
Filtering data
[6]: usa=data.query('country=="United States"')
     usa.head(3)
           country continent lifeExp
                                                gdpPercap
                                           pop
                                                                   gdp
year
                                68.44 157.553
                                                           2204.242423
1952 United States
                    Americas
                                                13.990482
1957 United States
                                69.49 171.984
                    Americas
                                                14.847127
                                                           2553.468311
1962 United States Americas
                                70.21 186.538 16.173146 3016.906282
[7]: usa=data[data['country']=='United States']
    usa.head(3)
           country continent lifeExp
                                                gdpPercap
                                           pop
                                                                   gdp
year
                                68.44 157.553
                                                13.990482
                                                           2204.242423
1952 United States
                    Americas
1957 United States
                    Americas
                                69.49 171.984
                                                14.847127
                                                           2553.468311
1962 United States
                                70.21 186.538 16.173146 3016.906282
                    Americas
[8]: gdp=usa['gdp']
    gdp.head()
year
1952
       2204.242423
1957
       2553.468311
1962
       3016.906282
1967
       3880.918003
1972
       4576.999720
Name: gdp, dtype: float64
Plotting data
[9]: import matplotlib.pyplot as plt
     gdp.plot(title='GDP of USA')
    plt.xlabel('Year')
    plt.ylabel('GDP')
    plt.show()
<Figure size 640x480 with 1 Axes>
```

2. Pandas Series Basics

2.1 Creating a Series Object

```
From list
```

```
[10]: import pandas as pd
      s=pd.Series([5,6,4])
0
     5
     6
1
2
     4
dtype: int64
[11]: s=pd.Series([5,6,4],index=['apple','orange','grape'])
      s
apple
          5
orange
          6
grape
          4
dtype: int64
From dictionary
[12]: s=pd.Series({'apple':5,'orange':6,'grape':4})
      s
apple
          5
orange
          6
grape
dtype: int64
From scratch
[13]: s=pd.Series()
      s['apple']=5
      s['orange']=6
      s['grape']=4
          5
apple
orange
          6
grape
          4
dtype: int64
2.2 Indexing a Series
```

Obtaining a single element

```
[14]: s[1]
6
[15]: s.iloc[1]
```

```
6
[16]: s.loc['orange']
6
Slicing a contiguous chunk.
[17]: s[:2]
apple
          5
orange
          6
dtype: int64
[18]: s.iloc[:2]
apple
          5
orange
dtype: int64
[19]: s.loc[:'orange']
apple
          5
orange
          6
dtype: int64
Fancy indexing: specifying a list of positions
[20]: s[[0,2]]
         5
apple
grape
         4
dtype: int64
[21]: s.iloc[[0,2]]
         5
apple
grape
         4
dtype: int64
[22]: s.loc[['apple','orange']]
apple
          5
orange
          6
dtype: int64
Boolean indexing: specifying whether to include each element
[23]: s[[True,False,True]]
apple
         5
         4
grape
dtype: int64
```

Q1-a: Create the following Series object using three ways.

```
Fritos 20
Cheetos 15
Lays 25
dtype: int64
```

Q1-b: Obtain the single element corresponding to "Lays" using five ways.

Q1-c: Obtain everything but the first element using at least four ways.

2.3 Manipulating a Series Object

Basic information

```
[36]: s.index
Index(['apple', 'orange', 'grape'], dtype='object')
[37]: s.values
array([5, 6, 4])
[38]: s.unique()
array([5, 6, 4])
[39]: s.describe()
count
         3.0
mean
         5.0
std
         1.0
         4.0
min
25%
         4.5
50%
         5.0
75%
         5.5
max
         6.0
dtype: float64
```

Arithmetics

