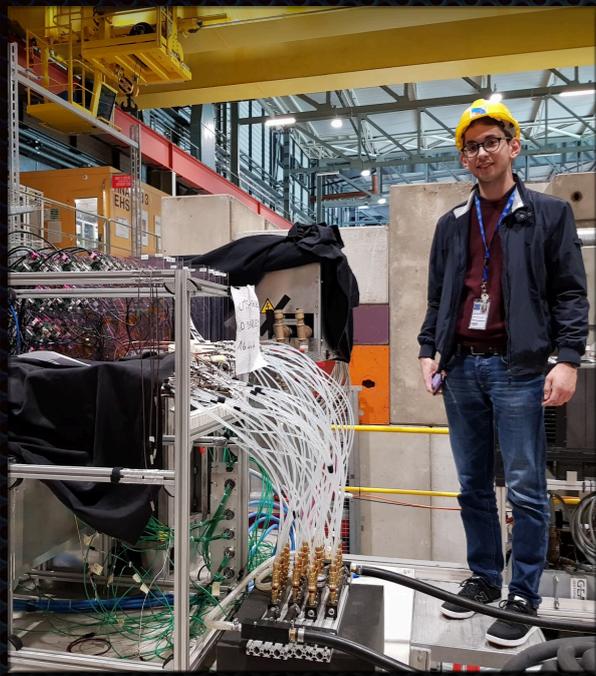


# SWAN for HGCAL Beam Tests Analysis

# Who am I?

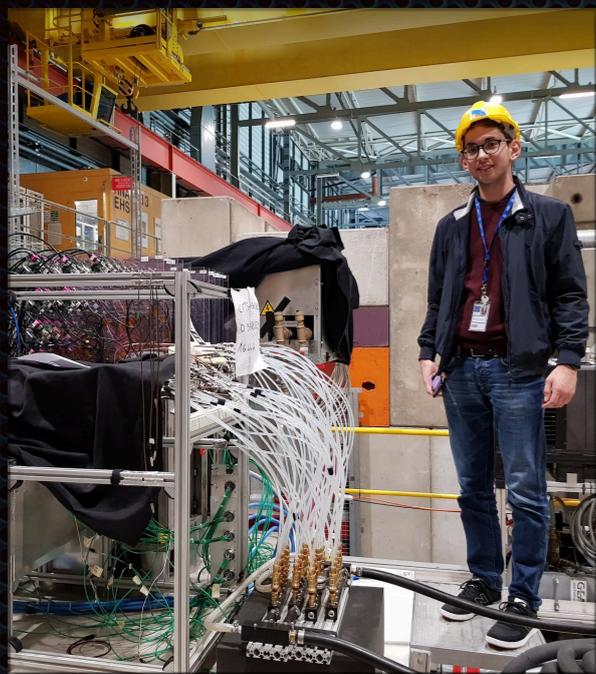
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- ✦ Matteo Bonanomi, 24 yo;
- ✦ Particle physics PhD student at École Polytechnique with the CMS experiment;
- ✦ Working on the CMS High Granularity Calorimeter (HCGAL) beam tests;
- ✦ ROOT User/addicted for my analyses ...

# Who am I?

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- ✦ Particle physics PhD student at École Polytechnique with the CMS experiment;
- ✦ Working on the CMS High Granularity Calorimeter (HCGAL) beam tests;
- ✦ ROOT User/addicted for my analyses ... until my PhD started (🤪🐍);
- ✦ SWAN User since ~1 year now.

