

Data Visualization

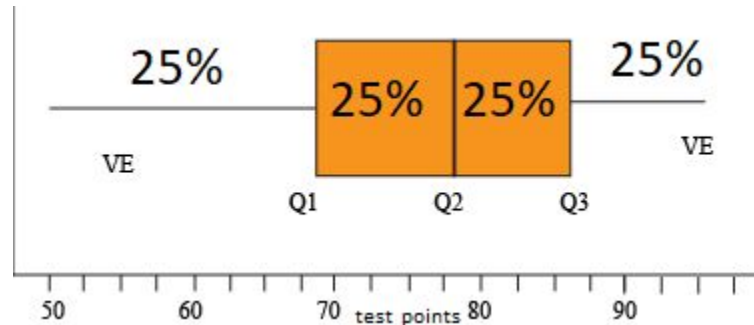
How to visualize data. Graphs, heatmap, pychart,...

How to call data.

Python Package named as **seaborn** (seaborn.pydata.org) statistical data visualization, based on **matplotlib**. You can install with `pip install seaborn` (documentation->Installing and getting started. It requires **numpy**, **scipy**, **matplotlib**, **pandas** package.

On gallery section of seaborn web we have a lot of different visualizations:

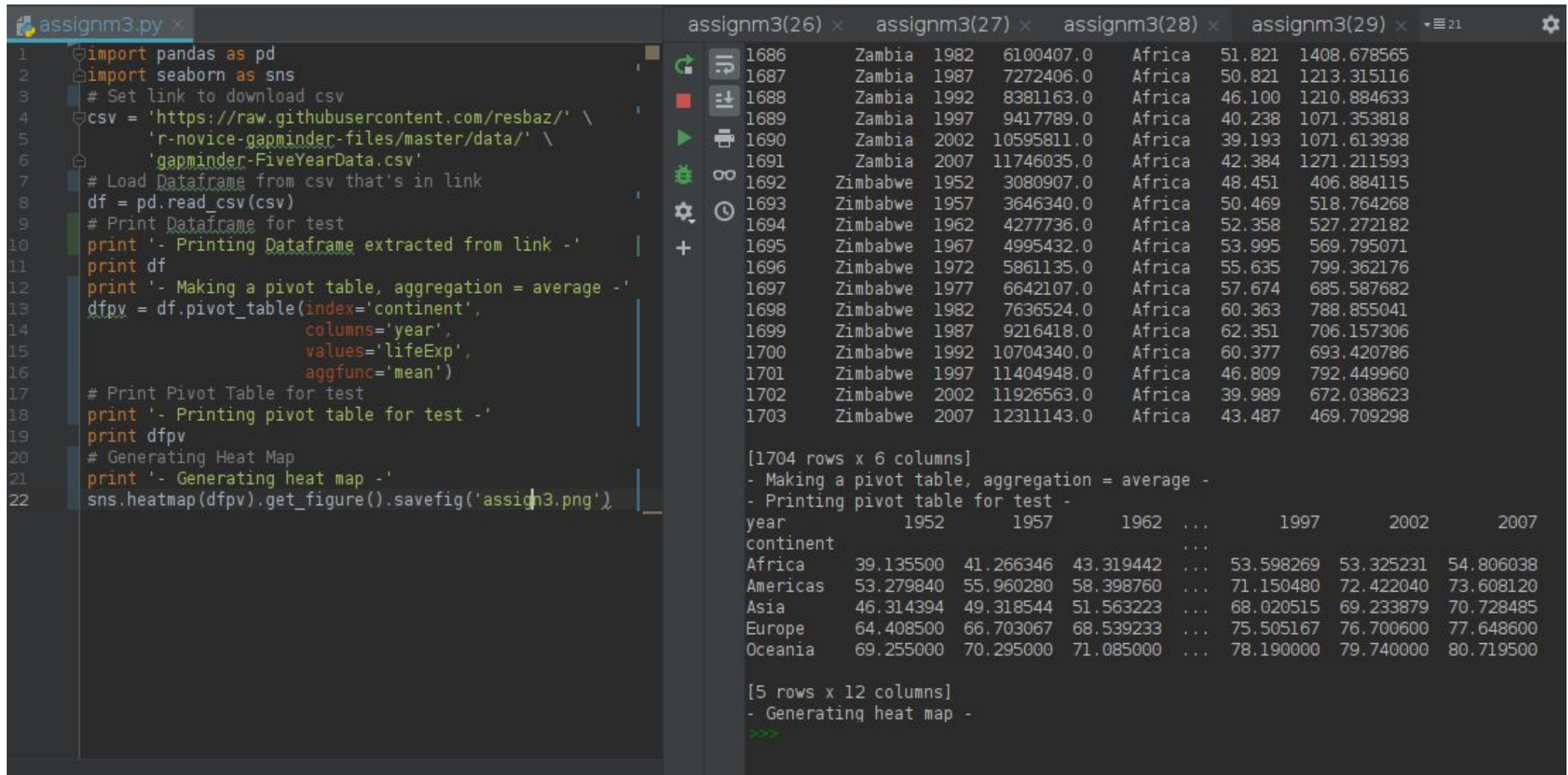
- ❑ Boxplots: source is sort the data and get the minimum, maximum, quartiles Q1(25%, lower median half), Q2(50% all data median) and Q3(75%, upper median half) and interquartile range ($IQR = Q3 - Q1$). Are a useful way to graph data divided into four quartiles, each with the same amount of values. The box diagram does not graph frequency or show individual statistics, but in them we can clearly see where half the data is. It is a good diagram to analyze the asymmetry in the data. Values below $Q1 - 1.5 \cdot IQR$ or higher than $Q3 + 1.5 \cdot IQR$ are considered outliers



