

Practical Work 01 – 24th February 2022

Let's get started

Objectives. The main objective of this Practical Work (PW) is to set up your computer environment and to bring you on track for the **Deep Learning** class. We are going to use Python as it offers many good libraries to handle data and to run machine learning tasks. This weeks' PW will also include tutorials and exercises to get familiar with this language. Finally, an introduction to Pytorch tensors will be given. We will also work with iPython Notebooks which is also a good format to hand in your PW. You can create, edit and run these notebooks from a web browser and write Python code or text (using Markdown) in so-called cells and also create nice output (e.g. tables, plots).

Moodle. Register on moodle at <https://moodle.msengineering.ch>. Navigate to *Home* → *Zürich* → *Technical scientific specialization modules* → *SPR22 TSM - Spring Semester 22* → *TSM_DeLearn* and use subscription key : `moodlemsekey`.

Hand-In of PW. Most of the time, you'll need to submit your PW reports by Wednesday, 10am of the following week. However, we may override this rule. In any cases, the dates indicated on Moodle for each PW are the ones you should follow. For this first PW, we expect you to submit a solution for exercise 4 about Pytorch tensor manipulation. The other exercises are optional. Submission is in the form of a *iPython Notebook* (follow the instructions in the iPython Notebook).

Python Installation on Your Computer

Skip this part if you are already set up with Python, Jupyter notebooks and your favorite IDE for Python. Otherwise, read the following options for you to install the tools.

Jupyter notebooks with manual installation in virtual environments

This is our **recommended** way to go. First install Python 3.8 from <https://www.python.org>. Versions 3.7 and 3.9 should also work for the class but be aware that 3.9 is still in debug mode. To be noted : the *virtual env* installation as explained here allows you to easily switch from one version to another by installing multiple *virtual env*.

Create a working directory and open a terminal in that directory :

- a) Install the virtual env : `python3.8 -m venv venv`
- b) Activate the virtual env on Mac : `source venv/bin/activate`
- c) Activate the virtual env on Windows : `C:\> <venv>\Scripts\activate.bat`
- d) Install Jupyter : `pip install jupyter`
- e) Install needed libs : `pip install numpy matplotlib pandas scikit-learn imageio`
- f) Install tensorflow : `pip install tensorflow`
- g) Install pytorch : `pip install torch torchvision`
- h) Launch Jupyter : `venv/bin/jupyter notebook`

Whenever you need to re-launch the notebooks, repeat steps b) and h). If you need to install other libraries, use `pip install LIB_NAME` once the virtual env is setup with step b).

Anaconda platform

Anaconda is a popular distribution of data science packages. Download and install from the [Anaconda web site](#). See the graphical installer on Figure 1 below.

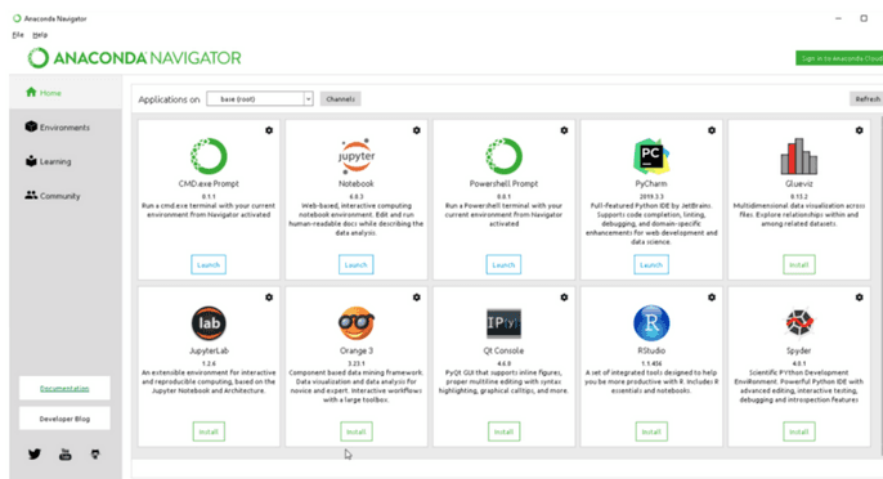


FIGURE 1 – Installing the Anaconda platform

Other options

You may run iPython notebooks directly from some integrated development environment (IDE).

