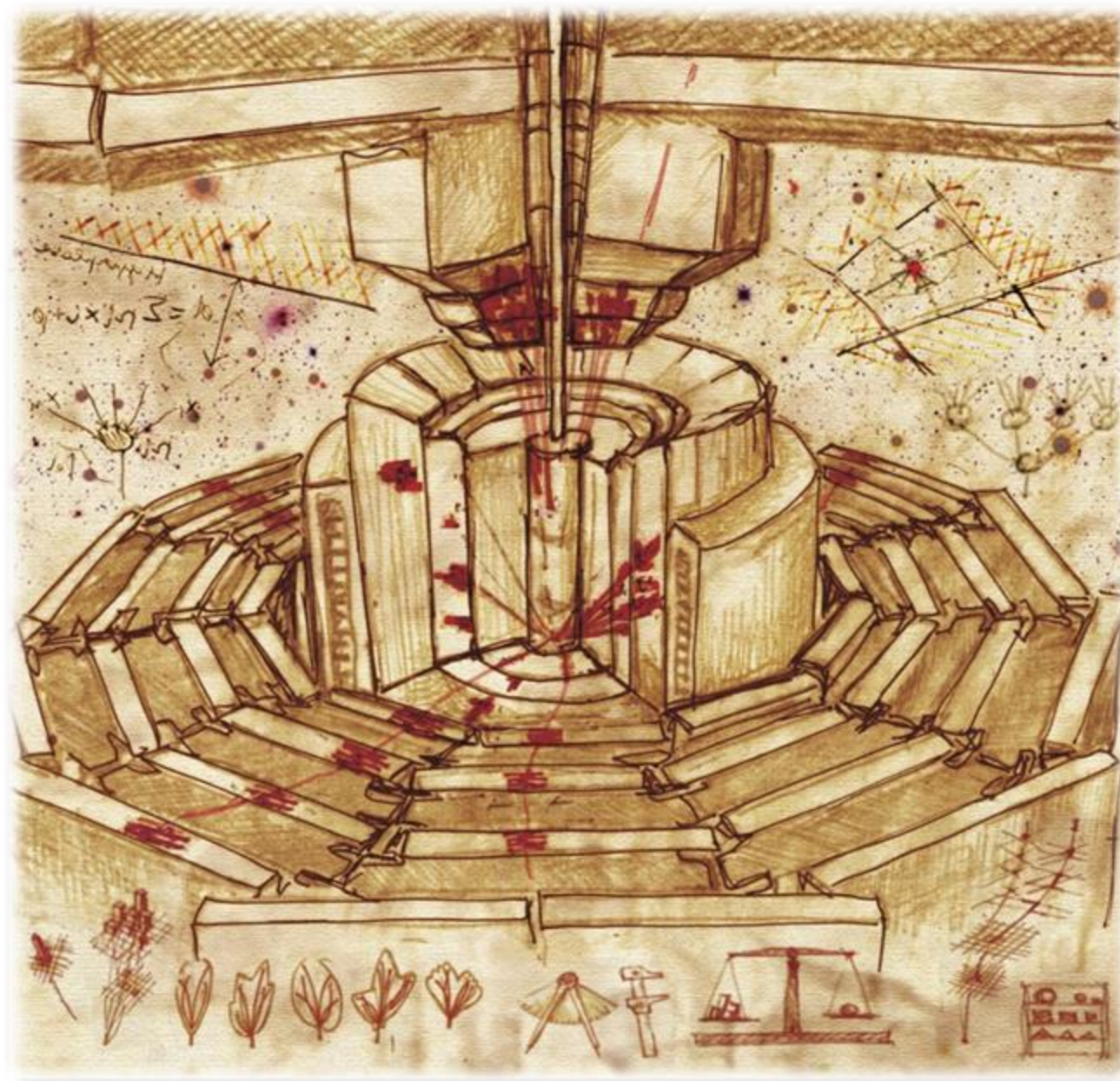


INnovative TRiggEr techniques for beyond the standard model Physics Discovery at the LHC

TWO YEAR JOURNEY



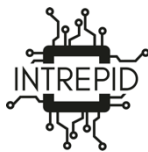
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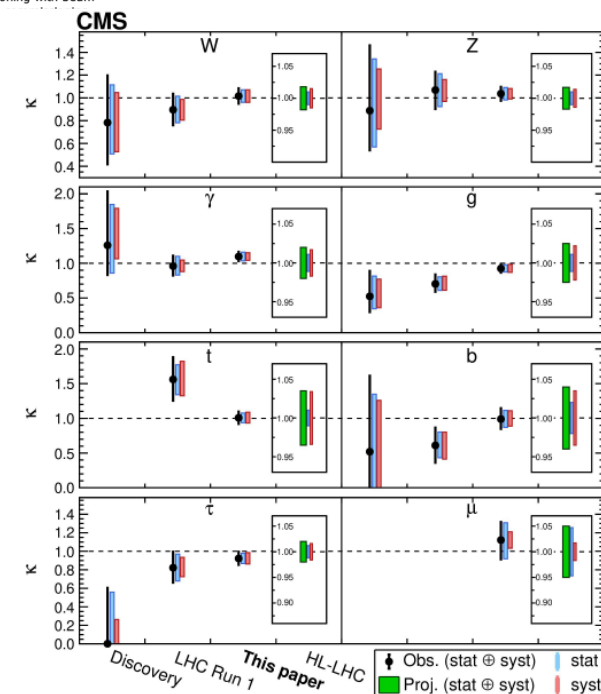
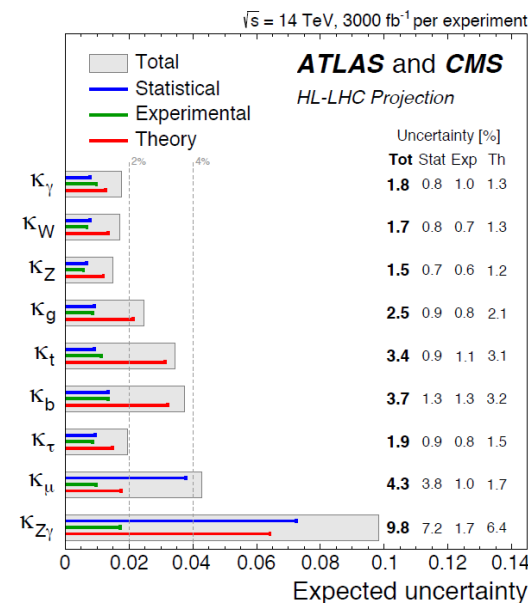
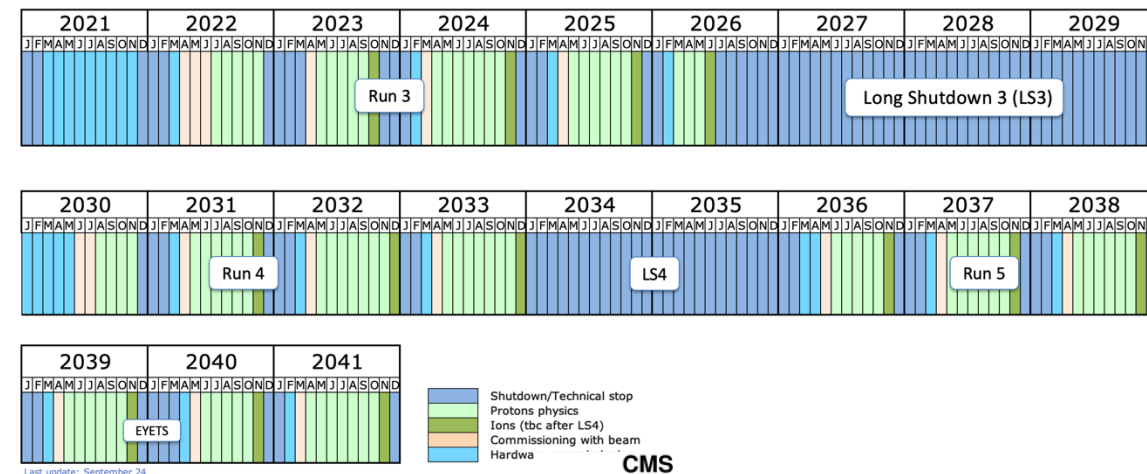
European Research Council
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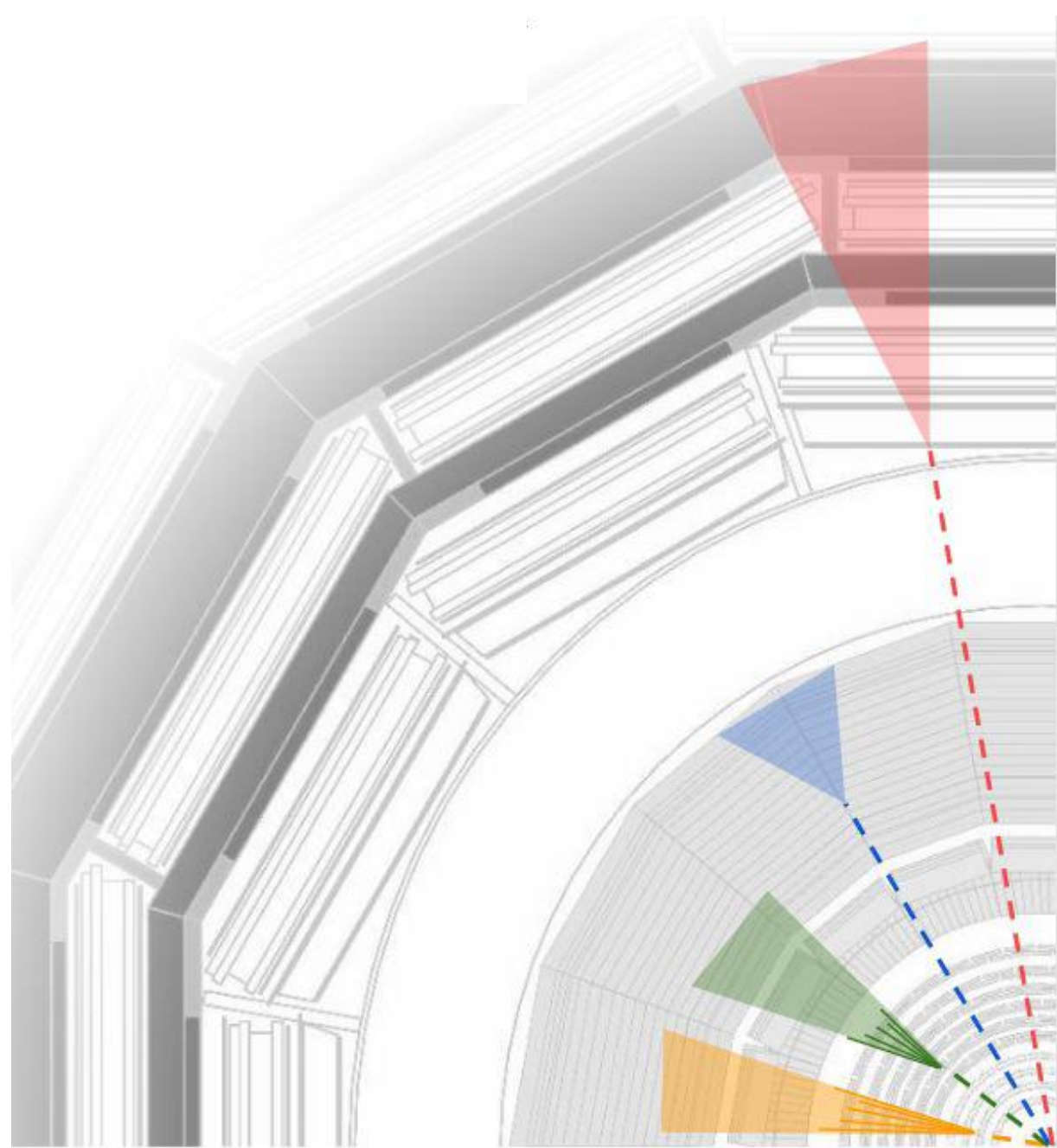
Towards the HL-LHC

