PYTHON / PANDAS FOR DATA ANALYSIS





Wat we gaan doen de komende 2 dagen

• Introductie:

- o waar werk je nu,
- o met welke software,
- o gebruik je nu Python,
- o en: wat wil je vandaag leren
- Highover uitleg Python, Pandas en belangrijkste Python data libraries
- Intro Jupyter Notebook + Python refresher + opstarten VM
- Data inlezen
 ---> pd.read csv()
- Data inspectie ---> df.info() df.head() df.describe()
- Data selectie ---> df[df.column == 'value'] df.loc[df.column == 'value', :]
- Data wrangling ---> df['column'].fillna() of df.drop_duplicates()
 - Data joinen ---> df.merge(df2)
- Data visualisatie ---> px.scatter(df, x, y)
- Eigen functies toepassen ---> df.apply(my_own_function)
- Installeren Jupyter + Python + Pandas ---> pip install pandas

Introductie

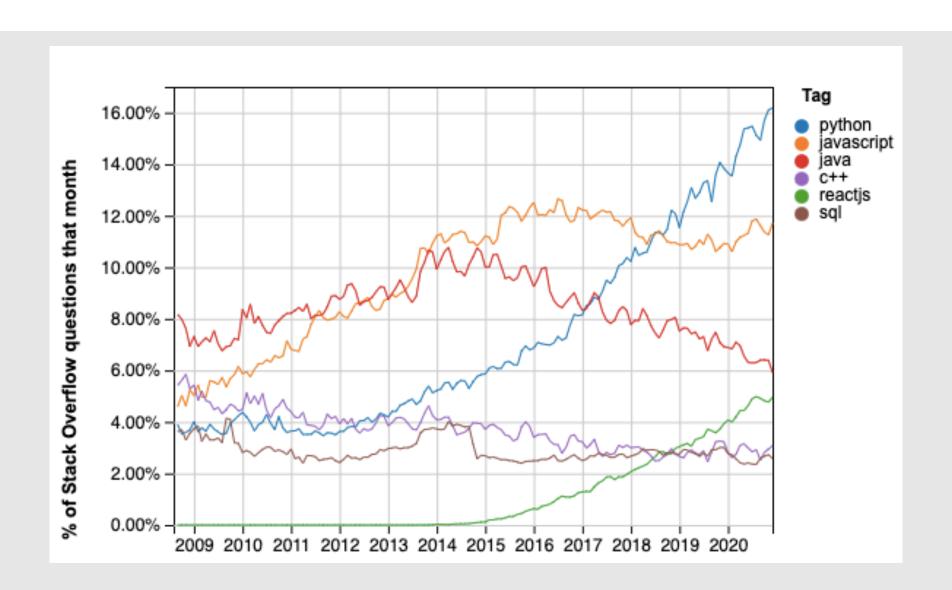
• Introductie:

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Highover uitleg

- Highover uitleg:
 - o Python
 - Pandas
 - o en belangrijkste Python data libraries

Python is eating the world



Why is it handy to know some python / pandas?

- veel gebruikt voor data analysis en machine learning
- steeds vaker ingezet voor data engineering: pandas en python zijn zeer flexibel voor data wrangling
- als je de pandas API enigszins kent, dan helpt dat bijv. ook weer met databricks / pyspark
- steeds groter Python data ecosystem, bijv. scheduling met Airflow
- makkelijk om met python API's aan te spreken
- websites scrapen
- zeer leesbare code (relatief)
- en hopelijk: leuk om wat nieuws te leren

Main disadvantages of python / pandas

 pandas doet alles in-memory, dus voor gedistribueerde oplossingen heb je bijv. Spark of Dask nodig

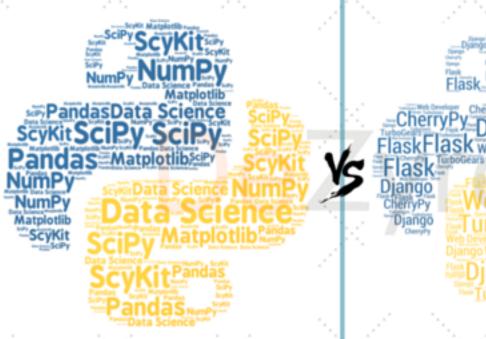
 higher-level language: gebouwd bovenop C. Omdat het higher-level is, is het ook langzamer

