Introduction to Data Science with Python

Chapter 2

December 6, 2022

Topics of the course

1. Chapter

2. Chapter

3. Chapter

Python Fundamentals

Basic concepts, Variables, basic data structures, functions

Data Wrangling & Simple visualizations

How to process data with pandas and visualize it with matplotlib

Visualizations & Modelling

More plots with matplotlib and seaborn and an introduction to modelling

What you learn in this Chapter...

... how to use Python libraries

... work with data in a DataFrame

... filter, merge and group your data

... visualize data with simple plots

Data Wrangling & Visualization Libraries

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,...)

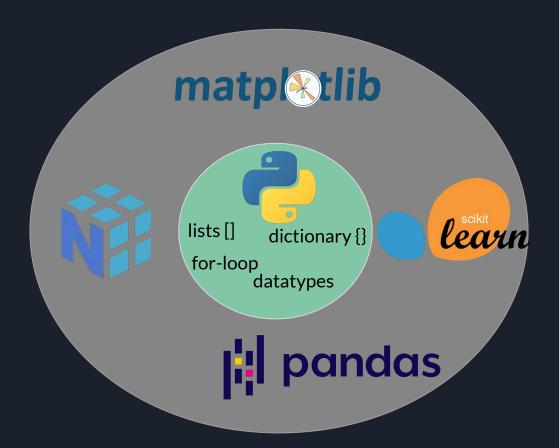


... creating machine learning models

Libraries



Libraries



Function & Methods

Functions

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

Function & Methods

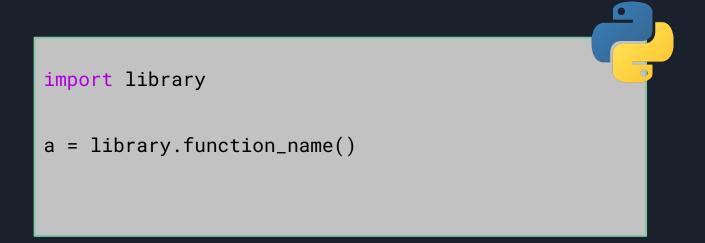
Functions

Methods

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

```
a = "a string!"
a = a.upper() # 'A STRING!'
```

Import Libraries



Import Libraries

```
import library as 1
a = 1.function_name()
```

Data Wrangling & Visualization Arrays with Numpy

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



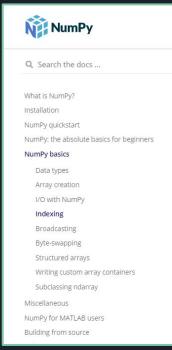
- Library for scientific computing
- Work with lists, matrices or higher dimensional structures
- NumPy lists have much more functionality than usual lists

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



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])# [1,2,3,4,5,6,7]
a[a<=2] # [1,2] (boolean indexing)
a.sum() # 28</pre>
```

Data Wrangling & Visualization **Tabular Data with Pandas**

pandas - library

```
import pandas as pd

data = {
```

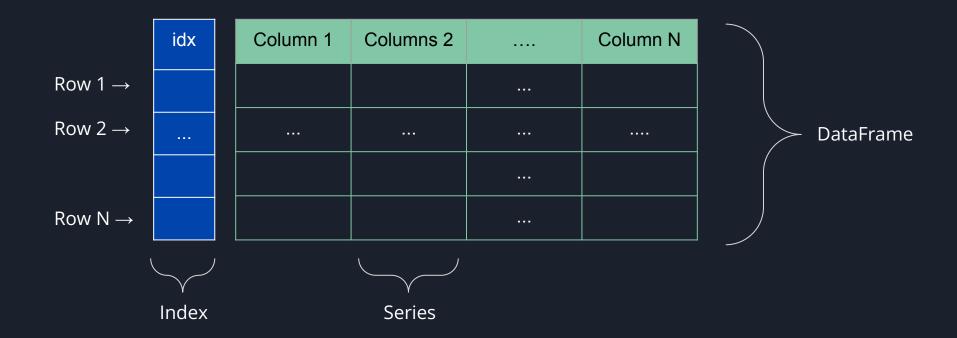
```
"Name":["Clara", "Tom", "Sarah", "John"],

"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



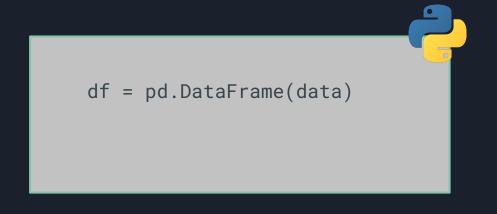
df.describe()