- **Preparing for the big upgrade** of the LHC detectors, starting 2026.
- HL-LHC upgrade offers an **unprecedented opportunity** to explore uncharted lands and achieve scientific progress.
 - 8 times more data to what we will have by the end of Run 3 will facilitate a rich physics program.
- **Improve current understanding of the SM and Higgs** sector by improving existing precision measurements and accessing rare decays ($H \rightarrow \mu\mu$) or production modes (HH) previously unseen at the LHC.
 - Search for deviations at high momenta (i.e. Effective Field Theories) or
- **Extend reach of new physics searches:** unexplored signatures (LLPs, HSCPs...) or regions of the phase-space will be within reach.



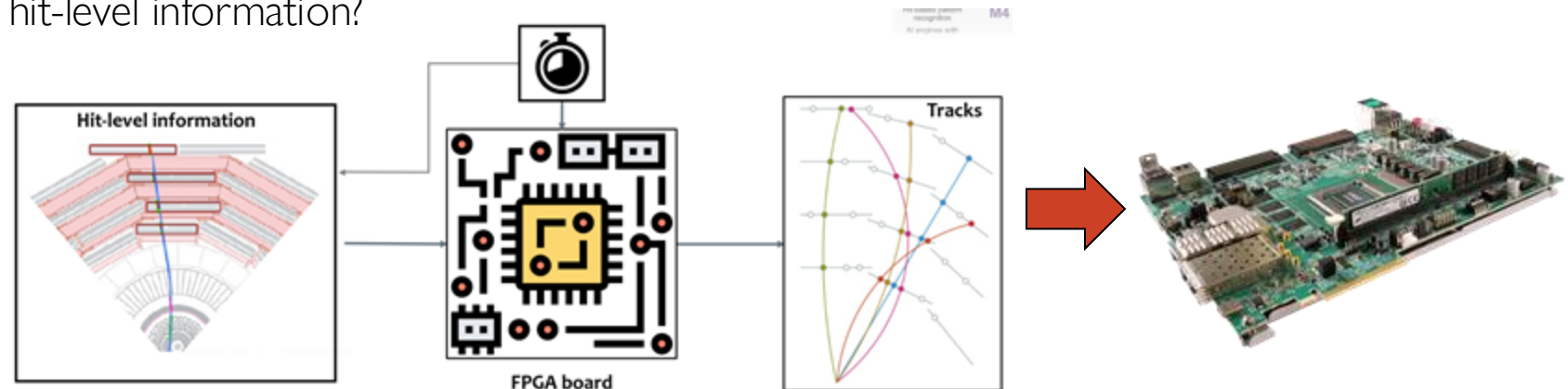
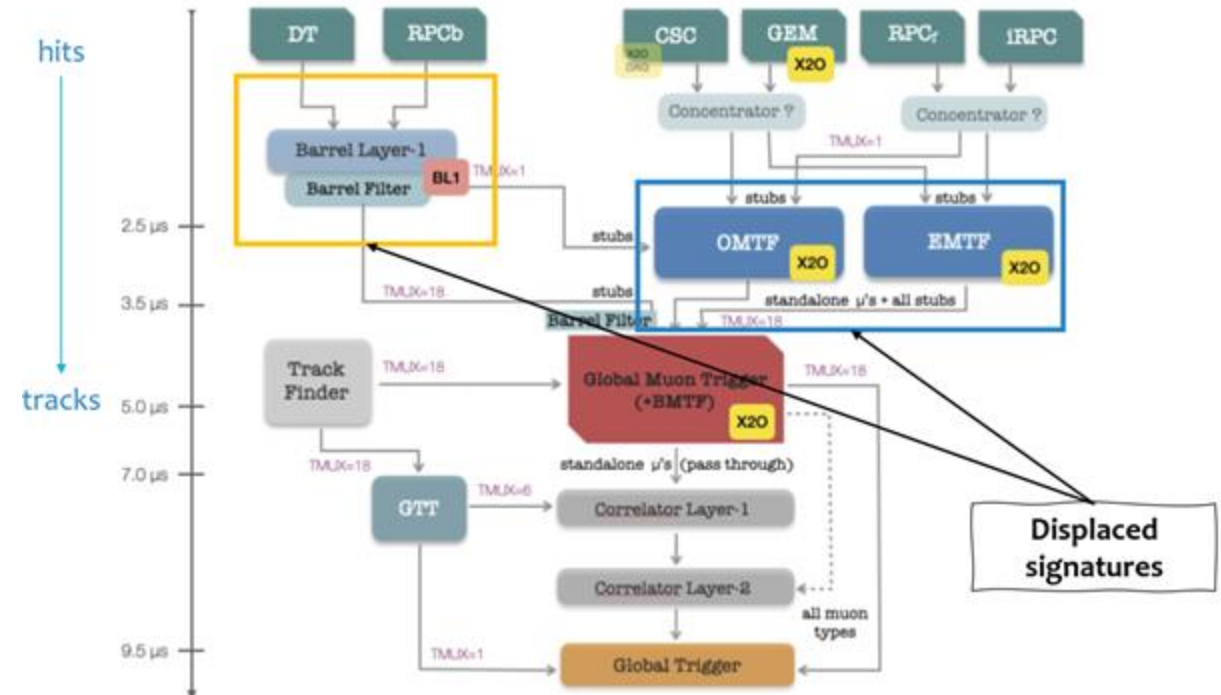
Yes but...

