

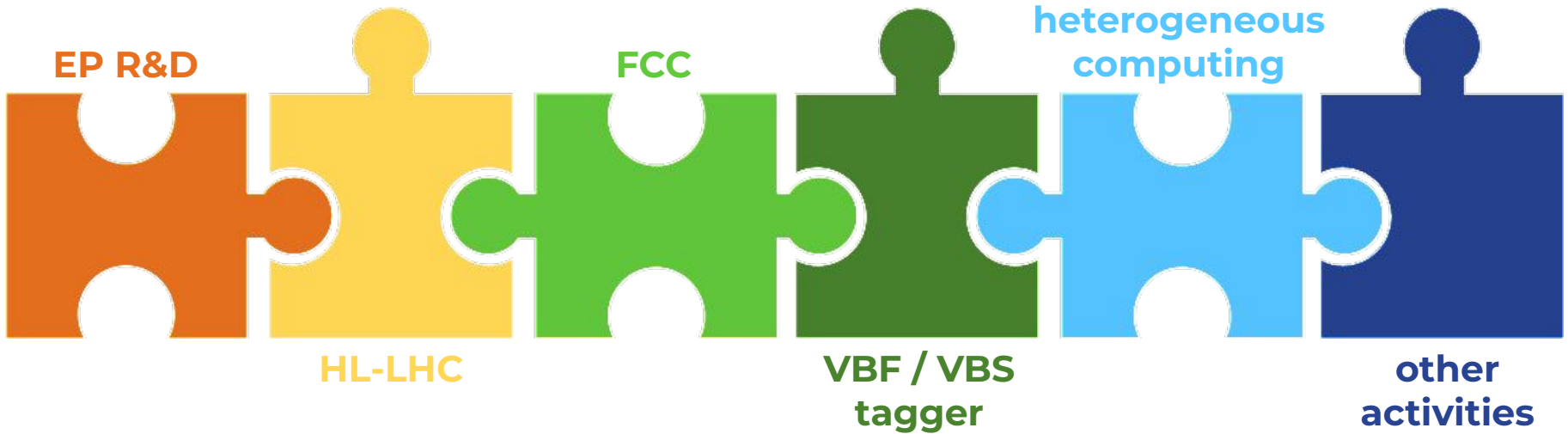


# Time information in event reconstruction at HL-LHC and future colliders and VBS/VBF tagger

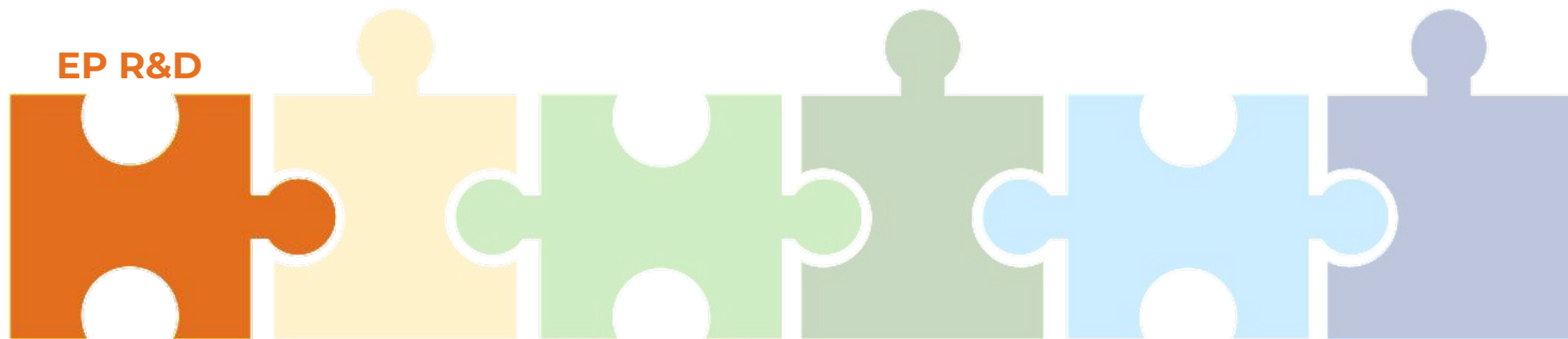
Aurora Perego

End of year PhD report - 20/09/24

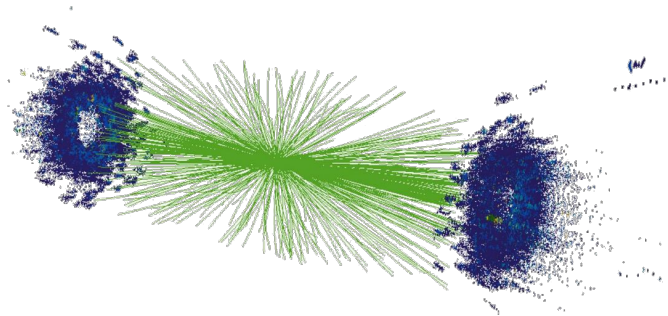
# Outline



EP R&D

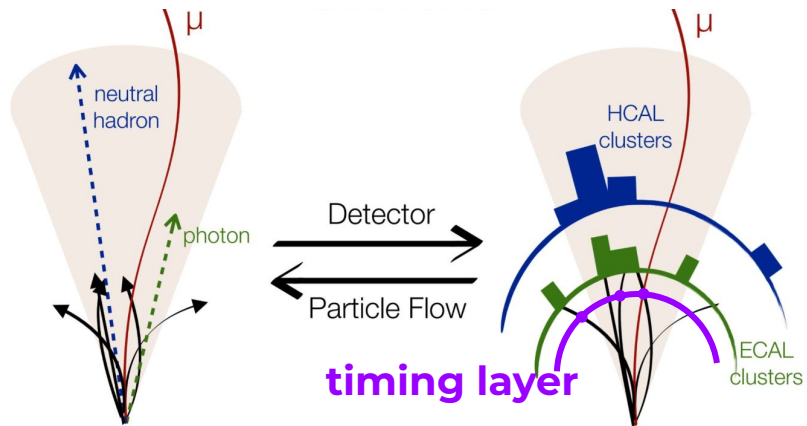


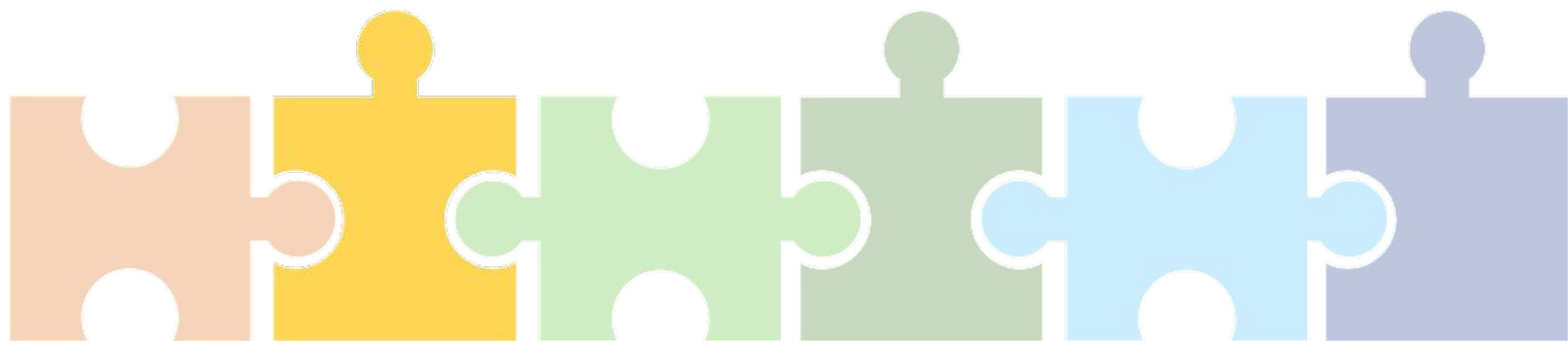
# PhD project within EP R&D at CERN



- My project at CERN is part of the **EP R&D** Phase 2 effort, specifically in the [software group](#)
- The goal of EP R&D is to address the technological challenges of future experiments
- My task is **calorimetry** reconstruction with the usage of **timing** information

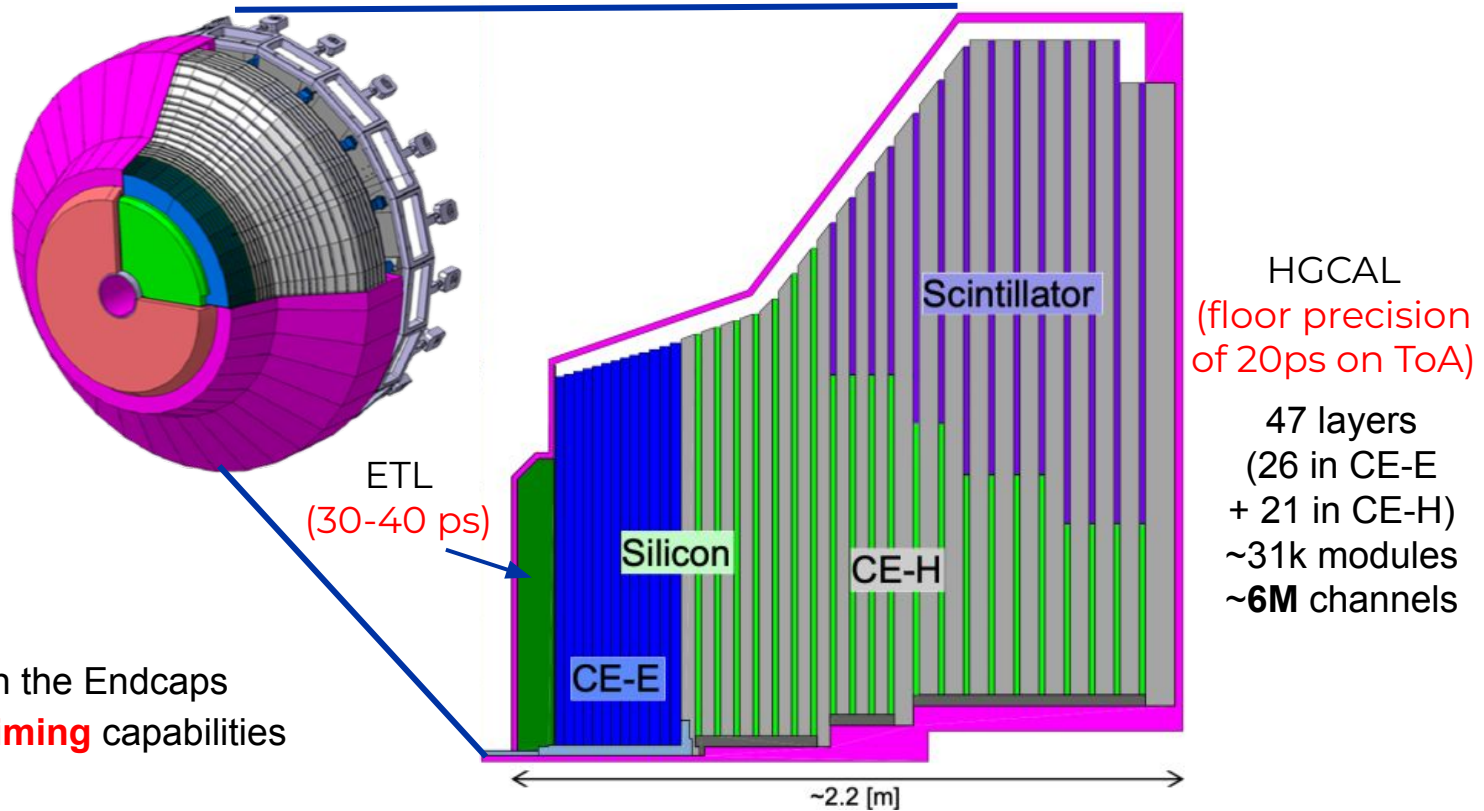
- Both **HL-LHC** and **future colliders** are covered in the project, with developments made for the former that can be adapted for the latter
- **Combination** with **tracker** information and **heterogeneous** software are also in the plan





