Applied Data Science with Python for Beginners Lecture 2 - 04 May 2021

## What you learn today...

... work with data in a DataFrame

... filter your data

... merge, group and aggregate your data

... visualize data with simple plots

#### Libraries

- A collection of functions is bundled in a **library**
- we import these libraries and can use the defined functions
- Some libraries come with a Python installation, some need to be installed



... for plotting and visualization



... for working with tabular data (Excel-files, csv-files,...)



... creating machine learning models

## Function & Methods

**Functions** 

```
a = function_name(parameter)
```

#### Function & Methods

**Functions** 

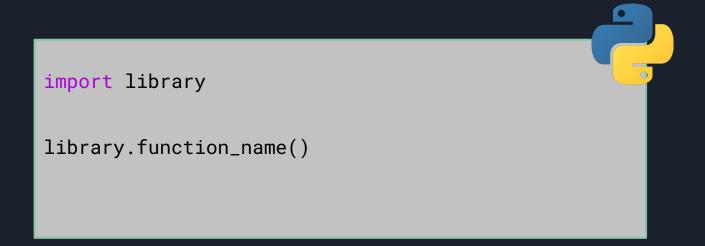
Methods

a = function\_name(parameter)

a = "a string"

a = a.upper() # 'A STRING'

# Import Libraries



# Import Libraries

```
import library as 1
l.function_name()
```

## NumPy - library



- Library for scientific computing
- Work with lists, matrices or higher dimensional structures
- NumPy lists have much more functionality than usual lists

```
import numpy as np
a = np.array([1,2,3,4])
```

## NumPy - library



- Library for scientific computing
- Work with lists, matrices or higher dimensional structures
- NumPy lists have much more functionality than usual lists

```
import numpy as np
a = np.array([1,2,3,4])
a.sum() # 10
a.mean() # 2.5
a.std() # 1.118...
```



## NumPy - library



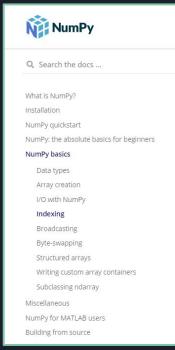
- Library for scientific computing
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- NumPy lists have much more functionality than usual lists

```
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a = np.array([1,2,3,4])
a.sum() # 10
a.mean() # 2.5
a.std() # 1.118...
```



#### Documentation

- Explanations to the methods and functions
- Often include examples and tutorials
- https://numpy.org/doc/



User Guide API reference Development

#### Single element indexing

Single element indexing for a 1-D array is what one expects. It work exactly like that for other standard Python sequences. It is 0-based, and accepts negative indices for indexing from the end of the array.

```
>>> x = np.arange(10)
>>> x[2]
2
>>> x[-2]
8
```

Unlike lists and tuples, numpy arrays support multidimensional indexing for multidimensional arrays. That means that it is not necessary to separate each dimension's index into its own set of square brackets.

```
>>> x.shape = (2,5) # now x is 2-dimensional
>>> x[1,3]
8
>>> x[1,-1]
```

Note that if one indexes a multidimensional array with fewer indices than dimensions, one gets a subdimensional array. For example:

## NumPy - Append



```
import numpy as np
a = np.array([1,2,3,4])
np.append(a, [5,6,7]) # array([1, 2, 3, 4, 5, 6, 7])
```

## NumPy - Append



```
import numpy as np
a = np.array([1,2,3,4])
np.append(a, [5,6,7]) # array([1, 2, 3, 4, 5, 6, 7])
a = np.append(a, [5,6,7])
```

#### Compute with arrays - Broadcasting

```
import numpy as np
a = np.array([1,2,3])
a * 2 # array([2, 4, 6])
a ** 2 # array([1, 4, 9])
a - 1 # array([0, 1, 2])
```