niet geschikt voor mobile development

Twee wegen die naar Python leiden

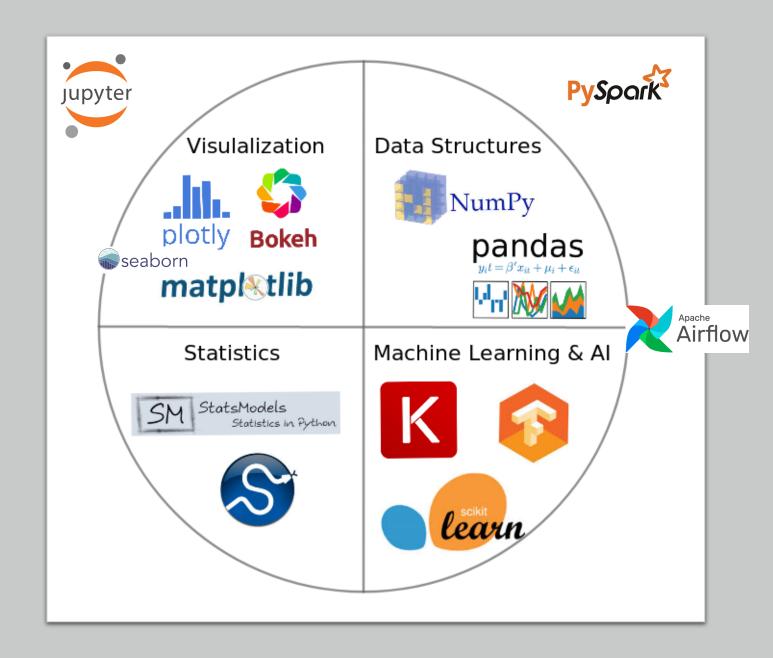
Data Science / Data Engineering

Backend Web Development





Python Data Science Libraries



Jupyter Notebooks + Python

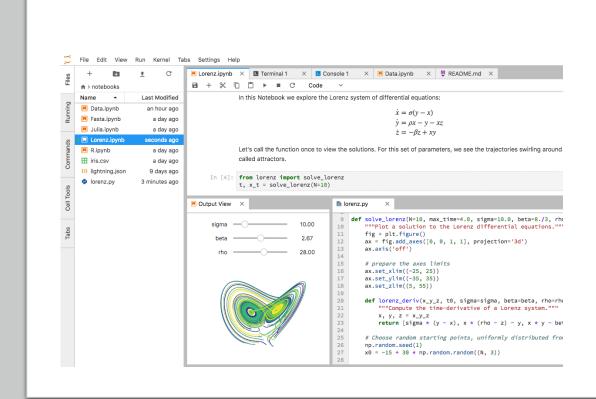
- Intro Jupyter Notebook
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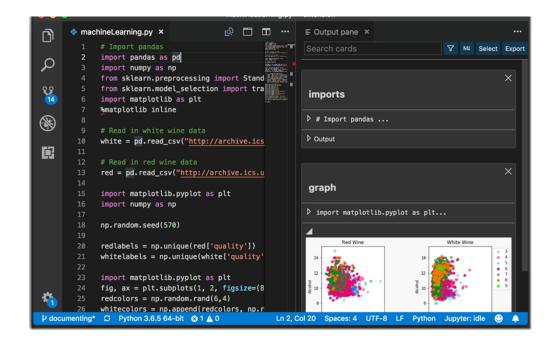
Python data types

Name	Туре	Description	
Integers	int	Whole numbers, such as: 3 300 200	
Floating point	float	Numbers with a decimal point: 2.3 4.6 100.0	
Strings	str	Ordered sequence of characters: "hello" 'Sammy' "2000" "楽しい"	
Lists	list	Ordered sequence of objects: [10,"hello",200.3]	
Dictionaries	dict	Unordered Key:Value pairs: {"mykey": "value", "name": "Frankie"}	
Tuples	tup	Ordered immutable sequence of objects: (10,"hello",200.3)	
Sets	set	Unordered collection of unique objects: {"a","b"}	
Booleans	bool	Logical value indicating True or False	

Jupyter Notebooks

or Visual Studio Code





Jupyter Notebooks tips and tricks

- Shift + Enter to run code
- Tab completion
- Nieuwe cell: Escape gevolgd door a (above) of b (below) of dd (delete)
- Shift Tab to see arguments and information about methods, functions or classes
- Magic commands, such as Is
- ? or ?? to get extra help and info

