Tutorial III

Python Programming with Jupyter Notebooks

Google Colab &

Python for Data Science



Google Colab

 Google is quite aggressive in Al research. Over many years, Google developed Al framework called TensorFlow and a development tool called Colaboratory.

- **TensorFlow** is an *open-sourced* end-to-end platform that makes it easy for you to build and deploy Machine Learning models.
- It is an *entire ecosystem* to help you solve challenging, real-world problems with machine learning.

Google Colab

- Colaboratory is now known as Google Colab or simply Colab.
- The introduction of Colab has eased the learning and development of machine learning applications.
- Another feature that Google offers to the developers is the use of GPU (graphical processing unit). Colab supports GPU and is free.
- The reasons for making it free for public was to make its software a standard in the academics for teaching machine learning and data science.



Google Colab

- Colab is a free Jupyter notebook environment that runs entirely in the cloud.
- Most importantly, it does not require a setup and the notebooks that you create can be simultaneously edited by your team members - just the way you edit documents in Google Docs.
- Colab *supports many popular machine learning libraries* which can be easily loaded in your notebook.



What Colab Offers You?

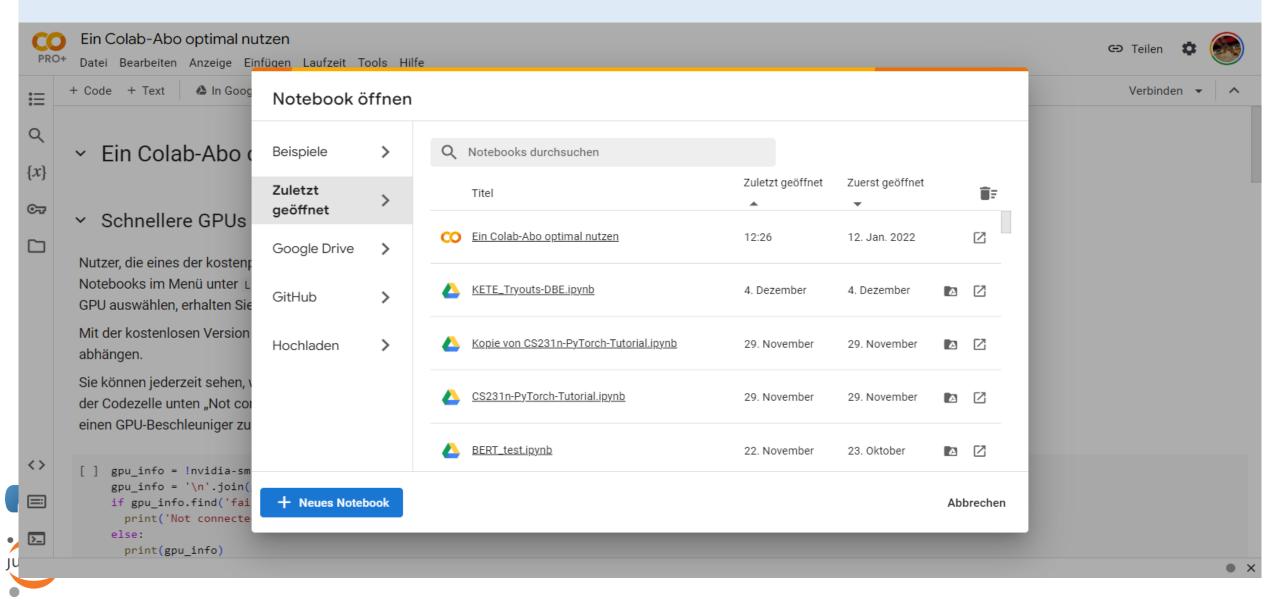
- As a programmer, you can perform the following using Google Colab
 - Write and execute code in Python
 - Document your code that supports mathematical equations
 - –Create/Upload/Share notebooks
 - -Import/Save notebooks from/to Google Drive
 - -Import/Publish notebooks from GitHub
 - -Import external datasets e.g. from Kaggle
 - —Integrate Machine Learning Packages like PyTorch, TensorFlow, Keras, OpenCV …
 - Free Cloud service with free GPU

How Google Colab Works?

