

1 Lecture 3b: Source code for plots from Lecture 3a + introduction to Seaborn and Plotly libraries

Data Visualization · 1-DAV-105

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This notebook contains the source code for all the plots shown in the first part of the lecture. It also introduces two new plotting libraries: Seaborn and Plotly.

1.1 Seaborn library

- [Seaborn](#) library is an extension of Matplotlib.
- Seaborn is more convenient for many types of plots; we will use it for more complex scatter plots and line plots, for bar plots, strip plots, histograms and heatmaps.
- In Seaborn functions, a whole DataFrame can be added using option `data=`. DataFrame column names are then used as `x`, `y`, `hue` (color), `col` (one of subfigures).
- Seaborn creates Matplotlib objects (e.g. figure, axes) which can be then modified using Matplotlib methods.
- The first example of this library is in section [Categorical variable via color](#)

1.2 Plotly library for interactive plots

- Another popular library is [Plotly](#).
- It provides some additional plot types and all plots are interactive.
- For example, in the [scatter plot](#), we can find information about each dot by hovering a mouse over it.
- We can also zoom into parts of the plot by selecting a rectangle.
- A menu with additional options appears in the top right corner of the plot.
- Plotly is also used the first time in section [Categorical variable via color](#).

1.3 Used libraries

```
[1]: import numpy as np
import pandas as pd
from IPython.display import Markdown
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
```

1.4 Importing World Bank data

Country indicators from World Bank, <https://databank.worldbank.org/home> under CC BY 4.0 license.

Country population, surface area in km squared, GDP per capita (current US\$), life expectancy at birth (years), fertility rate (births per woman); in years 2000, 2010, 2020.

```
[2]: url = 'https://bbrejova.github.io/viz/data/World_bank.csv'
countries = pd.read_csv(url).set_index('Country')
display(countries)
```

Country	ISO3	Region	Income Group \
Afghanistan	AFG	South Asia	Low income
Albania	ALB	Europe & Central Asia	Upper middle income
Algeria	DZA	Middle East & North Africa	Lower middle income
American Samoa	ASM	East Asia & Pacific	High income
Andorra	AND	Europe & Central Asia	High income
...
Virgin Islands	VIR	Latin America & Caribbean	High income
West Bank and Gaza	PSE	Middle East & North Africa	Upper middle income
Yemen	YEM	Middle East & North Africa	Low income
Zambia	ZMB	Sub-Saharan Africa	Lower middle income
Zimbabwe	ZWE	Sub-Saharan Africa	Lower middle income

Country	Population2000	Population2010	Population2020	Area \
Afghanistan	19542983.0	28189672.0	38972231.0	652860.0
Albania	3089026.0	2913021.0	2837849.0	28750.0
Algeria	30774621.0	35856344.0	43451666.0	2381741.0
American Samoa	58229.0	54849.0	46189.0	200.0
Andorra	66097.0	71519.0	77699.0	470.0
...
Virgin Islands	108642.0	108356.0	106291.0	350.0
West Bank and Gaza	2922153.0	3786161.0	4803269.0	6020.0
Yemen	18628701.0	24743945.0	32284046.0	527970.0
Zambia	9891135.0	13792087.0	18927715.0	752610.0
Zimbabwe	11834676.0	12839770.0	15669667.0	390760.0

Country	GDP2000	GDP2010	GDP2020	Expectancy2000 \
Afghanistan	NaN	562.499219	512.055098	55.298000
Albania	1126.683340	4094.349686	5343.037704	75.404000
Algeria	1780.376063	4495.921476	3354.153164	70.478000
American Samoa	NaN	10446.863206	15609.777220	NaN
Andorra	21620.465102	48237.890541	37207.238871	NaN
...
Virgin Islands	NaN	39905.128418	39411.045254	76.619512
West Bank and Gaza	1476.171850	2557.075624	3233.568638	70.388000
Yemen	519.591639	1249.063085	578.512010	62.588000
Zambia	364.026145	1469.361450	956.831729	45.231000
Zimbabwe	565.284390	937.840340	1372.696674	44.686000

Expectancy2010	Expectancy2020	Fertility2000 \
----------------	----------------	-----------------

Country			
Afghanistan	60.851000	62.575000	7.534
Albania	77.936000	76.989000	2.231
Algeria	73.808000	74.453000	2.566
American Samoa	NaN	NaN	NaN
Andorra	NaN	NaN	NaN
...
Virgin Islands	77.865854	79.819512	2.060
West Bank and Gaza	73.004000	74.403000	5.443
Yemen	67.280000	64.650000	6.318
Zambia	56.799000	62.380000	5.926
Zimbabwe	50.652000	61.124000	3.974

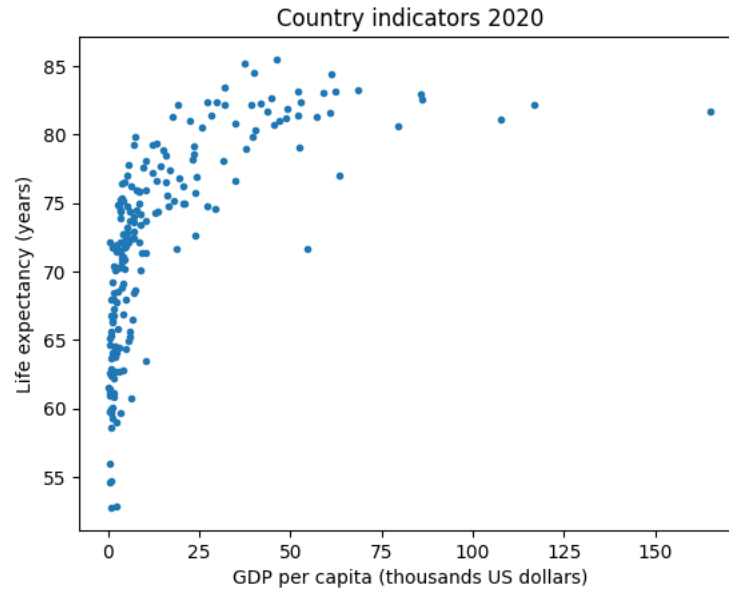
	Fertility2010	Fertility2020
Country		
Afghanistan	6.099	4.750
Albania	1.656	1.400
Algeria	2.843	2.942
American Samoa	NaN	NaN
Andorra	1.270	NaN
...
Virgin Islands	2.300	2.030
West Bank and Gaza	4.383	3.570
Yemen	4.855	3.886
Zambia	5.363	4.379
Zimbabwe	4.025	3.545

[217 rows x 16 columns]

1.5 A simple scatterplot

To create a simple scatterplot, commands from the previous lectures suffice. Note that we divide GDP by 1000 and add this information to the axis title. This makes the axis easier to read.

```
[3]: figure, axes = plt.subplots()
      axes.plot(countries.GDP2020 / 1000, countries.Expectancy2020, '.')
      axes.set_xlabel('GDP per capita (thousands US dollars)')
      axes.set_ylabel('Life expectancy (years)')
      axes.set_title('Country indicators 2020')
      pass
```



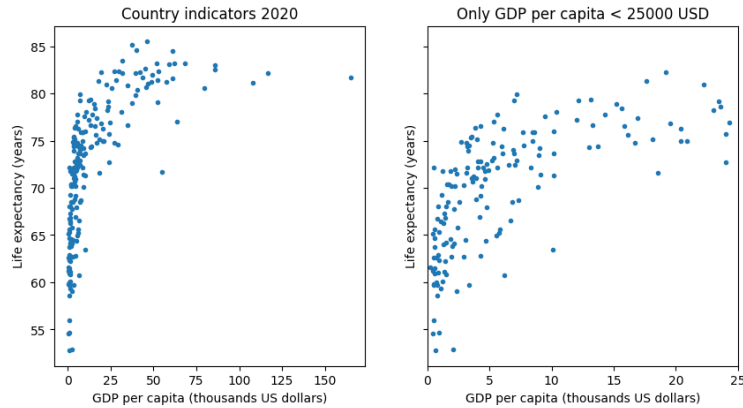
1.6 Zooming in

Limits on x axis are set using `set_xlim` method in order to zoom in on countries with lower GDP.

```
[4]: # create two subplots
figure, axes = plt.subplots(1, 2, figsize=(10, 5), sharey=True)

# the left subplot - full range of data
axes[0].plot(countries.GDP2020 / 1000, countries.Expectancy2020, '.')
axes[0].set_xlabel('GDP per capita (thousands US dollars)')
axes[0].set_ylabel('Life expectancy (years)')
axes[0].set_title('Country indicators 2020')

# the right subplot - smaller values of GDP
axes[1].plot(countries.GDP2020 / 1000, countries.Expectancy2020, '.')
axes[1].set_xlabel('GDP per capita (thousands US dollars)')
axes[1].set_ylabel('Life expectancy (years)')
axes[1].set_title('Only GDP per capita < 25000 USD')
axes[1].set_xlim(0, 25)
pass
```



