

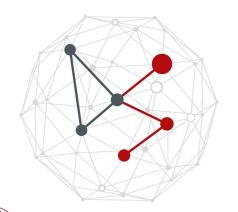
LAB 1 - MACHINE LEARNING TOOLS FOR DIMENSIONALITY REDUCTION: PCA AND CLUSTERING

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Lab classes



In this lab we will start with some basic tools







 We suggest you use Google Colab but if you are familiar with your local Python editor feel free to use it



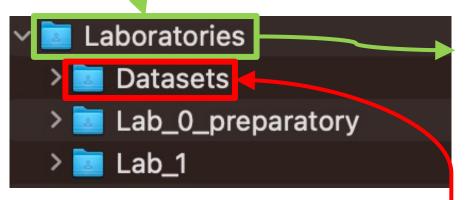
Note: Google Colab does not work with the University's Google accounts. You need to use your own Google account or create a new one.

Practical info – 1/2

Folder structure in Google Drive

Create a folder where you will create your own copy of the

labs



In the notebook we call it HDA_labs, feel free to chose your name and location, and then remember to modify the notebook accordingly

Only the first time

Create a shortcut to the folder of the datasets in your
 Google Drive – this folder is unique for all the laboratories
 → this will allow you to directly import the shared files without the need for copying them

Practical info – 2/2

✓ ■ Laboratories> ■ Datasets> ■ Lab_0_preparatory□ Lab_1

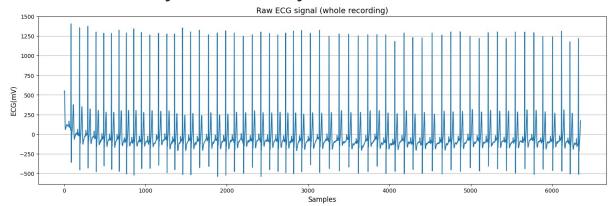
For each lab

 Create your own copy of the entire folder of the laboratory in your Google Drive (copy all the files, i.e., the .ipynb and .py files, and the folder of the images)



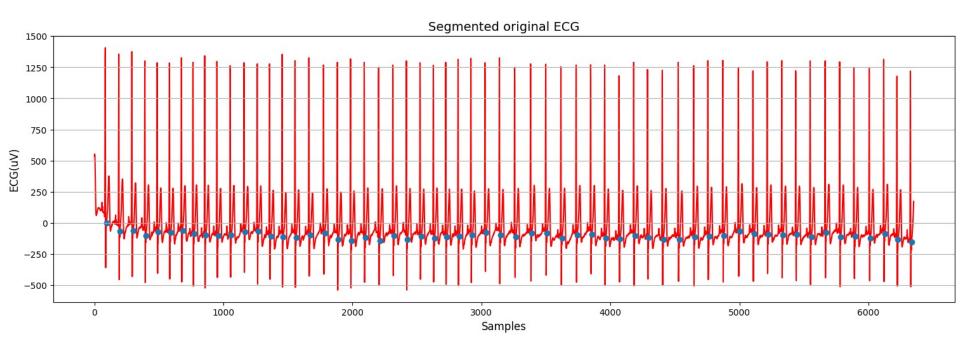
Lab 1 – Oct. 25

- Machine learning tools for dimensionality reduction: PCA and clustering
 - The challenge:
 - ECG signals dimensionality reduction
 - You will learn to:
 - implement a dimensionality reduction algorithm based on PCA from sketch and using the implementation from Python libraries
 - implement dimensionality reduction based on clustering
 - use Scikit-learn Python library



Preprocessing

- Split the ECG signal into the beats → dimensionality reduction will be applied to each beat
- Normalize the data → very important step to be applied at the input of machine learning algorithms!