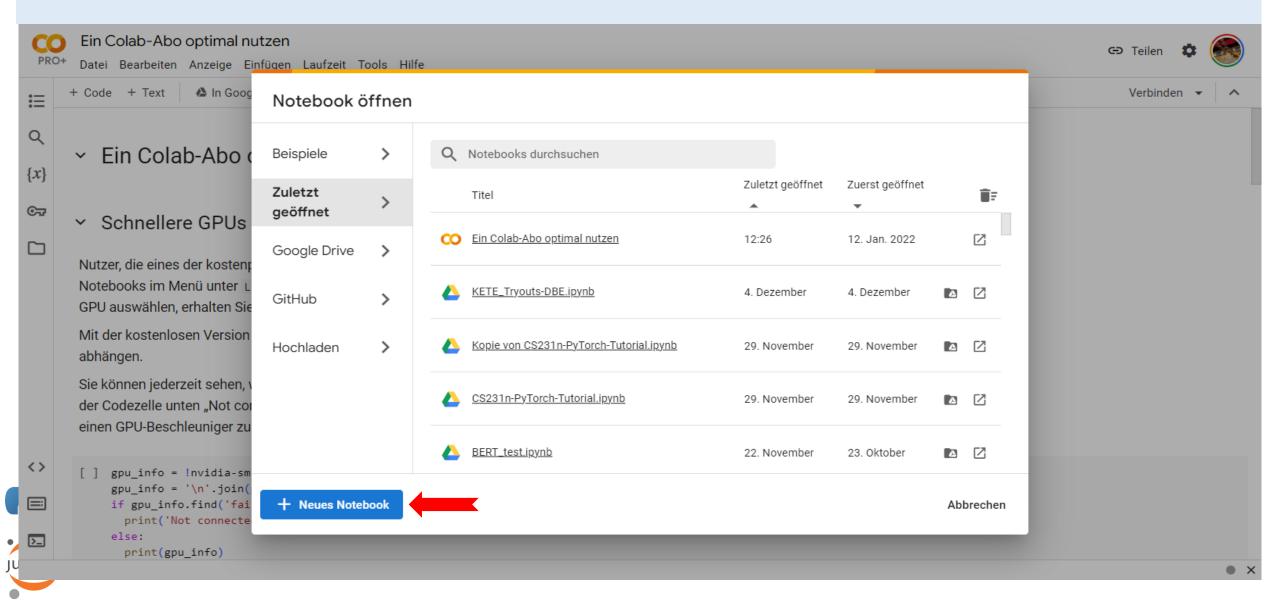
As Colab **implicitly uses Google Drive** for storing your notebooks, ensure that you are logged in to your **Google** Drive **account** before proceeding further.



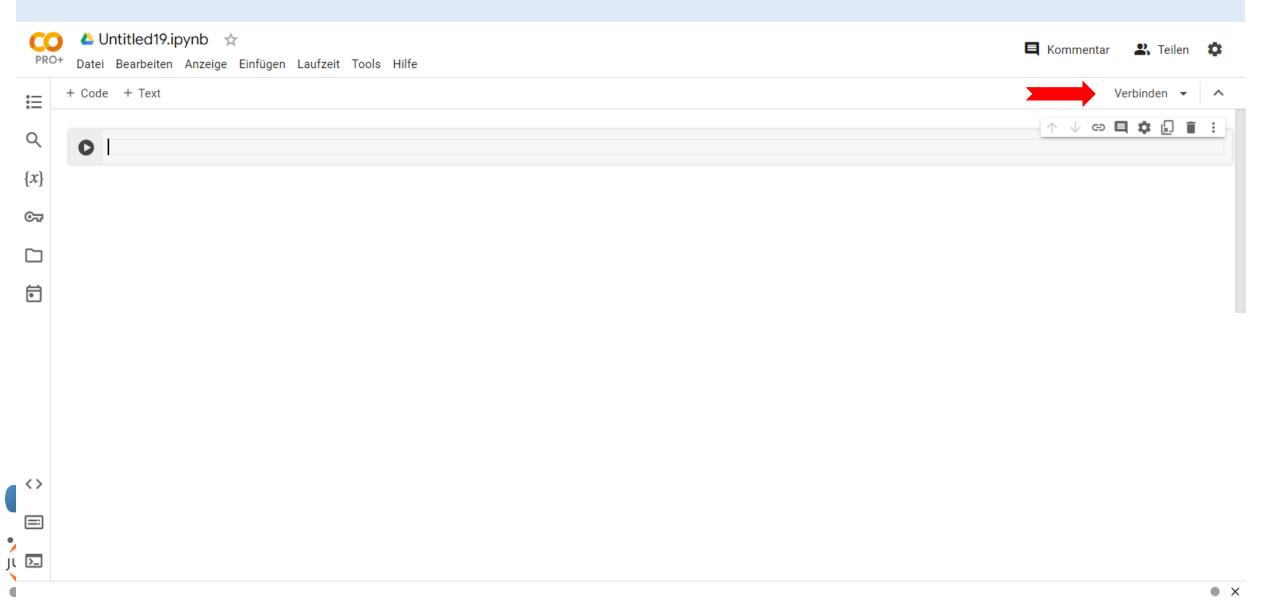
Step 0: Open the following URL in your browser: https://colab.research.google.com



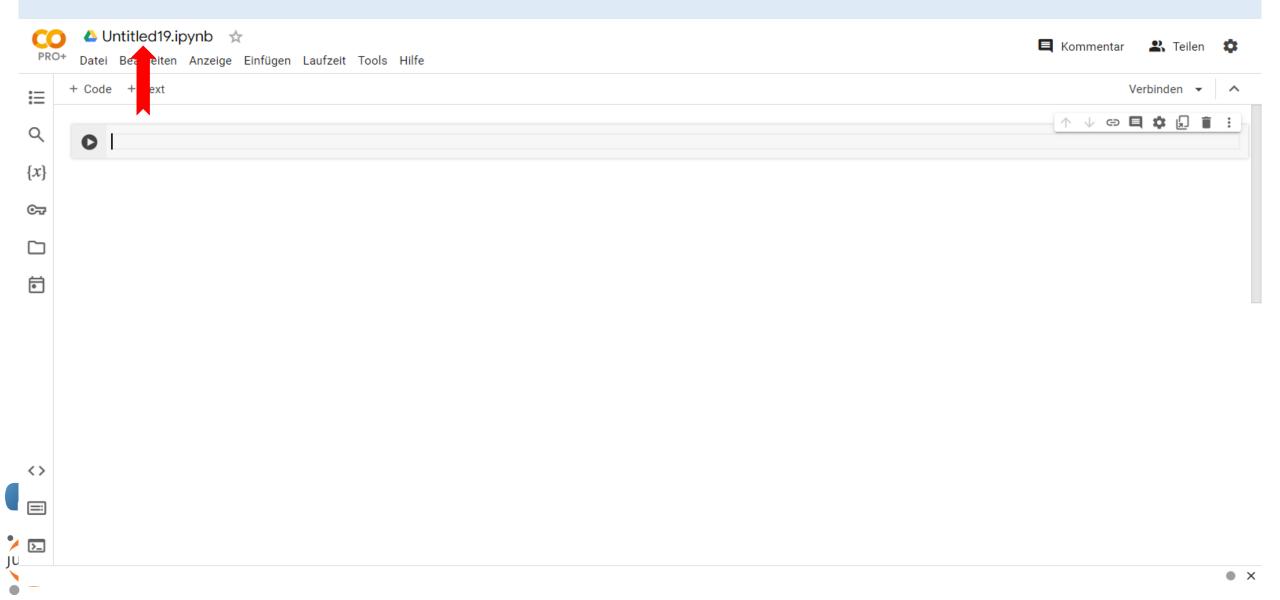
Step 1: Click on the **NEW NOTEBOOK** link at the bottom left of the opened window



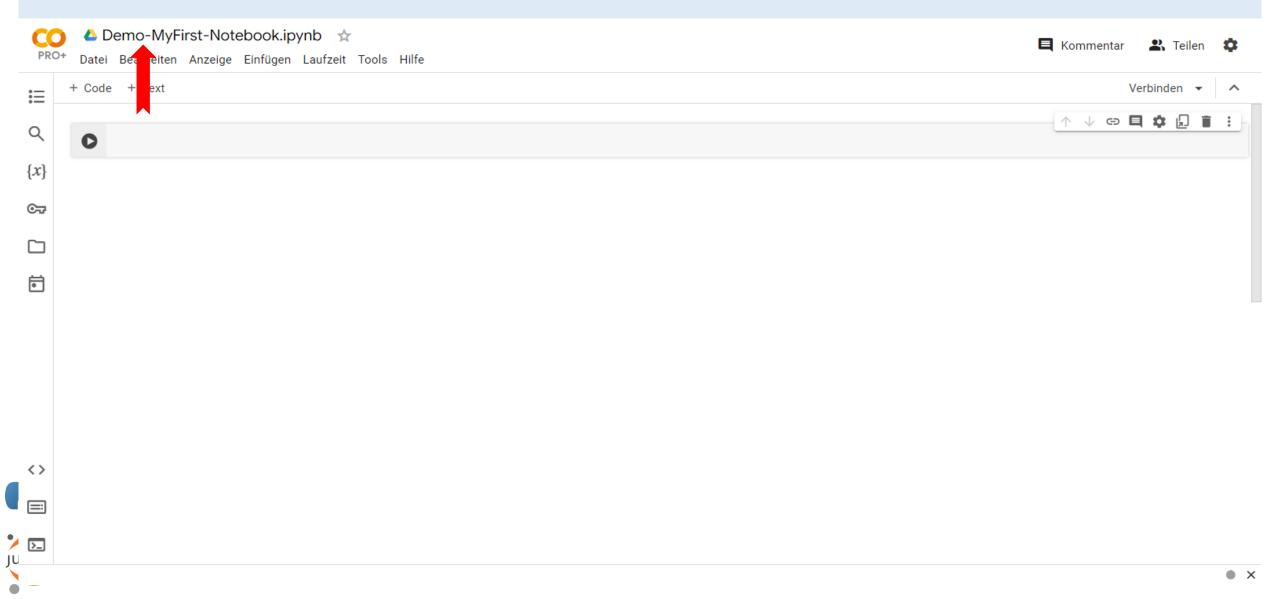
Step 3: Click on the connect at the top (right) of the screen



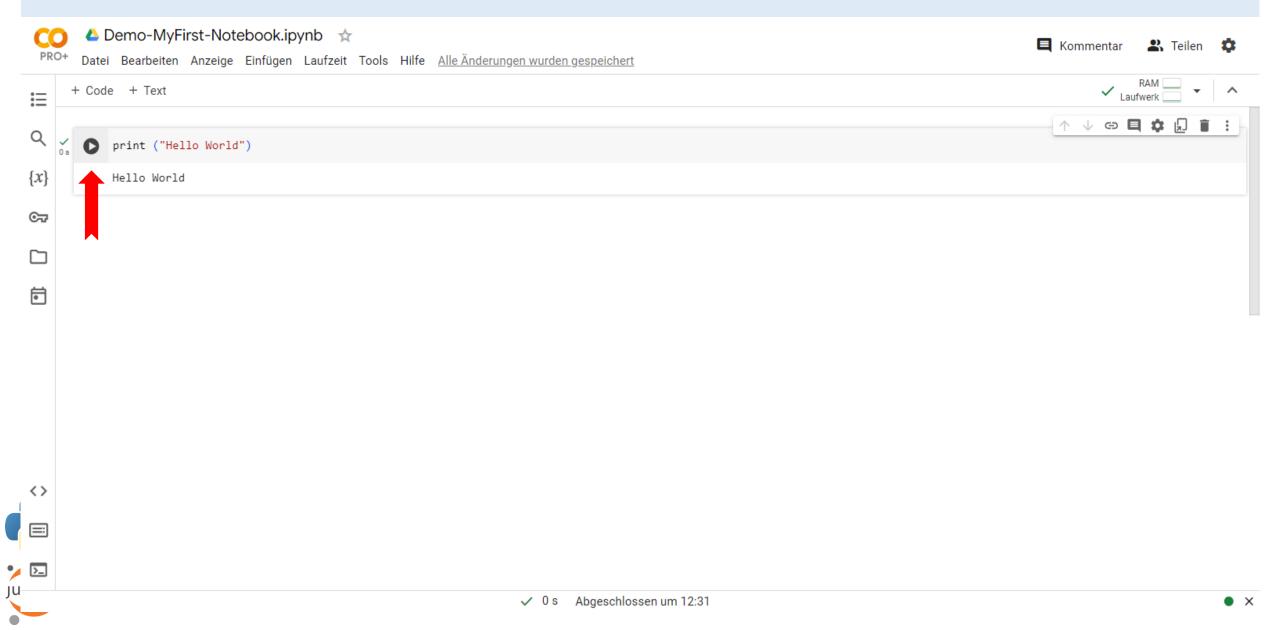
Step 4a: Click on the untitledxx.ipynb at the top (left) of the screen



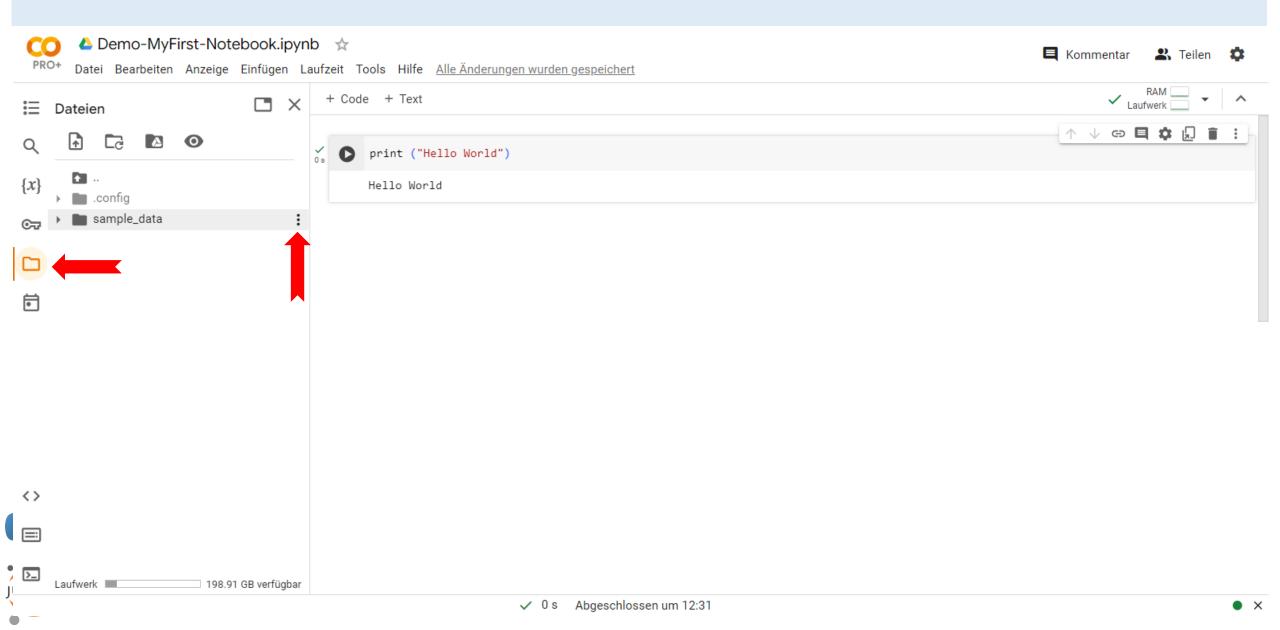
Step 4b: Replace untitledxx with an appropriate notebook name