	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

Select Data

Select data in a DataFrame

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



Select 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"]
```

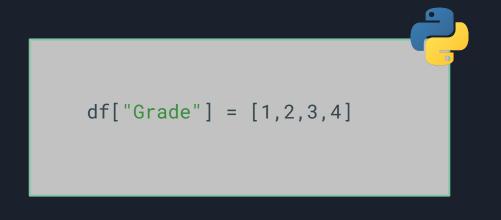
Select data in a DataFrame

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

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

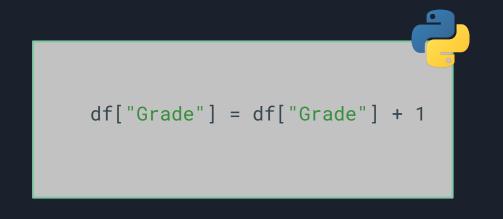
Add data to a DataFrame

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

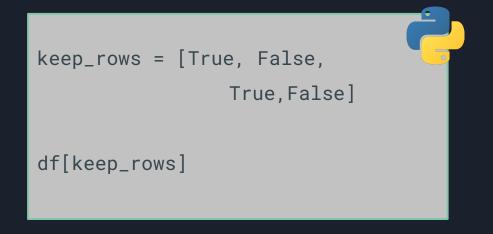
df = df.drop(column="Grade")

# Drop row

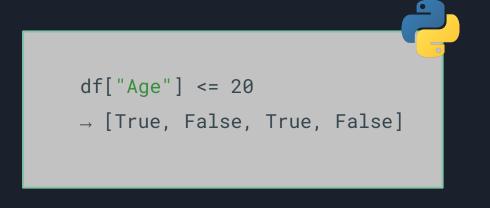
df.drop([0,1], inplace=True)
```

Filter data

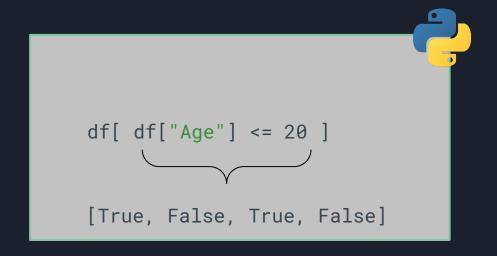
idx	Name	Age	Grade
0	Clara	20	2
2	John	21	5



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



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



idx	Name	Age	Grade
0	Clara	20	2

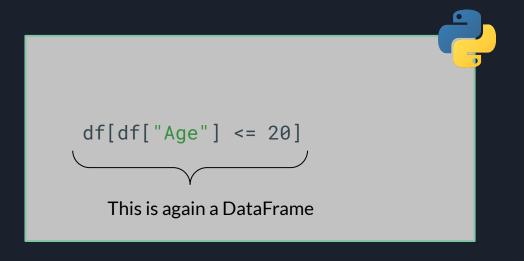
```
df[
    (df["Age"] <= 20) &
    (df["Grade"] < 3)
]</pre>
```