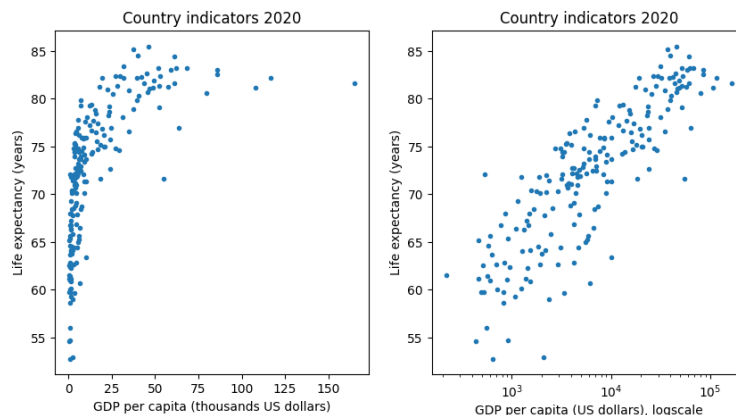
1.7 Log-scale plot

In this plot, the log-scale on the x-axis is switched on by `semilogx` method; similarly there is `semilogy` for the y-axis and `loglog` for both axes.

```
[5]: figure, axes = plt.subplots(1, 2, figsize=(10, 5))

# linear scale plot
axes[0].plot(countries.GDP2020 / 1000, countries.Expectancy2020, '.')
axes[0].set_xlabel('GDP per capita (thousands US dollars)')
axes[0].set_ylabel('Life expectancy (years)')
axes[0].set_title('Country indicators 2020')

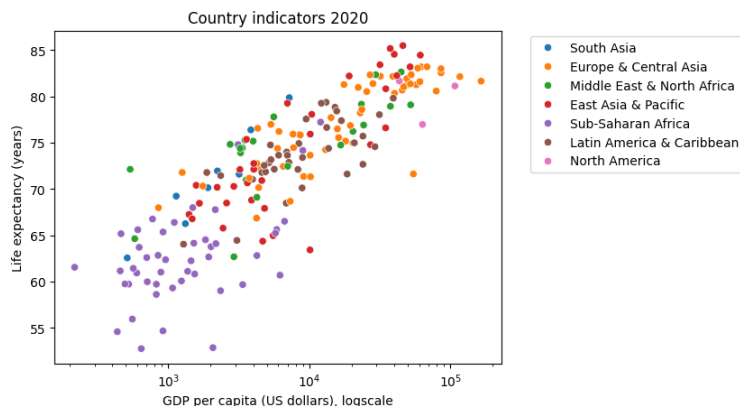
# log scale plot
axes[1].plot(countries.GDP2020, countries.Expectancy2020, '.')
axes[1].set_xlabel('GDP per capita (US dollars), logscale')
axes[1].set_ylabel('Life expectancy (years)')
axes[1].set_title('Country indicators 2020')
axes[1].semilogx()
pass
```



1.8 Categorical variable via color

Here we color countries by their region. Seaborn function `scatterplot` can do this easily via `hue` parameter. This function returns Matplotlib axes which can be then modified by familiar methods such as `set_xlabel`.

```
[6]: # create plot using Seaborn
axes = sns.scatterplot(data=countries, x='GDP2020', y='Expectancy2020',
                        hue='Region')
# set plot properties using methods from Matplotlib
axes.set_xlabel('GDP per capita (US dollars), logscale')
axes.set_ylabel('Life expectancy (years)')
axes.set_title('Country indicators 2020')
axes.semilogx()
# place legend outside the plot:
axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
pass
```



- The same plot in Plotly is even easier and interactive.
- Both Plotly and Seaborn automatically label axes with column names, such as `GDP2020`.
- Here we override such automated labels with longer ones using a dictionary `fig_labels`.

```
[7]: # we want to use index (country name) in the figure for tooltip info
# therefore we create a temporary table with column Country instead of index
temp_table = countries.reset_index()
# how to rename automated axis labels
fig_labels = {'GDP2020': 'GDP per capita (US dollars), logscale',
              'Expectancy2020': 'Life expectancy (years)'}
# create Plotly plot, add country name to tooltip data
fig = px.scatter(data_frame=temp_table,
                  x="GDP2020", y="Expectancy2020", color="Region",
```

```

        hover_data=['Country'], log_x=True,
        labels = fig_labels)
fig.show()

```

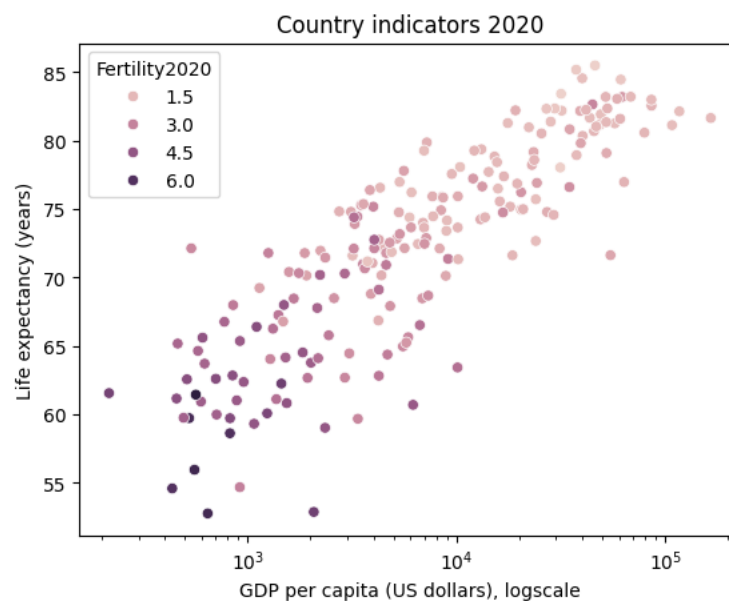
1.9 Numerical variable via color

Seaborn automatically detects if the column used as **hue** is a categorical or numerical variable. In the previous graph, regions were used as **hue** and Seaborn chose a color palette with a different color for each category. Here we have a numerical variable so a continuous palette with different shades of pink and purple is used by default. We will discuss color palettes later in the course.

```

[8]: axes = sns.scatterplot(data=countries, x='GDP2020', y='Expectancy2020',
                           hue='Fertility2020')
axes.set_xlabel('GDP per capita (US dollars), logscale')
axes.set_ylabel('Life expectancy (years)')
axes.set_title('Country indicators 2020')
axes.semilogx()
pass

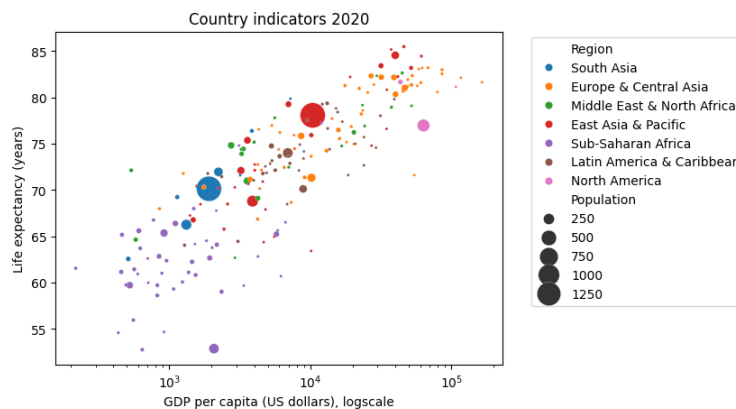
```



1.10 Numerical variable as point size

We will now use the population of each country as the size of each point (also called bubble), and we will color countries by regions. Sizing points according to the values in a specified table column is again simple to do in Seaborn using parameter **size** in `sns.scatterplot`. Parameter **sizes** sets the minimum and maximum point size to be used. For simplicity, population in millions is added as a new column to **countries**.

```
[9]: # add a column representing population in millions to table countries
countries['Population'] = countries['Population2020'] / 1e6
# create the plot
axes = sns.scatterplot(data=countries,
                        x='GDP2020', y='Expectancy2020', hue='Region',
                        size='Population',
                        sizes=(5, 400))
# set various plot properties
axes.set_xlabel('GDP per capita (US dollars), logscale')
axes.set_ylabel('Life expectancy (years)')
axes.set_title('Country indicators 2020')
axes.semilogx()
axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
pass
```



1.11 Categorical variable as marker type

- We add a new column named **Population change** with categories **increase**, **decrease** and **same** depending on how the population of a country changed between 2010 and 2020. Category **same** is applied to countries with population change less than 1% in either direction.
- This column is created using **apply** command, which applies a function (here a **lambda expression**) to **diff** Series containing relative change in population.
- This column is then used as argument **style** in **sns.scatterplot**. Size of markers is set to 100 (more than default) by argument **s**. Particular markers are selected by **markers** argument.
- Note that in the **scatterplot** we use both columns of **countries** table and separate Series.