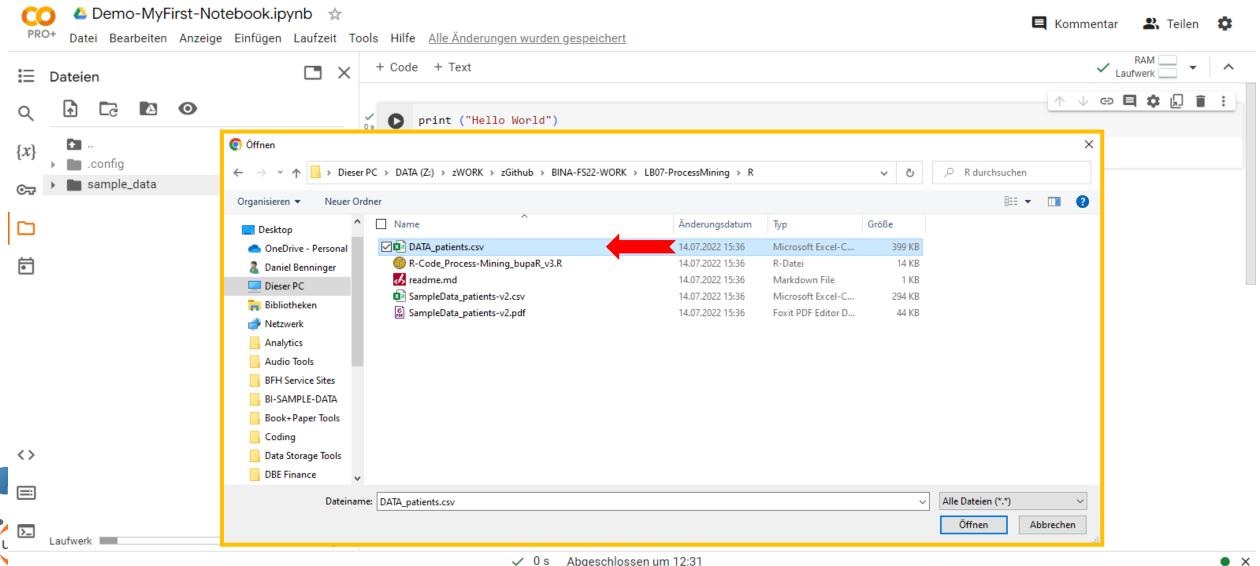
Step 5: Write and execute a single program statement



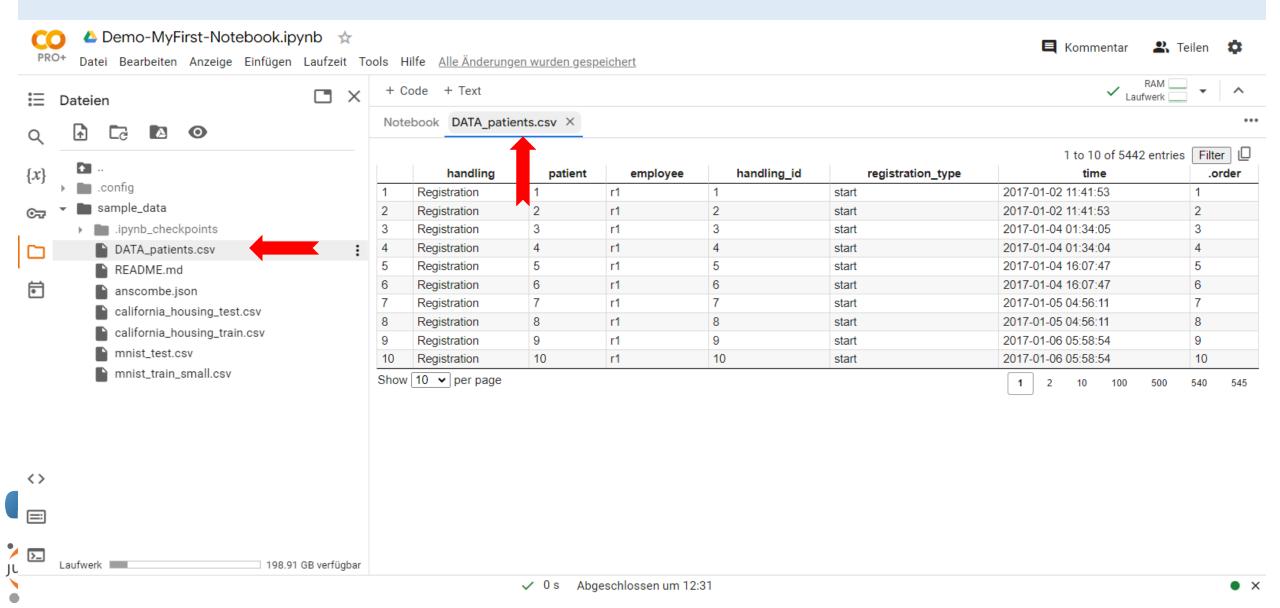
Step 6: Click on box at left side to import (upload) files