- **Prompt-Optimized Triggers Miss LLPs:** Current (Phase-1) L1 trigger system has been designed to select prompt objects from the primary vertex.
 - Searches are (mostly) limited for the L1 acceptance and they need to rely on inclusive triggers such as MET or cross-triggers. Some displaced muon trigger exist in Run-3
- Many models predict the existence of **long-lived particles** (LLPs)
 - Many Exotic scenarios not envisioned when the trigger system was being designed!
- LLPs transit layers at later times, timing information
- LLPs decay far from the interaction point and show displaced signatures
 - Dedicated trigger paths exploiting unique features
 - Displaced jets in the tracker, calorimeters, or muon systems
- **Run 3 is the perfect benchmark for “crazy” ideas for HL-LHC**

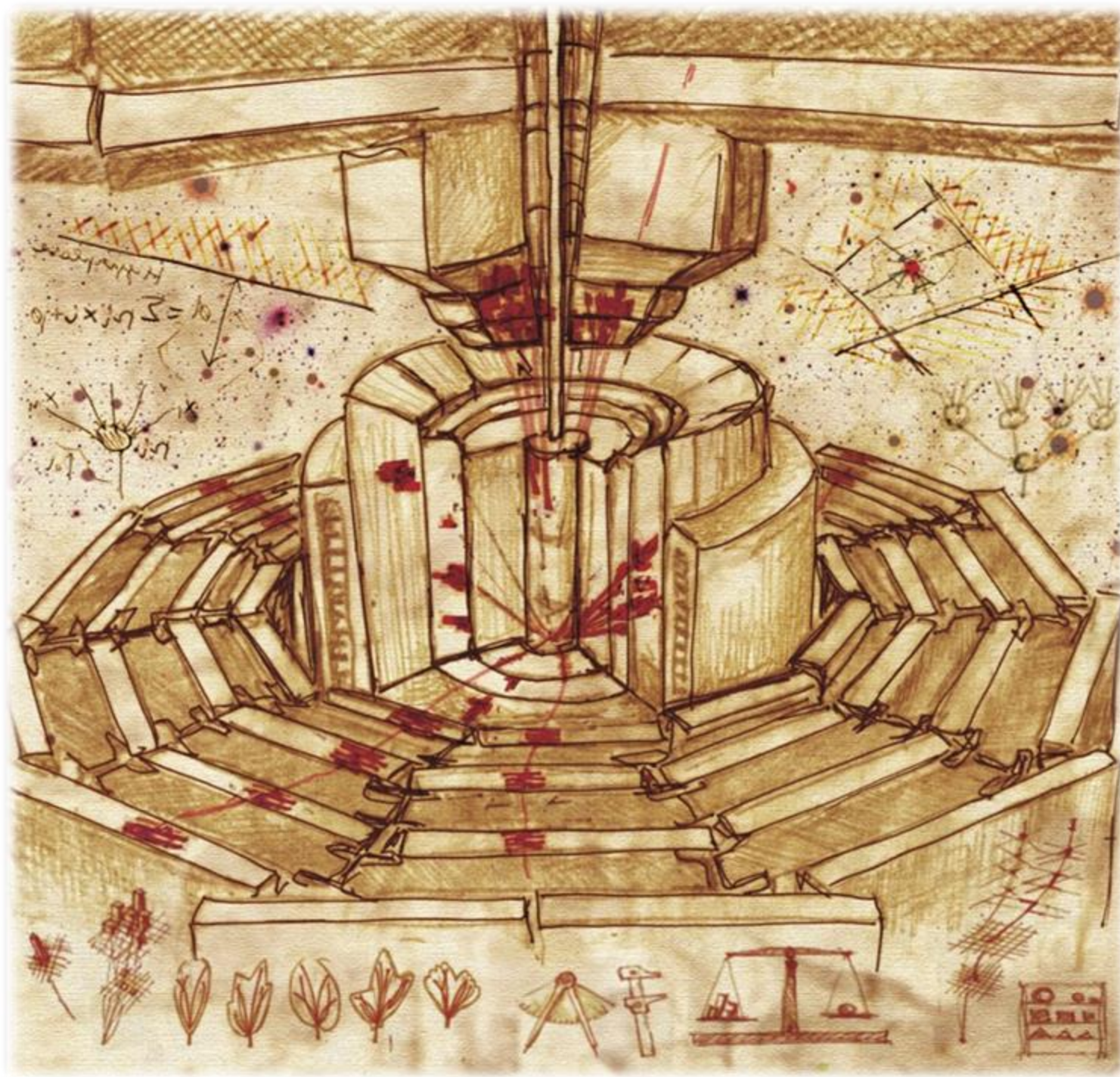


What are the goals?

- LLP signals might be easily missed or misinterpreted in LHC data.
- Optimizing the techniques within the current architecture to improve detection efficiency without requiring significant hardware upgrades.
- Explore alternative technologies and ideas which could not be otherwise investigated that could potentially lead to a significant breakthrough.
 - Can we profit from using hit-level information?
 - Use of 7nm technology?



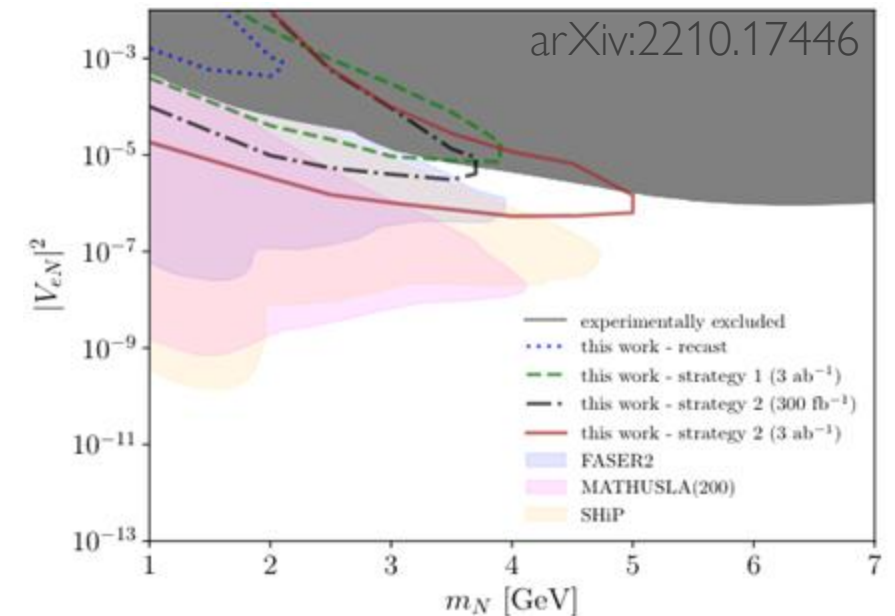
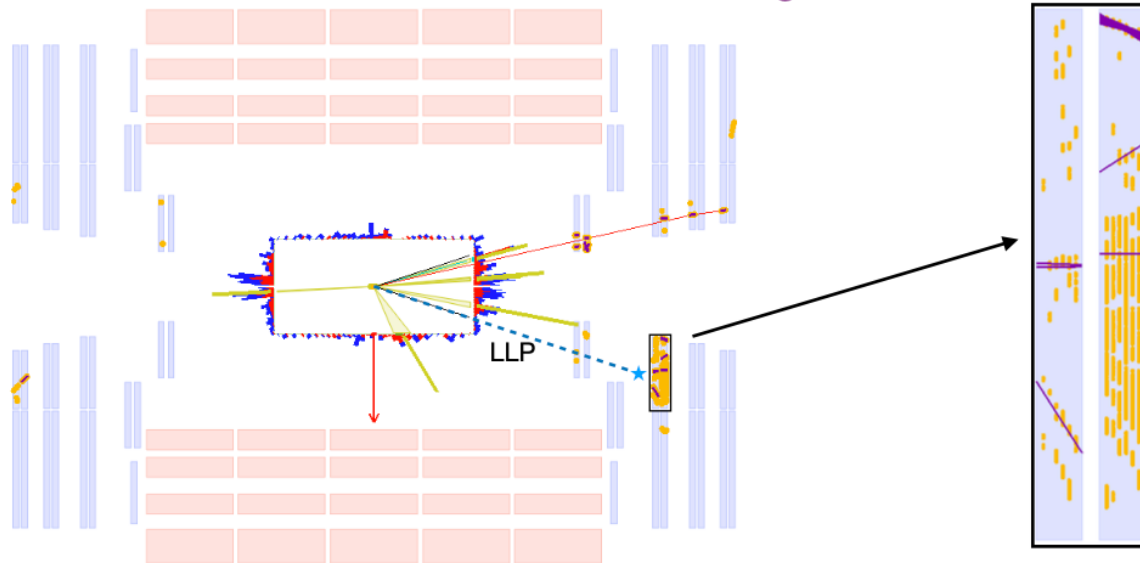
Using existing architecture



