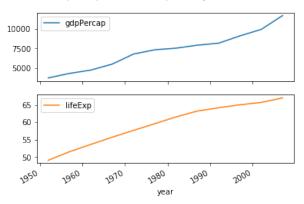
Session 14: Data Aggregation using Groupby (with Solutions)

Setting up

dtype=object)

```
[1]: import pandas as pd
    base='https://raw.githubusercontent.com/chendaniely/pandas_for_everyone/master/data/'
    filename='gapminder.tsv'
    gapminder=pd.read_csv(base+filename,sep='\t')
    gapminder.head()
      country continent year lifeExp
                                                   gdpPercap
                                             pop
0 Afghanistan
                                28.801
                   Asia 1952
                                         8425333 779.445314
1 Afghanistan
                   Asia 1957
                                30.332
                                         9240934 820.853030
                                31.997 10267083 853.100710
2 Afghanistan
                   Asia 1962
3 Afghanistan
                   Asia 1967
                                34.020 11537966 836.197138
4 Afghanistan
                   Asia 1972
                                36.088 13079460 739.981106
1. Grouping by One Column
[2]: gapminder.groupby('year')['lifeExp'].mean().head()
year
1952
       49.057620
1957
       51.507401
1962
       53.609249
1967
       55.678290
1972
       57.647386
Name: lifeExp, dtype: float64
[3]: gapminder.groupby('year')['lifeExp'].mean()\
         .plot(title='Average life expectancy of all countries')
<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e2726c18>
[4]: gapminder.groupby('year')[['gdpPercap','lifeExp']].mean().head()
       gdpPercap
                    lifeExp
year
1952 3725.276046 49.057620
1957 4299.408345 51.507401
1962 4725.812342 53.609249
1967 5483.653047 55.678290
1972 6770.082815 57.647386
[5]: gapminder.groupby('year')[['gdpPercap','lifeExp']].mean()\
         .plot(title='GDP per capita and life expectancy of all countries', subplots=True)
array([<matplotlib.axes._subplots.AxesSubplot object at 0x7fe3e237d0b8>,
       <matplotlib.axes._subplots.AxesSubplot object at 0x7fe3e2326668>],
```

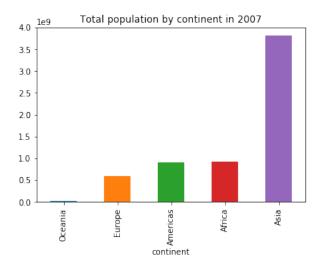




List of Pandas methods built into groupby:

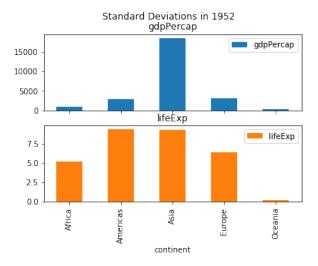
Method	Description
count	Number of elements not including NaN values.
size	Number of elements including NaNs.
sum	Total.
mean	Average.
median	Median.
var	Variance.
std	Standard deviation.
min	Minimum.
max	Maximum.
quantile(0.28)	28th percentile from the bottom.
describe	count, mean, std, min, 25%, 50%, 75%, and max
first	first non-null value.
last	last non-null value.
nth(3)	3rd value (does not skip null values).

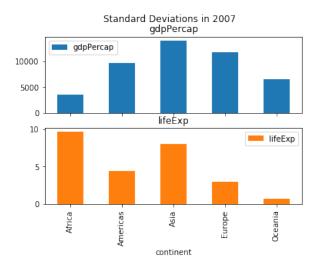
Q1: Create a bar plot comparing the total populations of each continent in 2007, as below. (Hint: First use "query" to filter for year being 2007, then group by the continent and compute the sum of the "pop" column. Then sort the result using "sort_values" and plot using "kind='bar'". All this can be chained together into one line.)



Q2: Create the following plots to compare the standard deviation in GDP per capita and life expectancy across countries within each continent, in 1952 and in 2007.

