

```
In [1]: # To call pandas library and use its built-in modules, need to import the lib.
import pandas as pd
```

```
In [2]: # Below the databases are read turn them from files into DataFrames.
fert = pd.read_csv('gapminder_total_fertility.csv', index_col=0)
life = pd.read_excel('gapminder_lifeexpectancy.xlsx', index_col=0)
popu = pd.read_excel('gapminder_population.xlsx', index_col=0)
cont = pd.read_csv('continents.csv', sep=';', index_col=1)

# To explore information of dataframes
fert.info()
life.info()
popu.info()
cont.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 260 entries, Abkhazia to Åland
Columns: 216 entries, 1800 to 2015
dtypes: float64(216)
memory usage: 440.8+ KB
<class 'pandas.core.frame.DataFrame'>
Index: 260 entries, Abkhazia to South Sudan
Columns: 217 entries, 1800 to 2016
dtypes: float64(217)
memory usage: 442.8+ KB
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 275 entries, Abkhazia to West Bank
Data columns (total 91 columns):
```

#	Column	Non-Null Count	Dtype
0	1800	229 non-null	float64
1	1810	229 non-null	float64
2	1820	229 non-null	float64
3	1830	229 non-null	float64
4	1840	229 non-null	float64
5	1850	229 non-null	float64
6	1860	229 non-null	float64
7	1870	229 non-null	float64
8	1880	229 non-null	float64
9	1890	229 non-null	float64
10	1900	229 non-null	float64
11	1910	229 non-null	float64
12	1920	229 non-null	float64
13	1930	229 non-null	float64
14	1940	229 non-null	float64
15	1950	256 non-null	float64
16	1951	256 non-null	float64
17	1952	256 non-null	float64
18	1953	256 non-null	float64
19	1954	256 non-null	float64
20	1955	256 non-null	float64
21	1956	256 non-null	float64
22	1957	256 non-null	float64
23	1958	256 non-null	float64
24	1959	256 non-null	float64
25	1960	256 non-null	float64
26	1961	256 non-null	float64
27	1962	256 non-null	float64
28	1963	256 non-null	float64
29	1964	256 non-null	float64
30	1965	256 non-null	float64
31	1966	256 non-null	float64
32	1967	256 non-null	float64
33	1968	256 non-null	float64
34	1969	256 non-null	float64

```

35 1970      256 non-null    float64
36 1971      256 non-null    float64
37 1972      256 non-null    float64
38 1973      256 non-null    float64
39 1974      256 non-null    float64
40 1975      256 non-null    float64
41 1976      256 non-null    float64
42 1977      256 non-null    float64
43 1978      256 non-null    float64
44 1979      256 non-null    float64
45 1980      256 non-null    float64
46 1981      256 non-null    float64
47 1982      256 non-null    float64
48 1983      256 non-null    float64
49 1984      256 non-null    float64
50 1985      256 non-null    float64
51 1986      256 non-null    float64
52 1987      256 non-null    float64
53 1988      256 non-null    float64
54 1989      256 non-null    float64
55 1990      256 non-null    float64
56 1991      256 non-null    float64
57 1992      256 non-null    float64
58 1993      256 non-null    float64
59 1994      256 non-null    float64
60 1995      256 non-null    float64
61 1996      256 non-null    float64
62 1997      256 non-null    float64
63 1998      256 non-null    float64
64 1999      256 non-null    float64
65 2000      256 non-null    float64
66 2001      256 non-null    float64
67 2002      256 non-null    float64
68 2003      256 non-null    float64
69 2004      256 non-null    float64
70 2005      256 non-null    float64
71 2006      256 non-null    float64
72 2007      256 non-null    float64
73 2008      256 non-null    float64
74 2009      238 non-null    float64
75 2010      238 non-null    float64
76 2011      233 non-null    float64
77 2012      233 non-null    float64
78 2013      233 non-null    float64
79 2014      231 non-null    float64
80 2015      231 non-null    float64
81 Unnamed: 82 0 non-null    float64
82 Unnamed: 83 0 non-null    float64
83 Unnamed: 84 0 non-null    float64
84 Unnamed: 85 0 non-null    float64
85 Unnamed: 86 0 non-null    float64
86 Unnamed: 87 0 non-null    float64
87 Unnamed: 88 0 non-null    float64
88 Unnamed: 89 0 non-null    float64
89 Unnamed: 90 0 non-null    float64
90 Unnamed: 91 0 non-null    float64
dtypes: float64(91)
memory usage: 197.7+ KB
<class 'pandas.core.frame.DataFrame'>
Index: 194 entries, Algeria to Venezuela
Data columns (total 1 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   continent    194 non-null    object
dtypes: object(1)
memory usage: 3.0+ KB