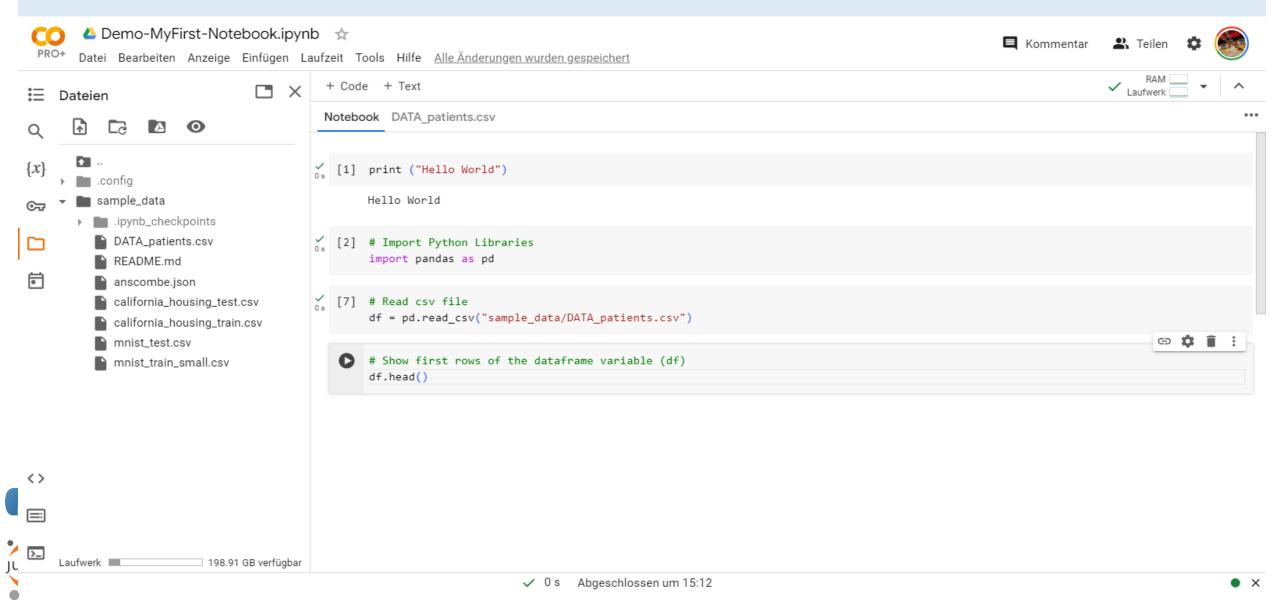
Step 7a: Select (local) file to upload



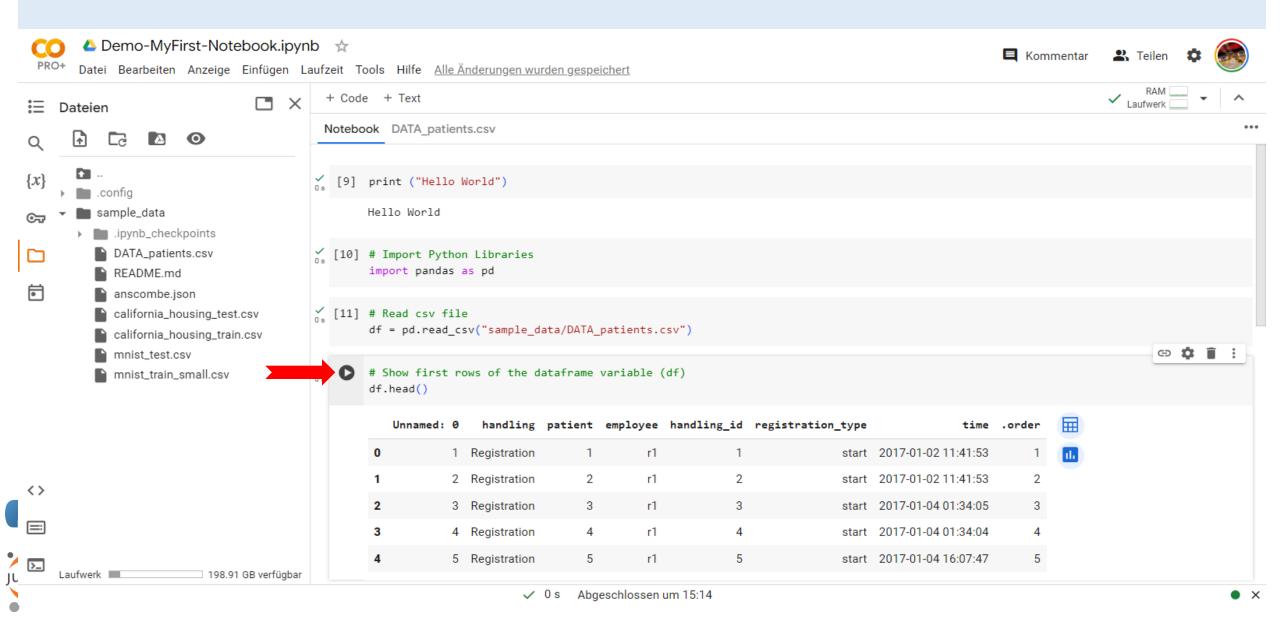
Step 7b: Show/Inspect uploaded file



Step 8a: Write program to load/read data from a csv file



Step 8b: Execute program to load/read data from a csv file



What is Python?

- **Python** is an *interpreted*, high level and general-purpose programming language.
- **Python** is a *general-purpose* coding language—which means that, unlike HTML, CSS, and JavaScript, it can be used for other types of programming and software development besides web development.



Why Python?

- Python works on **different platforms** (Windows, Mac, Linux, Raspberry Pi, etc).
- Python has a simple syntax similar to the English language with influence from mathematics.
- Python has **syntax** that allows developers to write programs with fewer lines than some other programming languages.
- Python runs on an interpreter system, meaning that code can be executed as soon as it is written. Thus prototyping can be very quick.
- Python can be treated in a procedural way, an object-oriented way or a functional way.



What can Python do?

- Python can be used on a server to create web applications.
- Python can be used alongside software to create workflows.
- Python can connect to database systems. It can also read and modify files.
- Python can be used to handle big data and perform complex mathematics.
- Python can be used for rapid prototyping, or for production-ready software development.



Multi Words Variable Names

Variable names with more than one word can be difficult to read.
 There are several techniques you can use to make them more readable:

Pascal Case

Each word starts with a capital letter:
 e.g. MyVariableName = "John"

Camel Case

Each word, except the first, starts with a capital letter:
 e.g. myVariableName = "John"

Snake Case