Displaced jets in the muon system

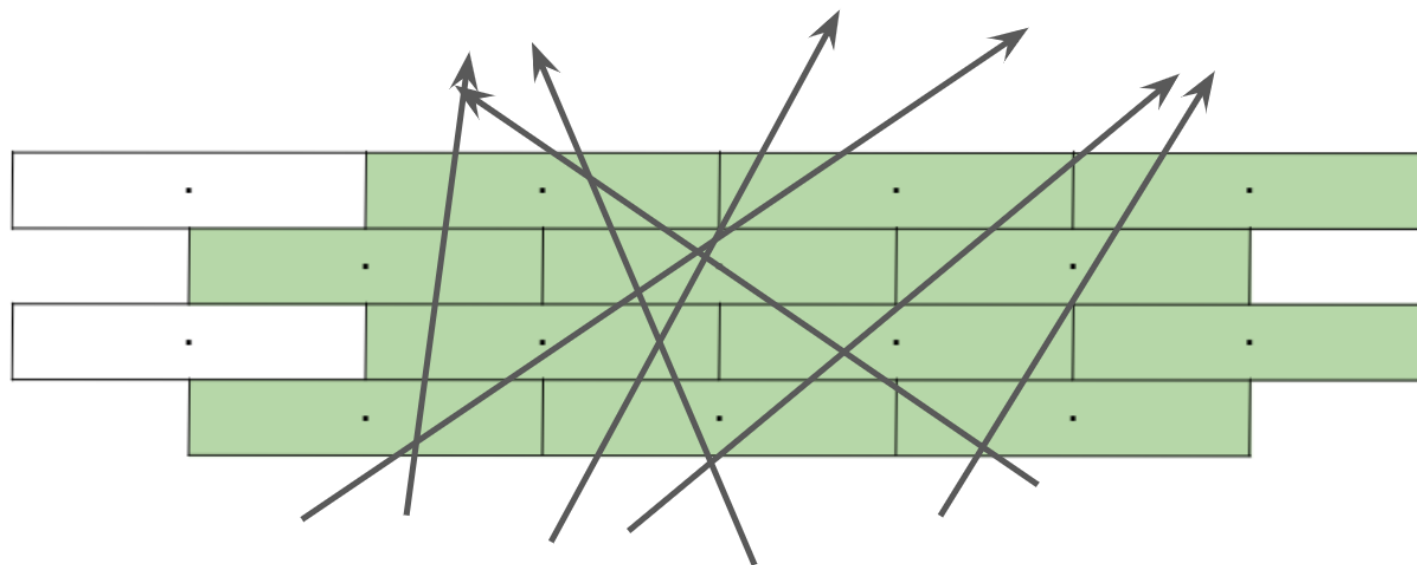
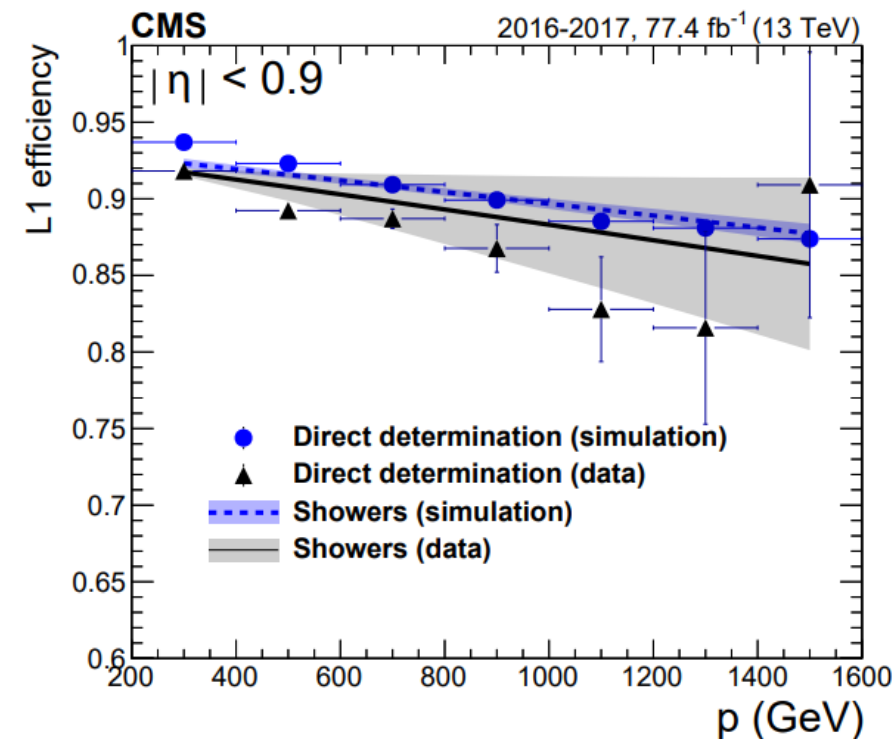
- Long-lived particles such as HNL could decay in the muon system if sufficient long lifetime (or low mass)
 - Hadronic showers in the muon system
 - Probe lower masses (< 10 GeV) or longer lifetimes $O(1\text{m})$ parameter space
- Nowadays this signatures escape detection due to reconstruction and trigger constrains
- Unprecedented opportunity for new physics discovery

~1100 rechits & 33 segments in ME-2/1



Triggering on muon showers

- Shower are detected as multiple hits in either the CSC or DT chambers
- Steel between muon stations can act as absorbers in a sampling calorimeter
 - Shielding of 12-27 interaction length (Background suppression factor $\sim 10^7$)
 - Unique feature of CMS muon system
- Sensitive to LLP with longer $c\tau \sim (1 - 10 \text{ m})$
- Working algorithm on the CSC (right) and extending it on the barrel for phase-2



Triggered on October 8th
2022

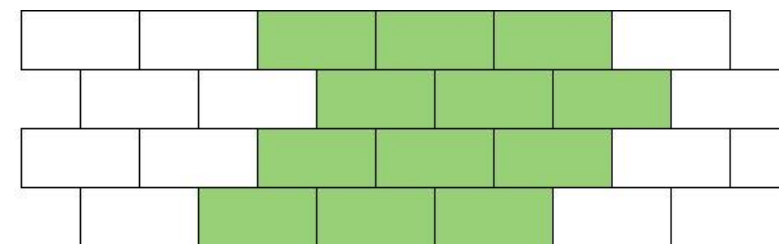
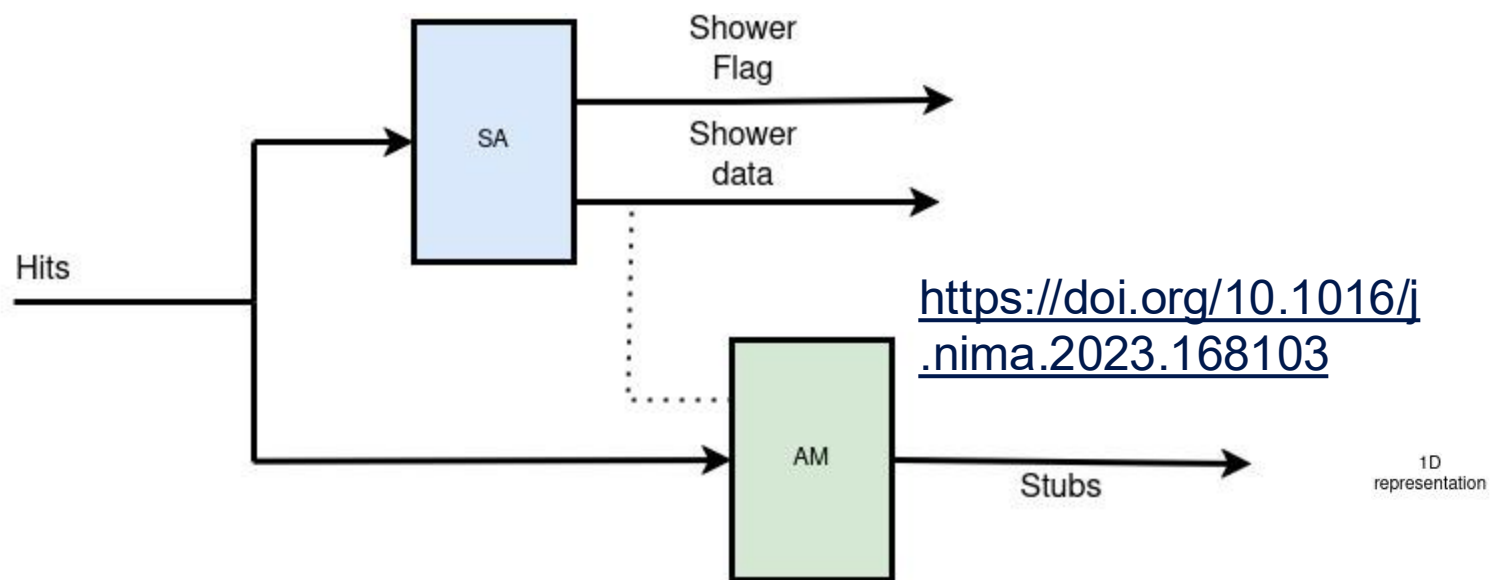
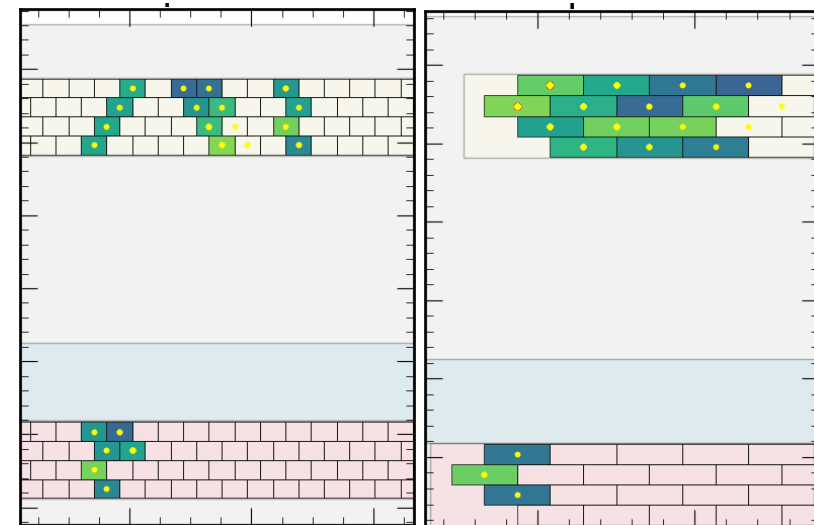