Pandas Introductie

```
 Data inlezen ---> pd.read_csv()
 Data inspectie ---> df.info() df.head() df.describe()
 Data selectie ---> df[df.column == 'value'] df.loc[df.column == 'value', :]
 Data wrangling ---> df['column'].fillna() df.drop_duplicates()
 Data joinen ---> df.merge(df2, how='inner', on='column_name')
 Data visualisatie ---> px.scatter(df, x, y)
```

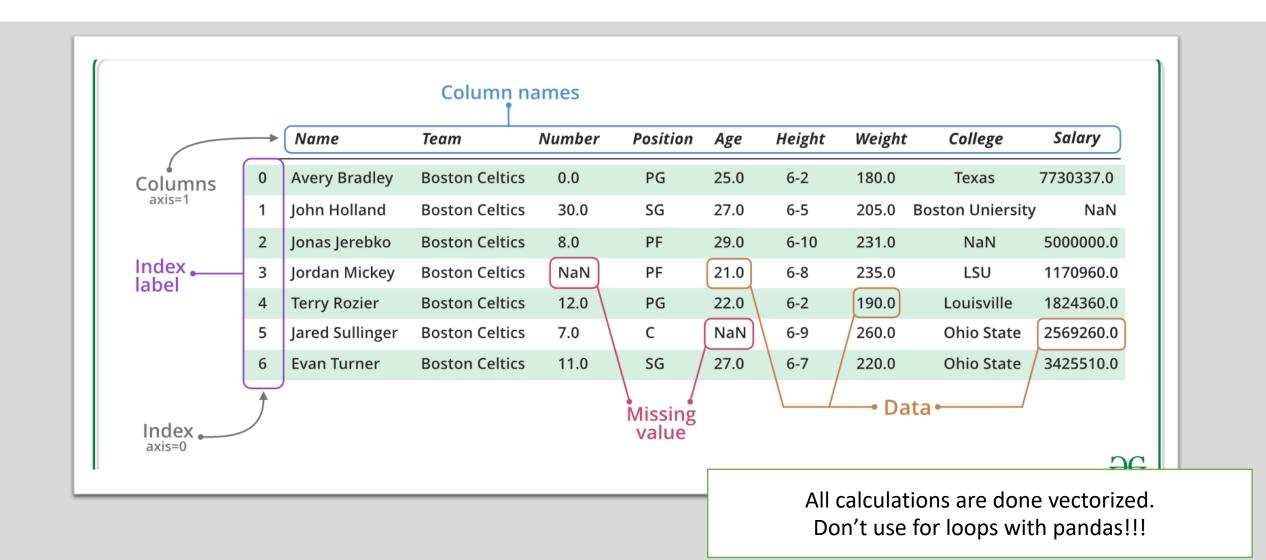
Wat is Pandas?