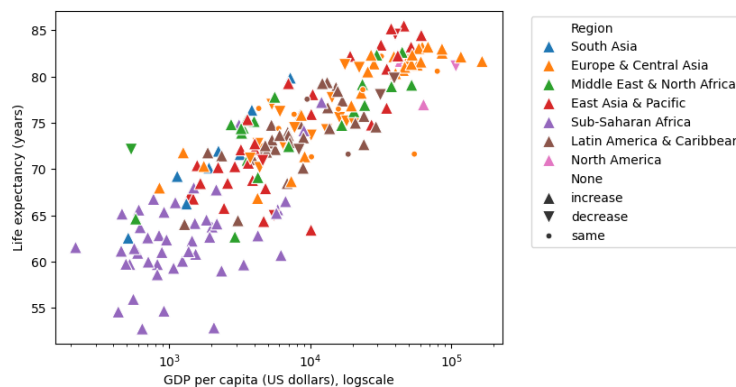
```
[10]: # compute relative differences in population between years 2010 and 2020
diff = (countries.Population2020 - countries.Population2010) / countries.
        Population2010
# new series with values 'increase', 'decrease' and 'same'
diff_class = diff.apply(lambda x : 'decrease' if x < -0.01
                        else 'increase' if x > 0.01 else 'same')
```



```
# create plot
axes = sns.scatterplot(data=countries,
                       x='GDP2020', y='Expectancy2020', hue='Region',
                       style=diff_class, s=100,
                       markers={'increase':'^', 'decrease':'v', 'same': '.'})

# plot settings
axes.semilogx()
axes.set_xlabel('GDP per capita (US dollars), logscale')
axes.set_ylabel('Life expectancy (years)')
axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
```

[10]: <matplotlib.legend.Legend at 0x7f69d9c0ecd0>



1.12 Importing Gapminder life expectancy

We import life expectancy data provided free by the [Gapminder foundation](https://bbrejova.github.io/viz/data/life_expectancy_years.csv) under the CC-BY license. The data set gives for each year and each country an estimate of how many years would newborn babies live on average if the trends in mortality of different age groups that were prevailing in the year of their birth would prevail through their entire life.

```
[11]: url="https://bbrejova.github.io/viz/data/life_expectancy_years.csv"
life_exp = pd.read_csv(url, index_col=0)
life_exp_years = life_exp.iloc[:, 1:]
display(life_exp)
```

	ISO3	1900	1901	1902	1903	1904	1905	1906	1907	\
Country										
Afghanistan	AFG	29.4	29.5	29.5	29.6	29.7	29.7	29.8	29.9	
Albania	ALB	35.4	35.4	35.4	35.4	35.4	35.4	35.4	35.4	
Algeria	DZA	30.2	30.3	30.4	31.4	25.4	28.1	29.6	29.5	
Angola	AGO	29.0	29.1	29.2	29.3	29.3	29.4	29.4	29.5	
Antigua and Barbuda	ATG	33.8	33.8	33.8	33.8	33.8	33.8	33.8	33.8	
...	
Venezuela	VEN	32.4	32.4	32.4	32.4	32.4	32.4	32.5	32.5	

Vietnam	VNM	31.2	31.1	31.1	31.1	31.1	31.0	31.0	31.0
Yemen	YEM	23.5	23.5	23.5	23.5	23.5	23.6	23.6	23.6
Zambia	ZMB	33.6	33.6	33.6	33.7	33.7	33.8	33.8	33.8
Zimbabwe	ZWE	34.1	34.1	34.1	34.1	34.1	34.1	34.1	34.1

		1908	...	2012	2013	2014	2015	2016	2017	2018	\
Country		...									
Afghanistan		29.9	...	60.8	61.3	61.2	61.2	61.2	63.4	63.081	
Albania		35.4	...	77.8	77.9	77.9	78.0	78.1	78.2	79.184	
Algeria		29.5	...	76.8	76.9	77.0	77.1	77.4	77.7	76.066	
Angola		29.6	...	61.3	61.9	62.8	63.3	63.8	64.2	62.144	
Antigua and Barbuda		33.8	...	76.7	76.8	76.8	76.9	77.0	77.0	78.511	
...	
Venezuela		32.5	...	75.2	75.2	75.0	75.0	75.3	75.3	71.979	
Vietnam		30.9	...	73.8	74.0	74.1	74.3	74.4	74.5	73.976	
Yemen		23.6	...	68.3	68.9	69.0	68.6	68.1	68.1	64.575	
Zambia		33.9	...	58.8	60.0	61.1	62.0	62.8	63.2	62.342	
Zimbabwe		34.2	...	54.9	56.8	58.5	59.6	60.5	61.4	61.414	

		2019	2020	2021
Country				
Afghanistan		63.565	62.575	61.982
Albania		79.282	76.989	76.463
Algeria		76.474	74.453	76.377
Angola		62.448	62.261	61.643
Antigua and Barbuda		78.691	78.841	78.497
...
Venezuela		72.161	71.095	70.554
Vietnam		74.093	75.378	73.618
Yemen		65.092	64.650	63.753
Zambia		62.793	62.380	61.223
Zimbabwe		61.292	61.124	59.253

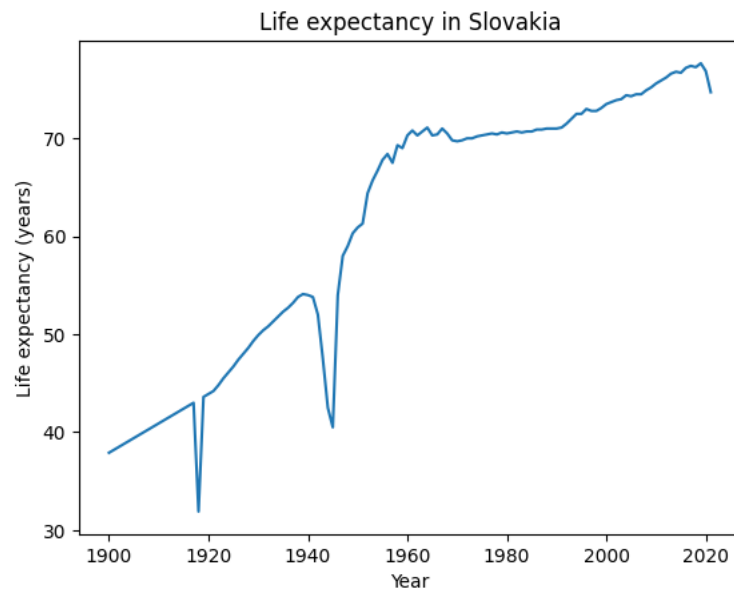
[184 rows x 123 columns]

1.13 A simple line graph

Here we use `plot` from `matplotlib` to plot life expectancy over the years for Slovakia. Years are column names which need to be converted from string to integer using Python list comprehension.

```
[12]: # list of numerical years from column names
years = [int(x) for x in life_exp_years.columns]
# simple plot for one row of the table
figure, axes = plt.subplots()
axes.plot(years, life_exp_years.loc['Slovak Republic'])
# plot settings
axes.set_xlabel('Year')
axes.set_ylabel('Life expectancy (years)')
```

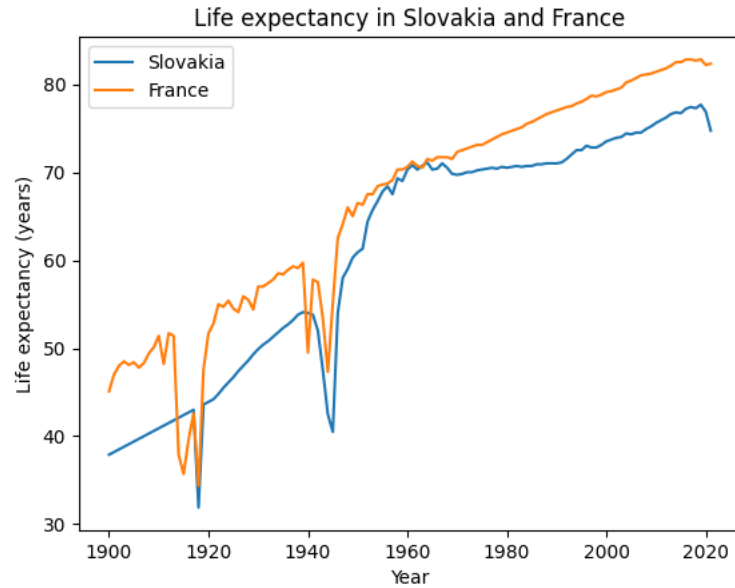
```
axes.set_title('Life expectancy in Slovakia')
pass
```



1.14 A line graph with multiple lines

Here we plot two lines, each by a separate call to `plot`. Each line has a label to show in the legend.

