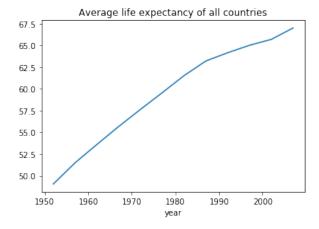
Session 14: Data Aggregation using Groupby

Setting up

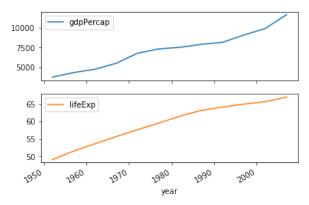
```
[2]: import pandas as pd
    base='https://raw.githubusercontent.com/chendaniely/pandas_for_everyone/master/data/'
     filename='gapminder.tsv'
    gapminder=pd.read_csv(base+filename,sep='\t')
    gapminder.head()
                                                   gdpPercap
      country continent year
                               lifeExp
                                             pop
                                28.801
0 Afghanistan
                   Asia
                         1952
                                         8425333 779.445314
1 Afghanistan
                   Asia 1957
                                30.332
                                         9240934 820.853030
2 Afghanistan
                   Asia 1962
                                31.997
                                        10267083 853.100710
3 Afghanistan
                                34.020
                                        11537966 836.197138
                   Asia 1967
4 Afghanistan
                   Asia 1972
                                36.088
                                        13079460 739.981106
```

1. Grouping by One Column

1972 6770.082815 57.647386





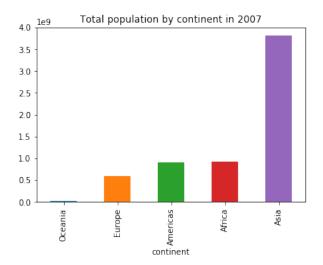


List of Pandas methods built into groupby:

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.)

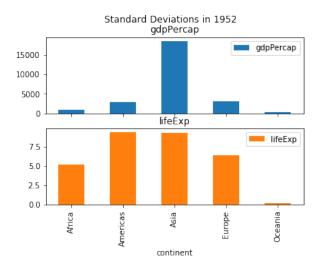
[7]:



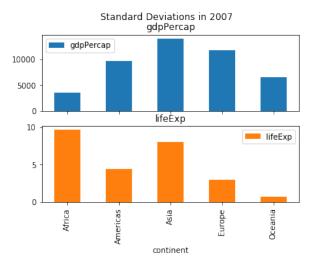
Q2: Create the following plots to compare the standard deviation in GDP per capita and life expectancy across countries within each continent, in 1952 and in 2007.

(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.)

[8]:

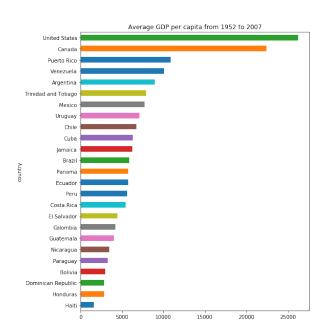


[9]:



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

[10]:



2. Grouping by Multiple Columns

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

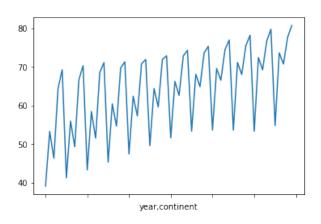
```
[11]: result=gapminder.groupby(['year','continent'])['lifeExp'].mean()
      result.head(10)
year continent
1952
                   39.135500
     Africa
      Americas
                  53.279840
      Asia
                   46.314394
      Europe
                   64.408500
                  69.255000
      Oceania
1957
     Africa
                   41.266346
      Americas
                  55.960280
      Asia
                   49.318544
     Europe
                   66.703067
      Oceania
                   70.295000
Name: lifeExp, dtype: float64
[12]: gapminder.groupby(['year','continent']).std().head()
                 lifeExp
                                   pop
                                           gdpPercap
year continent
                                          982.952116
1952 Africa
                5.151581 6.317450e+06
    Americas
                9.326082 3.234163e+07
                                         3001.727522
                9.291751 1.132267e+08 18634.890865
    Asia
    Europe
                6.361088 1.724745e+07
                                         3114.060493
                                          365.560078
    Oceania
                0.190919 4.735083e+06
```

```
[13]: gapminder.groupby(['year','continent'])\
          .agg({'lifeExp':'mean','gdpPercap':'median'}).head()
                  lifeExp
                              gdpPercap
year continent
1952 Africa
                39.135500
                             987.025569
     Americas
                53.279840
                            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()



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
  1952
          Africa 39.135500
0
  1952 Americas 53.279840
2 1952
            Asia 46.314394
 1952
          Europe 64.408500
3
4 1952
         Oceania 69.255000
[16]: result.reset_index().groupby('continent')['lifeExp'].std()
continent
Africa
           5.443393
Americas
           6.663477
Asia
           8.236002
Europe
           4.042574
Oceania
            3.843914
Name: lifeExp, dtype: float64
```

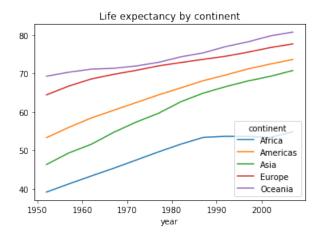
The "unstack" method moves one level of the MultiIndex to the columns, as below. The default behavior is to move the right most level.

[17]: result.unstack().head()

continent	Africa	Americas	Asia	Europe	Oceania
year					
1952	39.135500	53.27984	46.314394	64.408500	69.255
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
1972	47.450942	62.39492	57.319269	70.775033	71.910

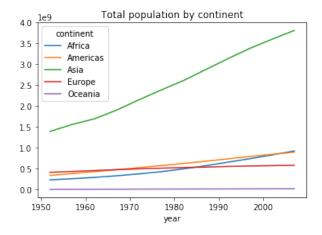
[18]: result.unstack(level=0).iloc[:5,:5]

year	1952	1957	1962	1967	1972
continent					
Africa	39.135500	41.266346	43.319442	45.334538	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



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

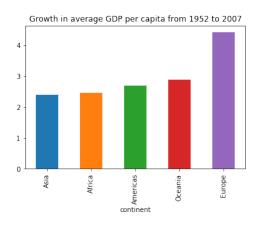
[20]:



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.)

[22]:



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]:

	year	continent	GDP
0	1952	Africa	3.115993e+11
1	1952	Americas	2.943475e+12
2	1952	Asia	1.125160e+12

[26]:

