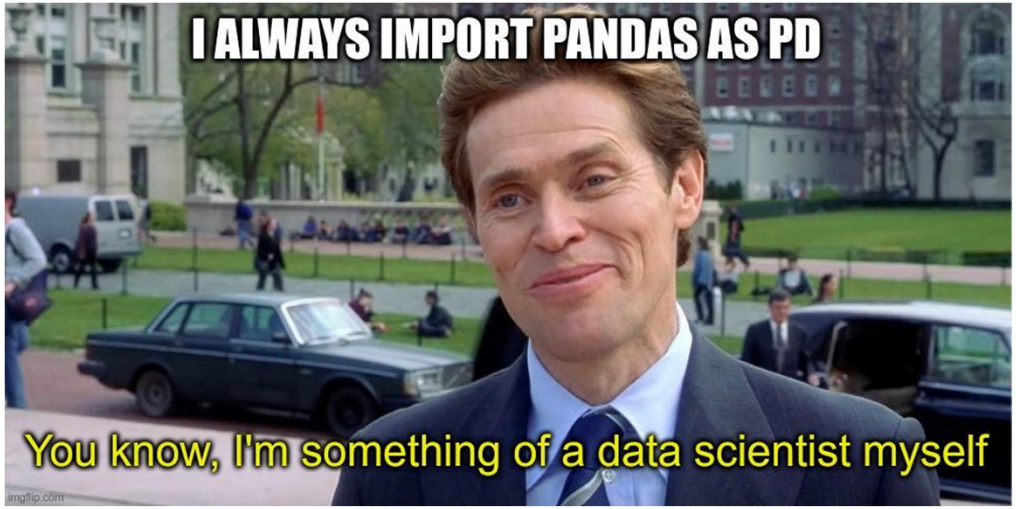
Data and Statistics with Python

So, you want to become a data scientist?

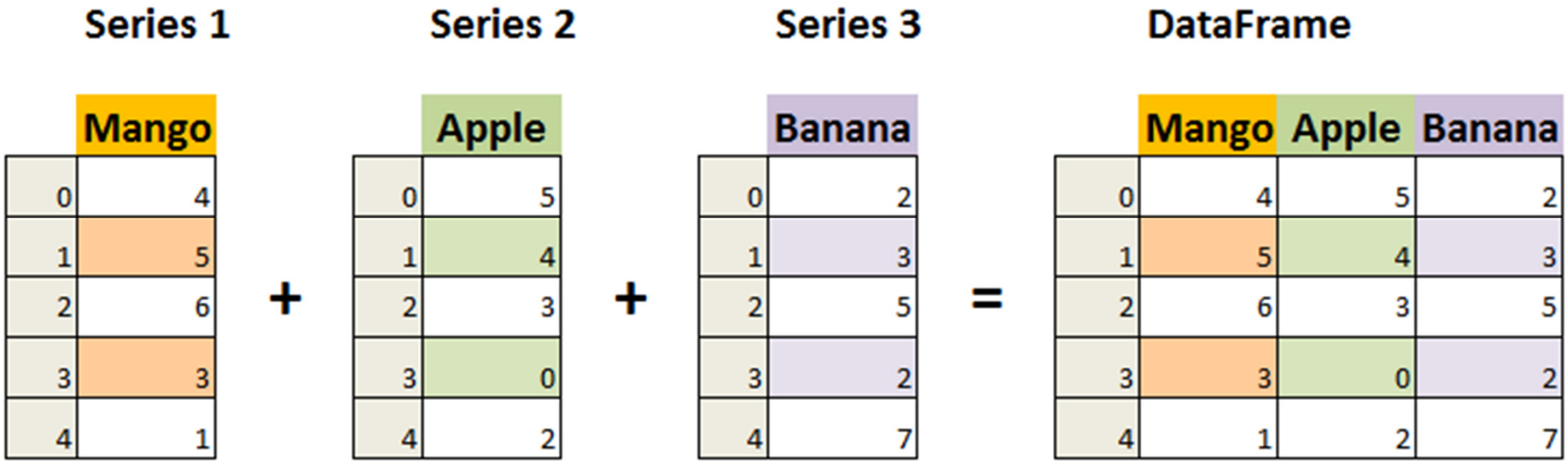


Requirement

pip install pandas bokeh

Creating Series and DataFrame objects

- Pandas objects can be thought of as enhanced versions of NumPy structured arrays in which the rows and columns are identified with labels rather than simple integer indices.
- Pandas provides a host of useful tools, methods, and functionality on top of the basic data structures, but nearly everything that follows will require an understanding of what these structures are. Thus, before we go any further, let's introduce these three fundamental Pandas data structures: the Series and DataFrame.