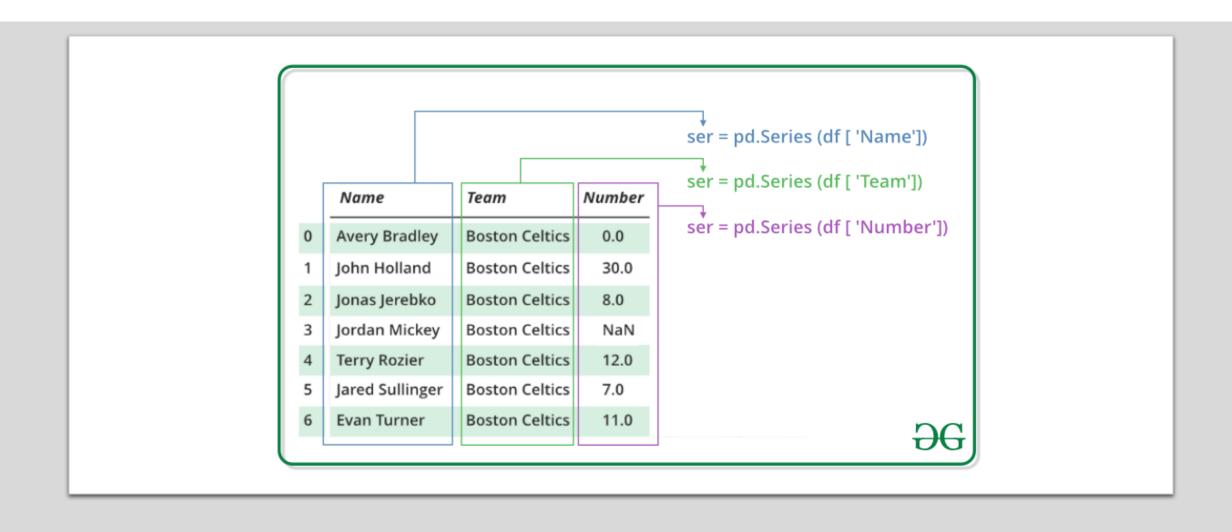
pandas

pandas is a fast, powerful, flexible and easy to use open source data analysis and manipulation tool, built on top of the Python programming language.

Most important concept is a DataFrame



Example of a Pandas Series



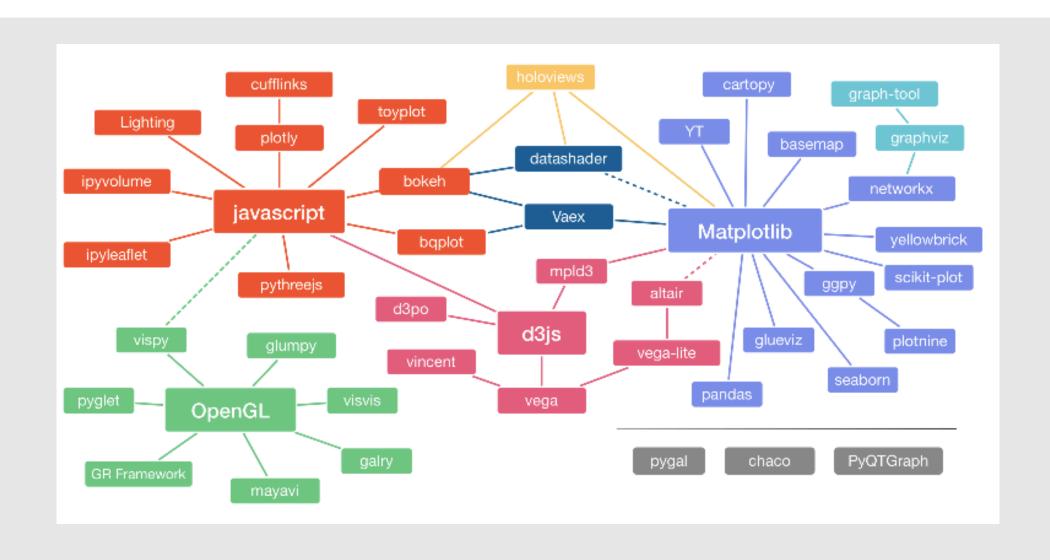
Plotting

• Data visualisatie

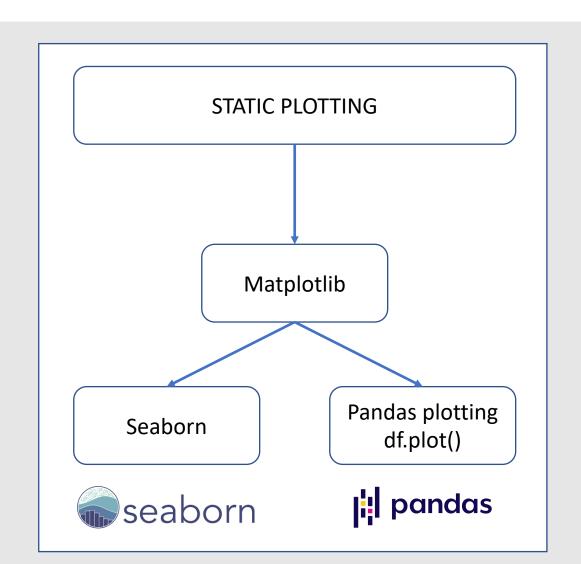
---> px.scatter(df, x, y)