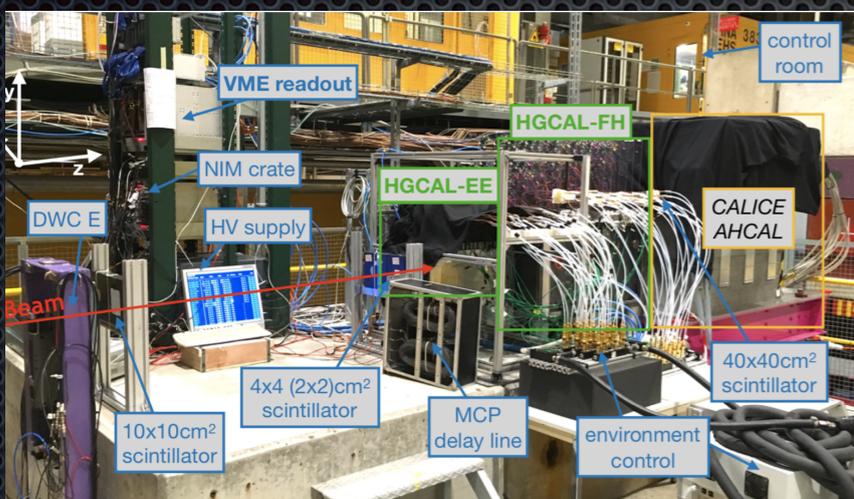
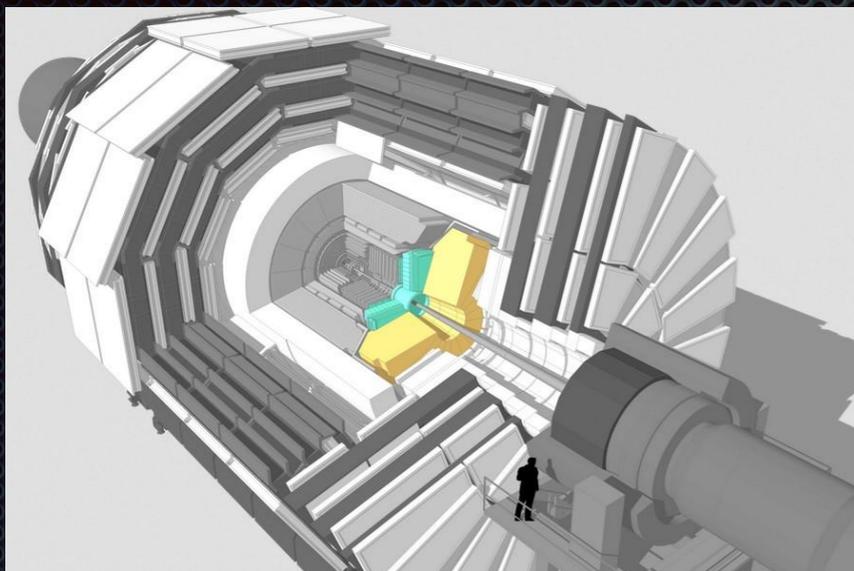
# In this talk

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- The HGCAL beam test analysis;
- From ROOT to pandas using uproot;
- SWAN in my everyday life;
- Feedbacks and conclusions.