```

In [18]:

```

# To check the shapes
display(fert.shape)

```

```
display(life.shape)
display(popu.shape)
display(cont.shape)

# To illustrate first two rows
display(fert.head(2))
display(life.head(2))
display(popu.head(2))
display(cont.head(2))
```

(56160, 3)
(56420, 3)
(275, 91)
(194, 2)

	country	year	fertility_rate
0	Abkhazia	1800	NaN
1	Afghanistan	1800	7.0

	country	year	Life expectancy
0	Abkhazia	1800	NaN
1	Afghanistan	1800	28.21

	1800	1810	1820	1830	1840	1850	1860
country							
Abkhazia	NaN	NaN	NaN	NaN	NaN	NaN	NaN
Afghanistan	3280000.0	3280000.0	3323519.0	3448982.0	3625022.0	3810047.0	3973968.0

2 rows x 91 columns

	country	continent
0	Algeria	Africa
1	Angola	Africa

```
In [19]: # To check columns of Dataframes
display(fert.columns)
display(life.columns)
display(popu.columns)
display(cont.columns)
```

```
Index(['country', 'year', 'fertility_rate'], dtype='object')
Index(['country', 'year', 'Life expectancy'], dtype='object')
Index([
      1800,      1810,      1820,      1830,
      1840,      1850,      1860,      1870,
      1880,      1890,      1900,      1910,
      1920,      1930,      1940,      1950,
      1951,      1952,      1953,      1954,
      1955,      1956,      1957,      1958,
      1959,      1960,      1961,      1962,
      1963,      1964,      1965,      1966,
      1967,      1968,      1969,      1970,
      1971,      1972,      1973,      1974,
      1975,      1976,      1977,      1978,
      1979,      1980,      1981,      1982,
      1983,      1984,      1985,      1986,
      1987,      1988,      1989,      1990,
```

```

1991, 1992, 1993, 1994,
1995, 1996, 1997, 1998,
1999, 2000, 2001, 2002,
2003, 2004, 2005, 2006,
2007, 2008, 2009, 2010,
2011, 2012, 2013, 2014,
2015, 'Unnamed: 82', 'Unnamed: 83', 'Unnamed: 84',
'Unnamed: 85', 'Unnamed: 86', 'Unnamed: 87', 'Unnamed: 88',
'Unnamed: 89', 'Unnamed: 90', 'Unnamed: 91'],
dtype='object')
Index(['country', 'continent'], dtype='object')

```

```

In [5]: # To delete columns starting with 'Unnamed' in Population dataframe
popu_clean = popu.drop(popu.iloc[:, -10:], axis=1)

```

```

In [6]: # To change the type of fertility's column and Population's column from object
fert.columns = fert.columns.astype(int)
popu_clean.columns = popu_clean.columns.astype(int)

```

```

In [7]: # To have one column with identical label in all dataframes
display(fert.index)
fert.index.name = 'country'
display(fert.tail(3))

display(life.index)
life.index.name = 'country'
display(life.tail(3))

display(popu_clean.index)
popu_clean.index.name = 'country'
display(popu_clean.tail(3))

# 'continent' dataframe has already a column labeled with 'country'
display(cont.index)
display(cont.head(3))