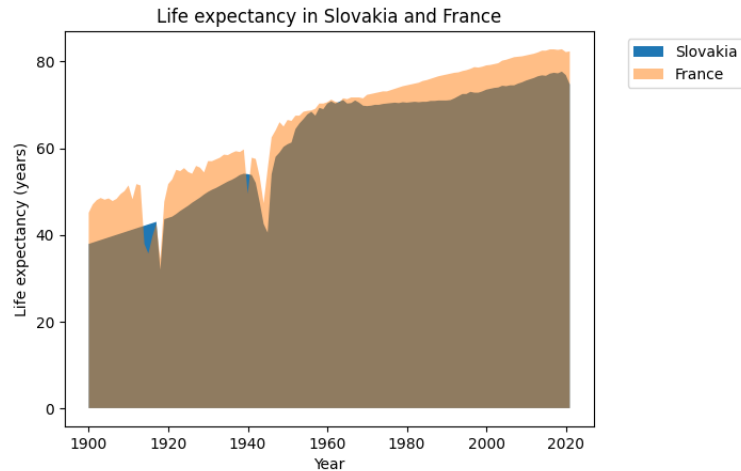
```
[13]: figure, axes = plt.subplots()
      # liot two lines
      axes.plot(years, life_exp_years.loc['Slovak Republic'], label='Slovakia')
      axes.plot(years, life_exp_years.loc['France'], label='France')
      # plot settings
      axes.set_xlabel('Year')
      axes.set_ylabel('Life expectancy (years)')
      axes.set_title('Life expectancy in Slovakia and France')
      axes.legend()
      pass
```



1.15 Area graph

Here we fill in the area between x-axis (value 0) and a table row using `fill_between` method. France is plotted on top and is set to be semi-transparent using `alpha=0.5`.

```
[14]: figure, axes = plt.subplots()
      # two filled areas, the second is semi-transparent
      axes.fill_between(years, 0, life_exp_years.loc['Slovak Republic'],
                        ↪label='Slovakia')
      axes.fill_between(years, 0, life_exp_years.loc['France'], label='France',
                        ↪alpha=0.5)
      # plot settings
      axes.set_xlabel('Year')
      axes.set_ylabel('Life expectancy (years)')
      axes.set_title('Life expectancy in Slovakia and France')
      axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
      pass
```



1.16 Line graph with many lines

- Here we want to plot lines for all countries alphabetically between Si and Sz and having at least million inhabitants.
- First we select such countries from `countries` to table `selection`.
- Using `intersection`, we get only countries from our selection that are also in Gapminder table (`life_exp`).
- Part of the Gapminder table for these countries is then stored as `life_exp_sel`.

```
[15]: selection = countries.query('Population2020 > 1e6 and Country >= "Si" and_
↳Country <= "Sz"')
life_exp_iso3 = life_exp.reset_index().set_index('ISO3')
life_exp_sel = life_exp_iso3.loc[life_exp_iso3.index.intersection(selection.
↳ISO3), :].set_index('Country')
display(life_exp_sel)
```

	1900	1901	1902	1903	1904	1905	1906	1907	1908	1909	\
Country											
Sierra Leone	27.4	27.5	27.5	27.6	27.7	27.8	27.9	27.9	28.0	28.1	
Singapore	34.2	34.2	34.2	34.2	34.2	34.2	34.2	34.2	34.2	34.2	
Slovak Republic	37.9	38.2	38.5	38.8	39.1	39.4	39.7	40.0	40.3	40.6	
Slovenia	36.4	36.4	36.4	36.4	36.4	36.4	36.4	36.4	36.4	36.4	
Somalia	31.2	31.2	31.3	31.4	31.4	31.5	31.5	31.6	31.7	31.7	
South Africa	34.5	34.5	34.5	34.6	34.6	34.6	34.6	34.7	34.7	34.8	
South Sudan	29.6	29.6	29.8	29.8	29.9	30.0	30.1	30.2	30.3	30.4	
Spain	34.7	35.6	36.4	37.2	38.0	38.9	39.7	40.5	41.4	41.0	
Sri Lanka	32.5	32.5	32.5	33.9	34.7	32.4	26.9	30.0	30.4	29.8	
Sudan	31.5	31.5	31.5	31.5	31.5	31.5	31.5	31.5	31.5	31.5	
Sweden	52.3	52.9	54.7	55.1	55.4	54.5	56.7	57.0	56.4	58.4	
Switzerland	47.5	49.0	50.5	50.1	49.2	49.7	50.8	51.3	52.3	51.7	
Syria	31.8	31.8	31.9	31.9	31.9	31.9	31.9	32.0	32.0	32.0	

	...	2012	2013	2014	2015	2016	2017	2018	\
Country	...								
Sierra Leone	...	56.9	57.9	57.1	58.5	59.8	60.4	59.796000	
Singapore	...	83.6	83.9	84.2	84.4	84.7	84.8	83.297561	
Slovak Republic	...	76.2	76.6	76.8	76.7	77.2	77.4	77.265854	
Slovenia	...	79.9	80.2	80.9	80.8	81.0	81.1	81.378049	
Somalia	...	56.8	57.4	57.9	58.3	58.5	58.5	56.375000	
South Africa	...	59.5	61.1	62.5	63.4	64.4	66.3	65.674000	
South Sudan	...	58.2	58.0	58.3	59.4	59.4	59.3	55.950000	
Spain	...	82.3	82.6	82.7	82.6	82.9	83.1	83.431707	
Sri Lanka	...	76.1	76.4	76.5	76.9	77.2	77.5	75.748000	
Sudan	...	68.4	68.7	69.1	69.6	69.8	70.3	65.681000	
Sweden	...	81.8	81.9	82.1	82.2	82.4	82.5	82.558537	
Switzerland	...	82.9	83.0	83.3	83.5	83.8	84.0	83.753659	
Syria	...	67.9	68.7	65.0	67.3	67.4	69.8	70.145000	

		2019	2020	2021
Country				
Sierra Leone	60.255000	59.763000	60.062000	
Singapore	83.595122	84.465854	83.441463	
Slovak Republic	77.665854	76.865854	74.714634	
Slovenia	81.529268	80.531707	80.875610	
Somalia	57.078000	55.967000	55.280000	
South Africa	66.175000	65.252000	62.341000	
South Sudan	55.912000	55.480000	54.975000	
Spain	83.831707	82.331707	83.178049	
Sri Lanka	76.008000	76.393000	76.399000	
Sudan	65.876000	65.614000	65.267000	
Sweden	83.109756	82.356098	83.156098	
Switzerland	83.904878	83.000000	83.851220	
Syria	71.822000	72.140000	72.063000	

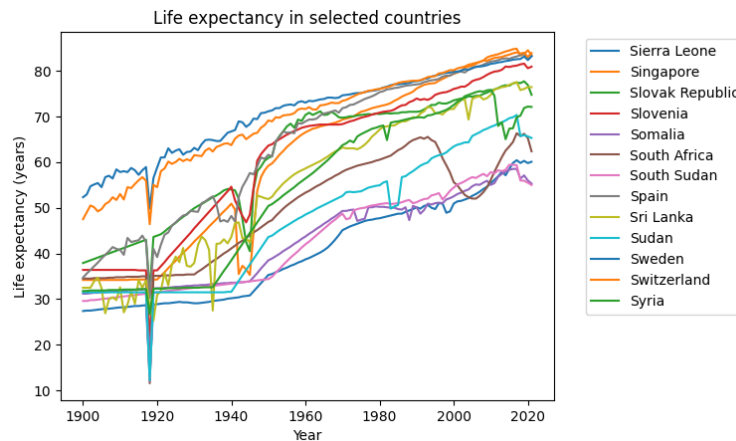
[13 rows x 122 columns]

- In Matplotlib, each country from `life_exp_sel` is plotted separately in a for-loop, similarly as for two countries above.
- Note that colors repeat because the default palette is not large enough.

```
[16]: figure, axes = plt.subplots()
      # loop over countries
      for country in life_exp_sel.index:
          axes.plot(years, life_exp_sel.loc[country], label=country)

      # plot settings
      axes.set_xlabel('Year')
      axes.set_ylabel('Life expectancy (years)')
      axes.set_title('Life expectancy in selected countries')
```

```
axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
pass
```



- To use Seaborn for the same plot, it is better to change `life_exp_sel` table from wide to long format using `melt` method. Year is converted from strings to integers.
- This creates a table with columns Country, Year, Expectancy.

```
[17]: life_exp_sel_long = (
        life_exp_sel.reset_index()
        .melt(id_vars=['Country'])
        .rename(columns={'variable': 'Year', 'value': 'Expectancy'})
        .astype({'Year': 'int32'})
    )
display(life_exp_sel_long)
```

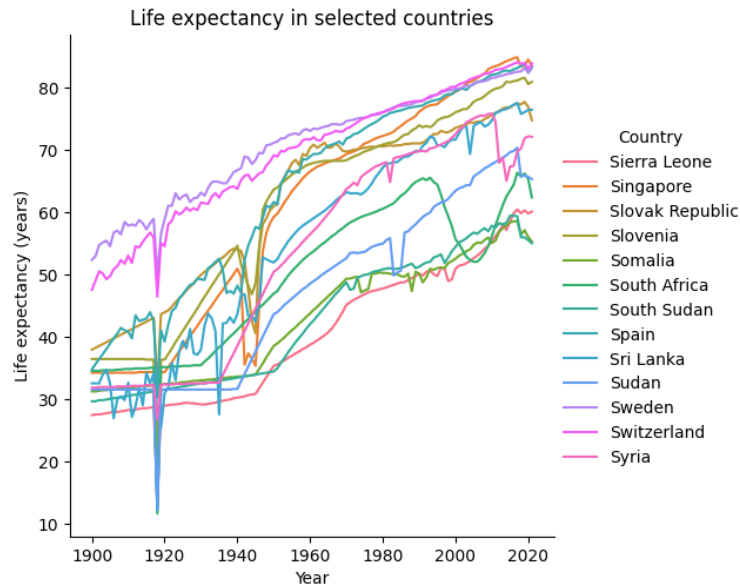
	Country	Year	Expectancy
0	Sierra Leone	1900	27.400000
1	Singapore	1900	34.200000
2	Slovak Republic	1900	37.900000
3	Slovenia	1900	36.400000
4	Somalia	1900	31.200000
...
1581	Sri Lanka	2021	76.399000
1582	Sudan	2021	65.267000
1583	Sweden	2021	83.156098
1584	Switzerland	2021	83.851220
1585	Syria	2021	72.063000

[1586 rows x 3 columns]

- Now we use Seaborn function `relplot`, setting parameters `x`, `y` and `hue` to column names in our long table and specifying that we want lineplot using `kind="line"`.