- PyCharm - an IDE for Python from [JetBrains](#). You may select Professional or Community edition - the Community edition should be enough for this class. Free licenses are usually given to students, use your `SCHOOL.ch` address for the registration by JetBrains.
- DataSpell - an IDE for Data Science supporting Python and notebooks, also from [JetBrains](#). Same remark as above for licenses.

- Visual Studio code - a general purpose IDE that support Python and notebooks from [Microsoft](#).

If you do not want to install anything on your computer, you may use Google Colab at <https://colab.research.google.com/>.

Exercise 1 OPTIONAL – Python Language in a Nutshell

We assume here that students are knowledgeable in other programming languages such as Java or C and that basic data structure concepts are known. If you know already Python and the concept of notebooks, then you can skip this exercise.

- Open Jupyter in your working directory launching `venv/bin/jupyter notebook` in the terminal or from the Anaconda Navigator. Jupyter Python notebooks run in your browser so, after launching Jupyter, a browser should show up.
- Open a new notebook from menu File and follow the User Interface Tour as illustrated in Figure 2.

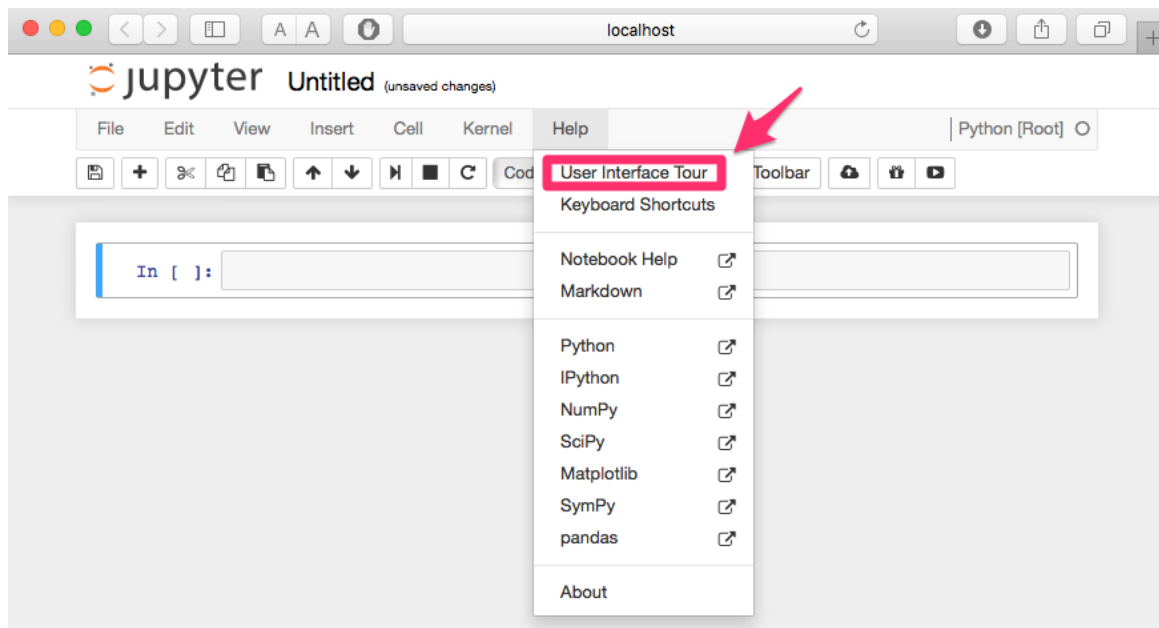


FIGURE 2 – First steps with Python notebooks.