```

```

Index(['Abkhazia', 'Afghanistan', 'Akrotiri and Dhekelia', 'Albania',
      'Algeria', 'American Samoa', 'Andorra', 'Angola', 'Anguilla',
      'Antigua and Barbuda',
      ...,
      'Vietnam', 'Virgin Islands (U.S.)', 'North Yemen (former)',
      'South Yemen (former)', 'Yemen', 'Yugoslavia', 'Zambia', 'Zimbabwe',
      'Åland', 'Åland'],
      dtype='object', name='Total fertility rate', length=260)

```

	1800	1801	1802	1803	1804	1805	1806	1807	1808	1809	...	2006	2007
country													
Zimbabwe	6.75	6.75	6.75	6.75	6.75	6.75	6.75	6.75	6.75	6.75	...	3.94	3.9
Åland	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN
Åland	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN

3 rows × 216 columns

```

Index(['Abkhazia', 'Afghanistan', 'Akrotiri and Dhekelia', 'Albania',
      'Algeria', 'American Samoa', 'Andorra', 'Angola', 'Anguilla',
      'Antigua and Barbuda',
      ...,
      'Vietnam', 'Virgin Islands (U.S.)', 'North Yemen (former)',
      'South Yemen (former)', 'Yemen', 'Yugoslavia', 'Zambia', 'Zimbabwe',

```

```
'Åland', 'South Sudan'],
dtype='object', name='Life expectancy', length=260)
```

	1800	1801	1802	1803	1804	1805	1806	1807	1808	1809	...	2007	20
country													
Zimbabwe	33.70	33.70	33.70	33.70	33.70	33.70	33.70	33.70	33.70	33.70	...	46.4	4
Åland	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN
South Sudan	26.67	26.67	26.67	26.67	26.67	26.67	26.67	26.67	26.67	26.67	...	55.5	5

3 rows × 217 columns

```
Index(['Abkhazia', 'Afghanistan', 'Akrotiri and Dhekelia', 'Albania',
      'Algeria', 'American Samoa', 'Andorra', 'Angola', 'Anguilla',
      'Antigua and Barbuda',
      ...,
      'British Indian Ocean Territory', 'Clipperton',
      'French Southern and Antarctic Lands', 'Gaza Strip',
      'Heard and McDonald Islands', 'Northern Marianas',
      'South Georgia and the South Sandwich Islands',
      'US Minor Outlying Islands', 'Virgin Islands', 'West Bank'],
dtype='object', name='Total population', length=275)
```

	1800	1810	1820	1830	1840	1850	1860	1870	1880	1890	...	2006	2007	;
country														
US Minor Outlying Islands	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	
Virgin Islands	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	
West Bank	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	

3 rows × 81 columns

```
Index(['Algeria', 'Angola', 'Benin', 'Botswana', 'Burkina', 'Burundi',
      'Cameroon', 'Cape Verde', 'Central African Republic', 'Chad',
      ...,
      'Brazil', 'Chile', 'Colombia', 'Ecuador', 'Guyana', 'Paraguay', 'Peru',
      'Suriname', 'Uruguay', 'Venezuela'],
dtype='object', name='country', length=194)
```

continent	
country	
Algeria	Africa
Angola	Africa
Benin	Africa

```
In [8]: # To move the row index into a column
fert = fert.reset_index()
display(fert.head(3))

life = life.reset_index()
display(life.head(3))

popu_clean = popu_clean.reset_index()
display(popu_clean.head(3))
```

```
cont = cont.reset_index()
display(cont.head(3))
```

	country	1800	1801	1802	1803	1804	1805	1806	1807	1808	...	2006	2007	2008
0	Abkhazia	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	NaN
1	Afghanistan	7.0	7.0	7.0	7.0	7.0	7.0	7.0	7.0	7.0	...	6.7	6.46	6.46
2	Akrotiri and Dhekelia	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	NaN