- The function returns `FacetGrid`, which potentially contains multiple axes, so we need to use slightly different methods to set labels.
- Seaborn created a sufficiently large color palette but some colors are then hard to distinguish.

```
[18]: grid = sns.relplot(data=life_exp_sel_long, x='Year', y='Expectancy',
                        hue='Country', kind="line")
grid.set_axis_labels('Year', 'Life expectancy (years)')
grid.set(title='Life expectancy in selected countries')
pass
```

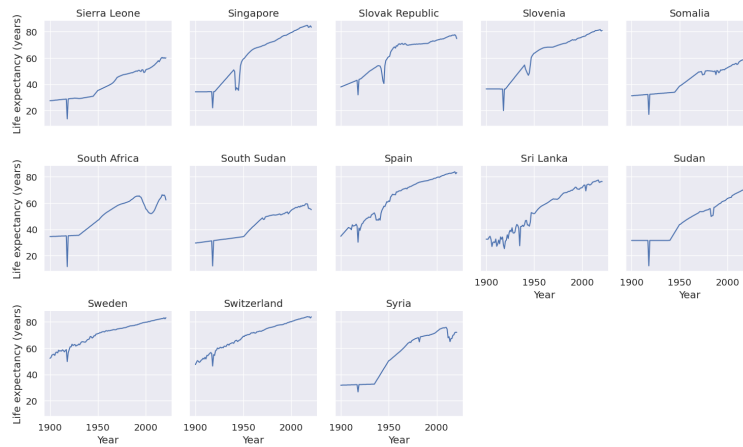


1.17 Small multiples

- Small multiples, with each country in our selection as a separate plot, is very easy to do in Seaborn from a long-format table using `relplot`, using column `Country` in option `col` which selects one of subplots for each data point.
- Option `col_wrap` selects how many subplots will be placed on one row of the overall figure.

```
[19]: # create grid of small multiple plots
sns.set_theme(font_scale=1.2)
grid = sns.relplot(data=life_exp_sel_long,
                  x='Year', y='Expectancy', col='Country',
                  col_wrap=5, kind="line", height=3, aspect=1)

grid.set_axis_labels('Year', 'Life expectancy (years)')
grid.set_titles("{col_name}") # title of each plot will be country name
pass
```

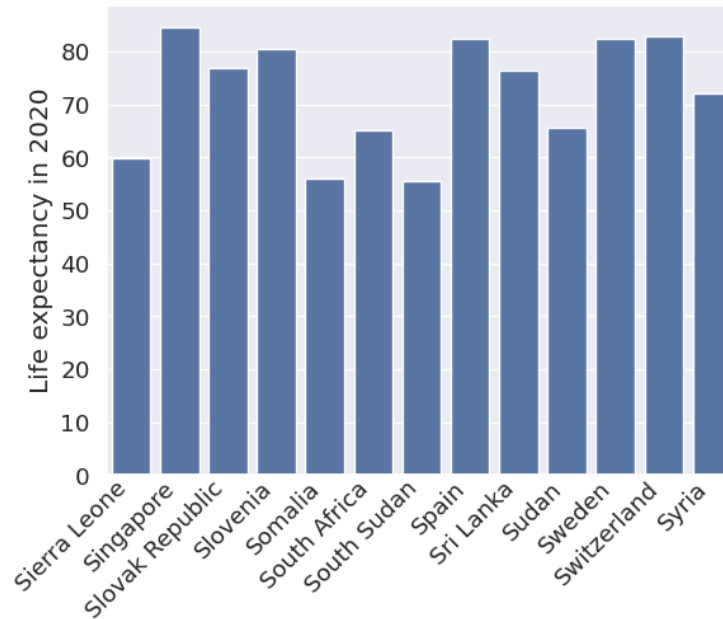



1.18 Bar graph

- We plot a bargraph of life expectancy in our selected countries by Seaborn function `barplot`.
- All bars are plotted by the same color using setting `color="C0"`.
- We rotate tick labels on the x axes to fit them in the given space.

```
[20]: def rotate_bar_labels(axes, angle=45):
    """Auxiliary function for rotating bar plot labels by 45 degrees"""
    axes.tick_params(axis='x', labelrotation=angle, pad=-5)
    plt.setp(axes.get_xticklabels(), ha='right')

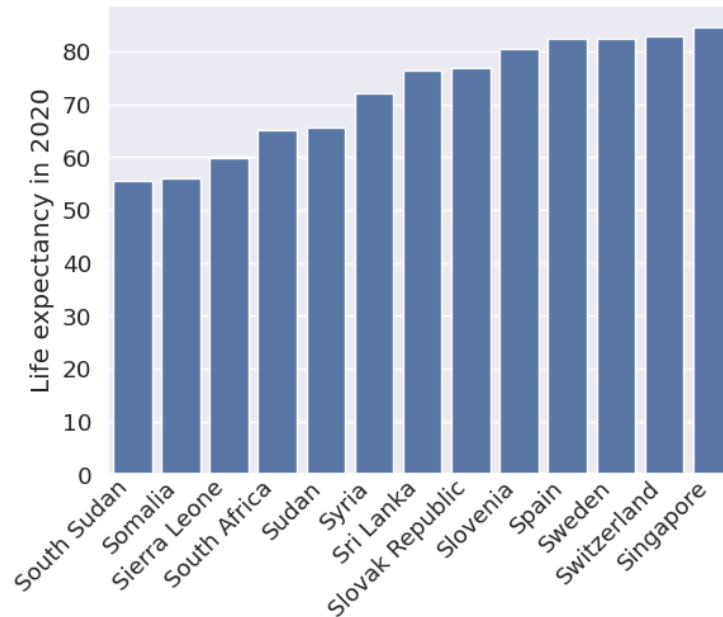
    # select one year from the long table
    life_exp_sel_2020 = life_exp_sel_long.query('Year==2020')
    # create barplot
    axes = sns.barplot(data=life_exp_sel_2020,
                       x='Country', y='Expectancy', color="C0")
    axes.set_ylabel("Life expectancy in 2020")
    axes.set_xlabel(None)
    rotate_bar_labels(axes)
    pass
```



1.19 Bar graph with sorted columns

Countries are sorted by value in preprocessing, then plotted as before.

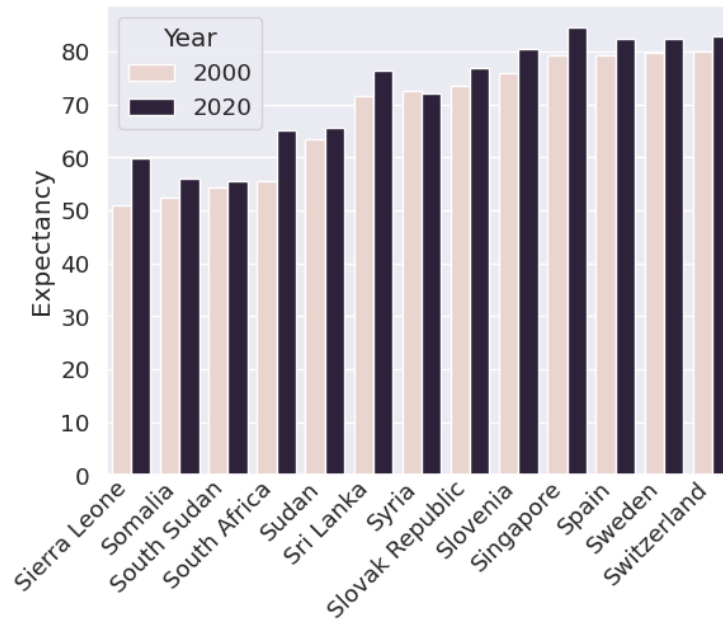
```
[21]: # sorting
life_exp_sel_2020_sorted = life_exp_sel_2020.sort_values('Expectancy')
# plotting
axes = sns.barplot(data=life_exp_sel_2020_sorted,
                    x='Country', y='Expectancy', color="C0")
axes.set_ylabel("Life expectancy in 2020")
axes.set_xlabel(None)
rotate_bar_labels(axes)
pass
```



1.20 Bar graph with multiple colors

- Now we compare life expectancy in two years in a bargraph with the colors of columns.
- After selecting appropriate rows of the long table, we use column **Year** in the **hue** parameter of **barplot**.