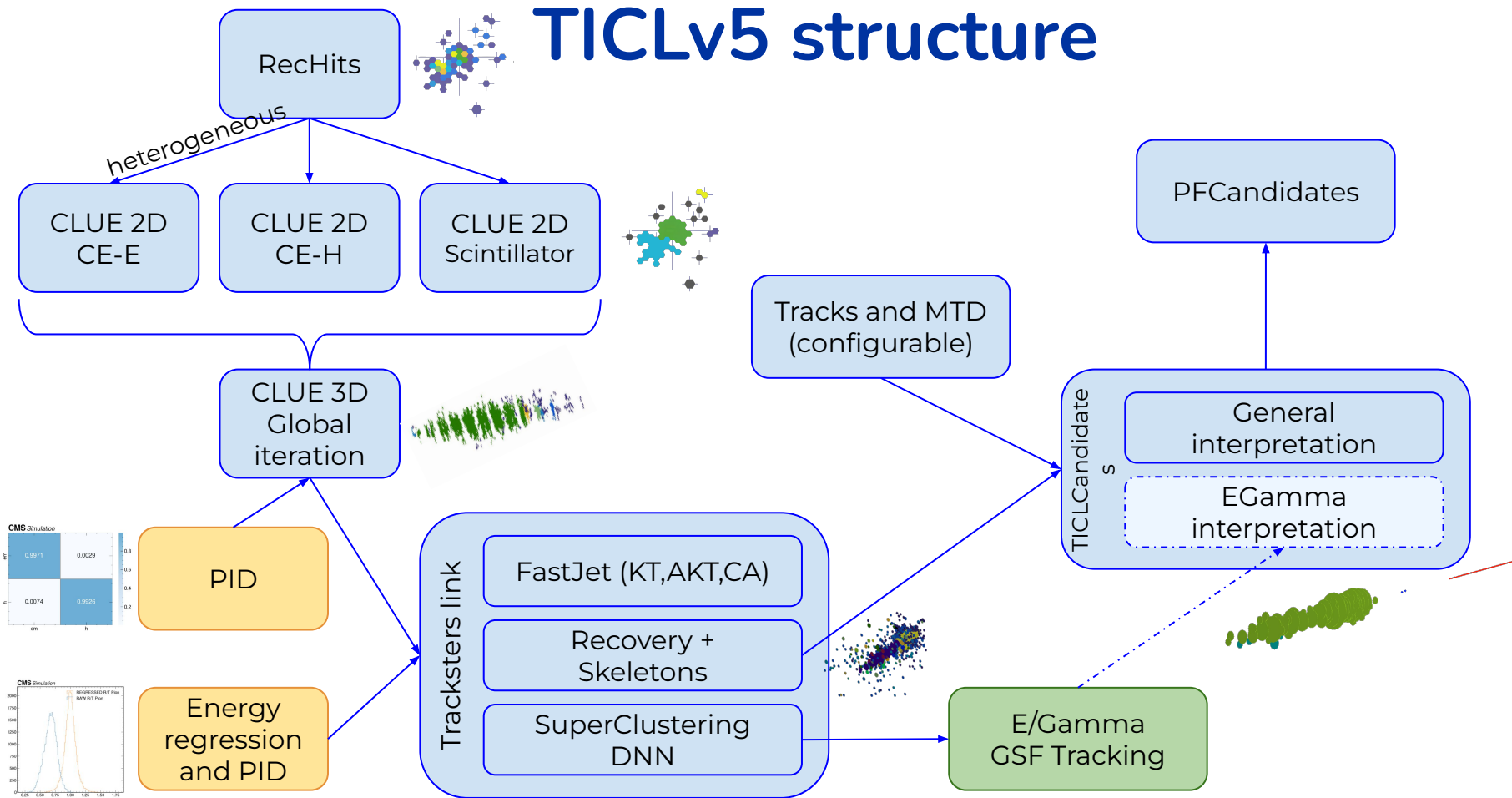
HL-LHC

# HL-LHC: CMS experiment



Phase-2 **upgrades** in the Endcaps  
New detectors with **timing** capabilities

# TICLv5 structure



# The Iterative CLustering (TICL)

TICL is a modular software framework integrated in CMSSW used to reconstruct particle showers in HGCal from hits to the final particle flow candidates

## TICLv4

- pure tracksters
  - room for improvement in hadron reconstruction

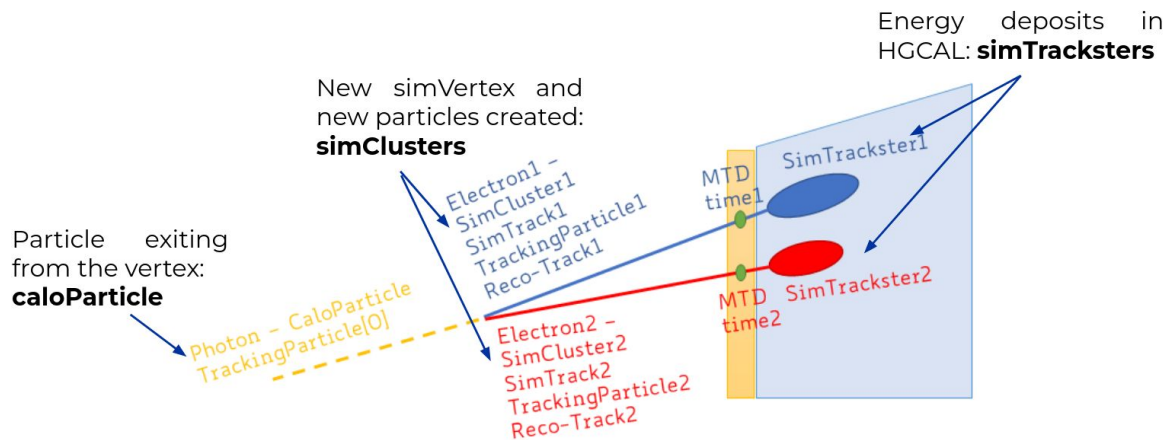
## TICLv5

- improved framework
- better usage of timing
- improved e/gamma and hadron reconstruction (new linking+recovery algorithm)
- split of algorithms parameters in different regions of HGCal
  - new ML models (DNN for superclustering and CNN for energy regression and particle Id)



# Simulation and validation

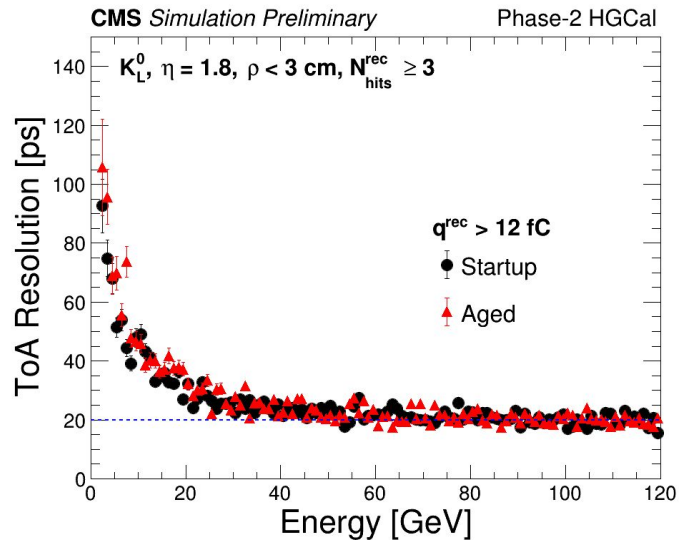
- For each particle that leaves hits in HGCal a **simTICLCandidate** is created
  - represents the **best reconstruction** that can be achieved
  - contains the **reconstructed track** that is best associated with the simulated particle
  - knows the **true** energy and pid of the particle, but also the **reconstructed energy** in HGCal
- hit by hit associators between the simulated and reconstructed TICLCandidates
- used to develop a dedicated validation to monitor the performance of the HGCal reconstruction and linking with the track



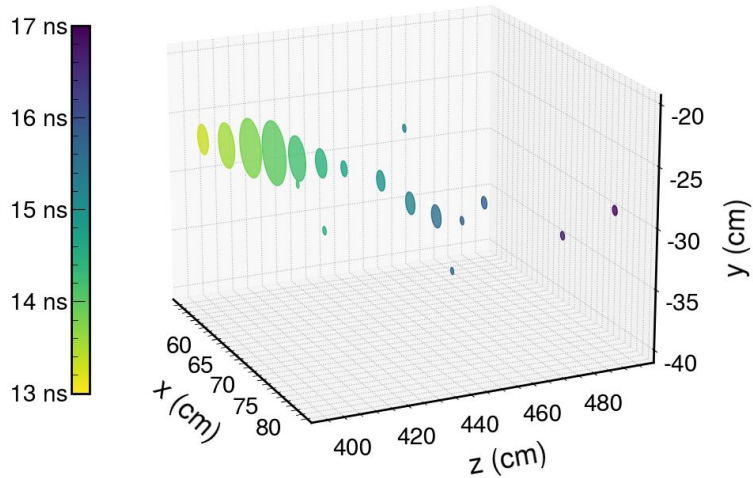
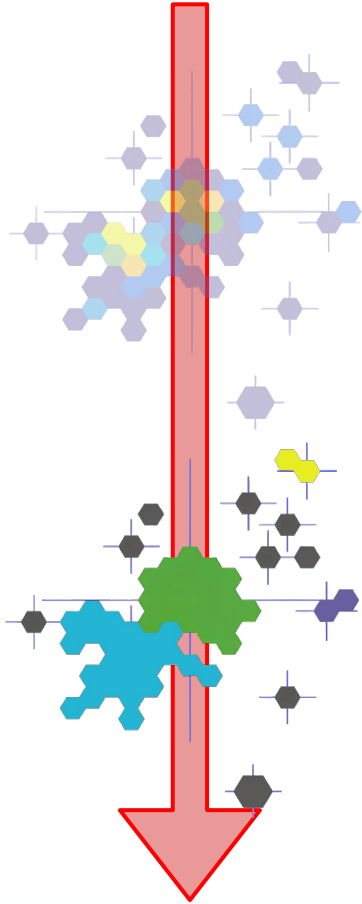
# Time information in TICL