```
[40]: s+1
apple
          6
          7
orange
grape
          5
dtype: int64
[41]: s>5
apple
          False
orange
          True
grape
          False
dtype: bool
[42]: s[s>5]
```

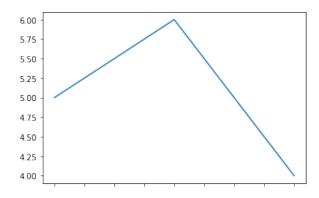
```
orange
          6
dtype: int64
[43]: (s>=6) | (s<=4)
          False
apple
orange
           True
grape
           True
dtype: bool
[44]: s[(s>=6) | (s<=4)]
          6
orange
grape
          4
dtype: int64
[45]: (s>=5) & (s<=6)
apple
           True
orange
           True
grape
          False
dtype: bool
[46]: s[(s>=5) & (s<=6)]
apple
          5
orange
          6
dtype: int64
```

Vectorized functions

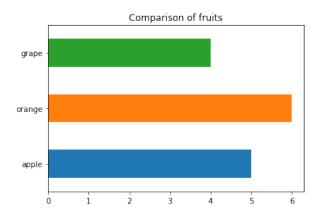
Plotting

[49]: s.plot()

<matplotlib.axes._subplots.AxesSubplot at 0x7f471e14f5f8>



<matplotlib.axes._subplots.AxesSubplot at 0x7f471e09e390>



Q2-a: Run the function describe on the Series gdp (from part 1 of this handout).

Q2-b: Write an expression divides the Series gdp by 1000 and round to 2 decimal places (using the round function).

Q2-c: Filter the Series gdp for values above 10000.

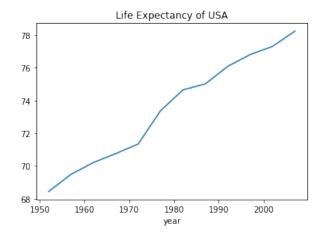
Q2-d: Obtain a Series corresponding to the life expectancy in USA when the GDP is above 10 trillion. (Hint: obtain the life expectancy column using usa['lifeExp'] and use boolean indexing on gdp1 as in Q2-b.)

Q2-e: Compute the average life expectancy in the data set for USA when the GDP is above 10 trillion. (Hint: call the function mean of the above Series.)

Q2-f: Plot the life expectancy of USA in the data set using a line plot.

[56]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f471e00ac50>



3. Pandas DataFrame Basics II

3.1 Creating a DataFrame Object

From list of lists

```
[57]: import pandas as pd
     df=pd.DataFrame([[5,3,'M'],[6,2,'M'],[4,1,'S']])
  0
     1
        2
  5 3 M
1 6 2 M
2 4 1 S
[58]: df=pd.DataFrame([[5,3,'M'],[6,2,'M'],[4,1,'S']],\
                      index=['apple','orange','grape'],\
                      columns=['Number','Rank','Size'])
     df
       Number Rank Size
apple
            5
                  3
orange
            6
                  2
                       Μ
                       S
grape
            4
                  1
```

From dictionary of columns

3.2 Indexing a DataFrame

grape

4

3

S

Obtaining a single element

```
[60]: df['Number'][0]
5
[61]: df.iloc[0,0]
5
[62]: df.loc['apple','Number']
5
```

Obtaining a column

```
[63]: df['Number']
apple
          5
orange
          6
grape
          4
Name: Number, dtype: int64
[64]: df.iloc[:,0]
apple
          5
orange
          6
grape
          4
Name: Number, dtype: int64
[65]: df.loc[:,'Number']
apple
          5
orange
          6
grape
Name: Number, dtype: int64
Obtaining a row
[66]: df.iloc[1,:]
Number
          6
Rank
          1
Size
          М
Name: orange, dtype: object
[67]: df.loc['orange',:]
Number
          6
Rank
          1
Size
          Μ
Name: orange, dtype: object
Obtaining particular rows
[68]: df.iloc[[0,2],:]
       Number Rank Size
apple
            5
                  2
            4
                  3
                       S
grape
[69]: df.loc[['apple','grape'],:]
       Number
               Rank Size
            5
                  2
apple
                       Μ
            4
                  3
                       S
grape
[70]: # Boolean indexing for Dataframes selects rows
      df[[True,False,True]]
```

	Number	Rank	Size
apple	5	2	M
grape	4	3	S

Q3-a: Obtain the second column of the DataFrame df in at least three ways.

Q3-b: Obtain the second and third row of the DataFrame df in at least five ways.

Q3-c: Obtain the rank of orange in at least four ways.

Q4-a: Obtain the set of unique continents in the DataFrame data. (Hint: use the function unique associated with the Series data['continent'].)

Q4-b: Filter for the rows of the DataFrame data for which the continent is "Americas", year is 2007, and GDP is at least 1000. (You can either use the query function associated with the DataFrame or boolean indexing.)

Q4-c: Compute the average gdpPercap of the countries in the Americas in 1952, and also in 2007. (No need to do population weighted average.)

Q4-d: Create a bar graphs of the gdpPercap of countries in the Americas for the year 2007. (Optional: sort the bars in descending order.)

[90]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f471e14f550>