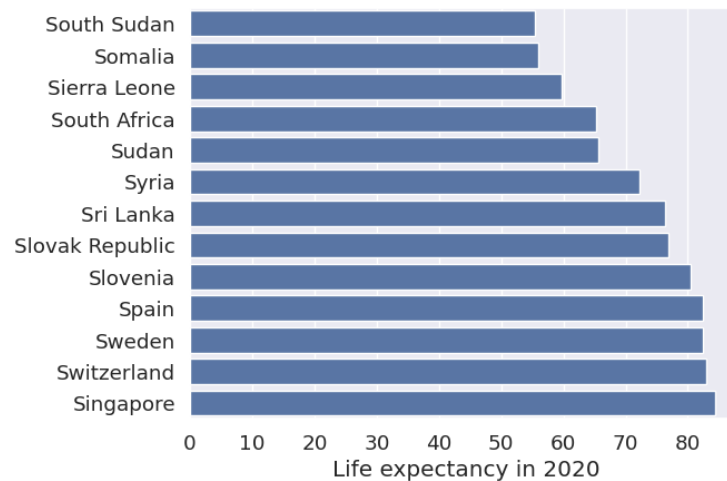
```
[22]: # select only years 2000 and 2020 from the table, sort
life_exp_sel_comp = life_exp_sel_long.query('Year==2020 or Year==2000').
    ↪sort_values('Expectancy')
# plotting
axes = sns.barplot(data=life_exp_sel_comp, x='Country', y='Expectancy',
    ↪hue='Year')
axes.set_xlabel(None)
rotate_bar_labels(axes)
pass
```



1.21 Horizontal bar graph

- Longer bar labels are easier to read in a horizontal barplot.
- In Seaborn, it is sufficient to switch x and y arguments.

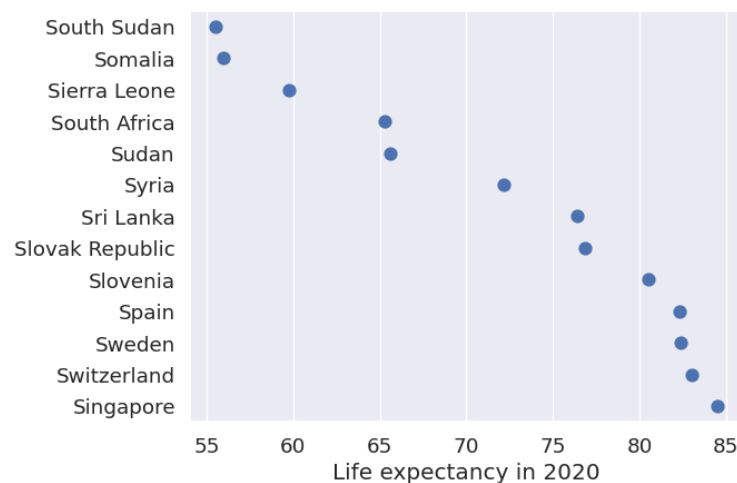
```
[23]: axes = sns.barplot(data=life_exp_sel_2020_sorted,
                        y='Country', x='Expectancy', color="C0")
axes.set_xlabel("Life expectancy in 2020")
axes.set_ylabel(None)
pass
```



1.22 Dot plot

- Dot plot shows only the end of each bar as a dot.
- Seaborn's `pointplot` joins these dots by lines by default, `linestyle='none'` prevents this.
- Note that in contrast to barplots, the x axis does not start at 0 (we could make it so by `set_xlim`).

```
[24]: axes = sns.pointplot(data=life_exp_sel_2020_sorted,
                           y='Country', x='Expectancy',
                           color="C0", linestyle='none')
axes.set_xlabel("Life expectancy in 2020")
axes.set_ylabel(None)
pass
```



1.23 Heatmap

- The goal is to create heatmap with countries as rows, several years as columns and life expectancy values as colors.
- We first need to create a DataFrame with these values in such an arrangement by selecting rows with appropriate years from our long table and pivoting the table by year to make it wide.
- Finally we sort the table by the expectancy in the last year.

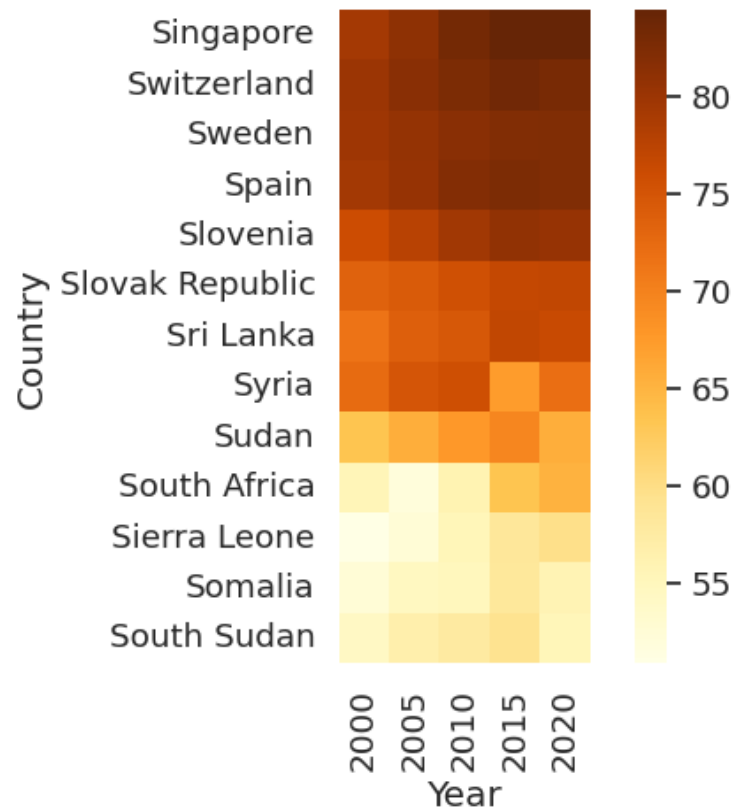
```
[25]: # set of years to be used
sel_years={2000, 2005, 2010, 2015, 2020}
# create desired wide table
life_exp_sel_wide = (life_exp_sel_long.query('Year in @sel_years')
                     .pivot(index='Country', columns='Year',
                             values='Expectancy')
                     .sort_values(2020, ascending=False))
# show the table
```

```
display(life_exp_sel_wide)
```

Year	2000	2005	2010	2015	2020
Country					
Singapore	79.3	81.1	83.2	84.4	84.465854
Switzerland	80.1	81.5	82.5	83.5	83.000000
Sweden	79.8	80.6	81.5	82.2	82.356098
Spain	79.4	80.5	82.0	82.6	82.331707
Slovenia	76.0	77.7	79.5	80.8	80.531707
Slovak Republic	73.5	74.3	75.6	76.7	76.865854
Sri Lanka	71.6	73.8	74.7	76.9	76.393000
Syria	72.5	75.0	75.8	67.3	72.140000
Sudan	63.4	65.7	67.7	69.6	65.614000
South Africa	55.6	52.0	56.1	63.4	65.252000
Sierra Leone	50.9	52.6	55.4	58.5	59.763000
Somalia	52.5	54.7	55.0	58.3	55.967000
South Sudan	54.4	56.7	57.8	59.4	55.480000

- Heatmap is plotted by `sns.heatmap` function.
- We have used options to set the shape of individual cells to square and change the color palette ('cmap').

```
[26]: axes = sns.heatmap(data=life_exp_sel_wide, square=True, cmap="YlOrBr")
pass
```



1.24 Pie chart

- To prepare data for pie chart, we use two features of Pandas which we will cover in a later lecture: converting the Income Group column to a categorical type and computing the number of countries in various income groups using `groupby`.
- In this way we create two Series: `groups` with counts for the whole world and `groups_asia` for just East Asian countries.

```
[27]: # creating a categorical type
income_categories = ["Low income", "Lower middle income",
                    "Upper middle income", "High income"]
cat_type = pd.api.types.CategoricalDtype(categories=income_categories,
                                         ordered=True)

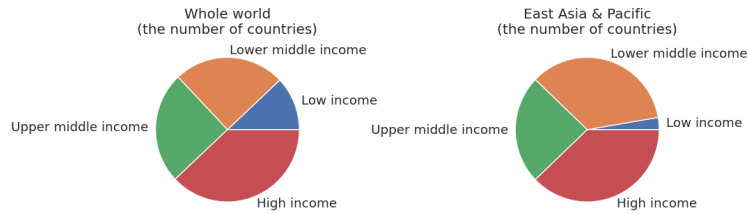
# converting Income Group column to cat_type
countries_cat = countries.astype({'Income Group': cat_type})
# aggregation using groupby
groups = countries_cat.groupby('Income Group').size().rename('Count')
# the same but only on countries selected by query
groups_asia = (countries_cat.query('Region=="East Asia & Pacific"')
               .groupby('Income Group').size().rename('Count'))

display(groups)
```

```
Income Group
Low income          26
Lower middle income  54
Upper middle income  54
High income         82
Name: Count, dtype: int64
```

- The plotting is done by the `pie` function from Matplotlib.
- It gets the series with counts as parameter `x` and country names as labels.