- Each reconstructed **hit** with a charge **deposit above** a certain **threshold** has a **time** associated with it
- The **error** on the time corresponds to the expected resolution based on its **S/N ratio** and is **inversely proportional** with respect to the rehit **energy**

$$\sigma_t = \sigma_{jitter} \oplus \sigma_{floor} = \frac{A}{S/N} \oplus C$$



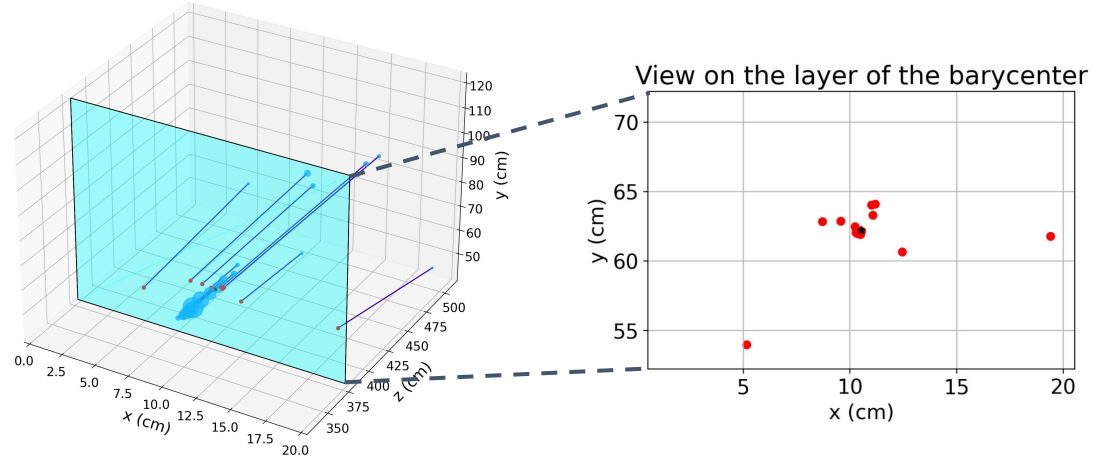
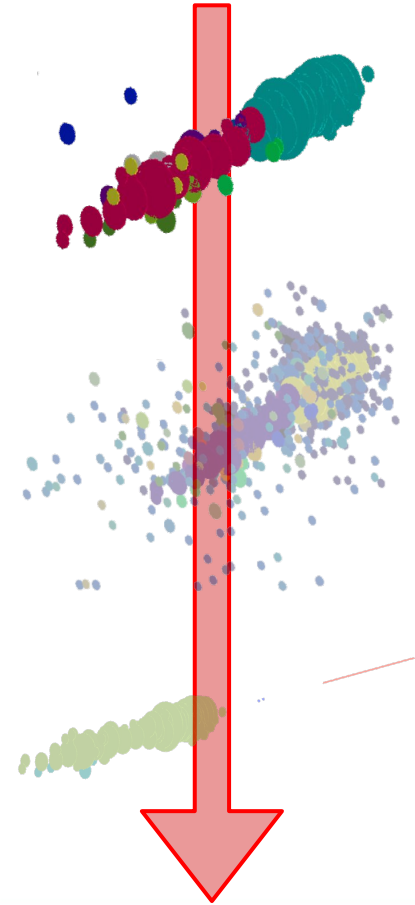
# Time information in TICL



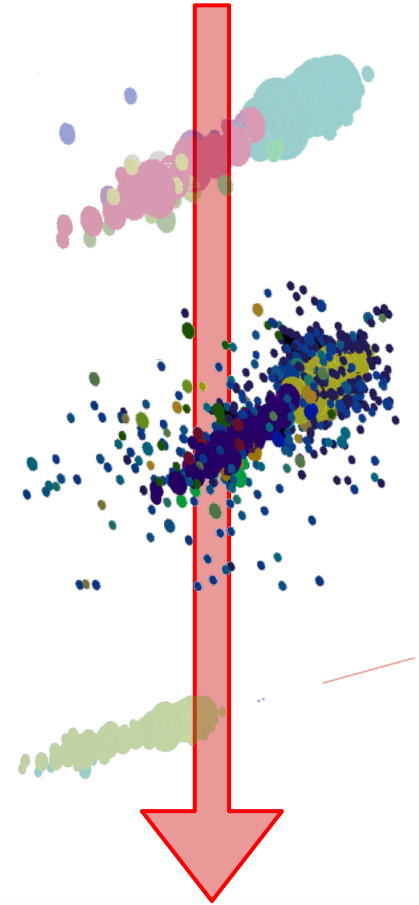
- The time for **layer clusters** with at least three hits with time information is computed as the **average** of the **time** of the **rechits** weighted on their error (a time window with the highest density of hits is used)
- Again, the **higher** the layer cluster **energy**, the **better** the **resolution** on the time

# Time information in TICL

- The layer clusters times are projected onto a **common surface** (HGCal layer where the barycenter is) and their **weighted average** is computed to obtain the trackster time. To improve the resolution, only clusters within 3 cm from the shower axis are considered.

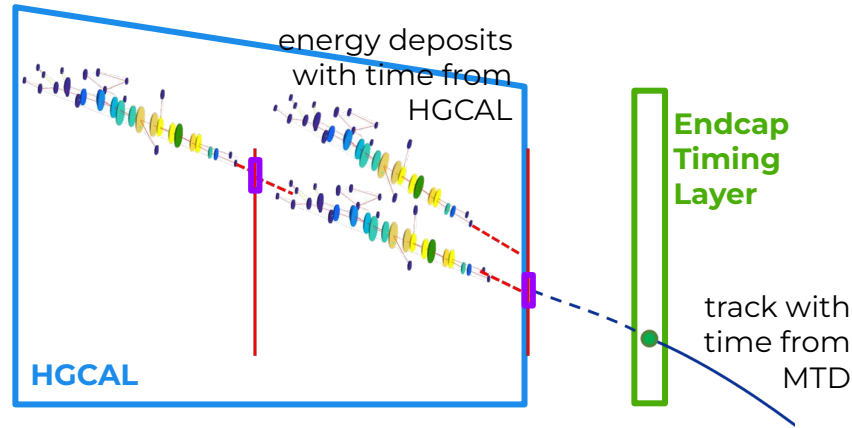
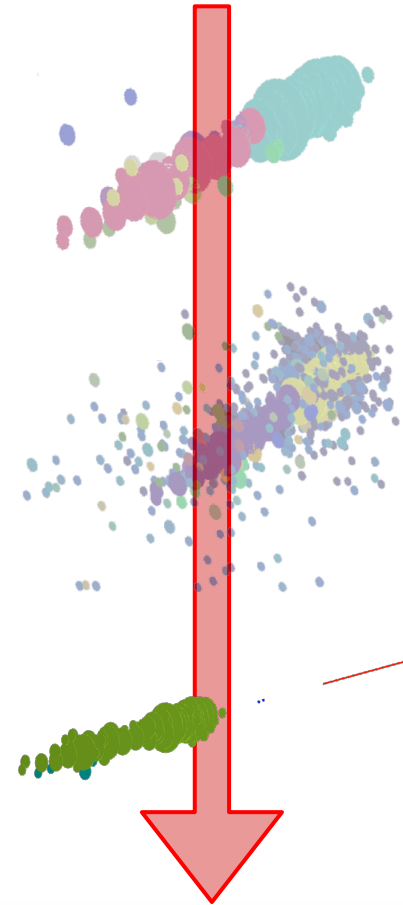


# Time information in TICL



- The same happens for the tracksters obtained with the linking algorithm (skeletons and recovery tracksters)

# Time information in TICL

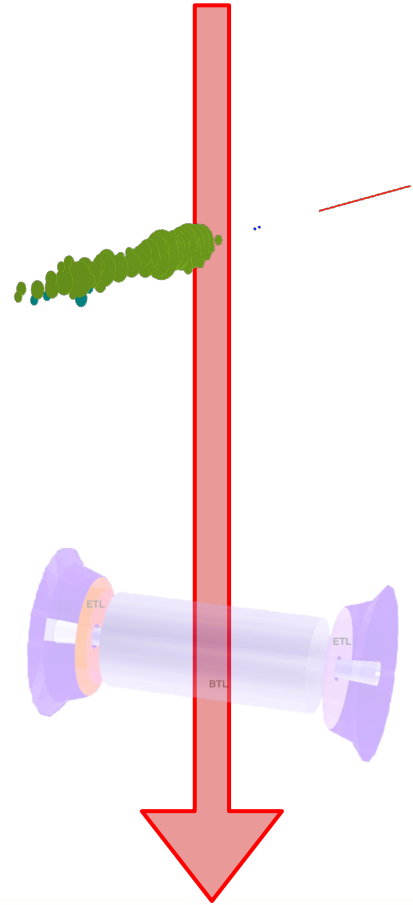


- During the linking with the track a time compatibility check is performed between the track and the trackster

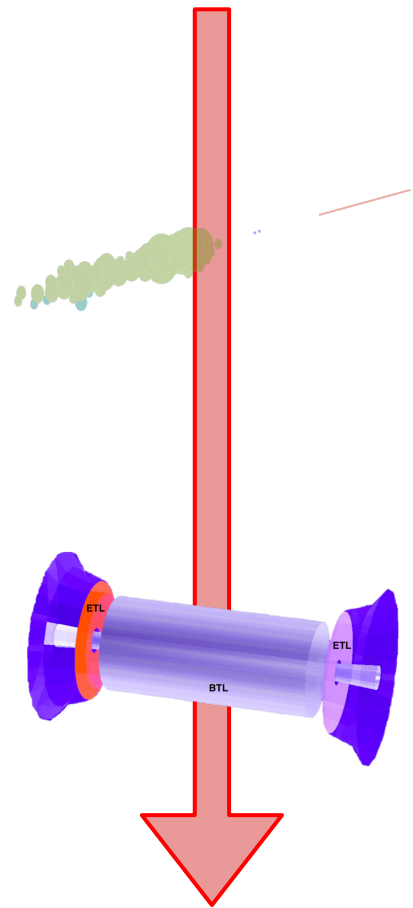
$$\left| \frac{d_{\text{MTD,trackster}}}{c} - (t_{\text{MTD}} - t_{\text{trackster}}) \right| < 3 \cdot \sqrt{\sigma_{t_{\text{MTD}}}^2 + \sigma_{t_{\text{trackster}}}^2}$$