- ❑ Heatmaps (anotated): color represents values, eg. pivot table. Color represents values of the data.
- ❑ Join Plot: Hexagonal bin plot with marginal distributions.
- ❑ Histograms: accurate representation of the distribution of numerical data, is an estimate of the probability distribution of a continuous variable. First step is to "bin" or "bucket" the range of values, that is, divide the entire range of values into a series of non overlapping intervals, that must be adjadcent, but not required of equal size; then count how many values fall into each interval. Histogram can be normalized to display relative frequencies; it shows the proportion of cases that fall into each of several categories, with the sum of the heights equaling 1. When bins are not equal width, vertical axis is not a frequency, is a frequency density.
- ❑ connell plots
- ❑

Assignment #3

It keeps simple: Download csv as dataframe, and make pivot table, after render this data with seaborn.



The screenshot displays a Jupyter Notebook interface with a code editor on the left and a console output on the right. The code in the editor performs the following steps:

- Imports `pandas` as `pd` and `seaborn` as `sns`.
- Defines a CSV link: `csv = 'https://raw.githubusercontent.com/resbaz/' \ 'r-novice-gapminder-files/master/data/' \ 'gapminder-FiveYearData.csv'`
- Loads the CSV into a DataFrame: `df = pd.read_csv(csv)`
- Prints the DataFrame for testing: `print df`
- Creates a pivot table with 'continent' as the index, 'year' as columns, and 'lifeExp' as values, using 'mean' for aggregation: `dfpv = df.pivot_table(index='continent', columns='year', values='lifeExp', aggfunc='mean')`
- Prints the pivot table for testing: `print dfpv`
- Generates a heatmap: `sns.heatmap(dfpv).get_figure().savefig('assign3.png')`

The console output shows the execution of these steps, including the DataFrame printout and the pivot table printout. The pivot table is a 5x7 matrix showing mean life expectancy for Africa, Americas, Asia, Europe, and Oceania across the years 1952, 1957, 1962, 1967, 1972, 1977, and 1982. The heatmap is also generated and saved as 'assign3.png'.

continent	1952	1957	1962	1967	1972	1977	1982
Africa	39.135500	41.266346	43.319442	44.3598269	45.325231	46.806038	48.06038
Americas	53.279840	55.960280	58.398760	60.71150480	62.422040	64.608120	66.608120
Asia	46.314394	49.318544	51.563223	53.68020515	55.233879	57.728485	59.728485
Europe	64.408500	66.703067	68.539233	70.75505167	72.7600600	74.648600	76.648600
Oceania	69.255000	70.295000	71.085000	72.190000	73.740000	75.719500	77.719500

A pivot table is an aggregation method to summarize info from several dimensions as rows and columns.