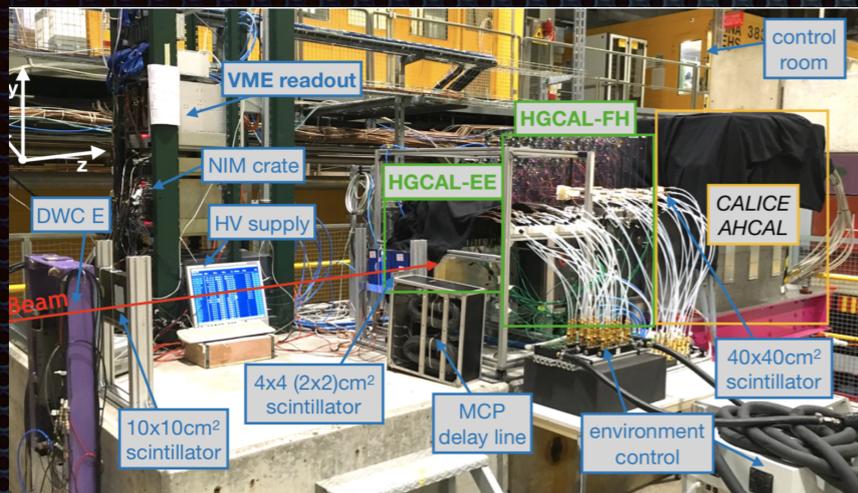


# HGCAL in a nutshell



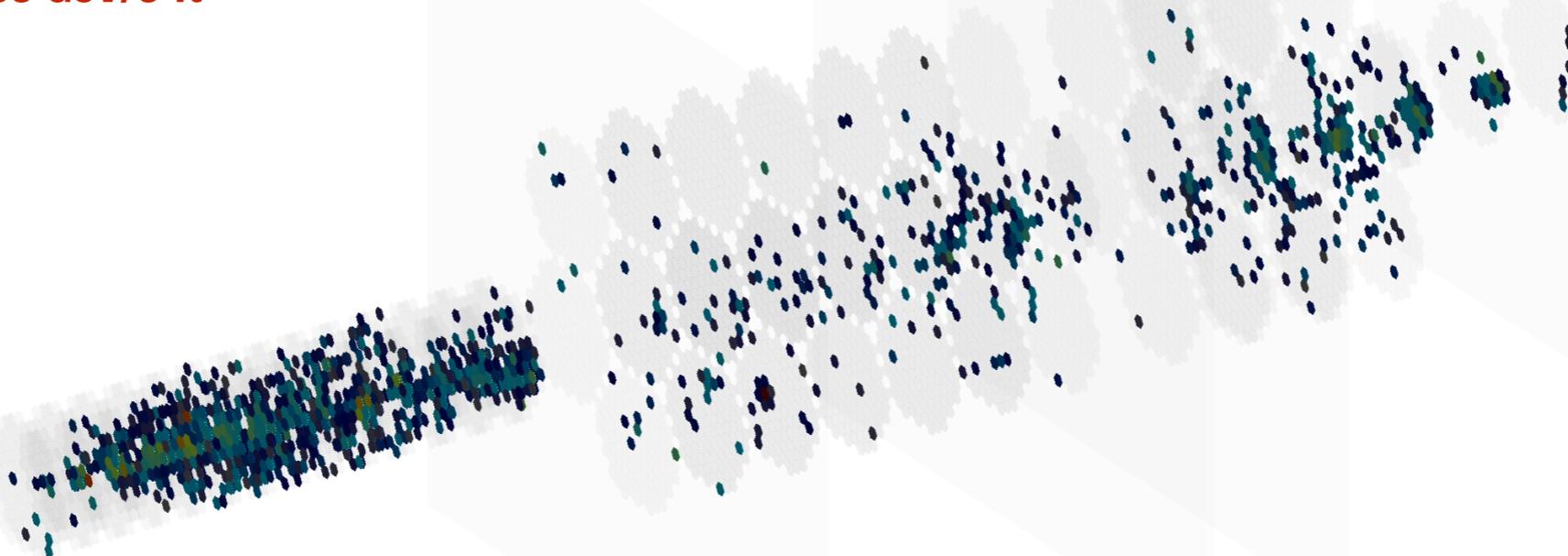
- ✦ **High Granularity Calorimeter (HGCAL);**
- ✦ Replacement of **CMS endcap calorimeters** for HL-LHC programme;
- ✦ First large scale test beam in October 2018: **0(1e6) data** to be analysed;
- ✦ **Analysis performed ~ entirely in SWAN** using the Jupiter notebook integration.

# HGCAL in a nutshell



- ✦ Tabular data: 1 row:1 **event**;
- ✦  **$O(1e6)$**  data to be analysed;