- Download the file `intro-python-3.ipynb` from moodle and open it from Jupyter in Anaconda. You need to navigate where you stored the ipynb file. See Figure 3 below.
- Go through the content of the `intro-python-3.ipynb` notebook and play with the cells. This document should give you a quick introduction to Python assuming that you are fluent with other programming languages. You may also want to get familiar with [Markdown syntax](#) if not known already.
- If you want a more fully fledged introduction to Python, read the official tutorial from <https://docs.python.org/3.8/tutorial/>.

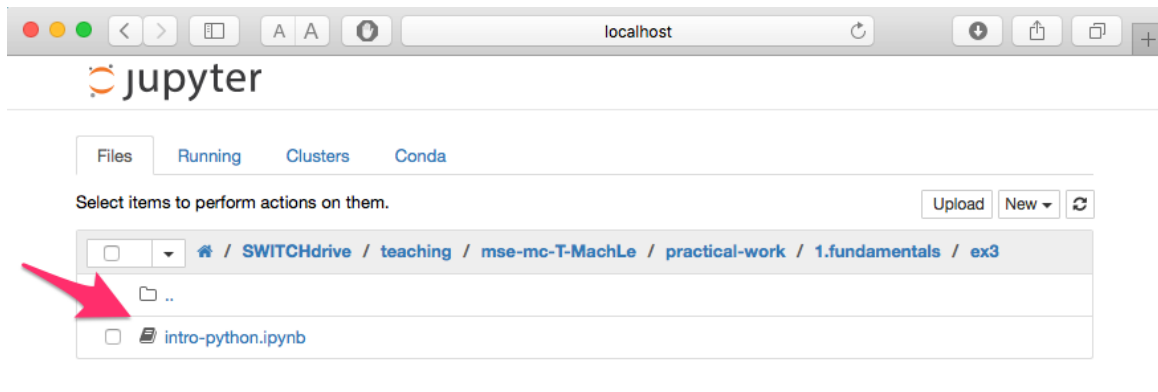


FIGURE 3 – Intro to Python language and Python notebooks.

Exercise 2 Numpy in a Nutshell

This exercise is to become more familiar with `numpy`. Read the content of the Jupyter notebook `numpy-tutorial-stud.ipynb` that you will find on Moodle. Pay special attention to the *broadcasting* section that allows to gain significant speedup when processing large numpy arrays. Regarding this, it is usually more efficient to use *broadcasting* instead of for loops in your code.

At the end of the tutorial, you have to complete some manipulations of images stored by arrays to obtain a result as in 4.

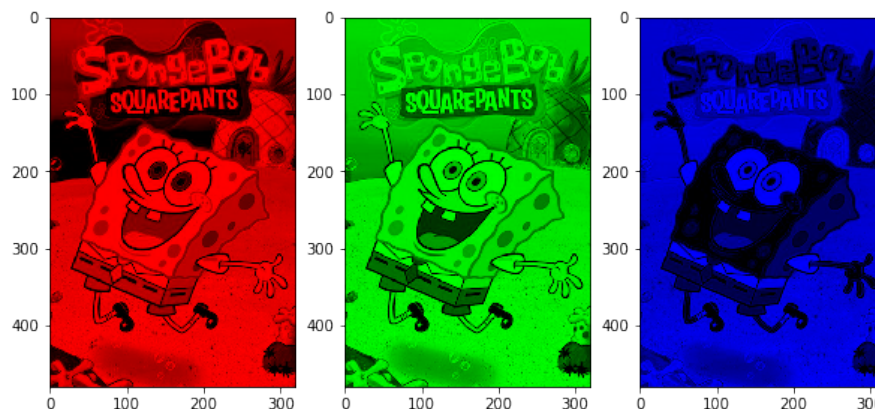


FIGURE 4 – Tri-color planes of Sponge Bob image.

Exercise 3 First hands on Pytorch tensors

- Read and listen to F. Fleuret class 1.4 *Tensor basics and linear regression* and class 1.5 *High dimension tensors* on <https://fleuret.org/dlc/>.
- Read the Pytorch tutorial on https://pytorch.org/tutorials/beginner/basics/tensorqs_tutorial.html and get familiar with basic tensor manipulations.