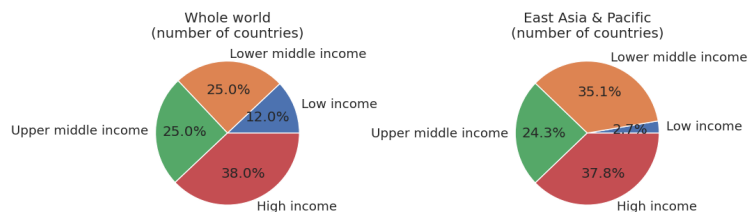
```
[28]: figure, axes = plt.subplots(1,2, figsize=(10,5))
axes[0].pie(x=groups, labels=groups.index)
axes[0].set_title('Whole world\n(the number of countries)')
axes[1].pie(x=groups_asia, labels=groups_asia.index)
axes[1].set_title('East Asia & Pacific\n(the number of countries)')
figure.subplots_adjust(wspace=1)
pass
```



1.25 Pie chart with labels

- Labels are added by `autopct` setting in `pie`. This setting provides a [formatting string](#) for the values, here we print one decimal place.

```
[29]: figure, axes = plt.subplots(1,2, figsize=(10,5))
axes[0].pie(x=groups, labels=groups.index, autopct="%.1f%%")
axes[0].set_title('Whole world\n(number of countries)')
axes[1].pie(x=groups_asia, labels=groups_asia.index, autopct="%.1f%%")
axes[1].set_title('East Asia & Pacific\n(number of countries)')
figure.subplots_adjust(wspace=1)
pass
```



1.26 Stacked bar graph instead of pie chart

- To prepare data for stacked bar graph, we need to combine our two count Series (`groups` and `groups_asia`) to one long table `groups_concat`.
- This is a DataFrame, because `Income Group` was moved from index to a column.
- We also add percentage column, which will be used in the plot. Percentage is computed by divided counts with the sum of all counts.
- We also add a column with region name, because we will consider two regions (East Asia and the whole world).

```
[30]: # first create DataFrame for East Asia
# add Income Group index as a column
temp_asia = groups_asia.reset_index()
# compute percentages and add as a new column
temp_asia['Percentage'] = temp_asia['Count'] * 100 / temp_asia['Count'].sum()
# add Region as a new column, filled with copies of the same string
```



```
temp_asia['Region'] = ["East Asia & Pacific"] * len(groups_asia)

# the same three steps for World
temp_world = groups.reset_index()
temp_world['Percentage'] = temp_world['Count'] * 100 / temp_world['Count'].sum()
temp_world['Region'] = ["World"] * len(groups)

# concatenate two DataFrames and display
groups_concat = pd.concat([temp_asia, temp_world], axis=0)
display(groups_concat)
```

	Income Group	Count	Percentage	Region
0	Low income	1	2.702703	East Asia & Pacific
1	Lower middle income	13	35.135135	East Asia & Pacific
2	Upper middle income	9	24.324324	East Asia & Pacific
3	High income	14	37.837838	East Asia & Pacific
0	Low income	26	12.037037	World
1	Lower middle income	54	25.000000	World
2	Upper middle income	54	25.000000	World
3	High income	82	37.962963	World

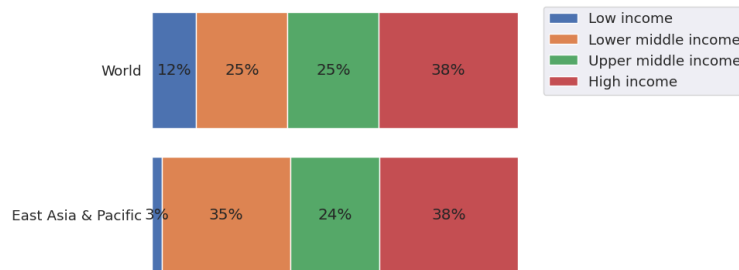
- Stacked bar graph is not very automated in Matplotlib.
- Left coordinate for each rectangle needs to be computed manually, then function `barh` is used (see also [tutorial](#)).
- Each bar is labeled with the percentage using `bar_label` function.

```
[31]: # list of regions and income groups
tmp_regions = groups_concat['Region'].unique()
tmp_groups = groups_concat['Income Group'].unique()
# the first rectangles start at 0
starts = pd.Series([0] * tmp_regions.shape[0])
# create plot
figure, axes = plt.subplots()

# iterate through income groups
for group in tmp_groups:
    # select data for this income group from both regions
    tmp_data = groups_concat.query("`Income Group` == @group")
    # plot
    rectangles = axes.barh(y=tmp_data['Region'], width=tmp_data['Percentage'],
    ↪left=starts, label=group)
    # add labels
    axes.bar_label(rectangles, label_type='center', fmt="%.0f%%")
    # move starts by the size of each rectangle
    starts += tmp_data['Percentage'].reset_index(drop=True)

axes.legend(bbox_to_anchor=(1, 1), loc=2)
# hide plot frame and x-axis ticks
```

```
axes.xaxis.set_visible(False)
axes.set_frame_on(False)
pass
```



- Stacked bar charts are much easier in Plotly using `px.bar` function.

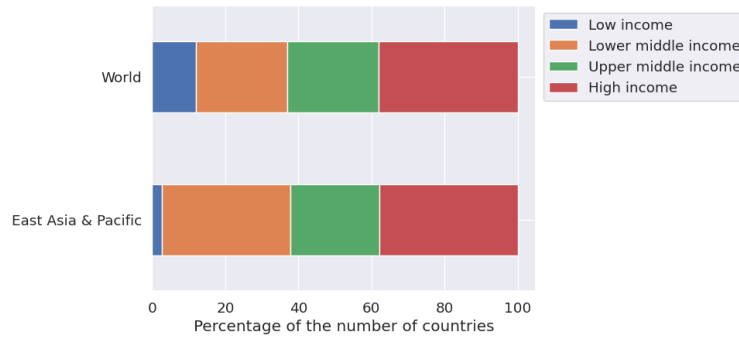
```
[32]: fig = px.bar(groups_concat, x="Region", y="Percentage", color="Income Group",
                  text="Percentage", text_auto=".0f")
fig.update_layout(font=dict(size=20), xaxis_title=None,)
fig.show()
```

It can also be drawn easily directly by Pandas, but values (sizes of rectangles) are not shown. The table is first converted to a wide form with different income groups as columns.

```
[33]: groups_concat_wide = groups_concat.pivot(columns='Income Group',
        ↪index='Region', values='Percentage')
display(groups_concat_wide)
axes = groups_concat_wide.plot(kind='barh', stacked=True)
axes.legend(bbox_to_anchor=(1, 1), loc=2)
axes.set_ylabel(None)
axes.set_xlabel('Percentage of the number of countries')
pass
```

Income Group	Low income	Lower middle income	Upper middle income	\
Region				
East Asia & Pacific	2.702703	35.135135	24.324324	
World	12.037037	25.000000	25.000000	

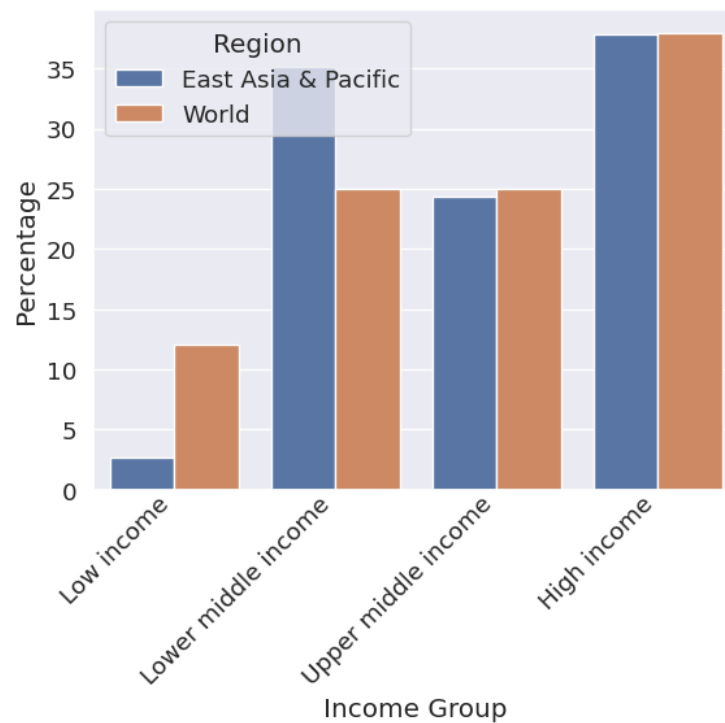
Income Group	High income
Region	
East Asia & Pacific	37.837838
World	37.962963



1.27 Colored bar graphs insteads of pie chart

- As we have seen before, colored bar graphs are easy in Seaborn from a long table.
- Therefore we use `groups_concat` DataFrame.

```
[34]: axes = sns.barplot(data=groups_concat,
                        x='Income Group', y='Percentage', hue='Region')
rotate_bar_labels(axes)
pass
```