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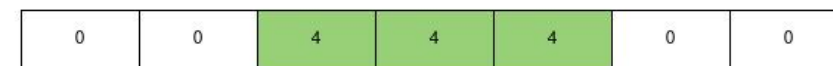


Triggering on muon showers

- Each hit is stored for 16BXs (400ns) to account for their drift times
- Above the threshold, hits are stored, and the station is labeled as a shower: BX, widthh, and 1D representation of n° hits.



1D
representation

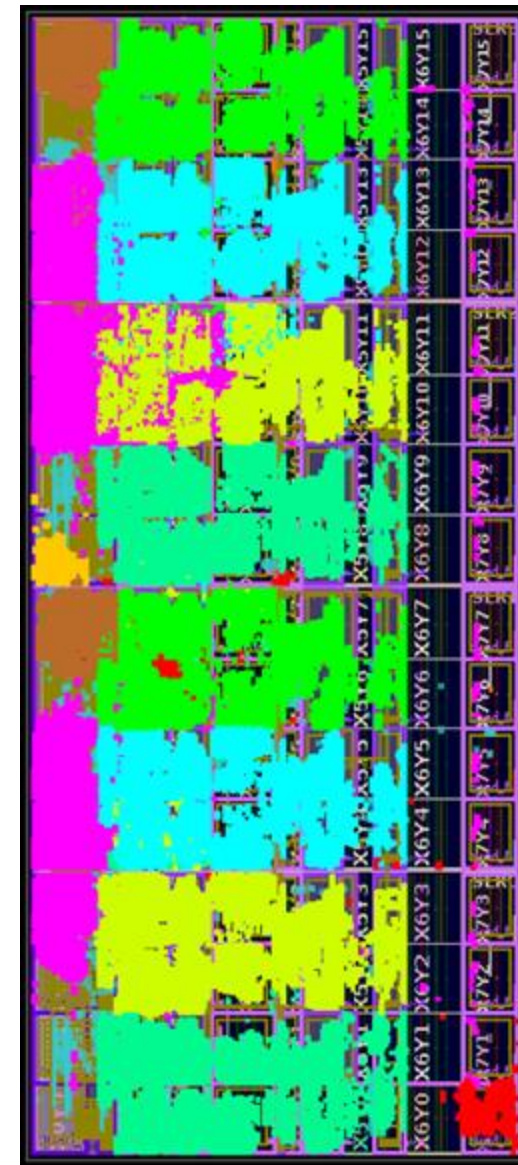
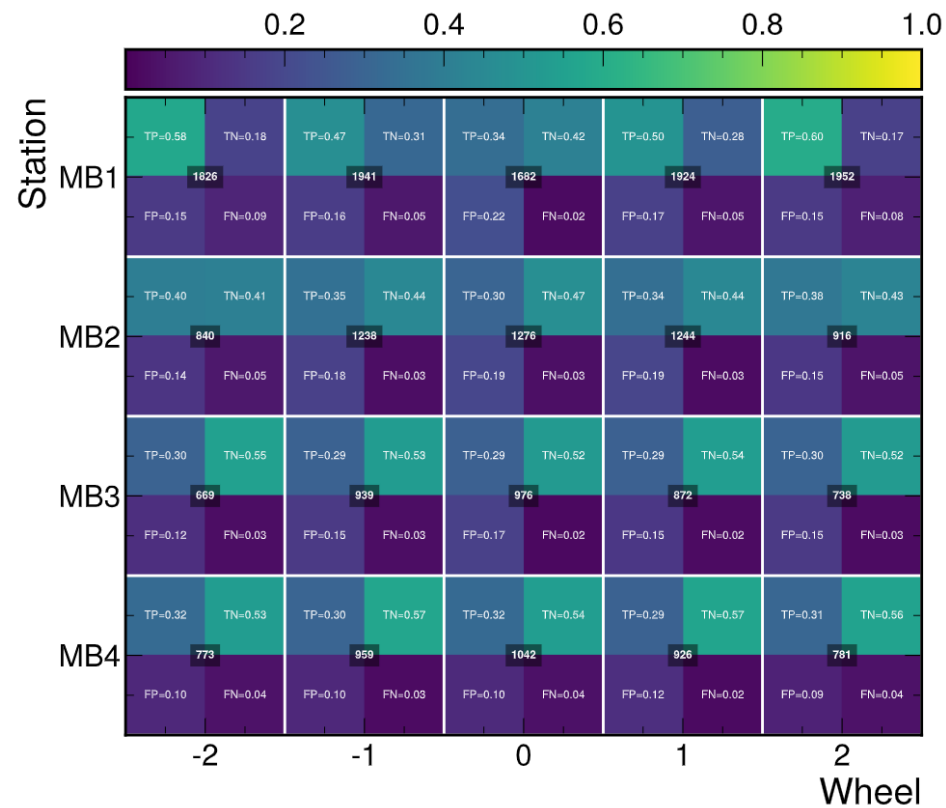
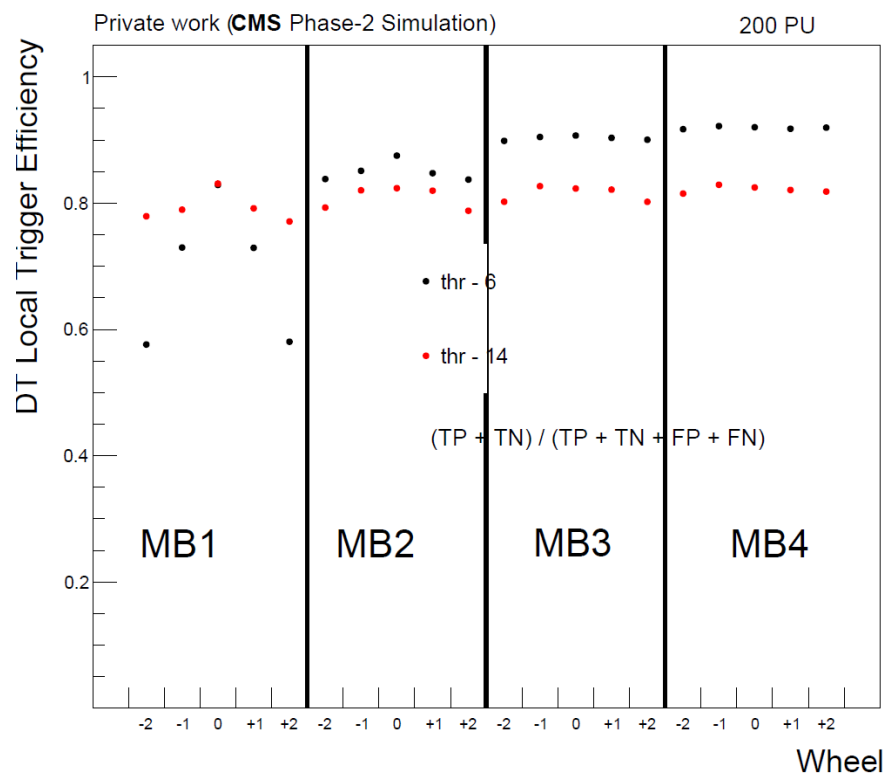


<https://doi.org/10.22323/1.476.0930> (ICHEP 2024)

<https://iopscience.iop.org/article/10.1088/1748-0221/20/03/C03024/pdf>

Triggering on muon showers

- A preliminary firmware algorithm has been developed, with less than 3K LUTS and less than 3K registers for each SuperLayer @480MHz
- Working with CIEMAT to integrate the algorithm with the current firmware and test in on the slice test this year.