3 rows × 217 columns

	country	1800	1801	1802	1803	1804	1805	1806	1807	1808	...	2007	2008	2009
0	Abkhazia	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	NaN
1	Afghanistan	28.21	28.2	28.19	28.18	28.17	28.16	28.15	28.14	28.13	...	52.4	52.8	52.8
2	Akrotiri and Dhekelia	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	...	NaN	NaN	NaN

3 rows × 218 columns

	country	1800	1810	1820	1830	1840	1850	1860
0	Abkhazia	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	Afghanistan	3280000.0	3280000.0	3323519.0	3448982.0	3625022.0	3810047.0	3973968.0
2	Akrotiri and Dhekelia	NaN	NaN	NaN	NaN	NaN	NaN	NaN

3 rows × 82 columns

	country	continent
0	Algeria	Africa
1	Angola	Africa
2	Benin	Africa

```
In [9]: # To convert wide format to long format (continent dataframe is already in long format)
fert = fert.melt(id_vars='country', var_name='year', value_name='fertility_rate')
life = life.melt(id_vars='country', var_name='year', value_name='Life expectancy')
popu_clean = popu_clean.melt(id_vars='country', var_name='year', value_name='Population')

display(fert)
display(life)
display(popu_clean)
display(cont)
```

	country	year	fertility_rate
0	Abkhazia	1800	NaN
1	Afghanistan	1800	7.00
2	Akrotiri and Dhekelia	1800	NaN
3	Albania	1800	4.60

	country	year	fertility_rate
4	Algeria	1800	6.99
...
56155	Yugoslavia	2015	NaN
56156	Zambia	2015	5.59
56157	Zimbabwe	2015	3.35
56158	Åland	2015	NaN
56159	Åland	2015	NaN

56160 rows × 3 columns

	country	year	Life expectancy
0	Abkhazia	1800	NaN
1	Afghanistan	1800	28.21
2	Akrotiri and Dhekelia	1800	NaN
3	Albania	1800	35.40
4	Algeria	1800	28.82
...
56415	Yugoslavia	2016	NaN
56416	Zambia	2016	57.10
56417	Zimbabwe	2016	61.69
56418	Åland	2016	NaN
56419	South Sudan	2016	56.10

56420 rows × 3 columns

	country	year	Total population
0	Abkhazia	1800	NaN
1	Afghanistan	1800	3280000.0
2	Akrotiri and Dhekelia	1800	NaN
3	Albania	1800	410445.0
4	Algeria	1800	2503218.0
...
22270	Northern Marianas	2015	NaN
22271	South Georgia and the South Sandwich Islands	2015	NaN
22272	US Minor Outlying Islands	2015	NaN
22273	Virgin Islands	2015	NaN
22274	West Bank	2015	NaN

22275 rows × 3 columns

country	continent
---------	-----------

	country	continent
0	Algeria	Africa
1	Angola	Africa
2	Benin	Africa
3	Botswana	Africa
4	Burkina	Africa
...
189	Paraguay	South America
190	Peru	South America
191	Suriname	South America
192	Uruguay	South America
193	Venezuela	South America

194 rows × 2 columns

```
In [10]: """To merge all four dataframes with common column(s) with same labels.
Below first fertility and lifeexpectancy are merged then the resulting data
to population dataframe, afterall continent dataframe is merged to resultin
df1 = fert.merge(life,on=['country','year']).merge(popu_clean,on=['country','year'])
df1
```

```
Out[10]:
```

	country	year	fertility_rate	Life expectancy	Total population	continent
0	Afghanistan	1800	7.00	28.21	3280000.0	Asia
1	Afghanistan	1810	7.00	28.11	3280000.0	Asia
2	Afghanistan	1820	7.00	28.01	3323519.0	Asia
3	Afghanistan	1830	7.00	27.90	3448982.0	Asia
4	Afghanistan	1840	7.00	27.80	3625022.0	Asia
...
14170	Zimbabwe	2011	3.64	51.60	14255592.0	Africa
14171	Zimbabwe	2012	3.56	54.20	14565482.0	Africa
14172	Zimbabwe	2013	3.49	55.70	14898092.0	Africa
14173	Zimbabwe	2014	3.41	57.00	15245855.0	Africa
14174	Zimbabwe	2015	3.35	59.30	15602751.0	Africa