## Compute with arrays

```
import numpy as np
a = np.array([1,2,3])
b = np.array([1,1,1])
a + b # array([2, 3, 4])
```

## Indexing

```
import numpy as np
a = np.array([1,2,3,4])
a[0:2] # array([1,2])
```

## Boolean Indexing

```
import numpy as np
a = np.array([1,2,3,4])
a[[True,True,False,False]] # array([1,2])
```

## Boolean Indexing

```
import numpy as np
a = np.array([1,2,3,4])
a <= 2 # [True, True, False, False]</pre>
```

## Boolean Indexing

```
import numpy as np

a = np.array([1,2,3,4])
a <= 2 # [True, True, False, False]
a[a<=2] # array([1,2])</pre>
```

## Exercise 1

```
import numpy as np

a = np.array([1,2,3,4])
a = np.append(a, [5,6,7]) # append
a[a<=2] # boolean indexing</pre>
```

# pandas DataFrame

# **pandas** - library

```
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```

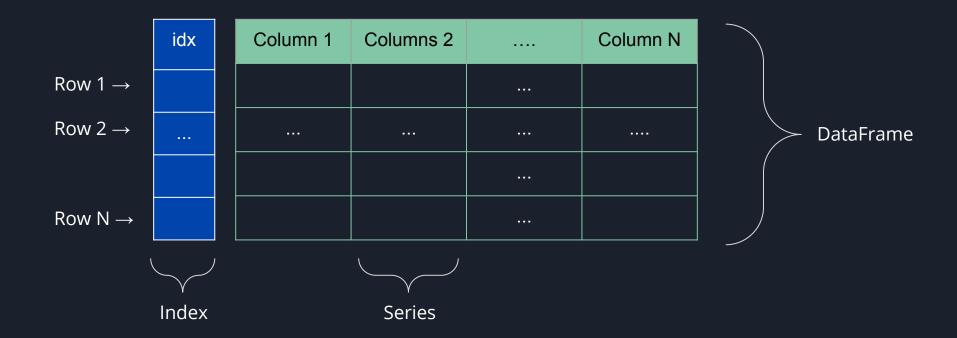
```
import pandas as pd

data = {
  "Name":["Clara", "Tom", "Sarah", "John"],
  "Age" :[20,24,19,21]}

df = pd.DataFrame(data)
```

	Name	Age
0	Clara	20
1	Tom	24
2	Sarah	19
3	John	21

#### DataFrame - data structure



#### Read data

- Pandas can read data from files
- Various data formats possible

```
df = pd.read_csv('path_to_file')
df = pd.read_excel('path_to_file')
df = pd.read_sql_table('postgres://db')
```

## First look at the data



df.head()

	id	price	neighbourhood_group_cleansed	latitude
0	28684898	\$50.00	Neukölln	52.473978
1	22607348	\$10.00	Treptow - Köpenick	52.468095
2	21019199	\$35.00	Neukölln	52.481810
3	21919556	\$99.00	Pankow	52.537269
4	4820648	\$39.00	Friedrichshain-Kreuzberg	52.491483



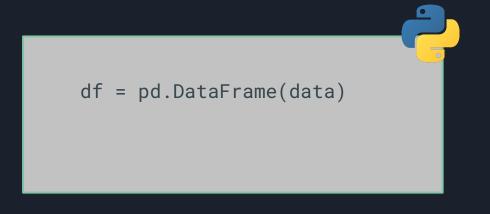
#### df.describe()

	id	price	latitude	longitude	bathrooms
count	1.353100e+04	13531.000000	13531.000000	13531.000000	13508.000000
mean	1.573089e+07	70.082625	52.509956	13.405871	1.095203
std	8.580394e+06	255.451132	0.030773	0.058517	0.335469
min	2.695000e+03	0.000000	52.346203	13.103557	0.000000
25%	8.041528e+06	30.000000	52.489082	13.374950	1.000000
50%	1.697254e+07	45.000000	52.509229	13.416764	1.000000
75%	2.264464e+07	70.000000	52.532808	13.439258	1.000000
max	2.986735e+07	9000.000000	52.651670	13.721671	8.500000

# Access data

#### Access data in a DataFrame

idx	Name	Age
0	Clara	20
1	Tom	24
2	Sarah	19
3	John	21



## Access data in a DataFrame

idx	Name	Age
0	Clara	20
1	Tom	24
2	Sarah	19
3	John	21