# Time information in TICL

- The time of the TICLCandidate is obtained **propagating** the time of the trackster(s) back to:
  - the **origin** in a straight line if no track has been linked
  - the **point of closest approach** to the beamspot following the track if presentat the speed of:
  - the particle, computed with MTD information
  - light, if MTD information is not available



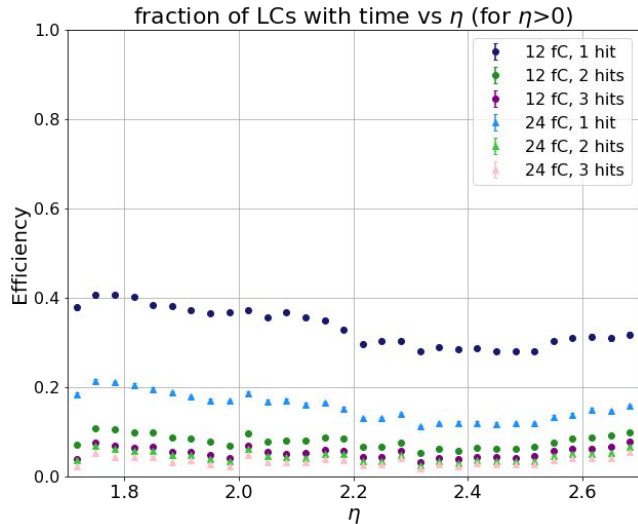
# Time information in TICL



- If **MTD** time is available and the **combination** is enabled the final TICLCandidate time and the MTD time at the vertex are combined
  - true by default, but configurable because the MTD time may not always be available (e.g. at the HLT)

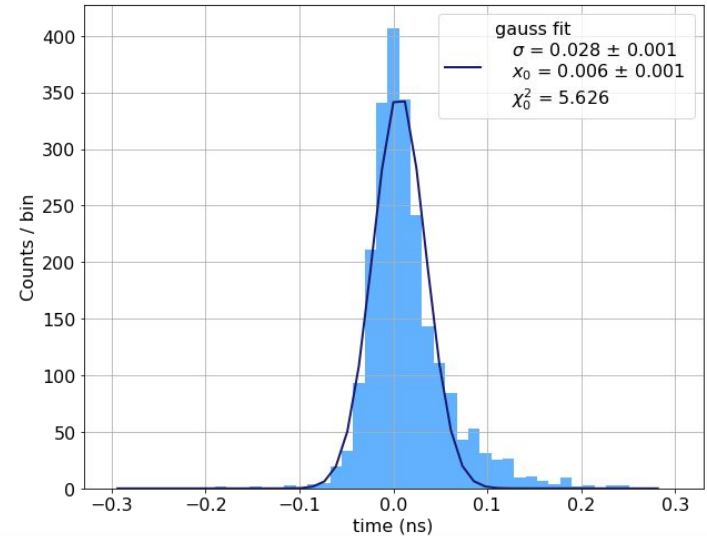


# Studies on tracksters time



- The threshold to trigger the **Time of Arrival** (ToA) in a sensor now is 12 fC, but it may change in the future
- The requirement on at least three **hits with time** in a cluster to compute the cluster's time could be optimized

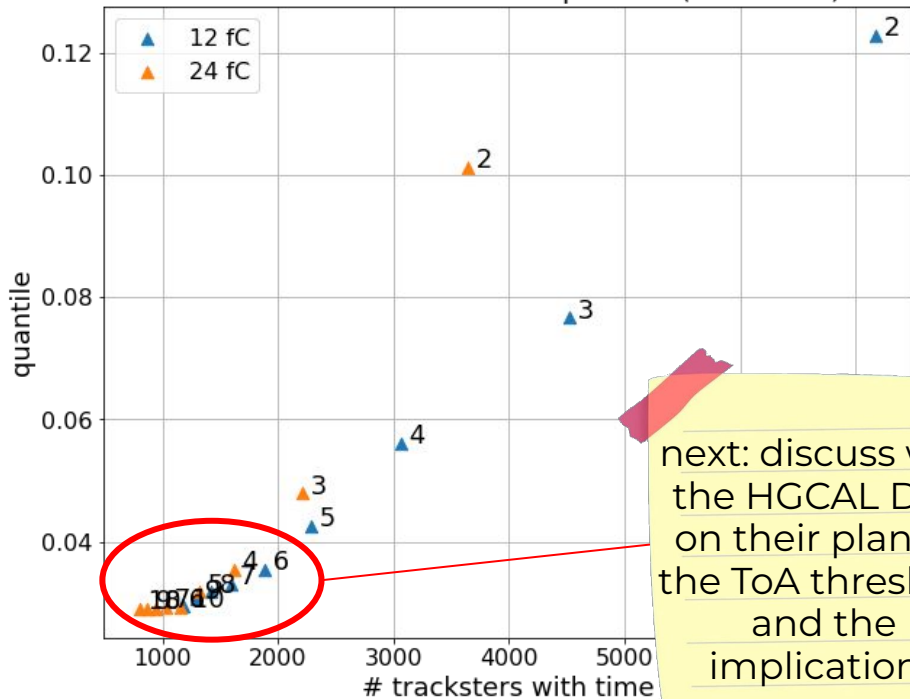
- Considering the CLUE3D tracksters, the difference between the trackster time and the simulated time has been computed for each combination of ToA (12 - 24 fC) and minimum number of hits (1 to 10)





# Studies on tracksters time

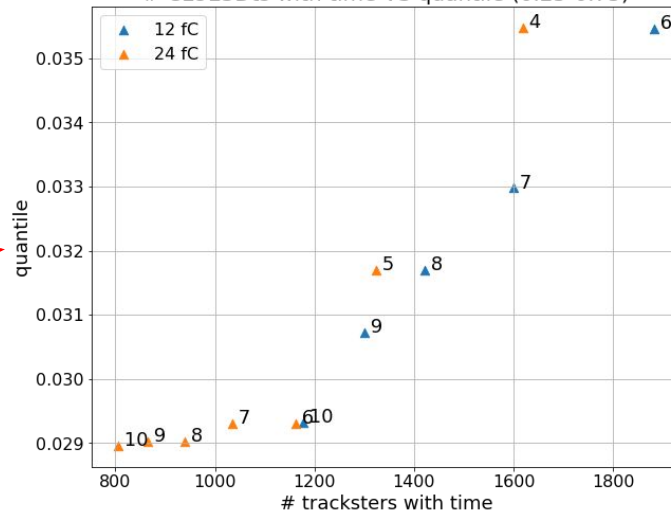
# CLUE3Dts with time VS quantile (0.25-0.75)



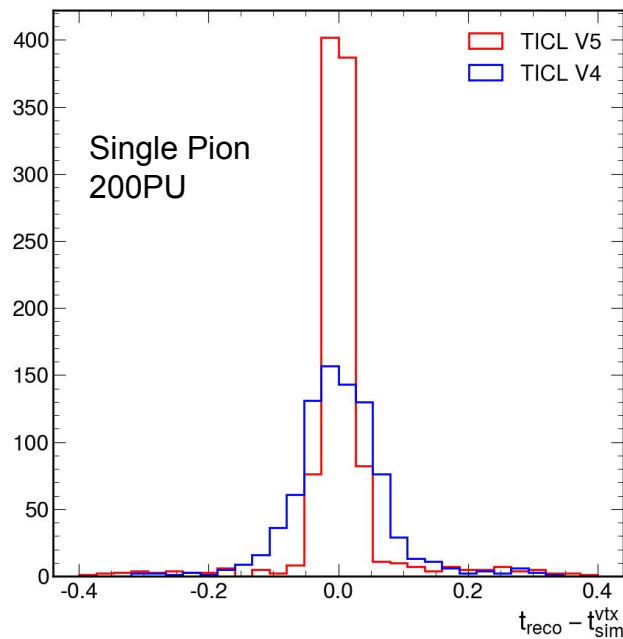
next: discuss with the HGCAL DPG on their plan for the ToA threshold and the implications

- number of tracksters with time against the difference between the 0.75 and 0.25 quantiles (to take into account the tails)
- the label is the minimum number of hits with time in LC

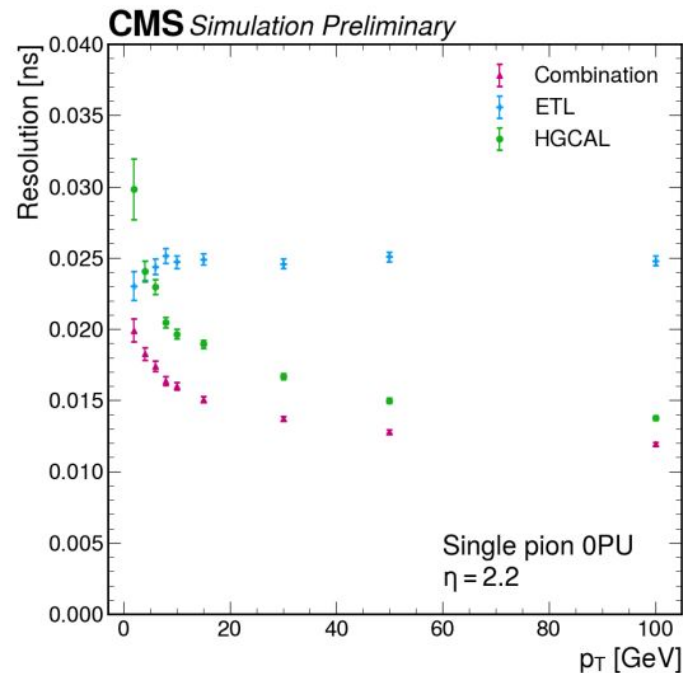
# CLUE3Dts with time VS quantile (0.25-0.75)