Each word is separated by an underscore character:
 e.g. my_variable_name = "John"



Data Type in Python

Built-in Data Types

- In programming, data type is an important concept.
- Variables can store data of different types, and different types can do different things.
- Python has the following data types built-in by default, in these categories:

Text Type:	str
Numeric Types:	int, float, complex
Sequence Types:	list, tuple, range
Mapping Type:	dict
Set Types:	set, frozenset
Boolean Type:	bool
Binary Types:	bytes, bytearray, memoryview



Data Type in Python

You can get the data type of any object by using the type() function

```
x = 5
print(type(x))
```



Setting the (implicit) Data Type

 In Python, the data type is set when you assign a value to a variable:

Example	Data Type
x = "Hello World"	str
x = 20	int
x = 20.5	float
x = 1j	complex
x = ["apple", "banana", "cherry"]	list
x = ("apple", "banana", "cherry")	tuple
x = range(6)	range
$x = {"name" : "John", "age" : 36}$	dict
x = {"apple", "banana", "cherry"}	set
<pre>x = frozenset({"apple", "banana", "cherry"})</pre>	frozenset
x = True	bool
x = b"Hello"	bytes
x = bytearray(5)	bytearray
<pre>x = memoryview(bytes(5))</pre>	memoryview



Setting the Specific Data Type

If you want to specify the data type, you can use the following constructor functions:

Example	Data Type
<pre>x = str("Hello World")</pre>	str
x = int(20)	int
x = float(20.5)	float
<pre>x = complex(1j)</pre>	complex
<pre>x = list(("apple", "banana", "cherry"))</pre>	list
<pre>x = tuple(("apple", "banana", "cherry"))</pre>	tuple
x = range(6)	range
<pre>x = dict(name="John", age=36)</pre>	dict
<pre>x = set(("apple", "banana", "cherry"))</pre>	set
<pre>x = frozenset(("apple", "banana", "cherry"))</pre>	frozenset
x = bool(5)	bool
x = bytes(5)	bytes
x = bytearray(5)	bytearray
x = memoryview(bytes(5))	memoryview



Some popular Python toolboxes/libraries:

- Data Processing and Modeling
 - NumPy
 - SciPy
 - Pandas
 - SciKit-Learn
 - TensorFlow
 - Keras
- Data Mining
 - scrapy

Visualization libraries

- matplotlib
- Seaborn
- plotly
- pydot



..... and many more

Python Libraries for Data Science Loading Python Libraries

```
In []: #Import Python Libraries
import numpy as np
import scipy as sp
import pandas as pd
import matplotlib as mpl
import seaborn as sns
```

Press Shift+Enter to execute the jupyter cell





NumPy

- NumPy (Numerical Python) is a perfect tool for scientific computing and performing basic and advanced array operations.
- introduces objects for multidimensional arrays and matrices, as well as functions that allow to easily perform advanced mathematical and statistical operations on those objects
- provides vectorization of mathematical operations on arrays and matrices which significantly improves the performance
- many other python libraries are built on NumPy



Link: http://www.numpy.org/









Pandas

- Pandas is a library created to help developers work with "labeled" and "relational" data.
- adds data structures and tools designed to work with table-like data (similar to Series and Data Frames in R)
- provides tools for data manipulation: reshaping, merging, sorting, slicing, aggregation etc.
- allows handling missing data



Link: http://pandas.pydata.org/

Python Libraries for Data Science Reading data using pandas

```
In [ ]: #Read csv file
    df = pd.read_csv("Salaries.csv")
```

Note: The data import process.

There is a number of pandas commands to read other data formats:

```
pd.read_excel('myfile.xlsx',sheet_name='Sheet1', index_col=None, na_values=['NA'])
pd.read_stata('myfile.dta')
pd.read_sas('myfile.sas7bdat')
pd.read_hdf('myfile.h5','df')
```





matplotlib

- A standard data science library that helps to *generate data* visualizations such as two-dimensional diagrams and graphs (histograms, line plots, pie charts, scatterplots).
- Matplotlib is one of those plotting libraries that are very useful in data science projects it provides an object-oriented API for embedding plots into applications.
- relatively low-level; some effort needed to create advanced visualization



Link: https://matplotlib.org/



Seaborn

- Seaborn is *based on Matplotlib* and serves as a useful Python machine learning *tool for visualizing statistical models* heatmaps and other types of visualizations that summarize data and depict the overall distributions.
- When using this library, you get to benefit from an extensive gallery of visualizations (including complex ones like time series, joint plots, and violin diagrams).
- provides high level interface for drawing attractive statistical graphics

Link: https://seaborn.pydata.org/





Plotly

- A web-based tool for data visualization that offers many useful out-of-box graphics – you can find them on the plotly webseite
- The library works very well in interactive web applications.
- Its creators are busy expanding the library with new graphics and features for supporting multiple linked views, animation, and crosstalk integration.



Link: https://plotly.com/



Scrapy

- One of the most popular Python data science libraries, Scrapy helps to *build crawling programs* (spider bots) that can retrieve structured data from the web for example, URLs or contact info.
- It's a great tool for scraping data used in, for example, Python machine learning models.
- Developers use it for gathering data from APIs. This full-fledged framework follows the Don't Repeat Yourself principle in the design of its interface. As a result, the tool inspires users to write universal code that can be reused for building and scaling large crawlers.



Link: https://scrapy.org/



SciPy:

- A large collection of algorithms for linear algebra, differential equations, numerical integration, optimization, statistics and more
- part of SciPy Stack
- built on NumPy



Link: https://www.scipy.org/scipylib/



SciKit-Learn:

- This is an industry-standard for data science projects based in Python.
- Scikits is a group of packages in the SciPy Stack that were created for specific functionalities for example, image processing. Scikit-learn uses the math operations of SciPy to expose a concise interface to the most common machine learning algorithms.
- Data scientists use it for handling *standard machine learning and data mining tasks* such as classification, clustering, regression, model selection, model validation, dimensionality reduction, and classification.
- built on NumPy, SciPy and matplotlib



Link: http://scikit-learn.org/



TensorFlow

- TensorFlow is a popular Python framework for machine learning and deep learning, which was developed at Google Brain.
- It's the best tool for tasks like object identification, speech recognition, and many others. It helps in working with artificial neural networks that need to handle multiple data sets.
- The library includes various layer-helpers (tflearn, tf-slim, skflow), which make it even more functional.
- TensorFlow is constantly expanded with its new releases including fixes in potential security vulnerabilities or improvements in the integration of TensorFlow and GPU.





Keras

- Keras is a great library for building neural networks and modeling.
- It's very straightforward to use and provides developers with a good degree of extensibility.
- The library takes advantage of other packages, (Theano or TensorFlow) as its backends. Moreover, Microsoft integrated CNTK (Microsoft Cognitive Toolkit) to serve as another backend.
- It's a *great pick if you want to experiment quickly* using compact systems the minimalist approach to design really pays off!