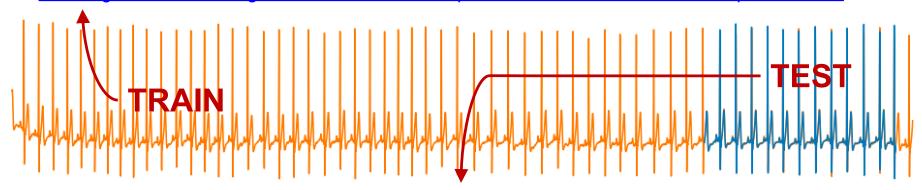
PCA

Obtain the principal components through two approaches:

- Implement PCA from sketch following the procedure detailed in L01_PCA (slides from Prof. Rossi) → will allow you to digest the procedure seen during the lectures
- Use the PCA algorithm already implemented in Python and offered as part of the Scikit-learn library

https://scikit-

learn.org/stable/modules/generated/sklearn.decomposition.PCA.html#sklearn.decomposition.PCA

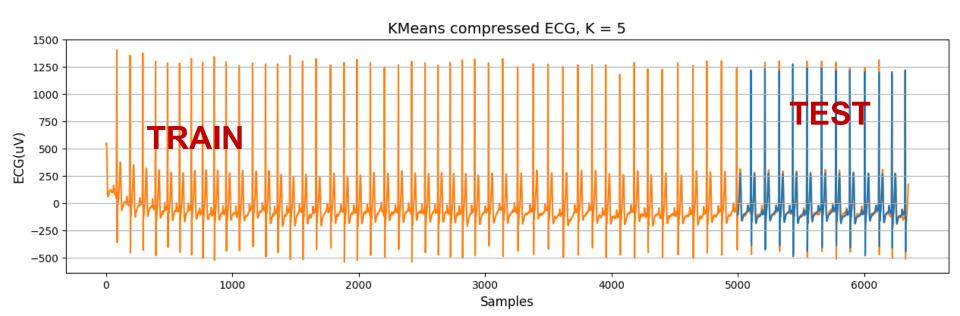


Evaluate the effectiveness of the components in representing the ECG beats (test on data not used during training!)

Clustering: K-means

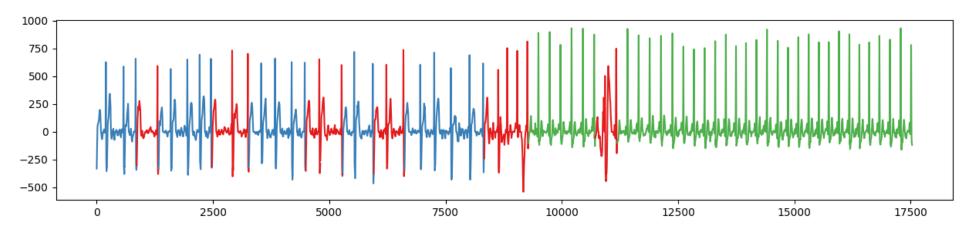
- Use K-means with 5 classes to obtain some representative beats (class centroids) to be used as codebook for dimensionality reduction
- Use the Scikit-learn library and evaluate the reconstruction error

https://scikit-learn.org/stable/modules/clustering.html#k-means



Clustering: DBSCAN

- You will use a different ECG signal composed of
 - 1. A first part collected from a patient with atrial fibrillation
 - 2. A second part with normal beats
- Implement DBSCAN using the Scikit-learn library
- Find the optimal Eps to separate the two classes (and outliers) using the procedure explained during the classes (see L03_DBSCAN)



How can I fill the notebooks?

- Read the text before the code: you can find the objetive of that portion of code and details about the procedure to implement
- 2. Click on the links of the documentation and read it
 - What does the function do? Is it the right function for your objective?
 - What are the inputs of the function? Matrix/array, shape, type...
 - What are the outputs of the function?
 - How should the function be imported at the beginning of the notebook? Usually we already do it for you but for the final project you will need to do it on your own, take a look at it
- 3. If the link to the documentation is not provided...search the documentation on your own ©

Import section

Two approaches: https://docs.python.org/3/reference/simple_stmts.html#import

```
# Scientific packages
import numpy as np
import scipy
import matplotlib
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from sklearn.cluster import KMeans
from sklearn.metrics import mean_squared_error
from sklearn.model_selection import train_test_split
from sklearn.cluster import DBSCAN
from kneebow.rotor import Rotor
from scipy.io import loadmat
from scipy.signal import medfilt
from scipy.signal import resample
# Plot packages
import matplotlib.pyplot as plt
from matplotlib.gridspec import GridSpec
# Custom functions
from hda_utils import load_dataset, load_dataset_DBSCAN, segment_ECG, segment_ECG2, matrix_to_signal
```