Series and DataFrame

```
+ import pandas as pd
+ import numpy as np
+ from numpy.random import default_rng

+ rng = default_rng(12345) #seed for random data
+ diff_data = rng.normal(0, 1, size=100)  #generate random data
+ cumulative = np.add.accumulate(diff_data)
+ data_series = pd.Series(diff_data) #create Series
+ print(data_series) # print data
+ data_frame = pd.DataFrame({
+     "diffs": data_series,
+     "cumulative": data_series.cumsum()
+ }) #create data frame
+ print(data_frame)
```

Loading and storing data from DataFrame

- Usually, we create a DataFrame object from the raw data in a Python session.
- In practice, the data will often come from an external source, such as an existing spreadsheet or CSV file, database, or API endpoint.
- For this reason, pandas provides numerous utilities for loading and storing data to file.
- Out of the box, pandas supports loading and storing data from CSV, Excel (xls or xlsx), JSON, SQL, Parquet, and Google BigQuery.
- This makes it very easy to import your data into pandas and then manipulate and analyze this data using Python.

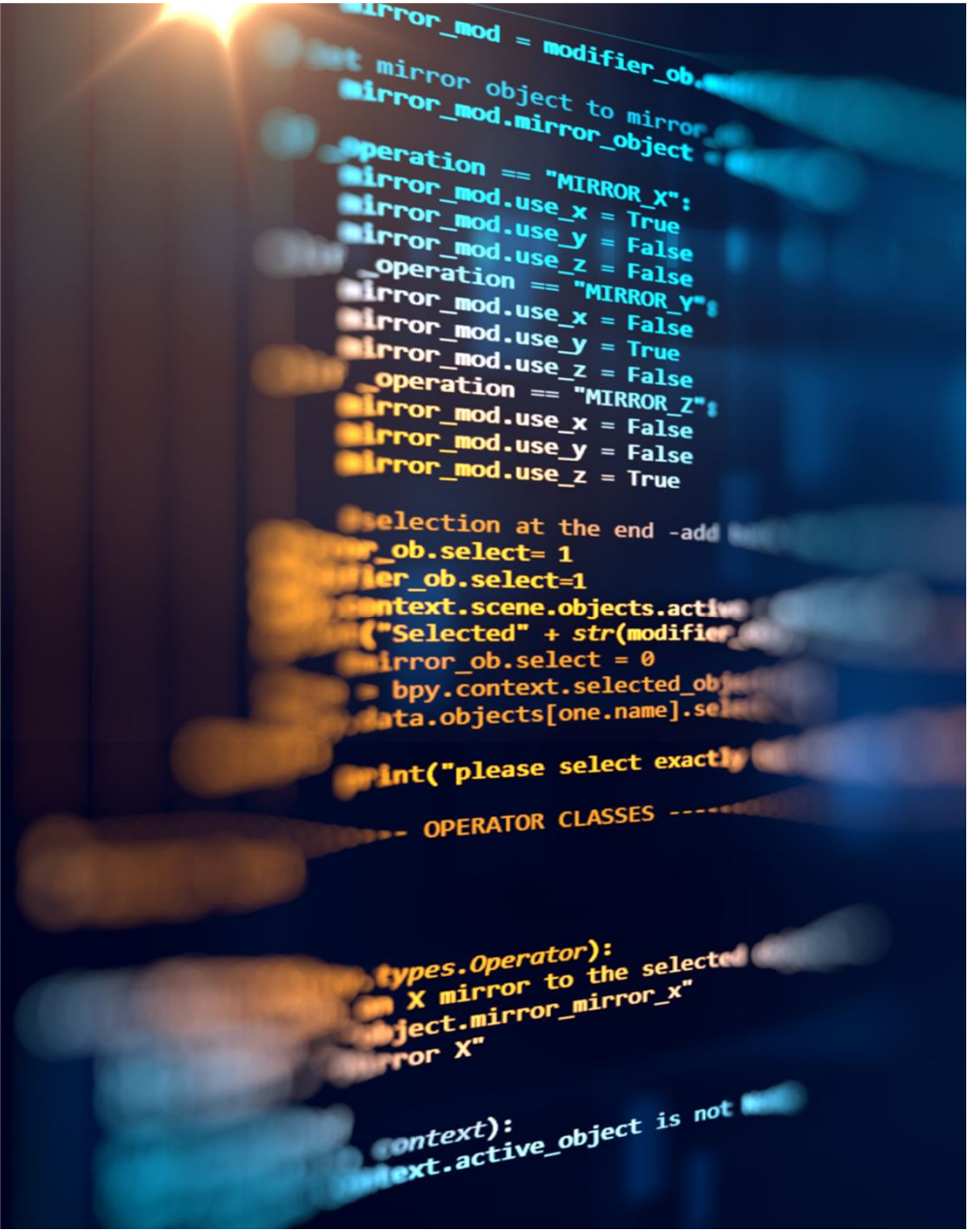
Save and Load

- Save DataFrame to CSV file

```
data_frame.to_csv("sample.csv", index=False)
```

- Load data from CSV file to DataFrame

```
df = pd.read_csv("sample.csv", index_col=False)
print(df)
```




```
import pandas as pd
import numpy as np
from numpy.random import default_rng
rng = default_rng(12345)

diffs = rng.normal(0, 1, size=100)
cumulative = np.add.accumulate(diffs)

data_frame = pd.DataFrame({
   "diffs": diffs,
   "cumulative": cumulative
})
print(data_frame)

data_frame.to_csv("sample.csv", index=False)
df = pd.read_csv("sample.csv", index_col=False)
print(df)
```

Manipulate DataFrame

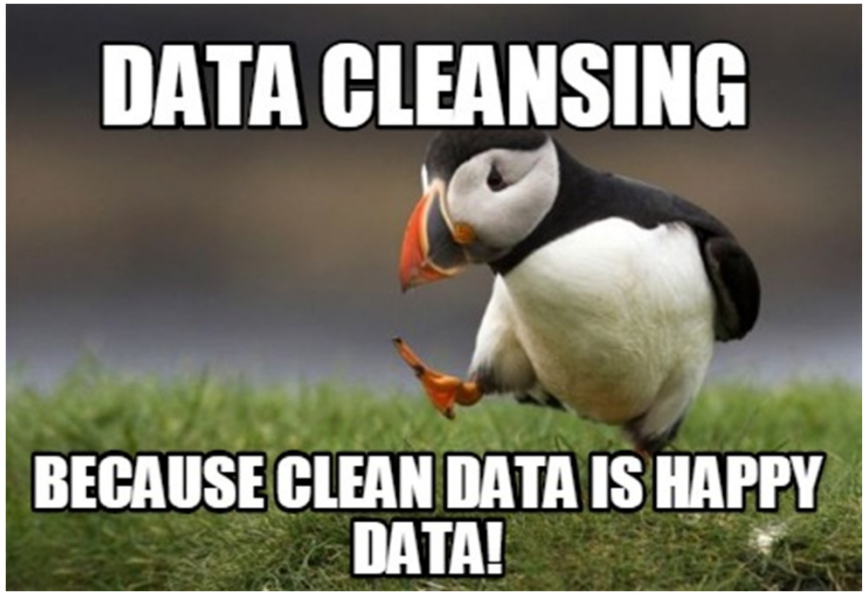
```
df = pd.DataFrame({'A': [1,2,3], 'B': [10,20,30] })
def plus_10(x):
        return x+10
df.apply(plus_10)
```