# TICLCandidates time resolution

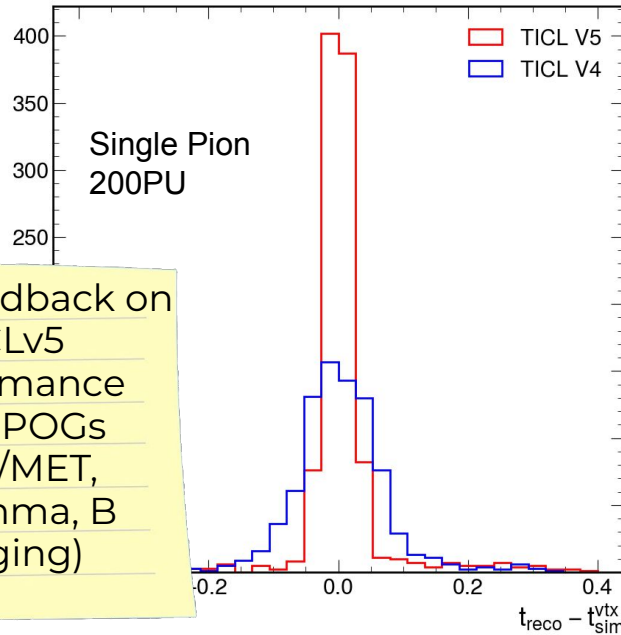


time residuals of charged candidates



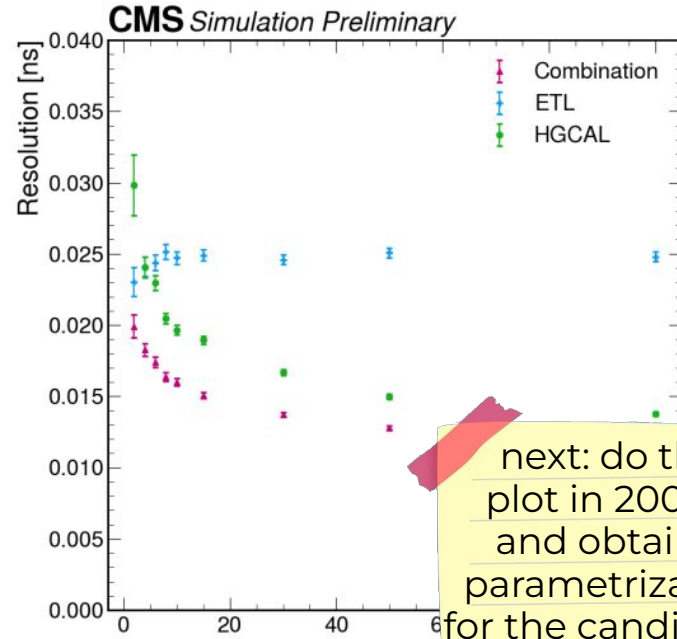
time resolution as a function of the simulated particle  $p_T$

# TICLCandidates time resolution



soon: feedback on TICLv5 performance from POGs (Jets/MET, e/gamma, B tagging)

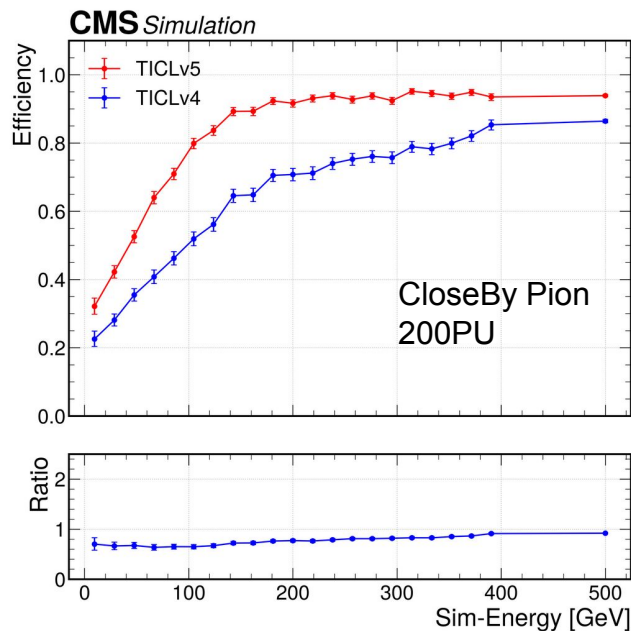
time residuals of charged candidates



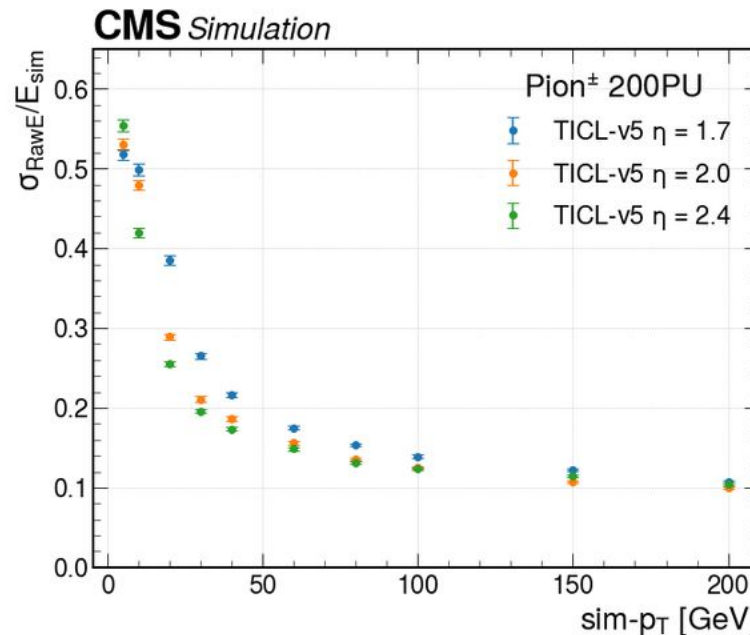
next: do this plot in 200PU and obtain a parametrization for the candidates time resolution

time resolution as a function of simulated particle  $p_T$

# TICLv5 first results

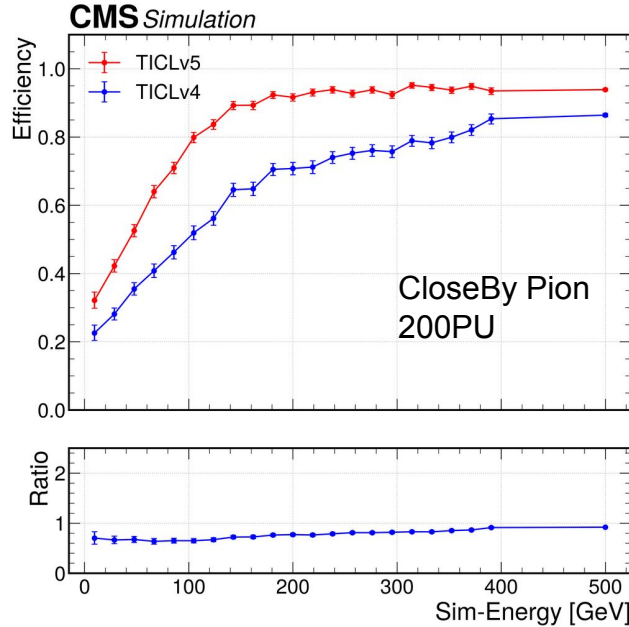


$$\text{Efficiency} = \frac{\text{simTracksters with a reco trackster that shares } > 50\% \text{ of the energy}}{\text{simTracksters}}$$

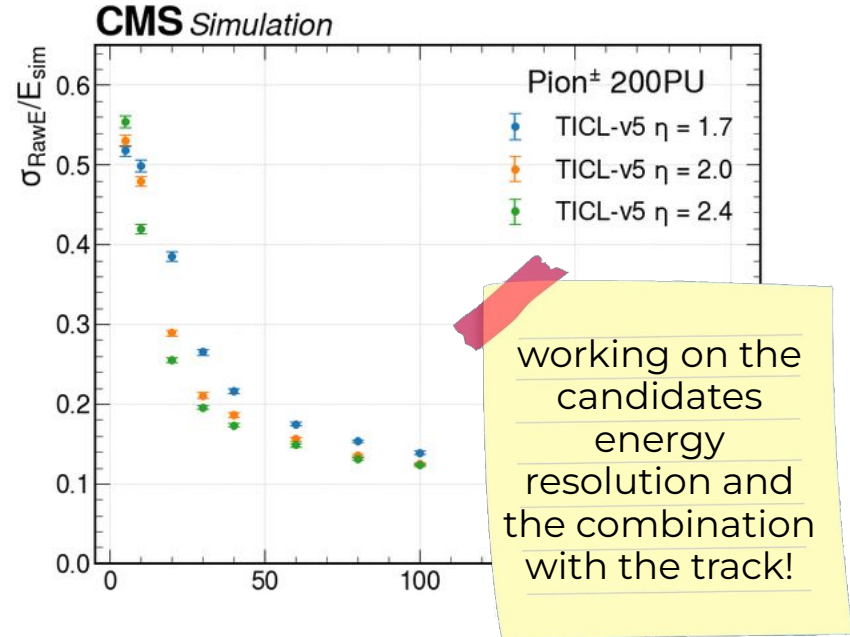


Energy resolution of the tracksters for different values of  $\eta$ ; each point is the effective  $\sigma$  of the distribution of the best (= shares > 50% of the energy with a simTrackster) reco tracksters raw energy

# TICLv5 first results



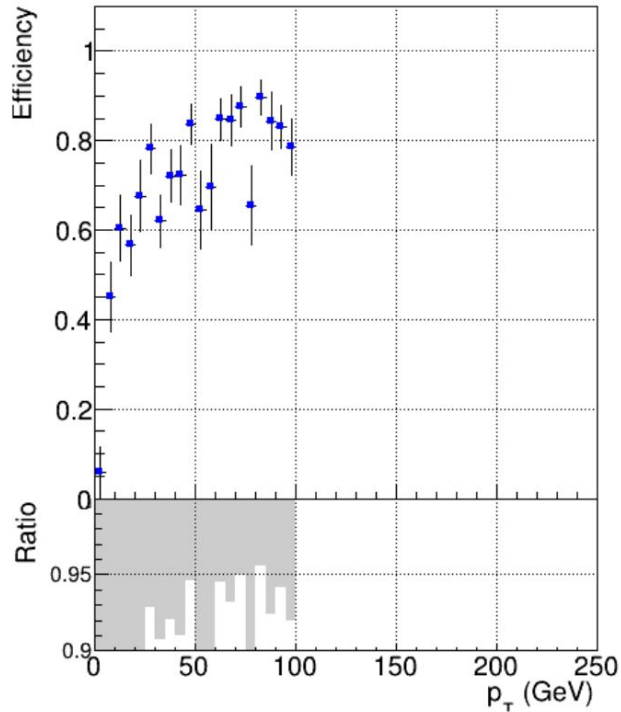
$$\text{Efficiency} = \frac{\text{simTracksters with a reco trackster that shares } > 50\% \text{ of the energy}}{\text{simTracksters}}$$



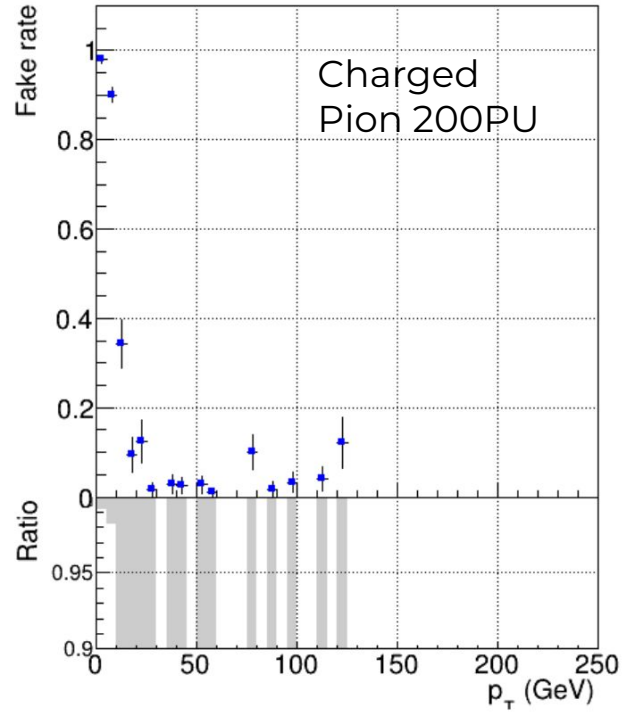
Energy resolution of the tracksters for different values of  $\eta$ ; each point is the effective  $\sigma$  of the distribution of the best (= shares > 50% of the energy with a simTrackster) reco tracksters raw energy