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

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

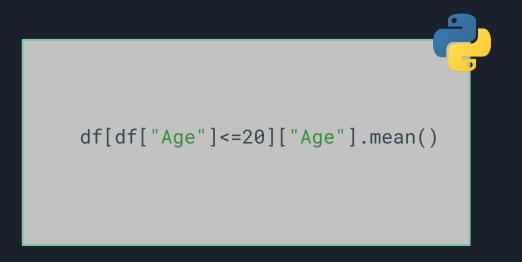
Combine filtering with other methods

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



Combine filtering with other methods

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



Exercise 2

```
import pandas as pd

df = pd.DataFrame(data) # create dataframe

df['new_column'] = [1,2,3]

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

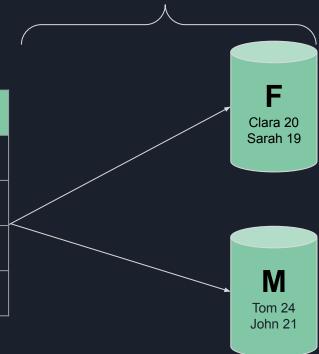
idx	Name	Age	Gender
0	Clara	20	F
1	Tom	24	М
2	Sarah	19	F
3	John	21	M

What is the average age per gender?

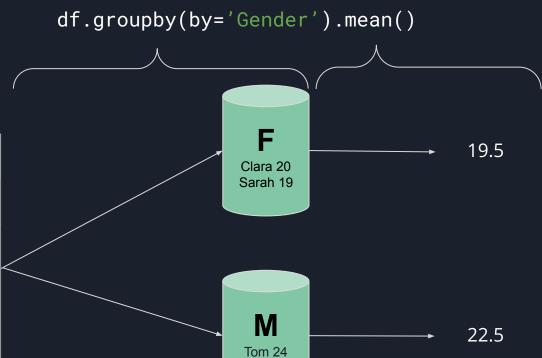
Group-By

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

Name	Age	Gender
Clara	20	F
Tom	24	М
Sarah	19	F
John	21	М
	Clara Tom Sarah	Clara 20 Tom 24 Sarah 19



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

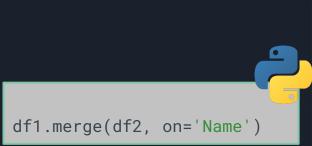
Name	Subject	
Sarah	Physics	
Tom	Politics	
John	English	

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

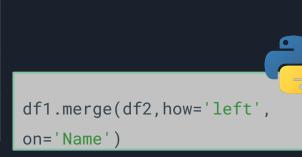
Name	Age	Grade
Clara	20	2
Tom	24	3
Sarah	19	4



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



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')

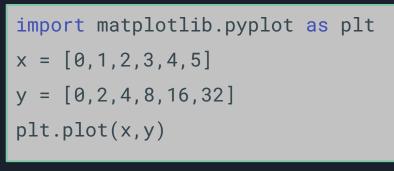
df = df.merge(df2,on='column_name',how='left')
```

Data Wrangling & Visualization

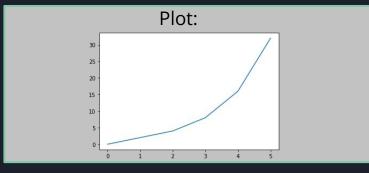
Visualizations with Matplotlib

matp (*) - library

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

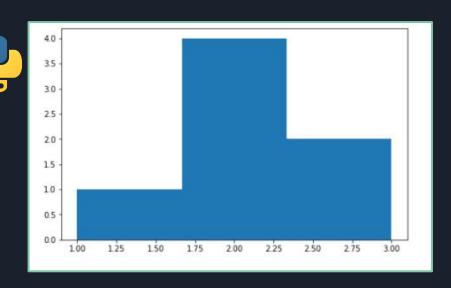






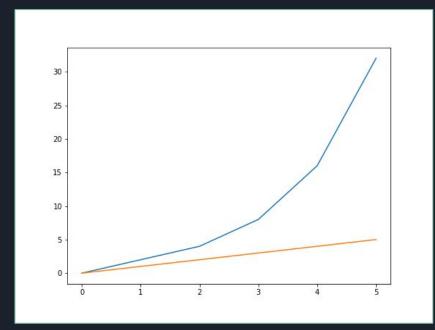
Histogram

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]

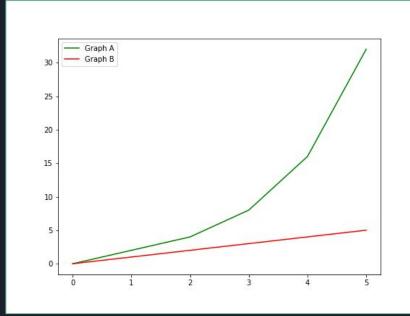
plt.plot(x,y)
plt.plot(x,y2)
```



```
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')

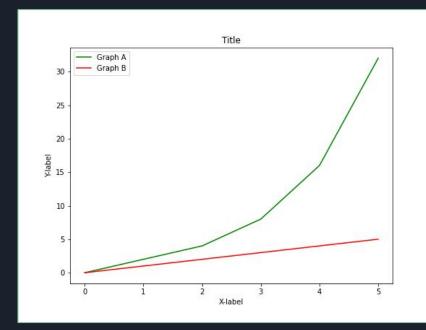
plt.legend()
```



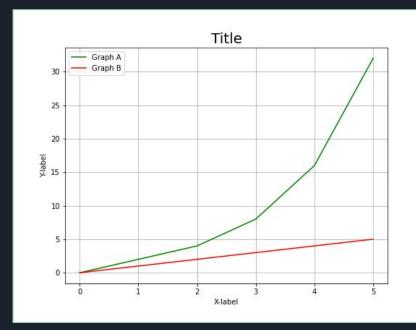
```
import matplotlib.pyplot as plt
x = [0,1,2,3,4,5]
y = [0,2,4,8,16,32]
y2 = [0,1,2,3,4,5]

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
df = pd.DataFrame(data)
plt.hist(df["Age"])
```