```
df = pd.DataFrame(data)
df["Name"]
```

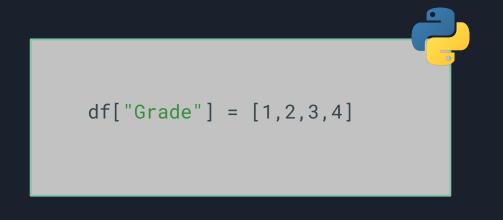
#### Access data in a DataFrame

idx	Name	Age
0	Clara	20
1	Tom	24
2	Sarah	19
3	John	21

```
df = pd.DataFrame(data)
df.iLoc[0:3]
```

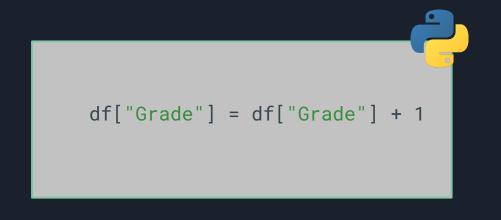
## Add data to a DataFrame

idx	Name	Age	Grade
0	Clara	20	1
1	Tom	24	2
2	Sarah	19	3
3	John	21	4



## Compute with data in a DataFrame

idx	Name	Age	Grade
0	Clara	20	2
1	Tom	24	3
2	Sarah	19	4
3	John	21	5



# Drop data

idx	Name	Age	Grade
0	Clara	20	1/
1	Tom	24	
2	Sarah	19	3
3	John	21	4

```
# Drop column

df = df.drop(column="Grade")
```

## Drop data

idx	Name	Age	Grade
2	Sarah	19	3
3	John	21	4

```
# Drop column

df = df.drop(column="Grade")

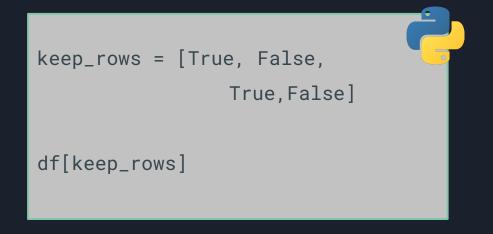
# Drop row

df.drop([0,1], inplace=True)
```

Filter data

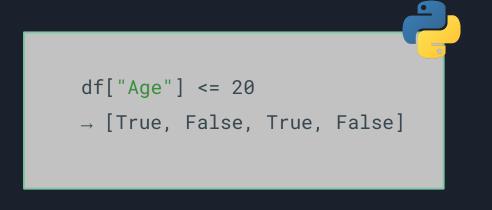
#### Filter data - select a subset of the data

idx	Name	Age	Grade
0	Clara	20	2
2	John	21	5



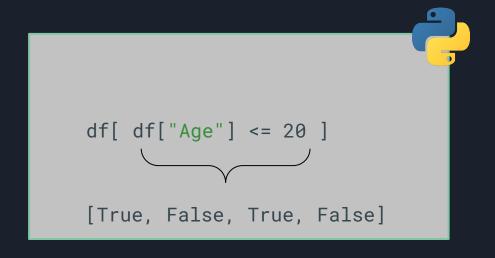
## Filter data - select a subset of the data

idx	Name	Age	Grade
0	Clara	20	2
1	Tom	24	3
2	Sarah	19	4
3	John	21	5



#### Filter data - select a subset of the data

idx	Name	Age	Grade
0	Clara	20	2
2	John	21	5



#### Filter data - select a subset of the data

idx	Name	Age	Grade
0	Clara	20	2

```
df[
    (df["Age"] <= 20) &
    (df["Grade"] < 3)
]</pre>
```

#### Filter data - select a subset of the data

idx	Name	Age	Grade
0	Clara	20	2
2	Sarah	19	4

```
df[
    (df["Age"] == 20) |
    (df["Grade"] == 4 )
]
```

### Exercise 2

```
import pandas as pd

df = pd.DataFrame(data) # create dataframe

df.sort_values(by='column_name') # sort

df[df['age']<=20] # filter data</pre>
```

# Pandas methods

### Apply pandas methods

- Pandas has large amount of commonly used methods
- Can be applied to single column or whole data frame