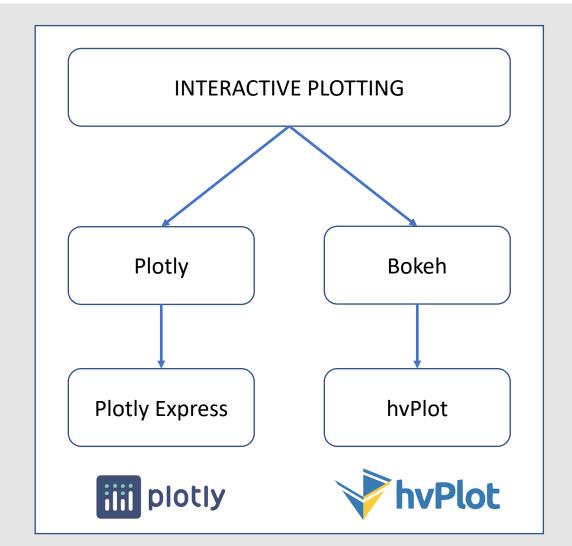
---> sns.scatterplot(df, x, y)

There are too many plotting packages



So my advice is: only focus on these



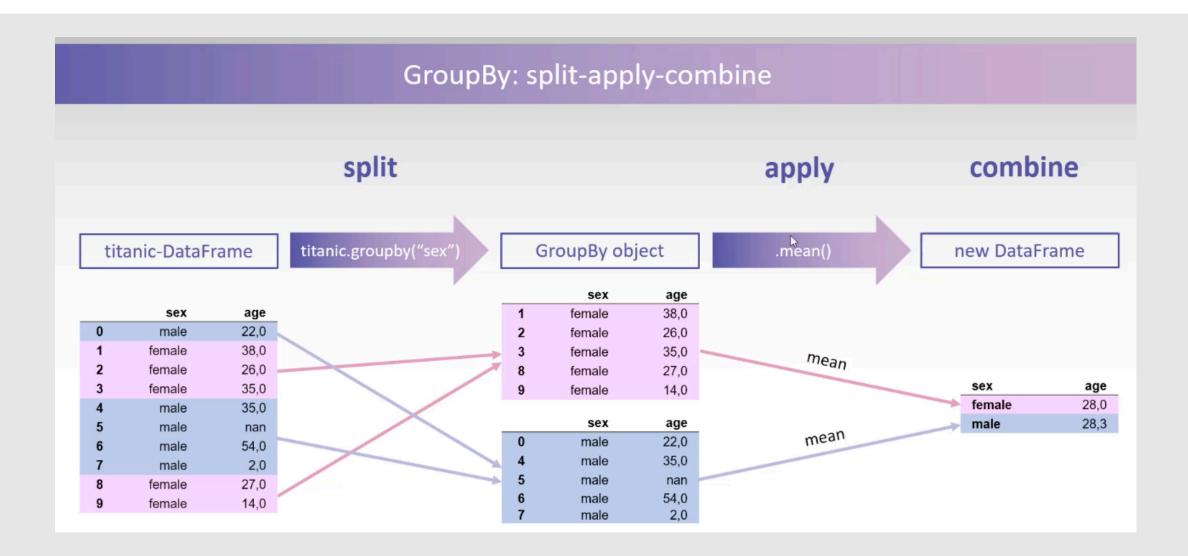


Data aggregation

Using groupby to aggregate:

---> df.groupby(['startYear'])[['averageRating']].mean()

How groupby() works

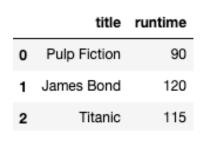


Joining tables

• Using df.merge() to join tables:

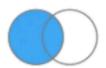
---> df.merge(df2, how='inner', on='mutual_column')

Joining tables with df.merge()



titlename0 Pulp FictionJohn Travolta1 Pulp FictionSamuel L. Jackson2 James BondSean Connery3 TerminatorArnold Schwarzenegger

left join



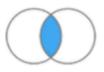
	title	runtime	name
0	Pulp Fiction	90	John Travolta
1	Pulp Fiction	90	Samuel L. Jackson
2	James Bond	120	Sean Connery
3	Titanic	115	NaN

right join



name	runtime	title	
John Travolta	90.000	Pulp Fiction	0
Samuel L. Jackson	90.000	Pulp Fiction	1
Sean Connery	120.000	James Bond	2
Arnold Schwarzenegger	NaN	Terminator	3

inner join



name	runtime	title	
John Travolta	90	Pulp Fiction	0
Samuel L. Jackson	90	Pulp Fiction	1
Sean Connery	120	James Bond	2

outer join



name	runtime	title	
John Travolta	90.000	Pulp Fiction	0
Samuel L. Jackson	90.000	Pulp Fiction	1
Sean Connery	120.000	James Bond	2
NaN	115.000	Titanic	3
Arnold Schwarzenegger	NaN	Terminator	4