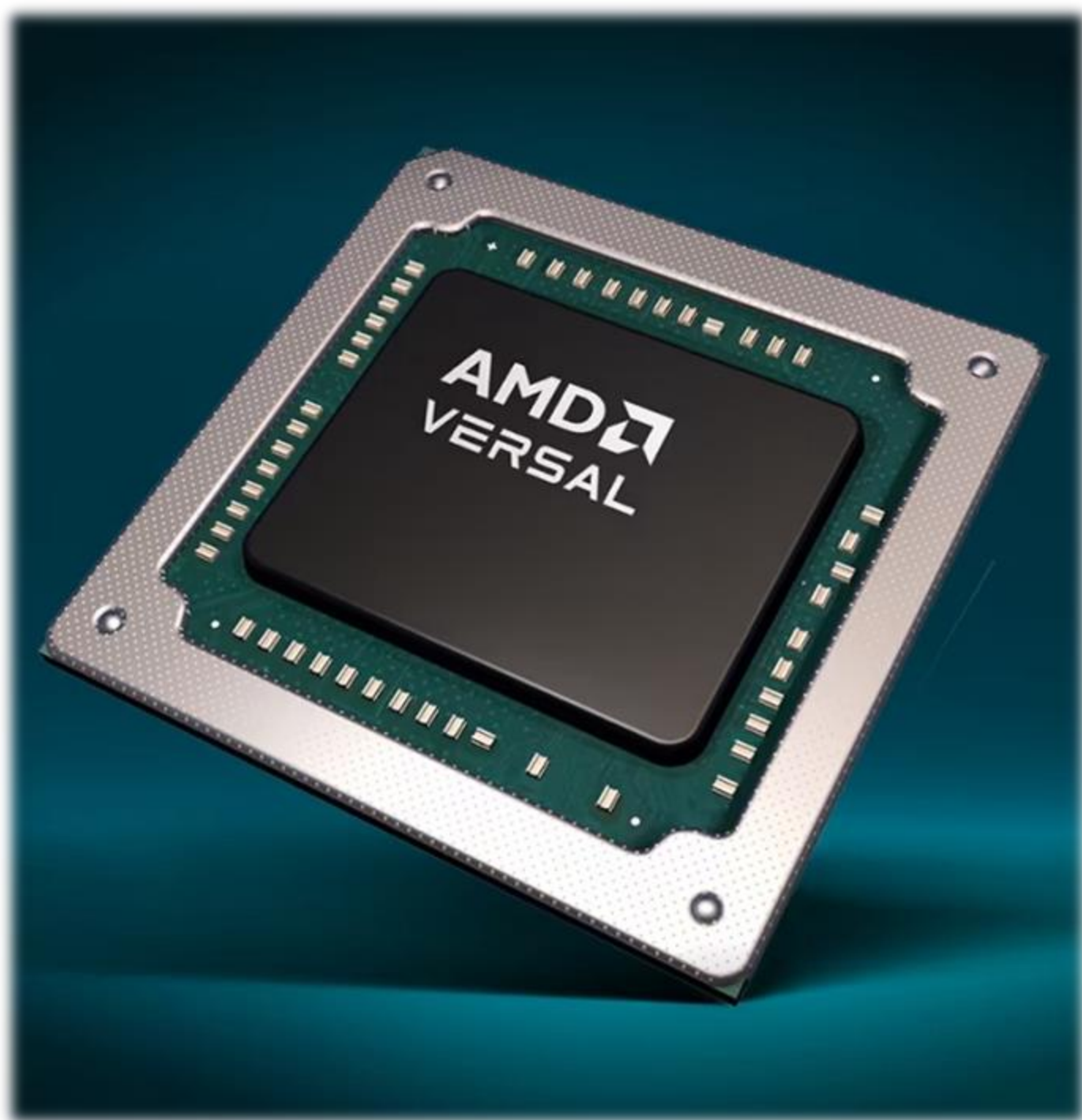


Contributing to the X2O manufacturing

- Gaining expertise in hardware assembly and testing to provide long-term support to the project.
- Javier Prado is now at UFL helping to assemble the first X2O batch.



What else we can do?



Displaced muons

- Muon detector are located 4-10m away from the interaction point. Unique signature to search for new physics.
- First result in Run-3 data (13.6 TeV) with a strong Spanish contribution.

First Run 3 result: Displaced dimuons at 13.6 TeV



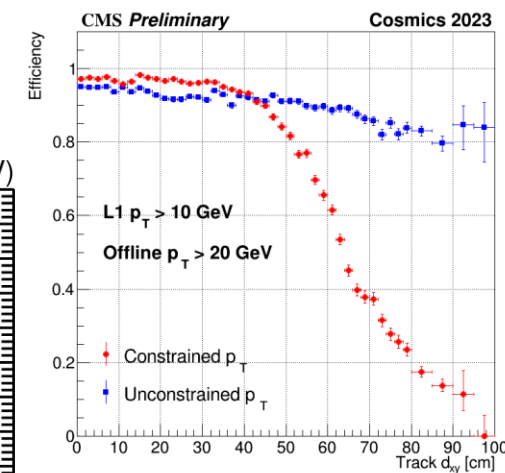
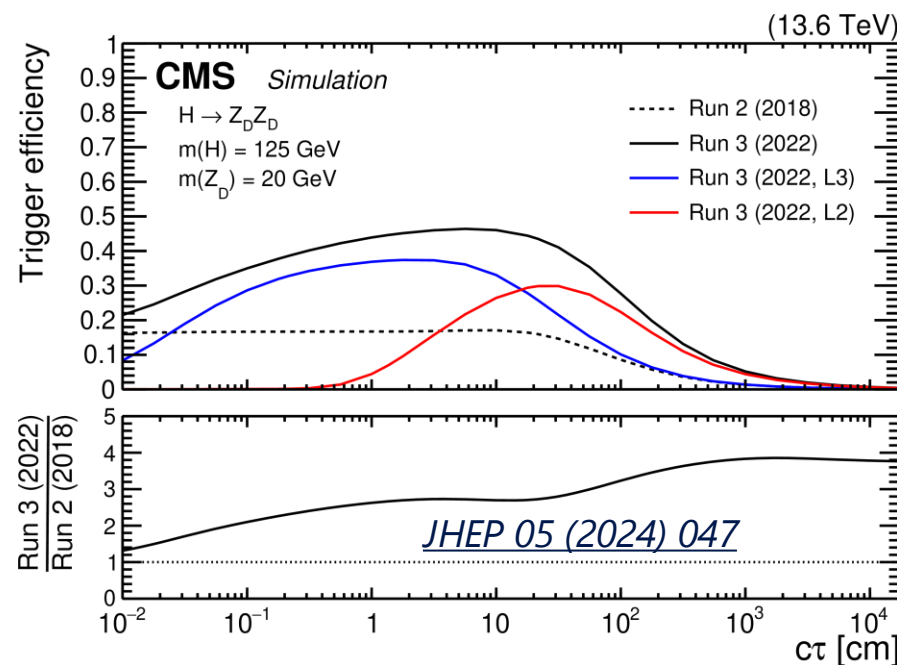
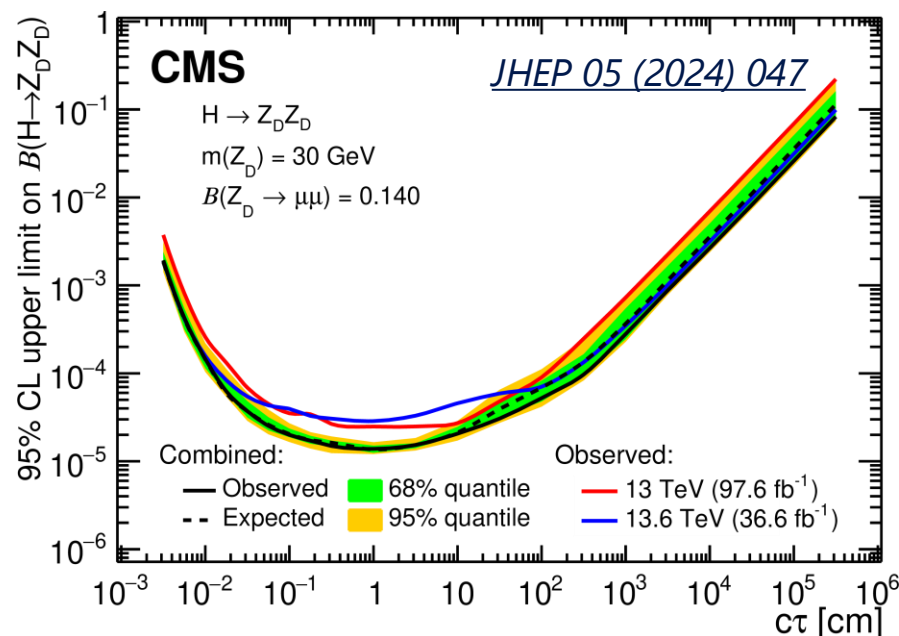
<https://cms.cern/news/long-lived-particles-light-lhc-run-3-data>

<https://home.cern/news/news/physics/cms-collaboration-cern-presents-latest-search-new-exotic-particles>

With a strong Spanish contribution:



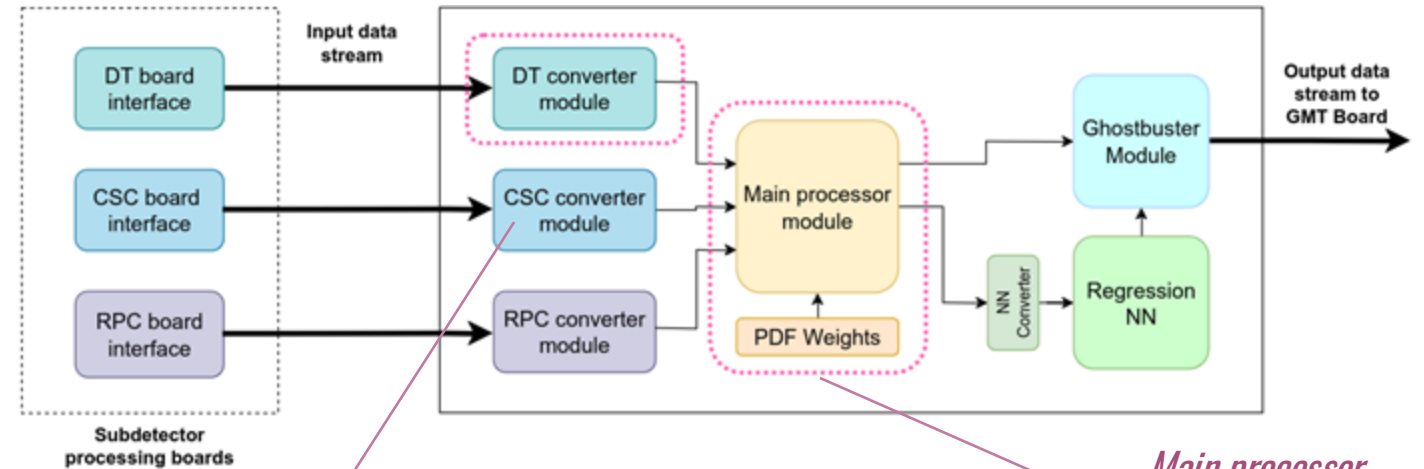
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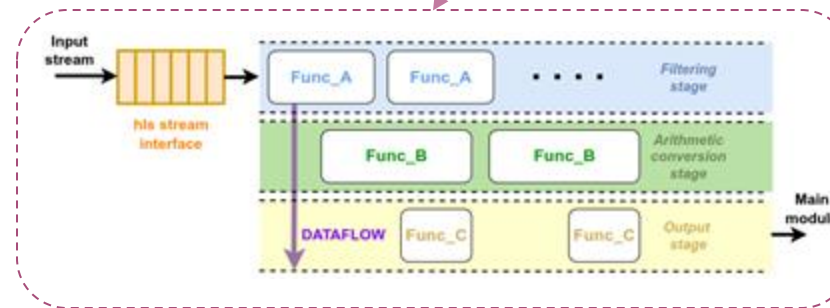
Firmware development

- Reduction of data load by implementing a zero-suppression approach (Batch processing instead of storing all the possible input data for each single event)
- We take advantage of the streamed inputs coming from the different subdetectors.

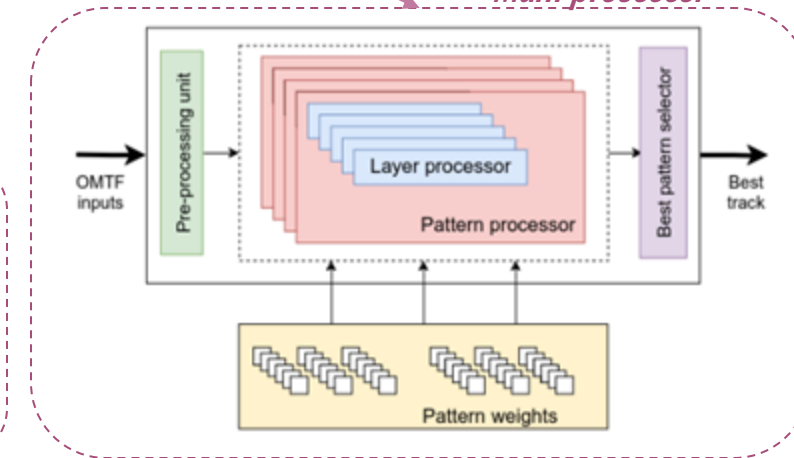
General scheme



Input converting modules



Main processor

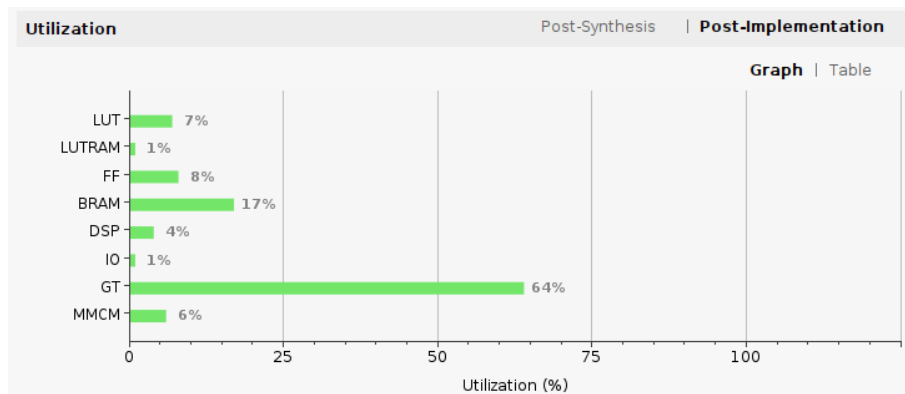
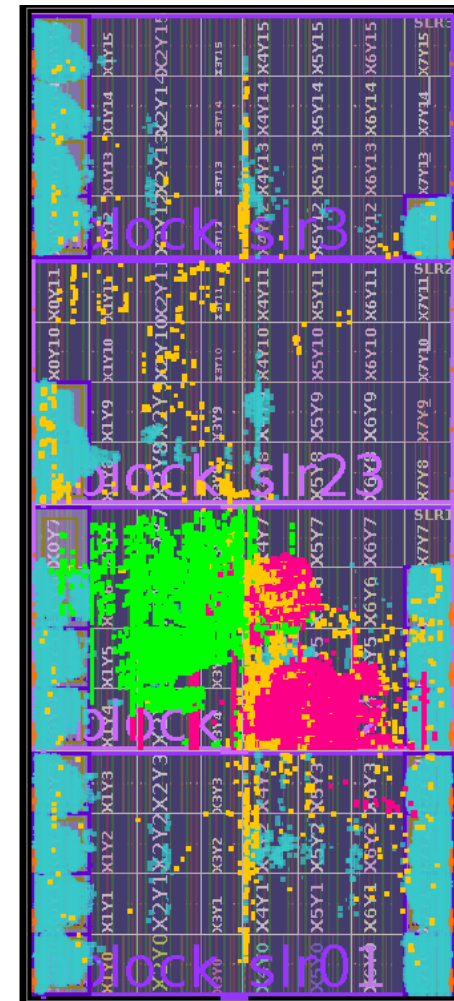
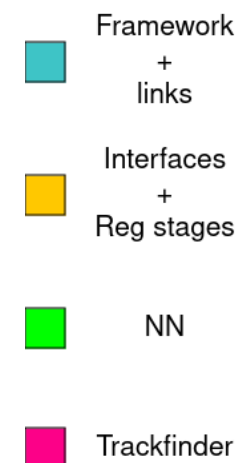


- > Manage streamed data.
- > Parse data frames.

- > Parallel “golden pattern” processing.
- > Stored weights reshaped for simultaneous availability.

Firmware development

- Latest implementation of the OMTF Phase-2 firmware on VU13P, including the 65 CSC links + 15 DT instantiated.
- No module is forced to be placed in a particular region
- Getting ready to do Slice tests at 904