1.28 Multiple bar graphs instead of pie chart

- In the next plot a separate bar graph for each region.
- This is also very simple in Seaborn using `catplot` with setting `col='Region'` and `kind='bar'`.
- Labels are rotated in each subplot using a for-loop.

```
[35]: #plotting
grid = sns.catplot(kind='bar', data=groups_concat,
                   x='Income Group', y='Percentage', col='Region')

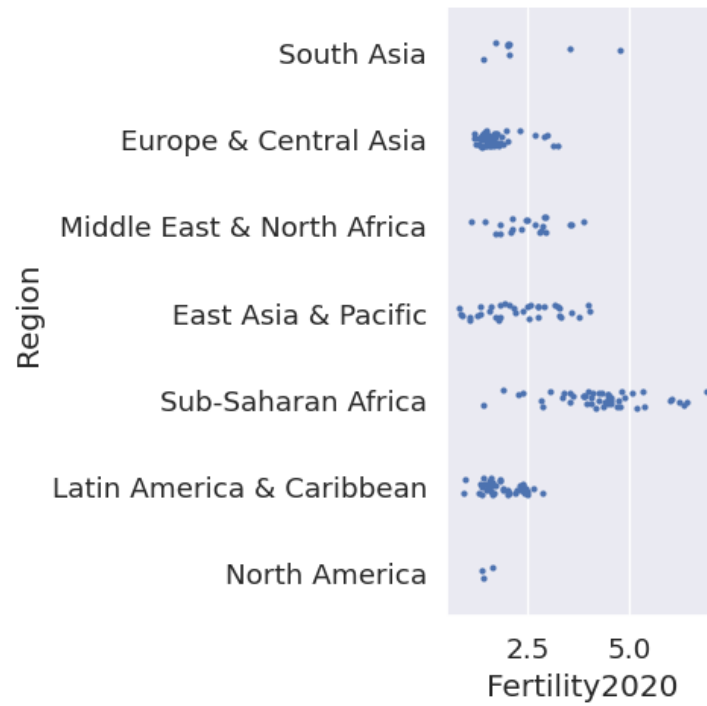
# label rotation
for which in [0,1]:
    rotate_bar_labels(grid.axes[0,which])
pass
```



1.29 Strip plot

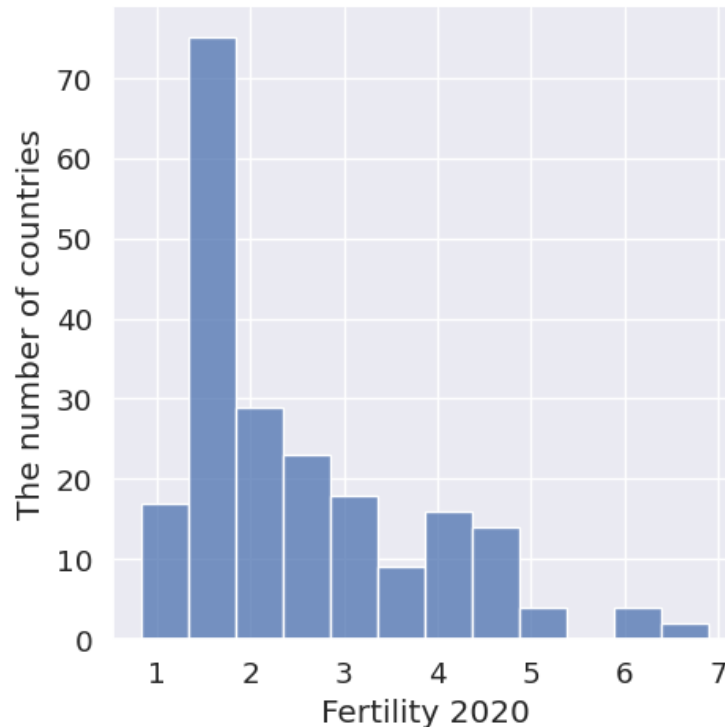
- Strip plot of fertility per region is also very simple in `sns.catplot`.
- Setting `kind='strip'` is default for `catplot`, so it is omitted here.

```
[36]: grid = sns.catplot(x="Fertility2020", y="Region", data=countries, size=3)
pass
```



1.30 Histogram

```
[37]: grid = sns.displot(countries, x="Fertility2020", binwidth=0.5)
      grid.set_axis_labels("Fertility 2020", "The number of countries")
      pass
```



1.31 Parallel coordinates

- We want to display various properties of individual countries as a parallel coordinate plot.
- We first create table `for_parallel` with selected columns and express all numbers as percentage of the maximum value.
- We add `selected` column which has `True` in row for Slovakia and `False` elsewhere. This is used to highlight Slovakia in the plot.
- Also ordering is changed to draw Slovakia the last.

```
[38]: # selecting columns
for_parallel_sel = countries.loc[:, ['Population2020', 'Area', 'GDP2020',
                                     'Expectancy2020', 'Fertility2020']]

# computing maximum in each column
for_parallel_max = for_parallel_sel.max(axis=0)
# dividing values by the maximum and multiplying by 100 to get percentage
for_parallel = for_parallel_sel.div(for_parallel_max, axis=1) * 100
# creating column of booleans called 'selected' which highlights Slovakia with
↳ True
for_parallel['selected'] = countries.index=="Slovak Republic"
# sort by 'selected' to put Slovakia last
for_parallel.sort_values('selected', inplace=True)
# show end of the table
display(for_parallel.tail())
```

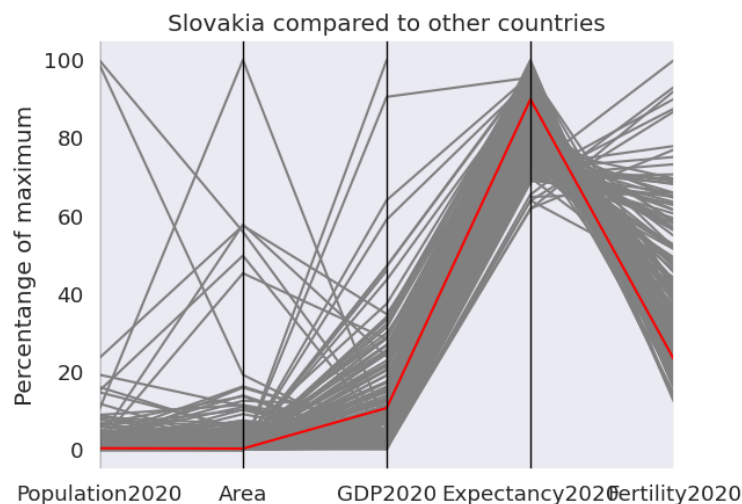
	Population2020	Area	GDP2020	Expectancy2020	\
Country					
Greece	0.758174	0.771775	9.651333	95.076168	
Greenland	0.003995	2.400538	29.962671	83.788155	
Fiji	0.065227	0.106853	2.638193	79.445541	
Zimbabwe	1.110458	2.285380	0.752009	71.492098	
Slovak Republic	0.386849	0.286754	10.711280	89.904148	

	Fertility2020	selected
Country		
Greece	20.168311	False
Greenland	29.294835	False
Fiji	36.201393	False
Zimbabwe	51.436448	False
Slovak Republic	23.070226	True

- Parallel coordinates are drawn using Pandas `parallel_coordinates` function, which internally calls Matplotlib and returns Axes object.

```
[39]: axes = pd.plotting.parallel_coordinates(for_parallel, class_column='selected',
                                             color=['gray', 'red'])

axes.get_legend().remove()
axes.set_ylabel("Percentage of maximum")
axes.set_title("Slovakia compared to other countries")
pass
```



1.32 Parallel categories

- We will use two categorical columns from the countries table, but more categorical columns could be easily added.
- We use the version of the table with a categorical income groups and sort countries by income.

- Now we use `parallel-categories` function from Plotly.
- This function orders each column of the figure by size. By calling `update_traces`, we reorder the first column by the same order as they first occur in our table.

```
[40]: for_parallel_cat = (countries_cat.loc[:,['Income Group','Region']]
        .sort_values('Income Group', ascending=False))
fig = px.parallel_categories(for_parallel_cat, width=800)
fig.update_traces(dimensions=[{"categoryorder": "array"}, {}])
fig.update_layout(font_size=20)
fig.update_layout(margin={'l':200,'r':200})
fig.show()
```

1.33 Radar chart

- Radar charts are not well supported in any of the used libraries.
- Below we compute angles of each axis manually, then use `plot` from Matplotlib.
- When creating axes, we specify polar coordinates `subplot_kw={'projection': 'polar'}`.
- We also use `set_thetagrids`

```
[41]: # skip 'selected' column, use only 3 countries
for_radar = for_parallel.loc[['India','China','United States'], :].iloc[:, 0:-1]
display(for_radar.head())

# setup plot with polar coordinates
figure, axes = plt.subplots(subplot_kw={'projection': 'polar'})
categories = list(for_radar.columns)
import math
angles = [ i * 2 * math.pi / len(categories) for i in range(len(categories))]
angles_deg = [x / math.pi * 180 for x in angles]
axes.set_thetagrids(angles_deg, labels=categories)

# for plotting, we will need to return to starting point in each line
angles.append(angles[0])

# for each country create list of values, add the starting point, plot
for country in for_radar.index:
    values = list(for_radar.loc[country, :])
    values.append(values[0])
    axes.plot(angles, values, label=country)

axes.legend(bbox_to_anchor=(1.05, 1), loc=2)
pass
```

	Population2020	Area	GDP2020	Expectancy2020	\
Country					
India	98.957347	19.225710	1.048125	82.049124	
China	100.000000	55.929174	5.702240	91.320734	
United States	23.493127	57.500095	34.803081	90.038227	

	Fertility2020
Country	
India	29.759141
China	18.586767
United States	23.817470