October 2018 run 517 - event 1  
**250 GeV/c  $\pi^-$**

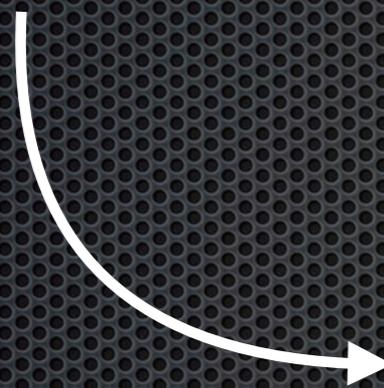


# Columnar data: from ROOT to data frames

```
*****
*Tree   :hits      : HGC rechits
*Entries :    6488 : Total =    591763506 bytes File Size = 106092451
*       :          : Tree compression factor =    5.58
*****
*Br    0 :event    : event/i
*Entries :    6488 : Total Size=    26725 bytes File Size =    13787
*Baskets :     4   : Basket Size=    44991 bytes Compression=    1.90
*.....*
*Br    1 :trigger_timestamp : trigger_timestamp/l
*Entries :    6488 : Total Size=    52781 bytes File Size =    3970
*Baskets :     4   : Basket Size=    44991 bytes Compression=   13.16
*.....*
*Br    2 :run       : run/i
*Entries :    6488 : Total Size=    26709 bytes File Size =    584
*Baskets :     4   : Basket Size=    44991 bytes Compression=   44.92
*.....*
```

- ROOT TTrees with arrays per event;
- Difficult to handle analysis cuts and to combine TTrees;
- Easily converted into **flattened pandas df** using **uproot**

`df = uproot.open(fname)[key].pandas.df(branches)`



```
In [10]: %%time
df_sim = do_df.do_df(energy, location_sim, isMC=True)

CPU times: user 13.3 s, sys: 6.93 s, total: 20.2 s
Wall time: 20.3 s

In [11]: df_sim.head()

Out[11]:
```

	rechit_chip	rechit_module	rechit_channel	rechit_energy	rechit_layer	ntracks	dwcReferenceType	b_x	b_y
event									
0	2.0	78.0	34.0	19.437500	1.0	1.0	13.0	-2.835938	1.137695
0	2.0	78.0	38.0	1.046875	1.0	1.0	13.0	-2.835938	1.137695
0	2.0	78.0	28.0	7.843750	1.0	1.0	13.0	-2.835938	1.137695
0	2.0	90.0	36.0	1.625977	2.0	1.0	13.0	-2.835938	1.137695
0	2.0	90.0	34.0	73.125000	2.0	1.0	13.0	-2.835938	1.137695



# The power of and (I)

## My SWAN configuration

- ✦ Software Stack: 96 **python 3**
- ✦ Platform: CentOS 7
- ✦ # cores: 4
- ✦ Memory: 10 GB
  - ✦ 16 GB not available on all instances

Matteo Bonanomi > HGCal TB analysis > Details

**H** **HGCal TB analysis** 

Project ID: 63270

 Add license  33 Commits  1 Branch  0 Tags  164 KB Files



Importing some user defined class for the HGCal TB analysis.

```
In [7]: from hgcalsw.data import read_ntuple as do_df
        from hgcalsw.utils import observables as obs
        from hgcalsw.fit import fit_functions as fit
```

# The power of and (II)

- ✓ SWAN offers a direct **access to eos** areas, along with a synchronisation with **CERN-Box**;
- ✓ Most of the python packages are already installed, but it is easy to add user defined or missing libraries;
- ✓ **Typical of a beam test analysis**: we are interested in **quick plots** for comparisons and to decide next analysis steps.
- ☹️ : Verbose **ROOT macros** difficult to modify and to keep under control when many plots are needed;
- 😊 : **Jupyter notebooks** to exploit python (`pandas df`) quickness in handling data and visualization.



# A day in life: beam test analysis

---

Let's have a look at this notebook to better understand my use-case of SWAN.



# The downside of

`df = uproot.open(fname)[key].pandas.df(branches)`

When accessing large files (~3GB) the conversion to pandas df causes crashes of the kernel:

One solution is to work only with **chunks of files**, but this would impact on the statistics.

Why this problem occurs? How the `.pandas.df()` command tweaks the memory?





# : What would be nice to have

---

- Could more memory help to solve the kernel dying issue?
- Notebook extensions for an improved usability;
- It would be nice to have *collaborative notebooks*, shared among analysers;
- Same performance in accessing files directly from eos and from `xrootd`;



# Let's wrap-up...

---

- ✦  +  in Jupyter notebooks are an extremely useful tool for analysers;
- ✦  Has many pros: synch with CERN-box, direct access to eos and “ready-to-go” environment ;
- ✦  Presents some downsides but none of them is limiting for the analysis.



# Let's wrap-up...

---

- 
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Looking forward to see next  steps and enjoy its improvements!