```
df["Grade"].mean()
df["Grade"].std()
df["Grade"].sum()
```

#### Often used methods

```
df["Grade"].mean()
df["Grade"].sum()
df["Grade"].value_counts()
df.sort_values(by="Grade")
df.groupby(by="Grade").sum()
df1.merge(df2)
df.drop(colums=["Grade"])
df["Grade"].replace(5, "Failed")
```

# Group-By

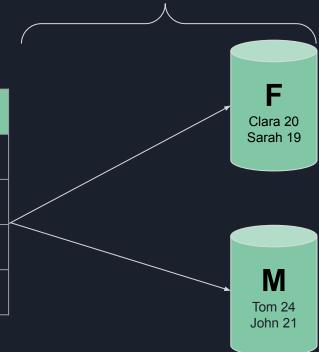
idx	Name	Age	Gender
0	Clara	20	F
1	Tom	24	M
2	Sarah	19	F
3	John	21	M

What is the average age per gender?

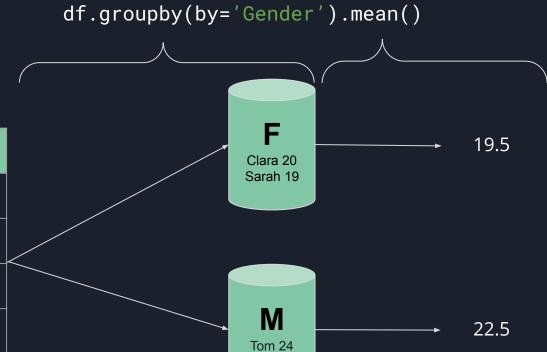
## Group-By

df.groupby(by='Gender').mean()

idx	Name	Age	Gender
0	Clara	20	F
1	Tom	24	М
2	Sarah	19	F
3	John	21	М



## Group-By



John 21

idx	Name	Age	Gender
0	Clara	20	F
1	Tom	24	М
2	Sarah	19	F
3	John	21	М

Name	Age	Grade
Clara	20	2
Tom	24	3
Sarah	19	4

Name	Subject
Sarah	Physics
Tom	Politics
John	English

Name	Age	Grade
Clara	20	2
Tom	24	3
Sarah	19	4

Name	Subject
Sarah	Physics
Tom	Politics
John	English

Name	Age	Grade	Subject
Tom	24	3	Politics
Sarah	19	4	Physics

Name	Age	Grade
Clara	20	2
Tom	24	3
Sarah	19	4

		0
df1.merge(df2,	on='Name')	

Name	Subject
Sarah	Physics
Tom	Politics
John	English

Name	Age	Grade	Subject
Tom	24	3	Politics
Sarah	19	4	Physics

Name	Age	Grade
Clara	20	2
Tom	24	3
Sarah	19	4

df1.merge(df2,how='left',
on='Name')

Name	Subject
Sarah	Physics
Tom	Politics
John	English

Name	Age	Grade	Subject
Tom	24	3	Politics
Sarah	19	4	Physics
Clara	20	2	-

### Exercise 3

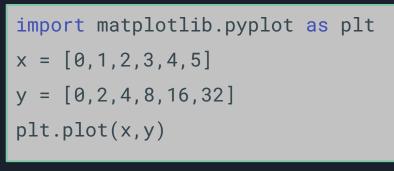
```
import pandas as pd

df = df.groupby(by='column_name'
df = df.merge(df2,on='column_name')
```

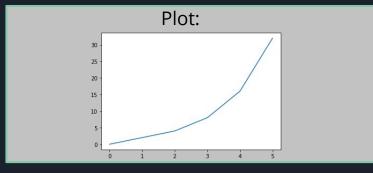
# Visualizations

# matp - library

- data visualization tool
- generate highly customizable plots
- good integrated with pandas

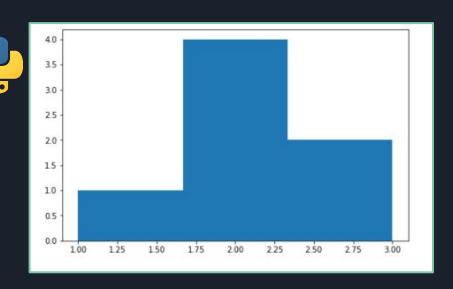






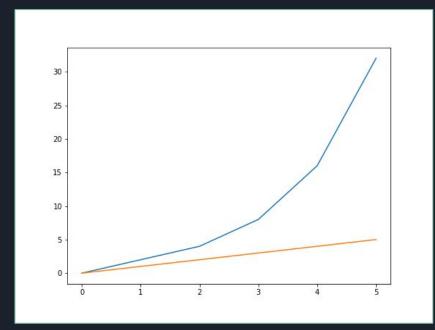
### Histogram

import matplotlib.pyplot as plt
x = [1,2,2,2,2,3,3]
plt.hist(x, bins=3)



