

Pandas - Series



Hands on!

```
In [1]:
```

```
import pandas as pd
import numpy as np
```

Pandas Series

We'll start analyzing "The Group of Seven (https://en.wikipedia.org/wiki/Group of Seven)". Which is a political formed by Canada, France, Germany, Italy, Japan, the United Kingdom and the United States. We'll start by analyzing population, and for that, we'll use a pandas.Series object.

```
In [2]:
```

```
# In millions
g7_pop = pd.Series([35.467, 63.951, 80.940, 60.665, 127.061, 64.511, 318.523])
```

In [3]:

```
g7_pop
```

Out[3]:

```
0 35.467
1 63.951
2 80.940
3 60.665
4 127.061
5 64.511
6 318.523
dtype: float64
```

Someone might not know we're representing population in millions of inhabitants. Series can have a <code>name</code> , to better document the purpose of the Series:

```
In [4]:
g7 pop.name = 'G7 Population in millions'
In [5]:
g7 pop
Out[5]:
0
      35.467
1
      63.951
2
      80.940
3
      60.665
4
     127.061
5
      64.511
     318.523
6
Name: G7 Population in millions, dtype: float64
Series are pretty similar to numpy arrays:
In [6]:
g7_pop.dtype
Out[6]:
dtype('float64')
In [7]:
g7_pop.values
Out[7]:
array([ 35.467, 63.951, 80.94 , 60.665, 127.061, 64.511, 318.52
3])
They're actually backed by numpy arrays:
In [8]:
```

```
type(g7 pop.values)
```

Out[8]:

numpy.ndarray

And they look like simple Python lists or Numpy Arrays. But they're actually more similar to Python dict s.

A Series has an index, that's similar to the automatic index assigned to Python's lists:

```
In [9]:
g7_pop
Out[9]:
0
      35.467
      63.951
1
2
      80.940
3
      60.665
4
     127.061
5
      64.511
6
     318.523
Name: G7 Population in millions, dtype: float64
In [10]:
g7_pop[0]
Out[10]:
35.467
In [11]:
g7_pop[1]
Out[11]:
63.951
In [12]:
g7 pop.index
Out[12]:
RangeIndex(start=0, stop=7, step=1)
In [13]:
1 = ['a', 'b', 'c']
```

But, in contrast to lists, we can explicitly define the index:

```
In [14]:
```

In [15]:

United States

Name: G7 Population in millions, dtype: float64

318.523

Compare it with the <u>following table (https://docs.google.com/spreadsheets/d/1llorV2-Oh9Da1JAZ7weVw86PQrQydSMp-ydVMH135il/edit?usp=sharing)</u>:

G7 Population			
(Expressed in millions)			
Canada	35.467		
France	63.951		
Germany	80.94		
Italy	60.665		
Japan	127.061		
United Kingdom	64.511		
United States	318.523		

We can say that Series look like "ordered dictionaries". We can actually create Series out of dictionaries:

In [16]:

```
pd.Series({
    'Canada': 35.467,  #this summarises the two steps above - can do everyth
ing at once
    'France': 63.951,
    'Germany': 80.94,
    'Italy': 60.665,
    'Japan': 127.061,
    'United Kingdom': 64.511,
    'United States': 318.523
}, name='G7 Population in millions')
```

Out[16]:

Canada	35.467		
France	63.951		
Germany	80.940		
Italy	60.665		
Japan	127.061		
United Kingdom	64.511		
United States	318.523		
Name: G7 Population	on in millions,	dtype:	float64

```
In [17]:
```

Out[17]:

```
Canada 35.467
France 63.951
Germany 80.940
Italy 60.665
Japan 127.061
United Kingdom 64.511
United States 318.523
Name: G7 Population in millions, dtype: float64
```

You can also create Series out of other series, specifying indexes:

```
In [18]:
```

```
pd.Series(g7_pop, index=['France', 'Germany', 'Italy', 'Spain'])

Out[18]:

France 63.951
Germany 80.940
Italy 60.665
Spain NaN
Name: G7 Population in millions, dtype: float64
```

Indexing

Indexing works similarly to lists and dictionaries, you use the index of the element you're looking for:

```
In [19]:
```

```
g7_pop
Out[19]:
Canada
                    35.467
France
                    63.951
Germany
                    80.940
Italy
                    60.665
Japan
                   127.061
                    64.511
United Kingdom
United States
                   318.523
Name: G7 Population in millions, dtype: float64
```

```
In [20]:
g7_pop['Canada']
Out[20]:
35.467
In [21]:
g7_pop['Japan']
Out[21]:
127.061
Numeric positions can also be used, with the iloc attribute:
In [22]:
g7_pop.iloc[0]
Out[22]:
35.467
In [23]:
g7_pop.iloc[-1]
Out[23]:
318.523
Selecting multiple elements at once:
In [24]:
g7_pop[['Italy', 'France']]
Out[24]:
           60.665
Italy
France
           63.951
Name: G7 Population in millions, dtype: float64
(The result is another Series)
In [25]:
g7_pop.iloc[[0, 1]]
Out[25]:
Canada
           35.467
France
           63.951
Name: G7 Population in millions, dtype: float64
```

Slicing also works, but **important**, in Pandas, the upper limit is also included:

```
In [28]:
```

```
g7_pop['Canada': 'Italy'] # unlike in a list, the upper limit IS included (unlike 1[:2])

Out[28]:

Canada 35.467
France 63.951
Germany 80.940
Italy 60.665
Name: G7 Population in millions, dtype: float64
```