	A	B
0	11	20
1	12	30
2	13	40

Manipulate single column

+ df['B_ap'] = df['B'].**apply(plus_10)**

	A	B	B_ap
0	1	10	20
1	2	20	30
2	3	30	40



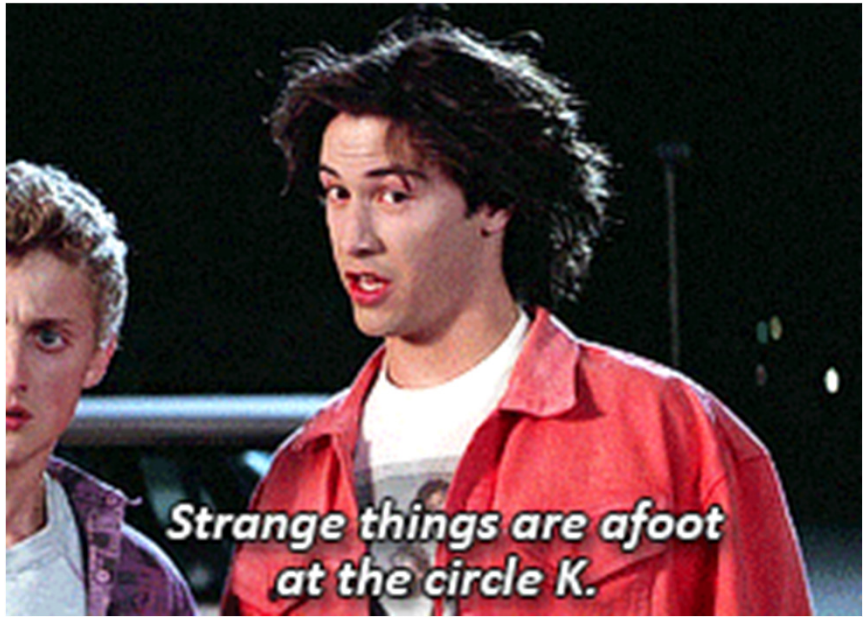
What about null vales?

+ df = data_frame.dropna()

NaN, NaT, and None

- If a column is **numeric** and you have a missing value that value will be a NaN. **NaN stands for *Not a Number*.**
- If a column is a **DateTime** and you have a missing value, then that value will be a NaT. **NaT stands for *Not a Time***.

A pandas object dtype column - the dtype for strings as of this writing - can hold None, NaN, NaT or all three at the same time!



What are these NaN values anyway?

+ NaN is a NumPy value. np.**NaN**
+ NaT is a Pandas value. pd.**NaT**
+ None is a vanilla Python value. **None**

Finding missing values

```
>>> print(df.isna())
       a      b
0  False  False
1  False   True
2  False  False
```

```
import pandas as pd
import numpy as np
#creating the DataFrame
df = pd.DataFrame({'a':[0,1,''],'b':['',None,3]})
print("------The DataFrame is----------")
print(df)
print("-------------------------------")
print(df.isna())
```

Plotting data from a DataFrame

- Step 1: Prepare the data
- Step 2: Create the DataFrame
- Step 3: Plot the DataFrame using Pandas

Step 1: Prepare the data

Unemployment_Rate	Stock_Index_Price
6.1	1500
5.8	1520
5.7	1525
5.7	1523
5.8	1515
5.6	1540
5.5	1545
5.3	1560
5.2	1555
5.2	1565

+ The following data will be used to create the scatter diagram. This data captures the relationship between two variables related to an economy:

Step 2: Create the DataFrame

```
import pandas as pd

data = {'Unemployment_Rate':
[6.1,5.8,5.7,5.7,5.8,5.6,5.5,5.3,5.2,5.2], 'Stock_Index_Price':
[1500,1520,1525,1523,1515,1540,1545,1560,1555,1565] }


df =
pd.DataFrame(data,columns=['Unemployment_Rate','Stock
_Index_Price'])


print (df)
```

	Unemployment_Rate	Stock_Index_Price
0	6.1	1500
1	5.8	1520
2	5.7	1525
3	5.7	1523
4	5.8	1515
5	5.6	1540
6	5.5	1545
7	5.3	1560
8	5.2	1555
9	5.2	1565