```
import matplotlib.pyplot as plt
x = [0,1,2,3,4,5]
y = [0,2,4,8,16,32]
y2 = [0,1,2,3,4,5]

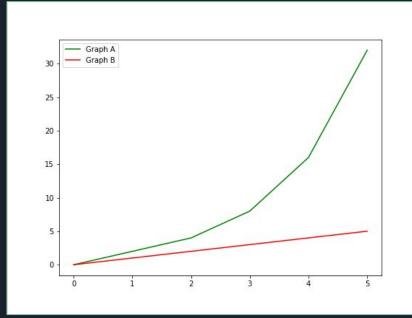
plt.plot(x,y)
plt.plot(x,y2)
```



```
import matplotlib.pyplot as plt
x = [0,1,2,3,4,5]
y = [0,2,4,8,16,32]
y2 = [0,1,2,3,4,5]

plt.plot(x,y, color='green', label='Graph A')
plt.plot(x,y2, color='red', label='Graph B')

plt.legend()
```



```
import matplotlib.pyplot as plt

x = [0,1,2,3,4,5]

y = [0,2,4,8,16,32]

y2 = [0,1,2,3,4,5]

plt.plot(x,y, color='green', label='Graph A')

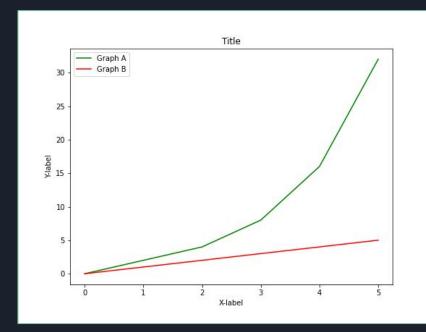
plt.plot(x,y2, color='red', label='Graph B')

plt.legend()

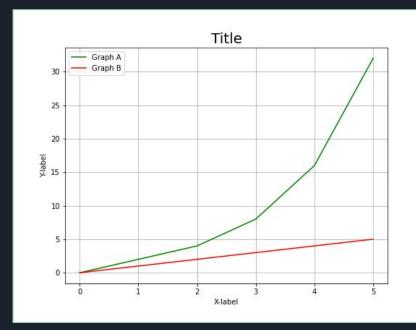
plt.xlabel("X-label")

plt.ylabel("Y-label")

plt.title("Title")
```



```
import matplotlib.pyplot as plt
x = [0, 1, 2, 3, 4, 5]
y = [0, 2, 4, 8, 16, 32]
y2 = [0,1,2,3,4,5]
plt.plot(x,y, color='green', label='Graph A')
plt.plot(x,y2, color='red', label='Graph B')
plt.legend()
plt.xlabel("X-label")
plt.ylabel("Y-label")
plt.title("Title",
{'fontname':'DejaVu Sans', 'size':'20'})
plt.grid()
```



### pandas & matplotlib

- Pandas and matplotlib work very well together
- We can pass columns of a DataFrame to matplotlib

```
import matplotlib.pyplot as plt
df = pd.DataFrame(data)
plt.hist(df["Age"])
```

### Exercise 4

```
import matplotlib.pyplot as plt
x = [0,1,2,3,4,5]
y = [0,2,4,8,16,32]

plt.plot(x,y, color='green', label='Graph A')

plt.legend()
plt.xlabel("X-label")
plt.ylabel("Y-label")
plt.title("Title")
```