Conditional selection (boolean arrays)

The same boolean array techniques we saw applied to numpy arrays can be used for Pandas Series:

```
In [31]:
```

```
g7_pop
Out[31]:
                    35.467
Canada
France
                    63.951
Germany
                    80.940
Italy
                    60.665
Japan
                   127.061
United Kingdom
                    64.511
United States
                   318.523
Name: G7 Population in millions, dtype: float64
In [32]:
g7_pop > 70
Out[32]:
Canada
                   False
France
                   False
Germany
                    True
Italy
                   False
Japan
                    True
United Kingdom
                   False
United States
                    True
Name: G7 Population in millions, dtype: bool
In [33]:
g7 pop[g7 pop > 70]
Out[33]:
Germany
                   80.940
Japan
                  127.061
                  318.523
United States
```

Name: G7 Population in millions, dtype: float64

```
In [34]:
g7 pop.mean()
Out[34]:
107.30257142857144
In [35]:
g7_pop[g7_pop > g7_pop.mean()]
Out[35]:
                 127.061
Japan
United States
                 318.523
Name: G7 Population in millions, dtype: float64
In [ ]:
g7_pop.std()
In [ ]:
~ not
  or
& and
In [ ]:
g7_pop[(g7_pop > g7_pop.mean() - g7_pop.std() / 2) | (g7_pop > g7_pop.mean() + g
7 pop.std() / 2)]
```

Operations and methods

Series also support vectorized operations and aggregation functions as Numpy:

```
In [29]:
g7_pop
Out[29]:
Canada
                    35.467
France
                    63.951
Germany
                    80.940
Italy
                    60.665
                   127.061
Japan
United Kingdom
                    64.511
United States
                   318.523
Name: G7 Population in millions, dtype: float64
```

```
In [30]:
g7_pop * 1_000_000
Out[30]:
Canada
                    35467000.0
France
                    63951000.0
Germany
                    80940000.0
Italy
                    60665000.0
Japan
                   127061000.0
United Kingdom
                    64511000.0
United States
                   318523000.0
Name: G7 Population in millions, dtype: float64
In [36]:
g7_pop.mean()
Out[36]:
107.30257142857144
In [37]:
np.log(g7_pop)
Out[37]:
Canada
                   3.568603
France
                   4.158117
Germany
                   4.393708
Italy
                   4.105367
                  4.844667
Japan
                  4.166836
United Kingdom
United States
                   5.763695
Name: G7 Population in millions, dtype: float64
In [38]:
g7_pop['France': 'Italy'].mean()
Out[38]:
```

Boolean arrays

68.5186666666666

(Work in the same way as numpy)

In [39]:

```
g7_pop
```

Out[39]:

 Canada
 35.467

 France
 63.951

 Germany
 80.940

 Italy
 60.665

 Japan
 127.061

 United Kingdom
 64.511

 United States
 318.523

Name: G7 Population in millions, dtype: float64

In [40]:

```
g7_pop > 80
```

Out[40]:

Canada False
France False
Germany True
Italy False
Japan True
United Kingdom False
United States True

Name: G7 Population in millions, dtype: bool

In [41]:

```
g7_pop[g7_pop > 80]
```

Out[41]:

Germany 80.940 Japan 127.061 United States 318.523

Name: G7 Population in millions, dtype: float64

In [42]:

```
g7_pop[(g7_pop > 80) | (g7_pop < 40)]
```

Out[42]:

 Canada
 35.467

 Germany
 80.940

 Japan
 127.061

 United States
 318.523

Name: G7 Population in millions, dtype: float64

```
In [43]:
```

```
g7\_pop[(g7\_pop > 80) \& (g7\_pop < 200)] # all the countries that have more than 8 0 and less than 200
```

Out[43]:

Germany 80.940 Japan 127.061

Name: G7 Population in millions, dtype: float64

Modifying series

```
In [44]:
q7 pop['Canada'] = 40.5 # assigning a variable - assigned canada
In [45]:
g7 pop
Out[45]:
Canada
                    40.500
France
                    63.951
Germany
                    80.940
                    60.665
Italy
Japan
                   127.061
United Kingdom
                    64.511
United States
                   318.523
Name: G7 Population in millions, dtype: float64
In [46]:
g7\_pop.iloc[-1] = 500 # the last country will have 500
In [47]:
g7_pop
Out[47]:
Canada
                    40.500
France
                    63.951
Germany
                    80.940
Italy
                    60.665
                   127.061
Japan
United Kingdom
                    64.511
                   500.000
United States
Name: G7 Population in millions, dtype: float64
```

In [48]:

```
g7_pop[g7_pop < 70] # all the countries that have less than 70
```

Out[48]:

 Canada
 40.500

 France
 63.951

 Italy
 60.665

 United Kingdom
 64.511

Name: G7 Population in millions, dtype: float64

In [49]:

 $g7_pop[g7_pop < 70] = 99.99 \# all the countries that have less than 70 will have 9$

In [50]:

g7_pop

Out[50]:

 Canada
 99.990

 France
 99.990

 Germany
 80.940

 Italy
 99.990

 Japan
 127.061

 United Kingdom
 99.990

 United States
 500.000

Name: G7 Population in millions, dtype: float64