



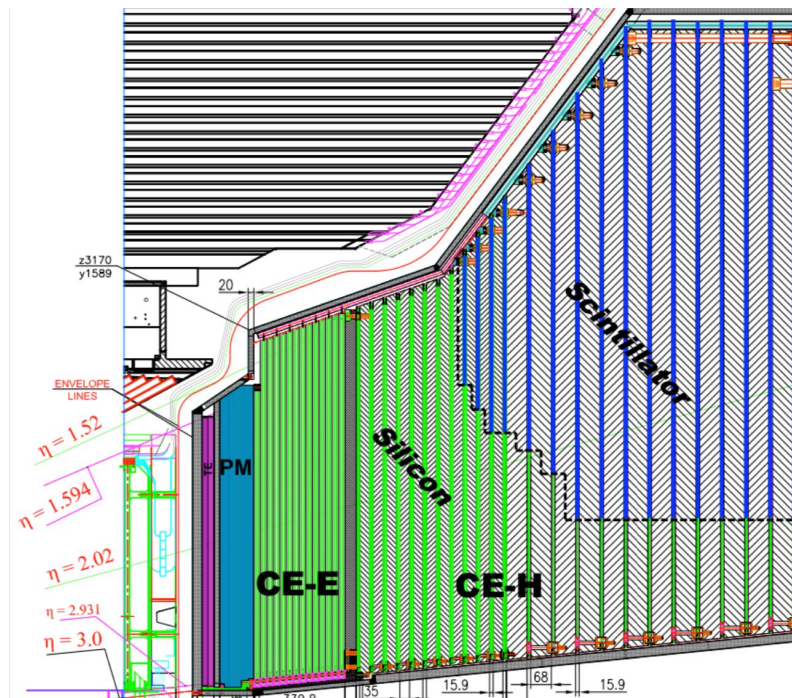
HIGH GRANULARITY CALORIMETER SIGNAL PROCESSING, WITH MACHINE LEARNING BASED RECONSTRUCTION IN THE CMS HGCAL

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Introduction

- High Granularity Calorimeter HGCaL will replace existing CMS endcap calorimeters (preparing for HL-LHC)
- Highly granular calorimeter with $\sim 6\text{M}$ detector channels from 50 layers
 - granular 5D (position(3), energy, time) signals
- Possibility of performing calorimetry with tracker like granular information.
- Challenge to fully utilize this information to extract physics objects out of the detector.



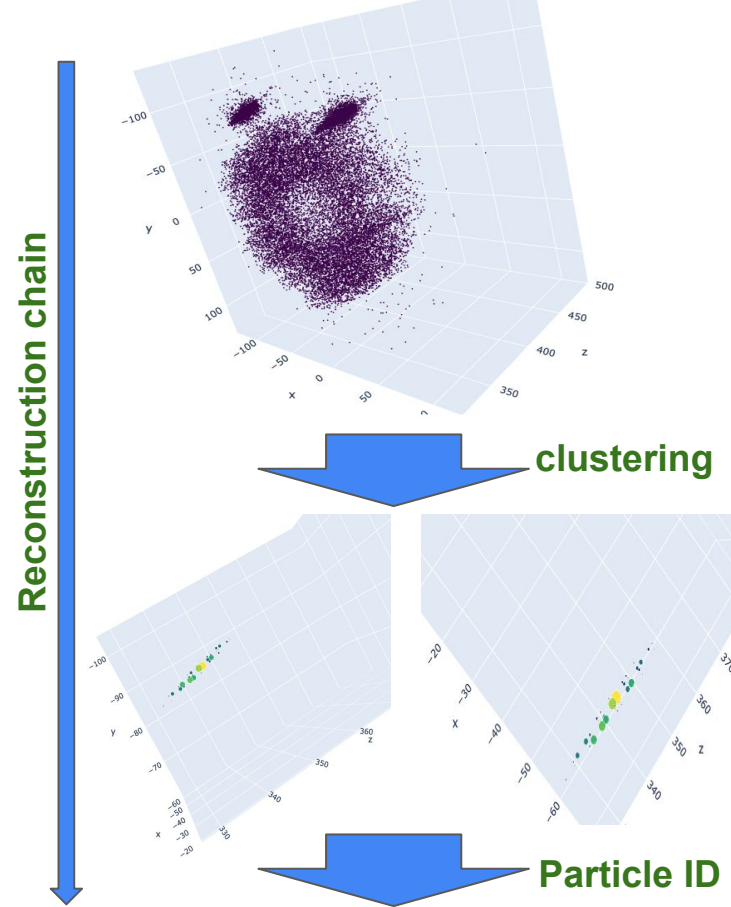
HGCaL detector to be discussed in details during Alexandre Zabi's talk on Thursday

Lab Objectives

- Visualise and explore signals corresponding to the shower of particles inside the detector
 - Simplest particles like photons and pions
- Assign properly the signals associated with individual particles with minimal contamination (**clustering**)
 - Single particles
 - Multiple close by particles with noise
- Properly utilise the signals attributed to a particle to identify the type of the particle (**particle ID**)
 - Photon vs pion