# TICLv5 first results

track efficiency for charged hadrons vs pt



track fake rate for charged hadrons vs pt



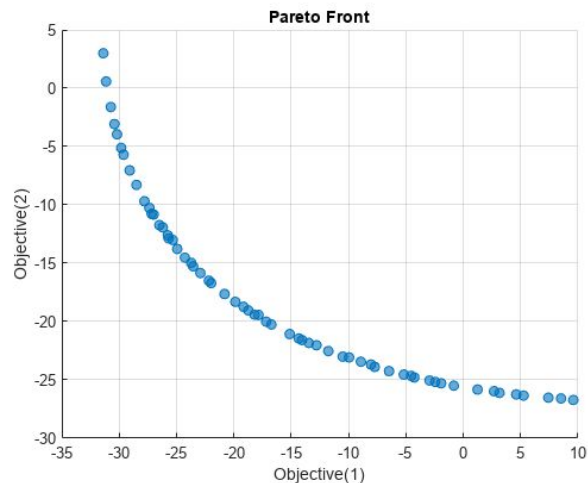
## TICLCandidates **validation**

- integrated in CMSSW
- set of plots to **monitor** the **performance**
- web visualization
- entire set of plots [here](#)

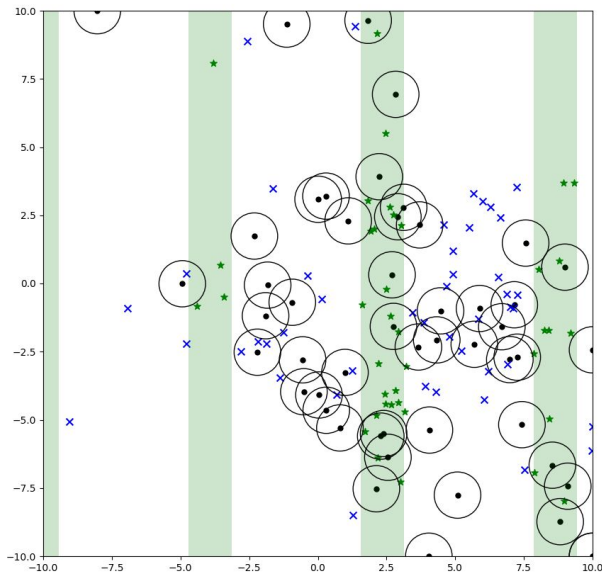


# Multi Objective Particle Swarm Optimization

- computational method employed to **optimize** a **problem** defined by multiple objective functions (metrics)
- **population** of *particles* explores the phase space of the solution to find the minimum
  - each particle updates its state at every iteration and knows the global best
- the result is a *pareto front* of solutions
- has been used in CMS to **optimize** some of the pixel reconstruction parameters and in **TICL** to tune the pattern recognition and linking **parameters**
  - successful, but requires a lot of time
  - idea: skip unnecessary evaluations



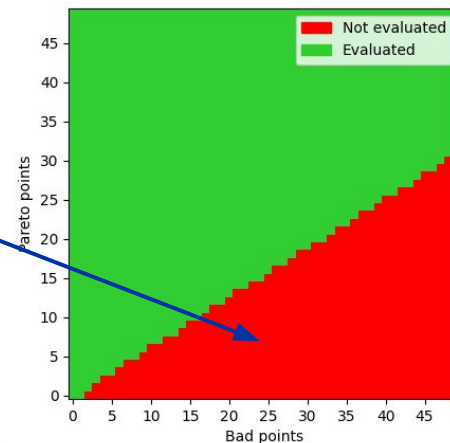
# MOPSO with RL



- strategy: use reinforcement learning to skip unnecessary evaluations
  - better than skipping them randomly
  - converges faster than standard MOPSO
- a point is bad if there are no pareto points in its surroundings
  - RL decides how many points to evaluate

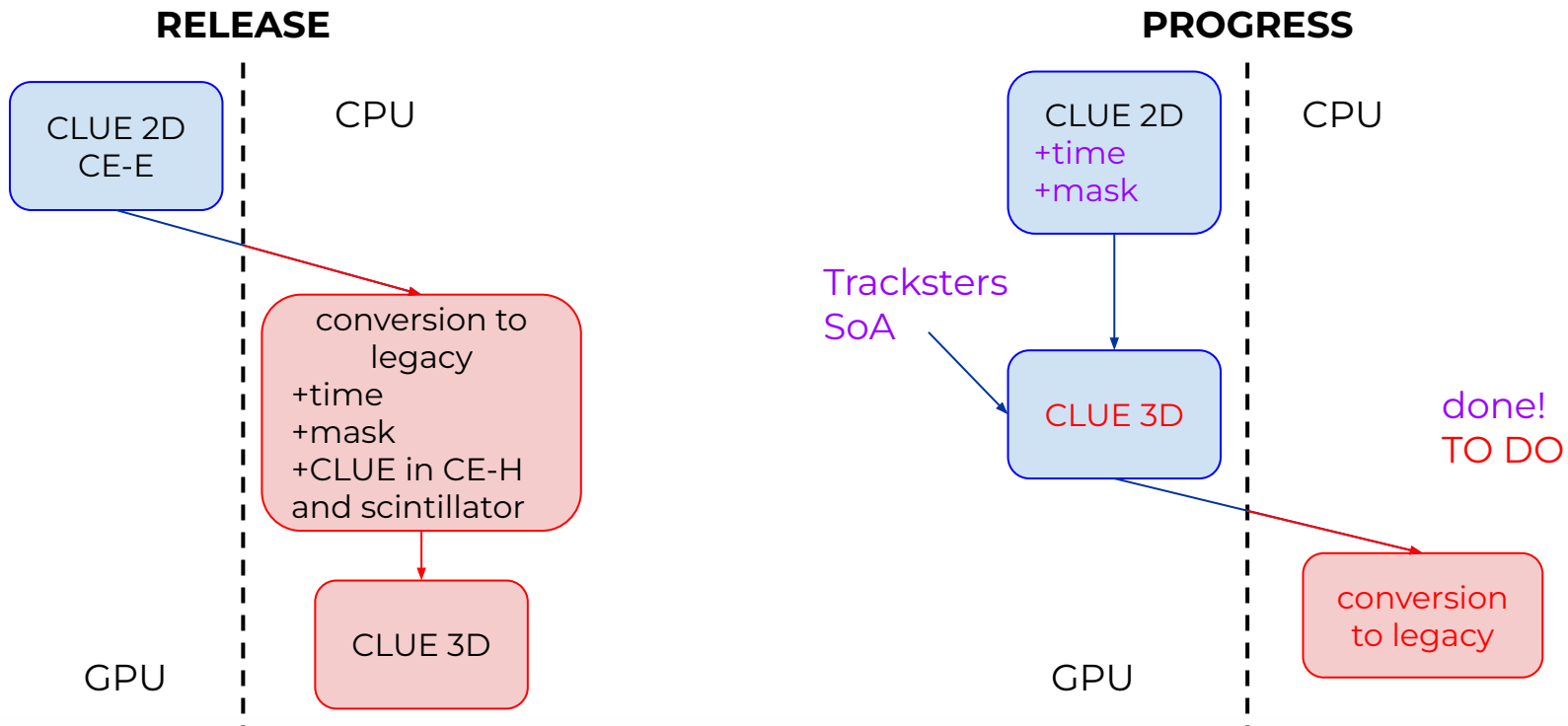
- Marco will test it on TICL for his thesis

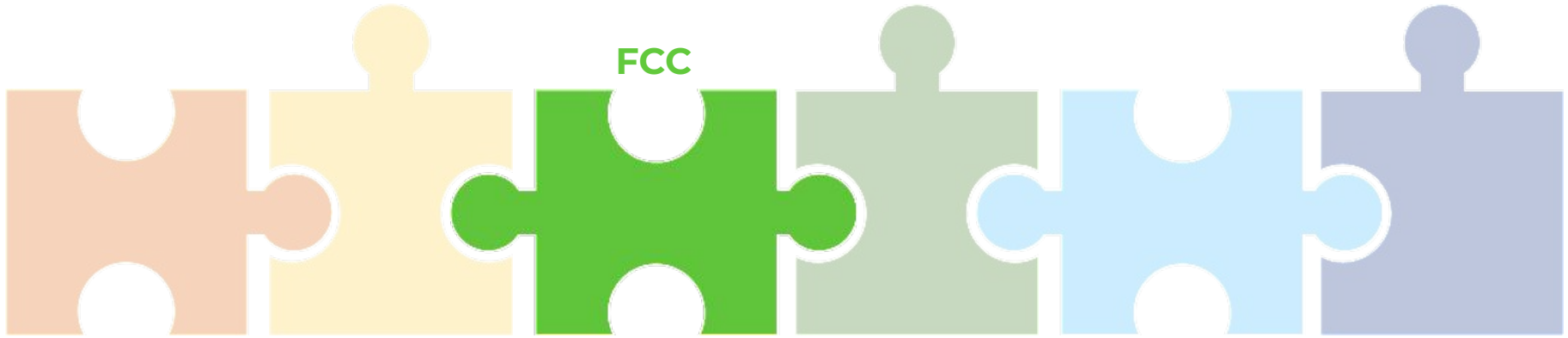
evaluations that  
will be skipped



# CLUE 3D @ alpaka

This summer I supervised a summer student who started the effort of writing the pattern recognition algorithm CLUE 3D with alpaka