14175 rows × 6 columns

```
In [20]: # To define a datafrmae for years between 1960 & 2015 (including 2015)
df2 = df1[df1['year'].between(1960,2016)]
df2
```

```
Out[20]:
```

	country	year	fertility_rate	Life expectancy	Total population	continent
25	Afghanistan	1960	7.67	31.94	8994793.0	Asia
26	Afghanistan	1961	7.67	32.47	9164945.0	Asia

	country	year	fertility_rate	Life expectancy	Total population	continent
27	Afghanistan	1962	7.67	33.01	9343772.0	Asia
28	Afghanistan	1963	7.67	33.53	9531555.0	Asia
29	Afghanistan	1964	7.67	34.07	9728645.0	Asia
...
14170	Zimbabwe	2011	3.64	51.60	14255592.0	Africa
14171	Zimbabwe	2012	3.56	54.20	14565482.0	Africa
14172	Zimbabwe	2013	3.49	55.70	14898092.0	Africa
14173	Zimbabwe	2014	3.41	57.00	15245855.0	Africa
14174	Zimbabwe	2015	3.35	59.30	15602751.0	Africa

9800 rows × 6 columns

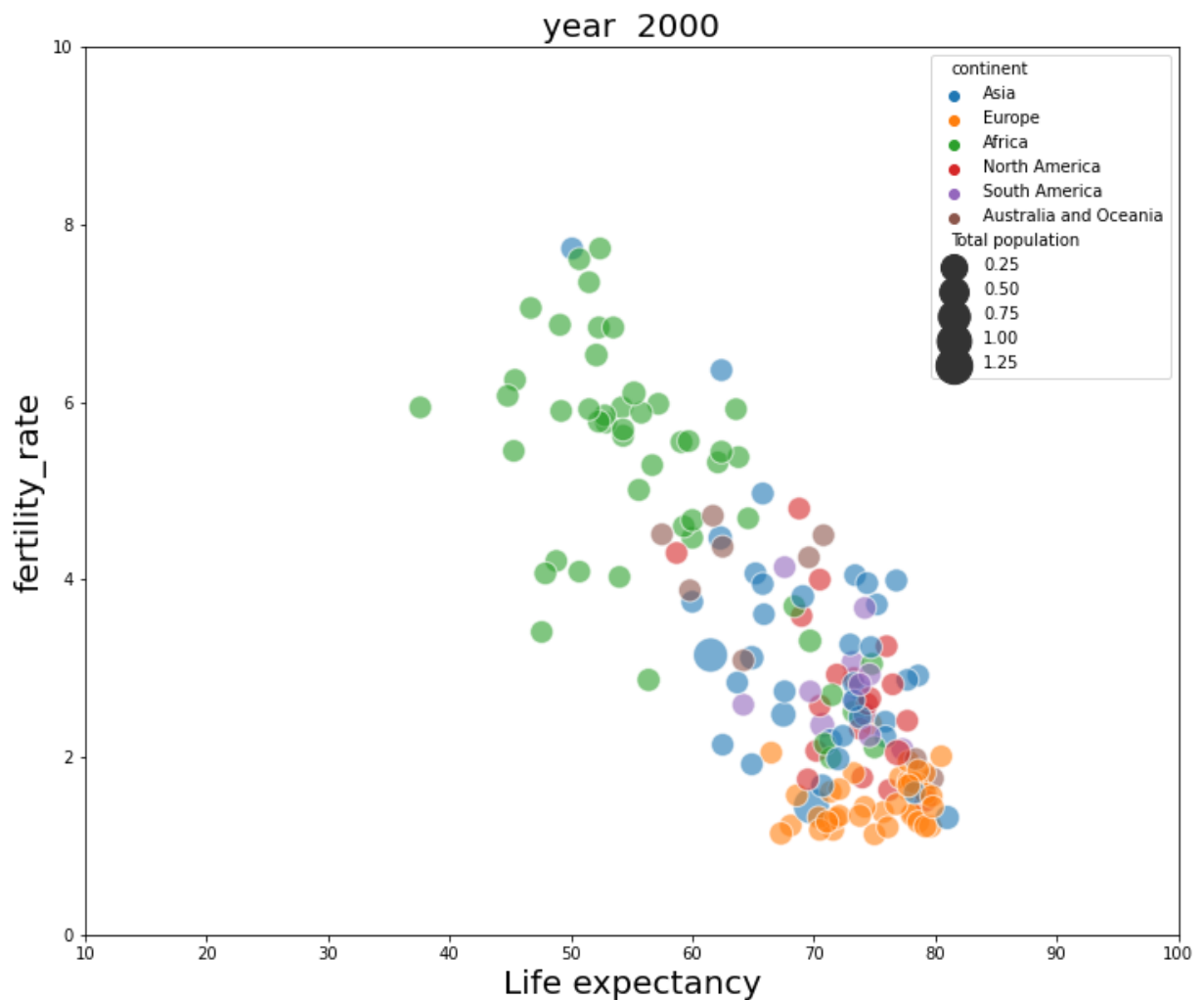
In [12]:

```
import matplotlib.pyplot as plt
import seaborn as sns

"""To plot relationship between ``Life expectancy`` and ``fertility_rate`` us
of the dataframe using the ``continent``, ``Total population`` parameters
Here the ``Total population`` is given as size of circle and ``continent``

df = df2.loc[df2['year'] == 2000]
plt.figure(figsize=(12,10))
sns.scatterplot(data=df, x='Life expectancy', y='fertility_rate', hue='continent',
                size='Total population',
                sizes=(200,500), palette=None, alpha=0.6, legend=True)
plt.xlabel('Life expectancy', fontsize=20)
plt.ylabel('fertility_rate', fontsize=20)