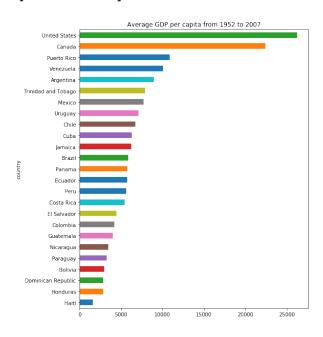
(Hint: for each graph, first use "query" to filter for the year, then group by the continent, and compute the standard deviation of both "gdpPercap" and "lifeExp". Then plot using "subplots=True". Each plot can be created using one line by chaining together commands.)





Q3: Plot the average GDP per capita over the years in the dataset for all countries in the continent "Americas", as below.

<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e2025320>



2. Grouping by Multiple Columns

Grouping by multiple columns is completely analogous, except for the result having multiple levels of index.

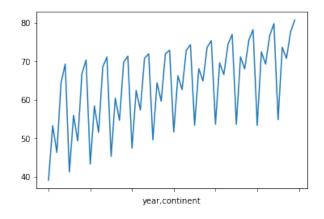
```
Americas
                   53.279840
                   46.314394
      Asia
     Europe
                   64.408500
      Oceania
                   69.255000
1957
     Africa
                   41.266346
      Americas
                   55.960280
      Asia
                   49.318544
                   66.703067
      Europe
      Oceania
                   70.295000
Name: lifeExp, dtype: float64
[12]: gapminder.groupby(['year','continent']).std().head()
                 lifeExp
                                           gdpPercap
                                   pop
year continent
1952 Africa
                5.151581 6.317450e+06
                                          982.952116
                9.326082 3.234163e+07
                                         3001.727522
     Americas
     Asia
                9.291751 1.132267e+08 18634.890865
     Europe
                6.361088 1.724745e+07
                                         3114.060493
     Oceania
                0.190919 4.735083e+06
                                          365.560078
[13]: gapminder.groupby(['year','continent'])\
          .agg({'lifeExp':'mean','gdpPercap':'median'}).head()
                  lifeExp
                              gdpPercap
year continent
1952 Africa
                39.135500
                             987.025569
                53.279840
     Americas
                            3048.302900
     Asia
                46.314394
                            1206.947913
     Europe
                64.408500
                            5142.469716
     Oceania
                69.255000
                          10298.085650
```

2.1 Working with MultiIndex Objects

The Series "result" has two levels of index, year and continent. This is an example of a Hierarchical Index or MultiIndex in Pandas. The data is the rightmost column above, while the two columns on the left are both row labels.

```
[14]: result.plot()
```

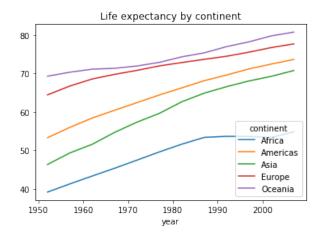
<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e1f095f8>



You can change a Series to a DataFrame by the command "reset_index", which get rid of the MultiIndex.