Step 3: Plot the DataFrame using Pandas

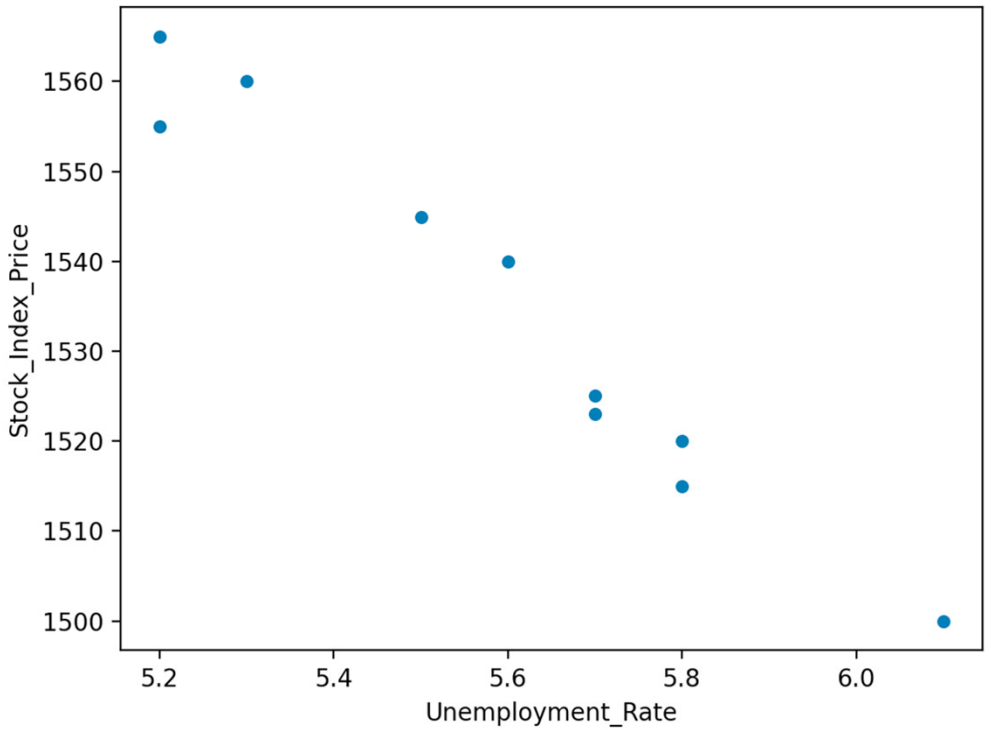
- df.plot(x ='Unemployment_Rate', y='Stock_Index_Price', kind = 'scatter')

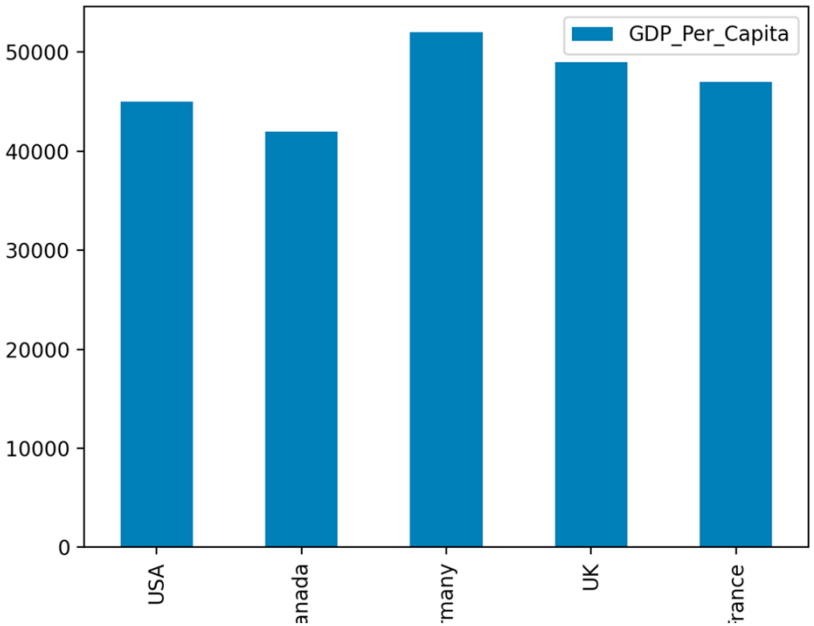
- Notice that you can specify the type of chart by setting **kind = 'scatter'**
- You'll also need to add the Matplotlib syntax to show the plot (ensure that the Matplotlib package is install in Python):
- **import matplotlib.pyplot as plt**