plt.title('year'+str(' 2000'), fontsize=20)

# To fix the coordinates of the plot
plt.axis((10, 100, 0, 10))
plt.savefig('2000.png', dpi=300, facecolor='white')
```



```
In [14]: # To go through each year and produce a scatterplot related to a year
for i in range(1960,2016,1):
    df = df2.loc[df2['year'] == i]
    plt.figure(figsize=(12,10))
    sns.scatterplot(data=df, x='Life expectancy', y='fertility_rate', size='Total population',
                    sizes=(200,500), palette=None, alpha=0.6, legend=True)
    plt.xlabel('Life expectancy', fontsize=20)
    plt.ylabel('fertility_rate', fontsize=20)
    plt.title('year'+str('')+str(i), fontsize=20)

    plt.axis((10, 100, 0, 10))
    plt.savefig('plot_'+str(i)+'.png', dpi=300, format='png', facecolor='white')
    plt.close()
```

```
In [15]: pip install imageio
```

```
Requirement already satisfied: imageio in /Users/Disalo/opt/anaconda3/lib/python3.8/site-packages (2.9.0)
Requirement already satisfied: pillow in /Users/Disalo/opt/anaconda3/lib/python3.8/site-packages (from imageio) (8.2.0)
Requirement already satisfied: numpy in /Users/Disalo/opt/anaconda3/lib/python3.8/site-packages (from imageio) (1.20.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [17]: # To generate a gif file from the generated scatterplots
import imageio

images = []
```

```
for i in range(1960, 2016, 1):  
    filename = 'plot_{}.png'.format(i)  
    images.append(imageio.imread(filename))  
  
imageio.mimsave('1.8. Long vs. Wide Format_Animated Scatterplot.gif', images,
```