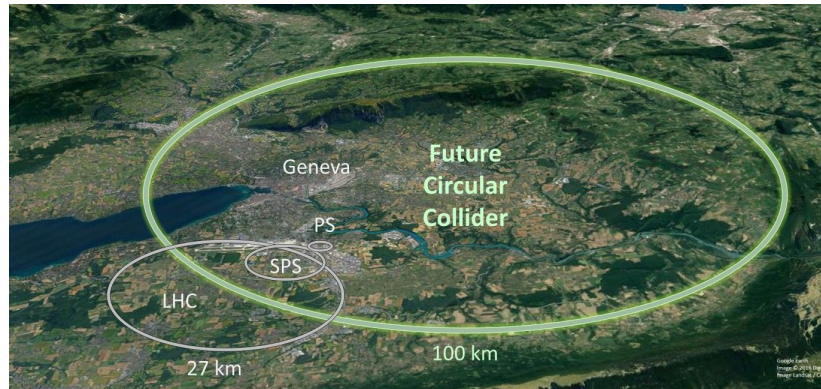




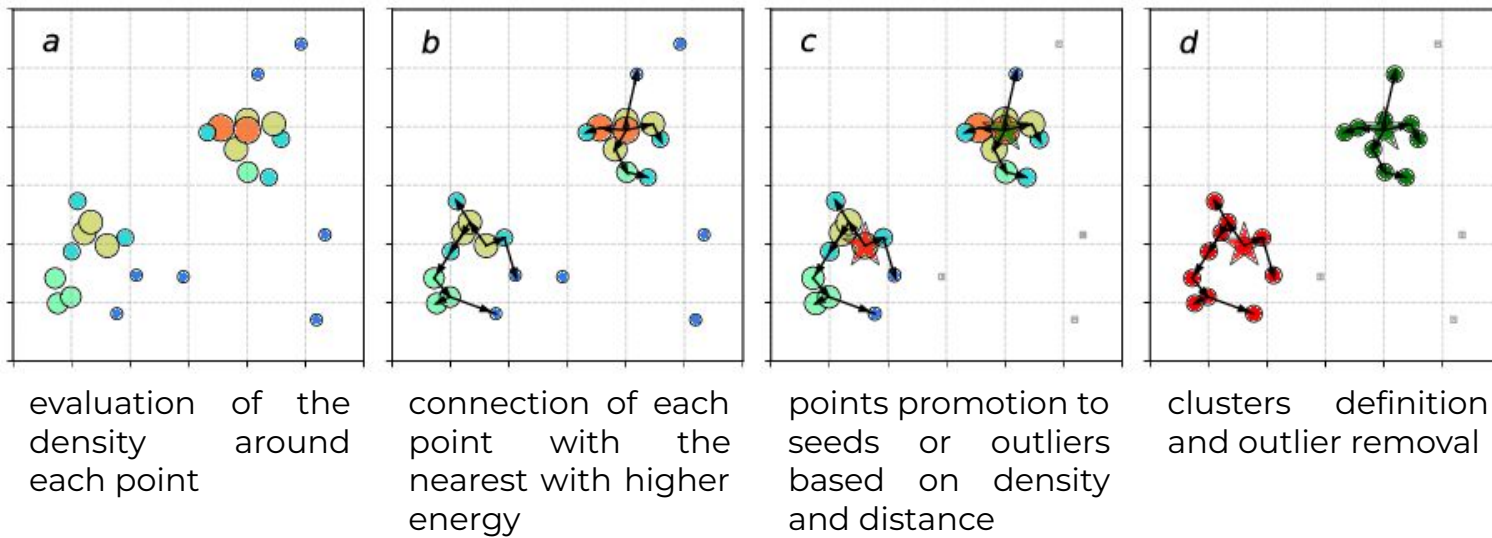
FCC

# plans for future colliders

- Part of my PhD is focused on **future colliders**
- 5 years plan:
  - explore the time information available in the FCC detectors
  - integrate **time** information in **clustering** and **pattern recognition** algorithms
  - use the timing of Particle Flow Candidates in Particle Flow algorithms
  - adapt the algorithms for execution in a **heterogeneous framework**
  - combination with the tracking reconstruction

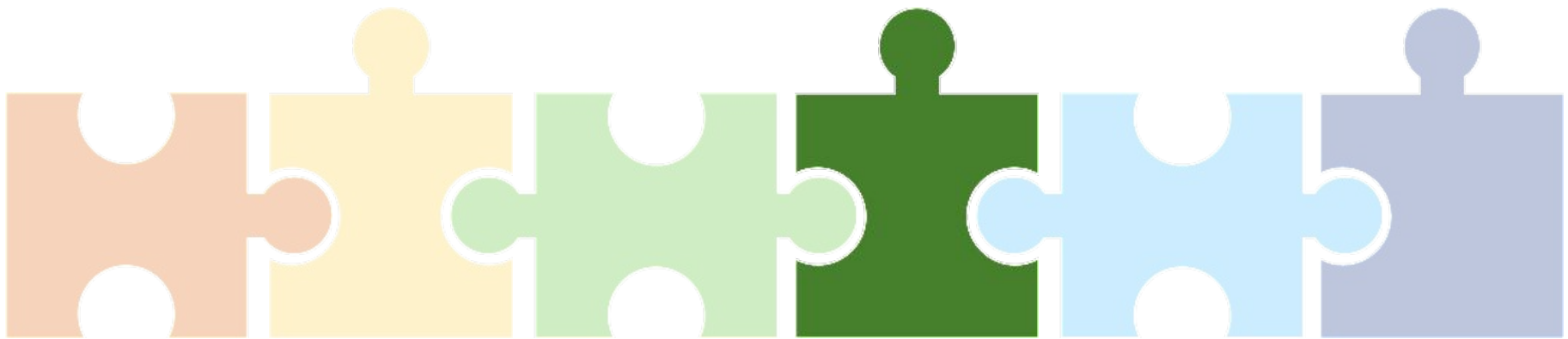


# n-dimensional CLUE



A more generic version of CLUE has been developed (already on GPU), able to do the clustering in N dimensions, where N can be:

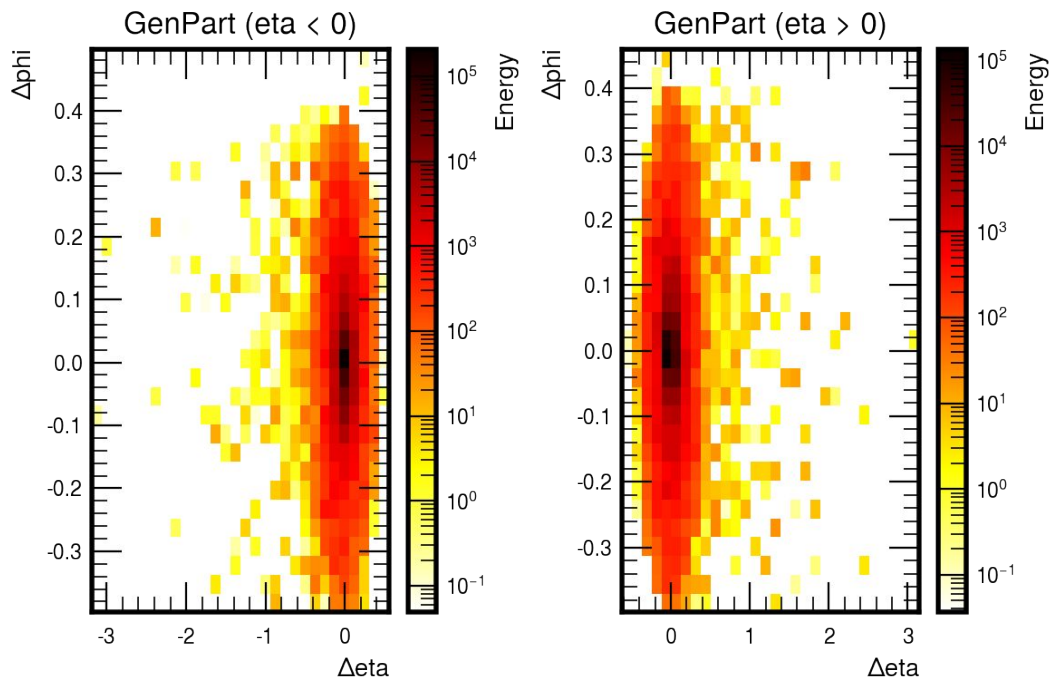
- 2: the current one
- 3: implementation ongoing, clustering directly the hits
- 4: with time (?)



**VBF / VBS  
tagger**

# physics motivation behind the tagger

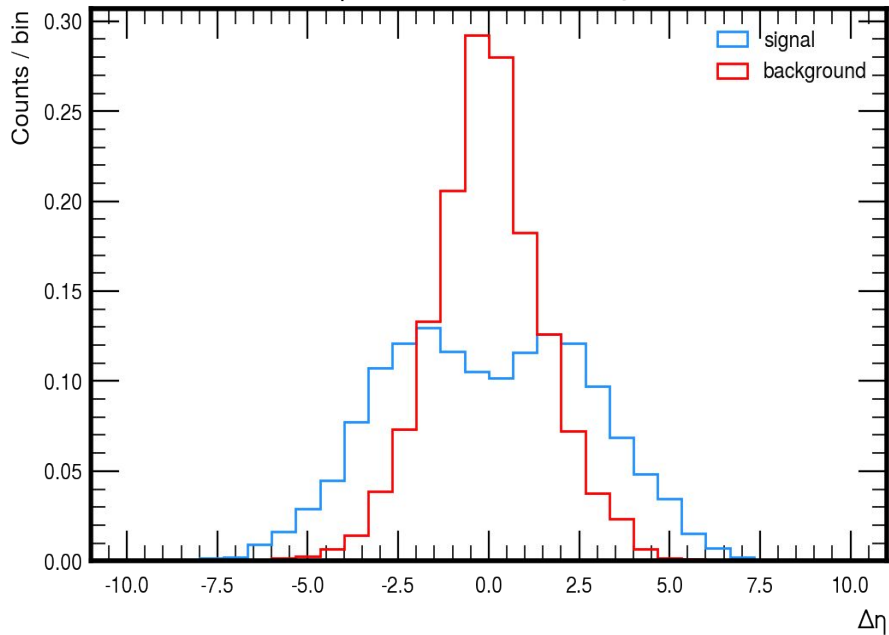
- based on the idea that the QCD activity between the tag jets and the beam is larger than in the space between jets
  - causes asymmetry in jets
  - from preliminary plots the asymmetry exists and is slightly more evident in the signal
- signal: VBF  
background: Drell Yan
- studies ongoing to find interesting variables for the tagger
  - jet pull
  - jet colour ring
  - Lund plane