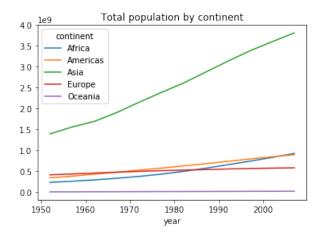
```
[15]: result.reset_index().head()
  year continent
                    lifeExp
          Africa 39.135500
 1952
 1952 Americas 53.279840
2 1952
            Asia 46.314394
3 1952
          Europe 64.408500
4 1952
         Oceania 69.255000
[16]: result.reset_index().groupby('continent')['lifeExp'].std()
continent
Africa
           5.443393
           6.663477
Americas
Asia
           8.236002
Europe
           4.042574
Oceania
           3.843914
Name: lifeExp, dtype: float64
 The "unstack" method moves one level of the MultiIndex to the columns, as below. The
default behavior is to move the right most level.
[17]: result.unstack().head()
continent
             Africa Americas
                                    Asia
                                             Europe Oceania
year
          39.135500 53.27984 46.314394 64.408500
                                                      69.255
1952
1957
          41.266346 55.96028 49.318544 66.703067
                                                      70.295
1962
          43.319442 58.39876 51.563223 68.539233
                                                      71.085
1967
          45.334538 60.41092 54.663640 69.737600
                                                      71.310
          47.450942 62.39492 57.319269 70.775033
1972
                                                      71.910
[18]: result.unstack(level=0).iloc[:5,:5]
year
                                                           1972
                1952
                          1957
                                     1962
                                                 1967
continent
          39.135500 41.266346 43.319442 45.334538
Africa
                                                      47.450942
Americas
          53.279840 55.960280 58.398760 60.410920
                                                      62.394920
Asia
          46.314394 49.318544 51.563223 54.663640
                                                      57.319269
Europe
          64.408500 66.703067 68.539233 69.737600
                                                      70.775033
Oceania
          69.255000 70.295000 71.085000 71.310000 71.910000
[19]: gapminder.groupby(['year','continent'])['lifeExp'].mean()\
          .unstack().plot(title="Life expectancy by continent")
```

<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e1ecb0f0>



Q4: Plot the trend in total population of each continent as below.

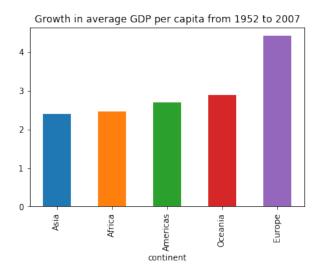
<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e1ebfb38>



Q5: Compute the average GDP per capita for each continent in 1952 and 2007, and plot the ratio.

(Hint: a quick way is to first group by the continent and year and compute the average GDP Per capita for each combination, then unstack it so that the years are the columns, similar to in Out[30]. Then you can compute the desired ratio by dividing the column for 2007 by the column for 1952.)

```
1952
                                   2007
year
continent
Africa
            1252.572466
                           3089.032605
Americas
            4079.062552
                          11003.031625
Asia
            5195.484004
                          12473.026870
                          25054.481636
Europe
            5661.057435
Oceania
           10298.085650
                          29810.188275
```



Q6. Create a plot over time of the difference in total GDP between the richest and the poorest continent, as below.

(Hint: first add a "GDP" column in the gapminder DataFrame by multiplying the "gdpPercap" and "pop" columns. Then create a "gdpSum" DataFrame by grouping by the year and continent, and summing the GDPs. See below for what this DataFrame looks like. Using this DataFrame, you can compute a Series called "maxGDP" by grouping by the year and finding the max GDP, and similarly compute a Series called "minGDP". Both of these are indexed by year. Finally, subtract maxGDP by minGDP and plot the result.)

```
[23]: gapminder['GDP']=gapminder['gdpPercap']*gapminder['pop']
      gapminder.groupby(['year','continent'])['GDP'].sum().reset_index().head()
  year continent
  1952
           Africa 3.115993e+11
  1952 Americas 2.943475e+12
2
  1952
             Asia 1.125160e+12
3
  1952
           Europe 2.549140e+12
  1952
          Oceania 1.083144e+11
[24]: gapminder.groupby(['year','continent'])['GDP'].sum().reset_index()\
          .groupby('year')['GDP'].max()
year
1952
        2.943475e+12
1957
        3.520427e+12
1962
        4.228827e+12
1967
        5.446688e+12
        6.703979e+12
1972
1977
        8.102135e+12
1982
        9.082850e+12
1987
        1.098619e+13
1992
        1.224750e+13
1997
        1.456733e+13
2002
        1.653122e+13
2007
        2.070795e+13
Name: GDP, dtype: float64
```

<matplotlib.axes._subplots.AxesSubplot at 0x7fe3e1c51278>