Putting everything together:

```
import pandas as pd
import matplotlib.pyplot as plt
data = {'Unemployment_Rate': [6.1,5.8,5.7,5.7,5.8,5.6,5.5,5.3,5.2,5.2],
'Stock_Index_Price': [1500,1520,1525,1523,1515,1540,1545,1560,1555,1565] }
df = pd.DataFrame(data,columns=['Unemployment_Rate','Stock_Index_Price'])
df.plot(x ='Unemployment_Rate', y='Stock_Index_Price', kind = 'scatter')
plt.show()
```

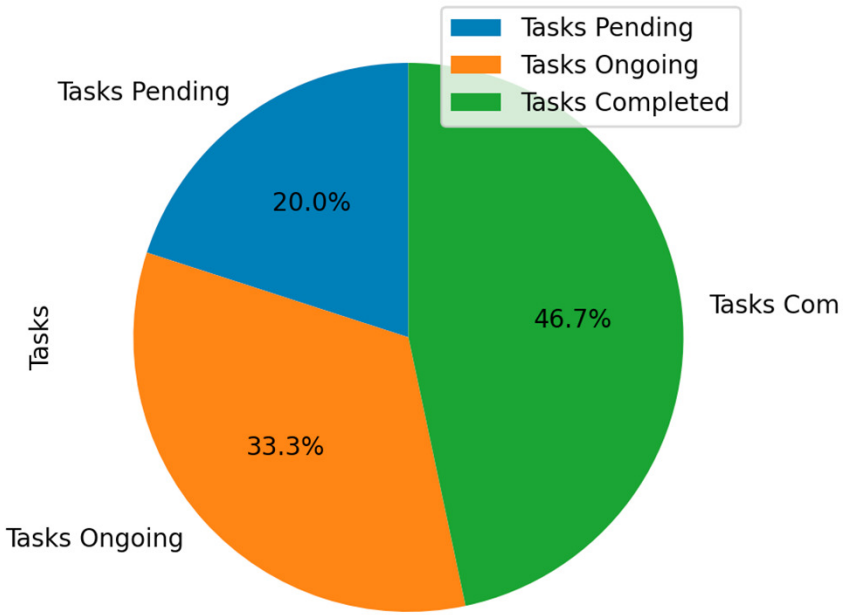
Output


1560
1550
1540
1530
1520
1510
1500
Stock_Index_Price
5.2
5.4
5.6
5.8
6.0
Unemployment_Rate



Plot a Bar Chart using Pandas

```
import pandas as pd

import matplotlib.pyplot as plt

data = {'Country':
['USA','Canada','Germany','UK','France'],
'GDP_Per_Capita': [45000,42000,52000,49000,47000]
}

df =
pd.DataFrame(data,columns=['Country','GDP_Per_Capita'])

df.plot(x ='Country', y='GDP_Per_Capita', kind = 'bar’)

plt.show()
```



Plot a Pie Chart using Pandas

```
import pandas as pd

import matplotlib.pyplot as plt

 data = {'Tasks': [300,500,700]}

df =
pd.DataFrame(data,columns=['Tasks'],index
= ['Tasks Pending','Tasks Ongoing','Tasks
Completed’])

df.plot.pie(y='Tasks',figsize=(5,
5),autopct='%1.1f%%', startangle=90)

plt.show()
```

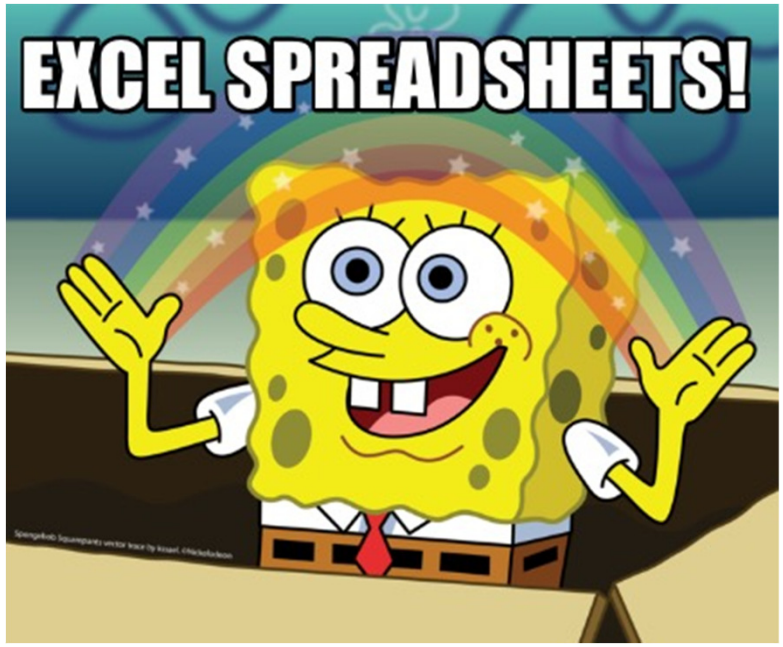
There are other libs in python for plotting graphs!