Exercise 4 Tensors – Linear regression

Start from the notebook `linear-regression-stud.ipynb` available on Moodle. We will use a small data set of apartment rent prices. The objective is to build a linear model to predict the rent price from the surface (m2) of the apartment. By linear model we mean here the equation of a line.

We are then in the case of a monovariable linear regression where we want to discover the parameters θ of a linear model defined with

$$h_{\theta}(\mathbf{x}) = \theta_0 + \theta_1 x \quad (1)$$

The “best” θ ’s are the ones that will minimise the squared difference between the gotten outputs and the target outputs defined with

$$J(\theta) = \frac{1}{2N} \sum_{n=1}^N (h_{\theta}(\mathbf{x}_n) - y_n)^2 \quad (2)$$

A closed form solution to this problem is given by the following normal equation :

$$\theta = (X^T X)^{-1} X^T \vec{y} \quad (3)$$

With X and \vec{y} defined as

$$X = \begin{pmatrix} 1 & x_{1,1} & \dots & x_{1,D} \\ 1 & & \ddots & \\ 1 & \vdots & & x_{n,d} & \vdots \\ 1 & & & \ddots & \\ 1 & x_{N,1} & \dots & x_{N,D} \end{pmatrix} \quad (4)$$

$$y = \begin{pmatrix} y_1 \\ \vdots \\ y_N \end{pmatrix} \quad (5)$$

- Following the instructions in the notebook, implement Equation 3 assuming that x is the living area and y is the renting price. Use `numpy` for the vector operations. Plot the computed line on top of the scatter plot.
- Compute the overall cost value according to Equation 2.

From the numpy solution, provide a solution to the problem using pytorch.

Exercise 5 OPTIONAL – Data Visualization

To train a bit yourself, we propose you to do a visualization workout with the Iris data set https://en.wikipedia.org/wiki/Iris_flower_data_set. Download the file `iris.txt` from moodle or from <http://www.statlab.uni-heidelberg.de/data/iris/>. Put it in a Python data structure and attempt to reproduce a plot close to the one in Figure 5. Advice : start by working with individual plots and then move to a grid of plots with `subplot` or take the shortcut of using the library `pandas`.