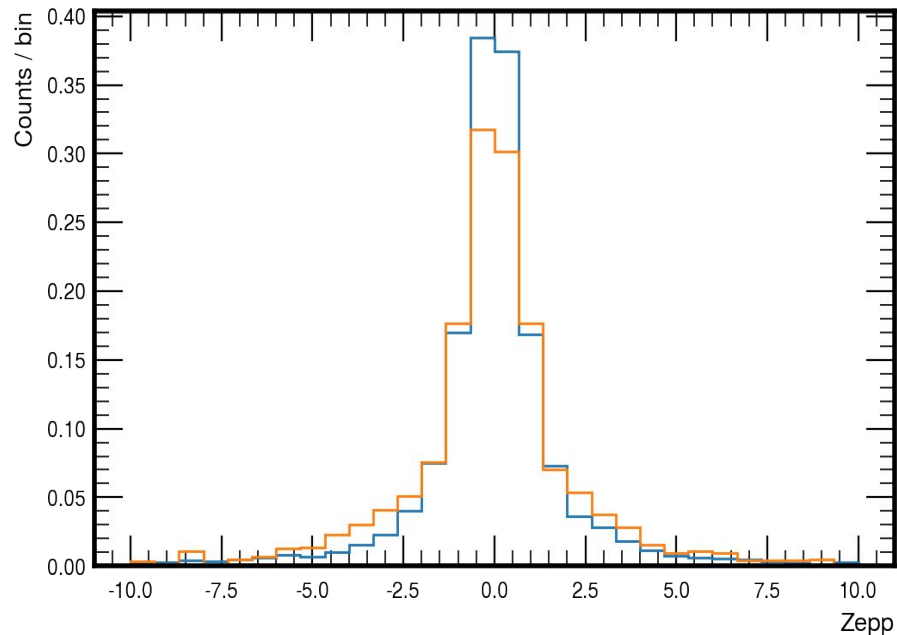


# (preliminary) comparisons

$\Delta\eta$  between the first two jets

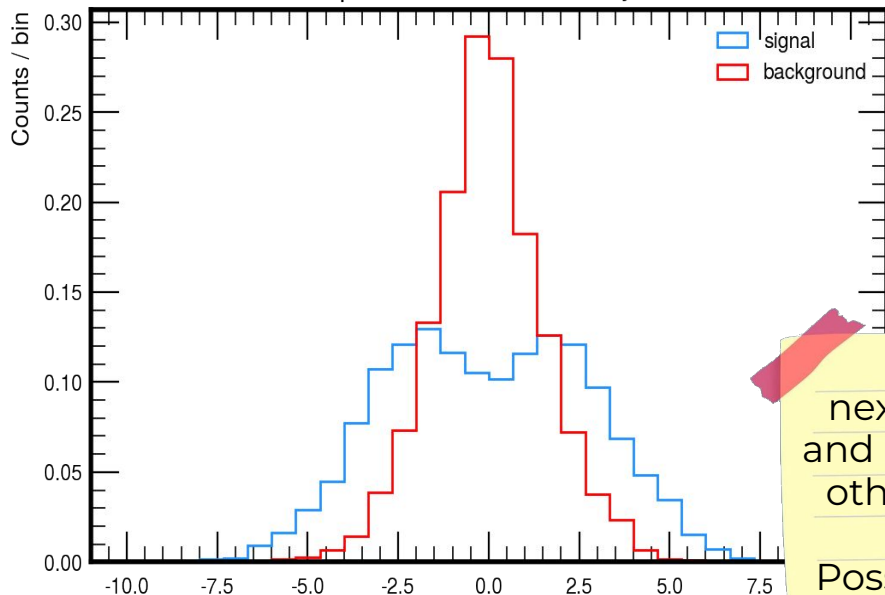


$$Zepp = \frac{\eta_j - \frac{\eta_1 + \eta_2}{2}}{|\eta_2 - \eta_1|}$$

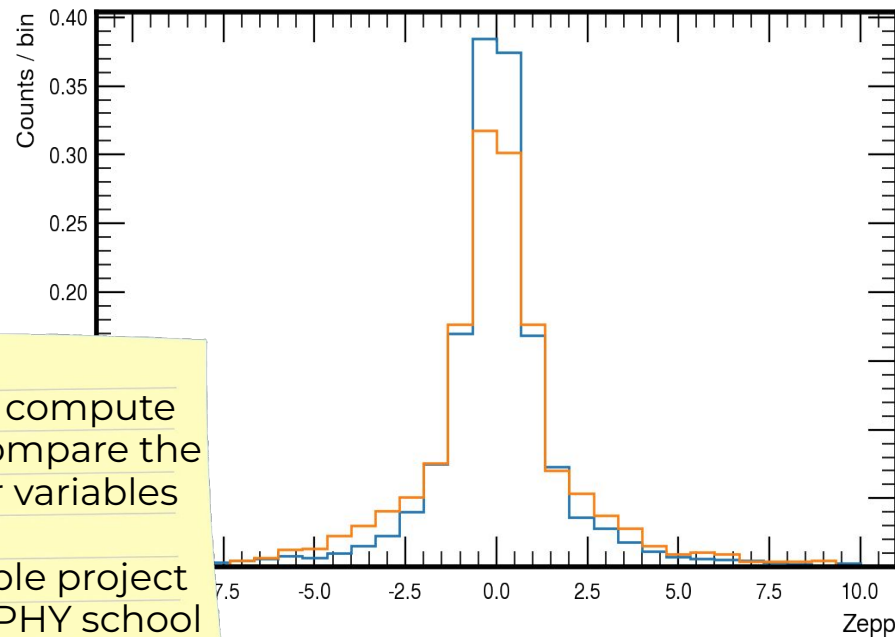


# (preliminary) comparisons

$\Delta\eta$  between the first two jets

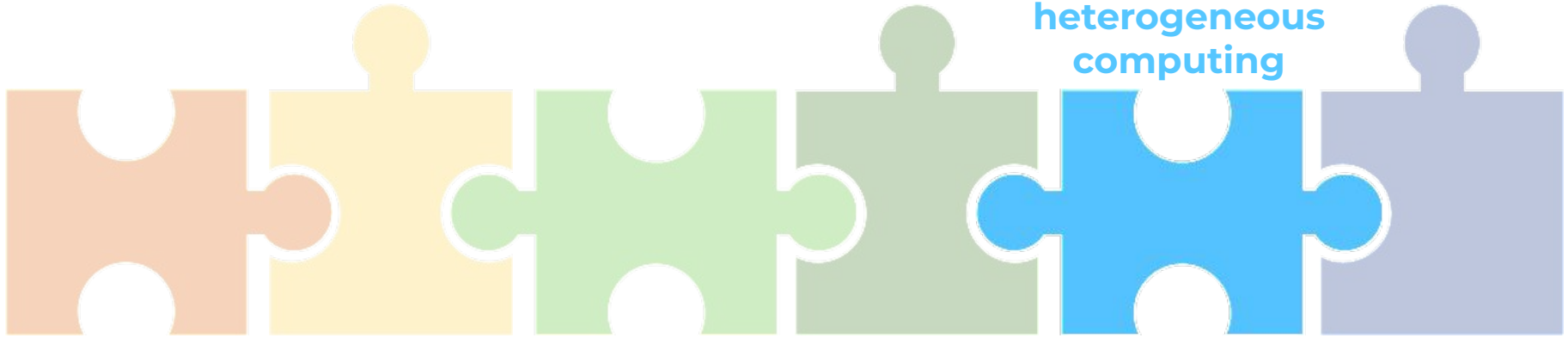


$$\text{Zepp} = \frac{\eta_j - \frac{\eta_1 + \eta_2}{2}}{|\eta_2 - \eta_1|}$$



next: compute  
and compare the  
other variables

Possible project  
for AI-PHY school



**heterogeneous  
computing**

# the alaska library

- is a header-only C++17 abstraction library for accelerator development
- aims to provide performance portability across accelerators through the abstraction of the underlying levels of parallelism
- has several backends:
  - serial and parallel for CPUs
  - CUDA for NVIDIA GPUs
  - HIP for AMD GPUs
  - oneAPI for CPUs, Intel GPUs and FPGAs



## latest updates:

v1.0.0: rewritten part of the oneAPI backend (USM, SYCL 2020)  
v1.1.0: warp-level primitives  
v1.2.0 (soon released): device global variables

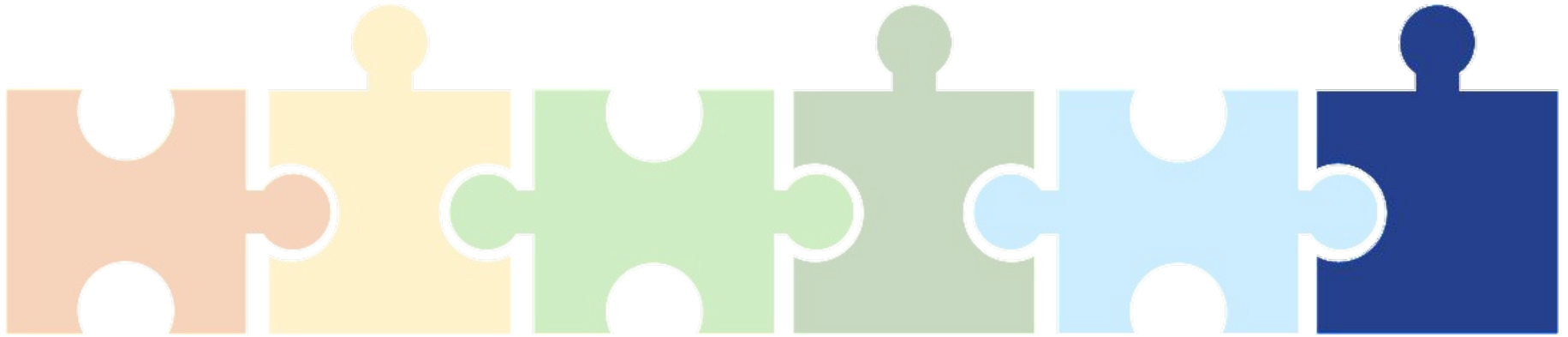
## future plans:

- add SYCL CUDA / HIP backend
- rewrite the shared memory implementation, currently very slow
- test it on LUMI supercomputer
- (maybe) try the FPGA backend

the oneAPI backend now is fully supported!

# heterogeneous computing in CMSSW

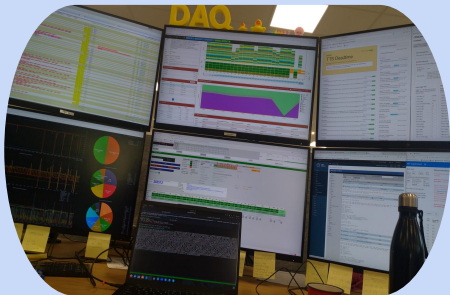
- the **CMSSW** framework already uses **alpaka** to achieve **performance portability** across CPUs and NVIDIA / AMD GPUs
  - the High Level Trigger farm is equipped with NVIDIA T4 and L4 GPUs
  - the fallback on CPU is automatic and transparent
  - the HLT reconstruction has been successfully tested on AMD GPUs
- with the complete **support** for **Intel GPUs** in alpaka, the plan is to extend their support also in CMSSW
  - ✓ write some simple tests in CMSSW with oneAPI and adapt the existing alpaka tests to use the oneAPI backend
  - (ask to ✓) implement in SCRAM the rules to compile for the oneAPI backend
  - add the support in CMSSW framework
  - test it!



**other  
activities**

# Other activities

central  
DAQ shifts



for CMS data  
taking

PhD course  
Scientific  
computing  
with Python

Tutor of  
“Laboratorio  
di Calcolo e  
Statistica” of  
the second  
year of  
bachelor

Conferences



Upgrade week  
CALOR