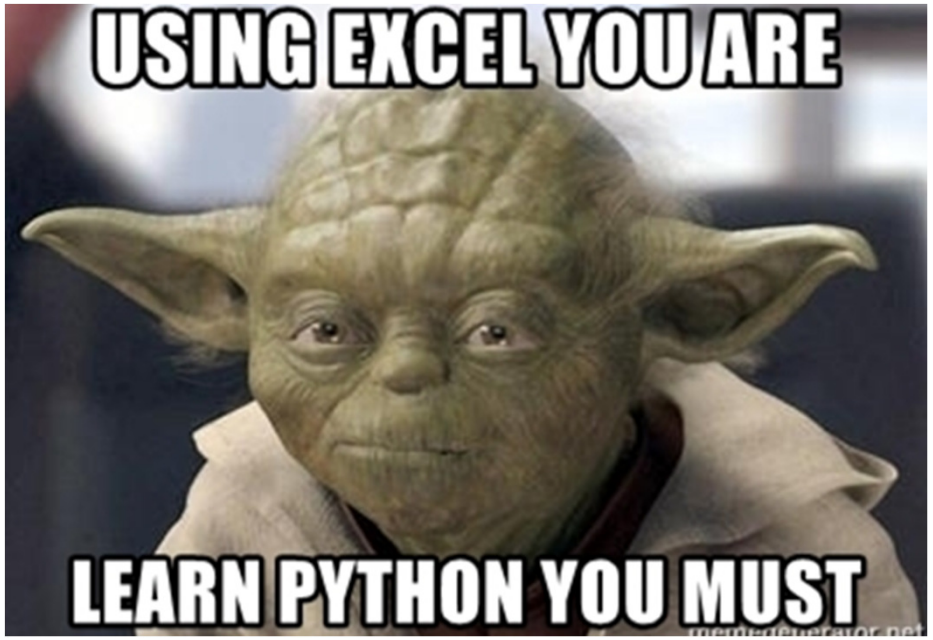
Descriptive statistics

- Descriptive statistics summarizes or describes the characteristics of a data set.
- Descriptive statistics consists of two basic categories of measures: measures of central tendency and measures of variability (or spread).
- Measures of central tendency describe the center of a data set.
- Measures of variability or spread describe the dispersion of data within the set.

```
import pandas as pd
import numpy as np

#Create a Dictionary of series
d = {'Name':pd.Series(['Tom','James','Ricky','Vin','Steve','Smith','Jack',
   'Lee','David','Gasper','Betina','Andres']),
   'Age':pd.Series([25,26,25,23,30,29,23,34,40,30,51,46]),
   'Rating':pd.Series([4.23,3.24,3.98,2.56,3.20,4.6,3.8,3.78,2.98,4.80,4.10,3.65])
}

#Create a DataFrame
df = pd.DataFrame(d)
print (df. describe(include='all'))
```

	Name	Age	Rating
count	12	12.000000	12.000000
unique	12	NaN	NaN
top	Tom	NaN	NaN
freq	1	NaN	NaN
mean	NaN	31.833333	3.743333
std	NaN	9.232682	0.661628
min	NaN	23.000000	2.560000
25%	NaN	25.000000	3.230000
50%	NaN	29.500000	3.790000
75%	NaN	35.500000	4.132500
max	NaN	51.000000	4.800000





You decide!

But Python is better, and it is free!