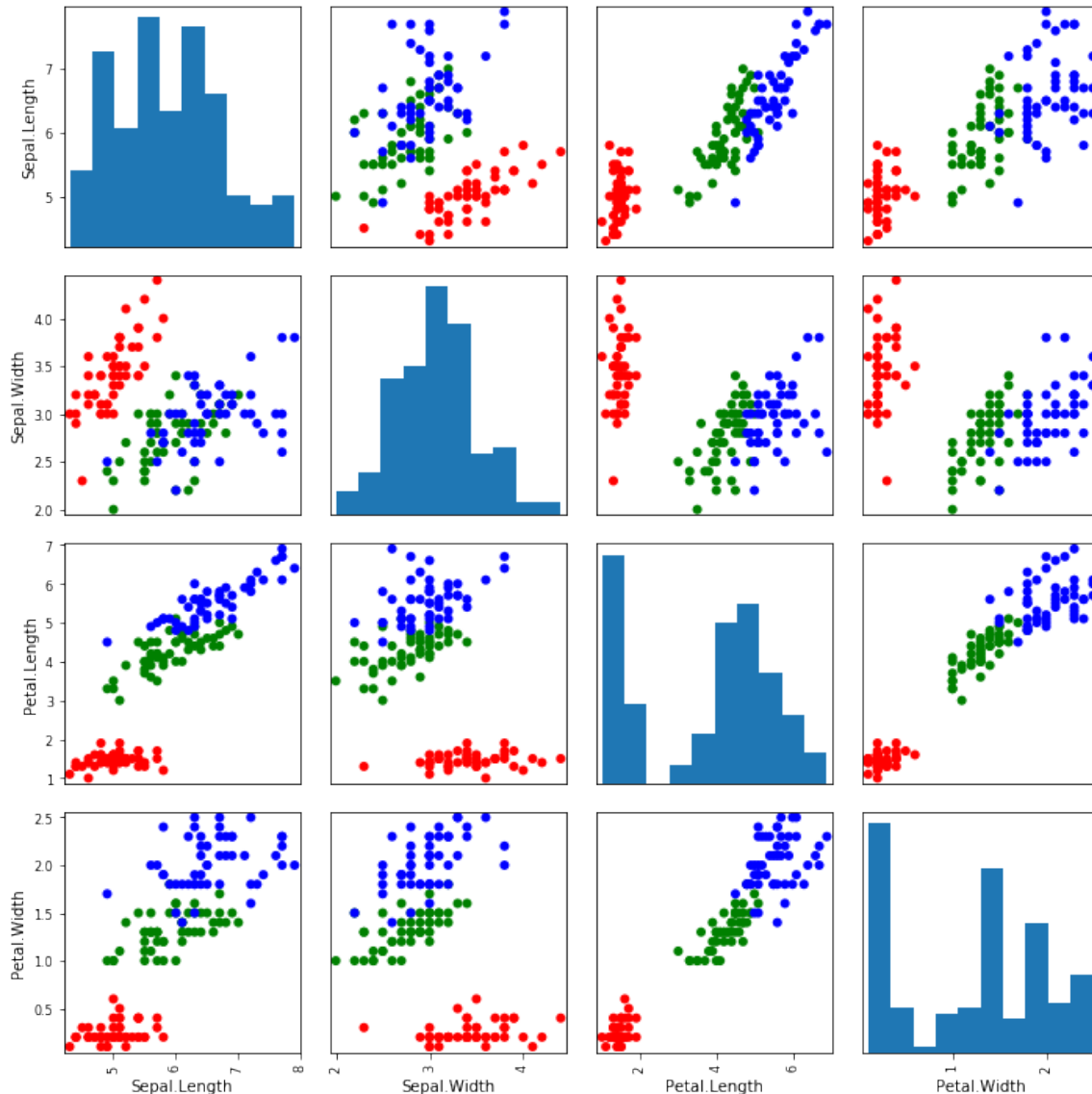


FIGURE 5 – Visualization of the Iris data as a pairwise scatter plot. The diagonal plots the marginal histograms of the 4 features. The other cells contain scatterplots of all possible pairs of features. Red circle = setosa, green diamond = versicolor, blue star = virginica.

The iris species classification task is a classical one. The goal is to distinguish three kinds of iris : setosa, versicolor and virginica. In the data set, a botanist has extracted 4 *features* : sepal length, sepal width, petal length and petal width.

What can you say regarding class separation? One class seems easier to distinguish from the others, which one? How would you separate the other two?

Exercise 6 OPTIONAL : Review Questions

1) Supervised vs. unsupervised systems

Of the following examples, which one would you address using a supervised or an unsupervised learning algorithm? Give some explanations for your answers.

- a) Given email labeled as spam/not spam, learn a **spam filter**.
- b) Given a set of news articles found on the web, group them into sets of **related articles**.
- c) Given a database of customer data, automatically discover **market segments** and group customers into different market segments.
- d) Given a dataset of patients diagnosed as either having **glaucoma** or not, learn to classify new patients as having glaucoma or not.

2) Classification vs. regression systems

Can we transform a regression problem into a classification problem? What would be the benefits of doing so?

3) Numpy arrays

- a) What are the main differences between a Python *list* and a numpy *array*?
- b) Can we mix different types of data in a numpy array?
- c) Define what is the *rank* in a numpy array.
- d) If we load a gray-level image in a numpy array, what is the rank of this array? Explain the different dimensions. Give a potential value of the **shape** attribute for a square images.
- e) Regarding the previous point, what would be the difference with a color image?

Exercise 7 OPTIONAL – Reading Assignments

— Read the first chapter of Murphy’s book “Machine Learning”.

Build your own summary of these chapters by doing a taxonomy of machine learning problems. You can find the pdf on Moodle.