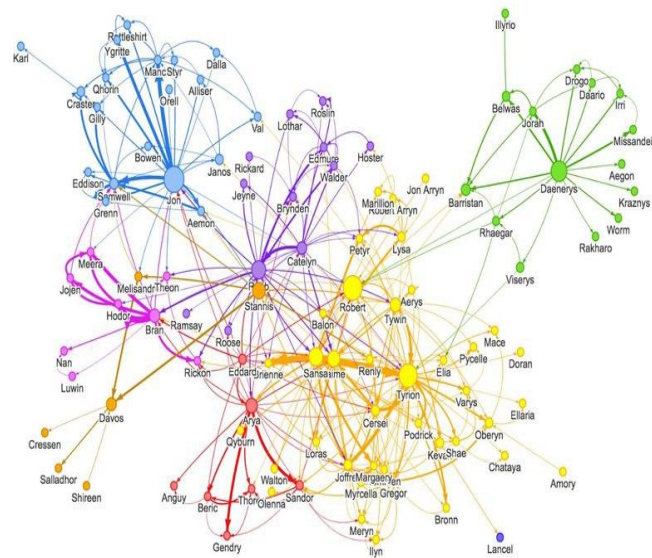
Ideas can be ported to your favourite experiment as well!

Hits in the detector with noise



Algorithms explored

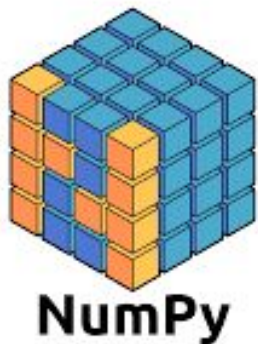
- Clustering : State of the art tools to be discussed for clustering
 - **K-means** clustering
 - **DBSCAN**
 - **Custom energy-density based clustering**(similar to that used in CMS)
- Particle ID : State of the art **Graph Neural Networks(GNNs)** to be considered
 - Hand crafted variables representing shower information has limitations
 - Use machine learning to extract maximal information
 - Use GNNs which excel at handling the 5D information the detector provides and building powerful representations of the showers
 - GNNs handle irregular geometries (hexagonal pixel size:a difficult task for out-of-the box Convolutional Neural networks) naturally.



Social media network
classification using GNNs:
CNNs/ RNNs do not work on
this kind of data

We will use powerful and popular graph network library: pytorch geometric

Fun and powerful packages: Hope to see you there!



Installation guide:

- Create a fresh conda environment
 - `conda create --name py3613 python=3.6.13` (this python version is preferred)
- Activate the new environment and install all the other packages inside this environment
 - `conda activate py3613`
- Install pytorch according to your system from:
 - <https://pytorch.org/>
- Install scipy
 - `conda install -c anaconda scipy`
- Install scikit-learn according to your system from:
 - <https://scikit-learn.org/stable/install.html>
- Install matplotlib
 - `conda install -c conda-forge matplotlib`
- Install tqdm
 - `conda install -c conda-forge tqdm`
- Install pytorch geometric according to your system from:
 - <https://pytorch-geometric.readthedocs.io/en/latest/notes/installation.html>
- Install jupyter
 - `conda install -c anaconda jupyter`
- `wget https://shghosh.web.cern.ch/shghosh/INFIERI2021/allschoolfinalmaterial.tar.gz` OR just download by clicking
 - All the notebooks and data are inside this tarball
- `tar -xzvf allschoolfinalmaterial.tar.gz`
 - Cd inside the untarred directory
- `jupyter notebook --port=8088 --no-browser`

Installation contd...

Or in linux systems with conda installed

- Get the list of all packages needed
 - `wget https://shghosh.web.cern.ch/shghosh/INFIERI2021/torchtest.yml`
- `conda env create --file torchtest.yml`
- `conda activate torchtest`
- `conda install jupyter`
- Download and untar to get the data and the notebooks
 - `jupyter notebook --port=8088 --no-browser`

Also the notebooks can be found in github at:

<https://github.com/sameasy/Infieri2021>

The data has to be downloaded from the link mentioned above

Some references:

- Calorimetry
 - Calorimetry: Energy Measurement in Particle Physics Book by R. Wigmans
 - https://indico.cern.ch/event/855973/contributions/3602188/attachments/1979913/3296643/calorimetry_sauvan_esipap2020.pdf : Awesome lecture 1
 - https://indico.cern.ch/event/124299/contributions/93835/attachments/74344/106642/EDIT_Calo_lecture_lr.pdf : Awesome lecture 2
- Clustering
 - <https://cs.wmich.edu/alfugaha/summer14/cs6530/lectures/ClusteringAnalysis.pdf>
- Graph Neural Networks:
 - Pytorch_geometric examples and tutorials : refer website
 - <https://arxiv.org/abs/1812.08434> : Review article(a bit old)
 - <https://arxiv.org/abs/1801.07829> : DGCNN paper
 - <https://arxiv.org/abs/2003.08013> : DRN paper