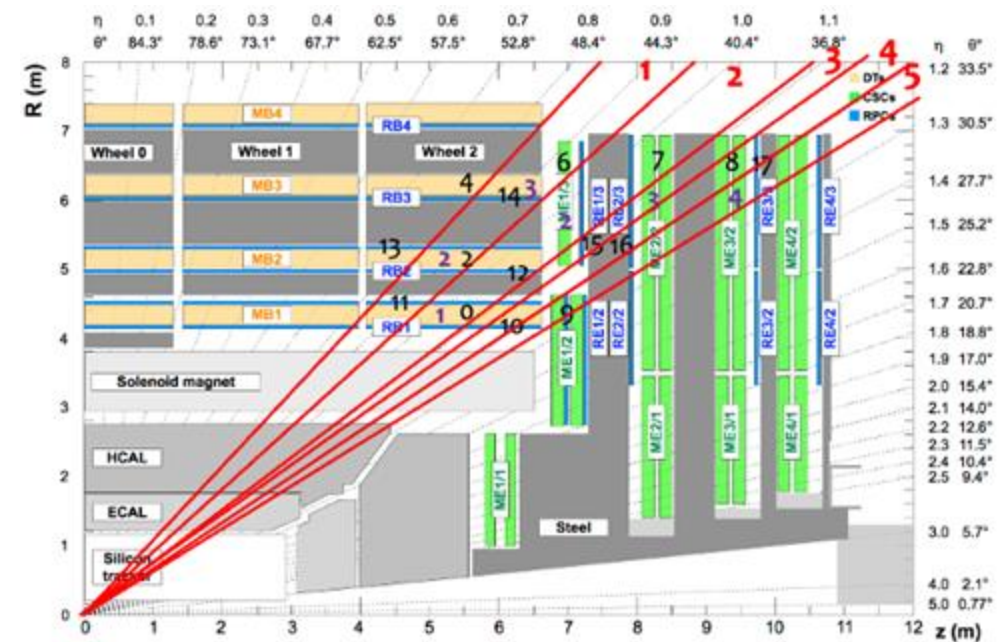
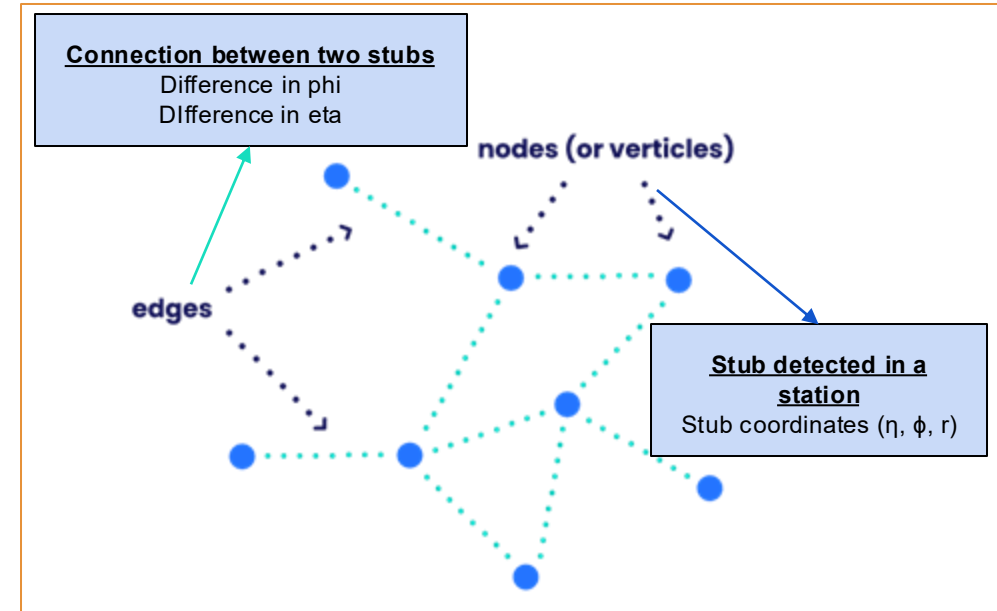


Resource	Utilization	Available	Utilization %
LUT	113848	1728000	6.59
LUTRAM	4522	791040	0.57
FF	280379	3456000	8.11
BRAM	447.50	2688	16.65
DSP	450	12288	3.66
IO	2	448	0.45
GT	82	128	64.06
MMCM	1	16	6.25

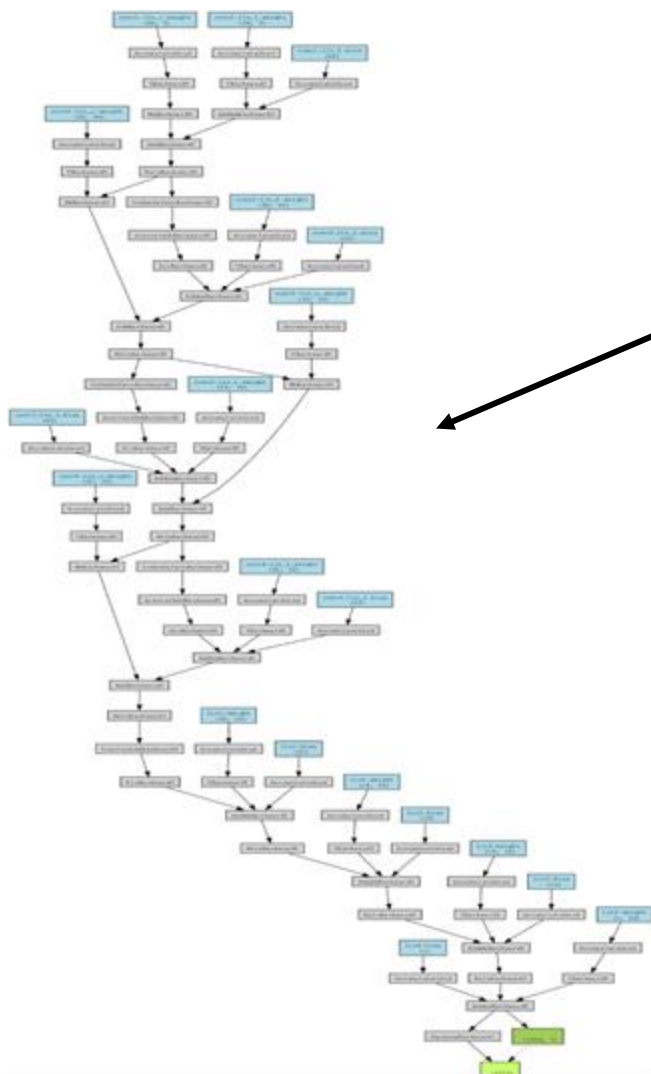
Conceptual design

Designing a basic network:

- Each node is a stub on a detected layer (DT, RPC, CSC).
- Edges represent the difference in η - ϕ between stubs.
- For now only regress momentum, either $q \cdot p_T$ or q/p_T
- Target OMTF as it has the most complex detector topology



Model configuration. Designing a “basic” network

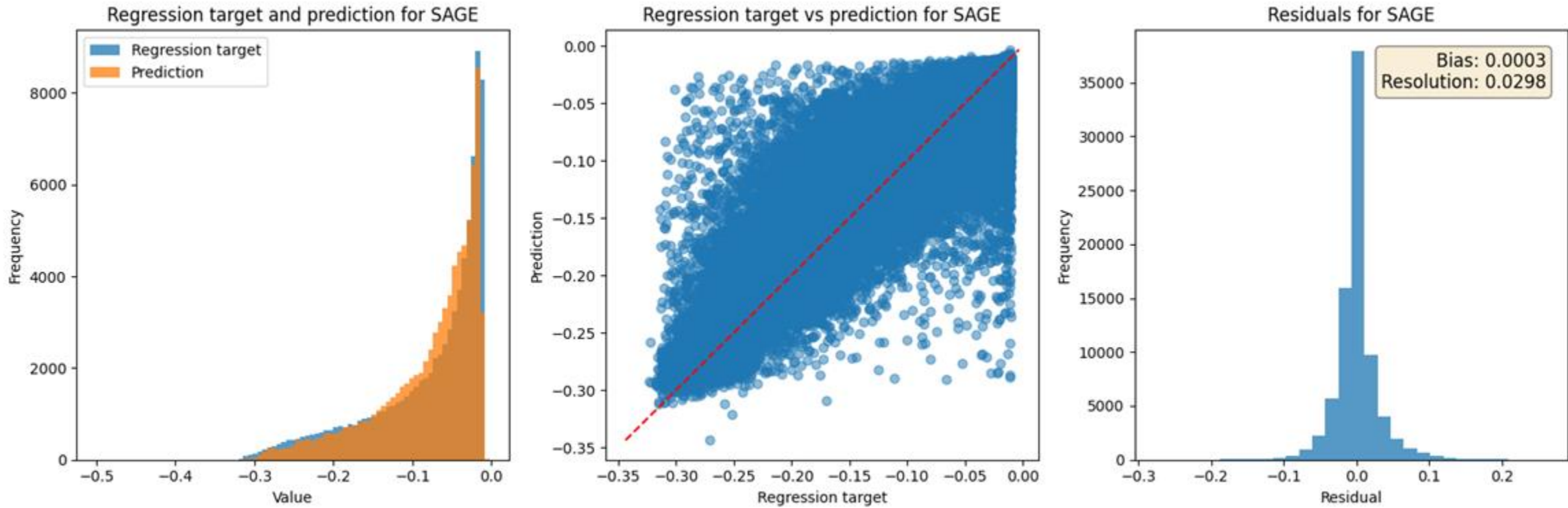


```

GraphSAGEModel(
  (conv1): SAGEConv(3, 128, aggr=mean)
  (conv2): SAGEConv(128, 64, aggr=mean)
  (conv3): SAGEConv(64, 64, aggr=mean)
  (conv4): SAGEConv(64, 64, aggr=mean)
  (lin1): Linear(in_features=64, out_features=64, bias=True)
  (lin2): Linear(in_features=64, out_features=32, bias=True)
  (lin3): Linear(in_features=32, out_features=32, bias=True)
  (lin4): Linear(in_features=32, out_features=1, bias=True)
)
MPLNNRegressor(
  (conv1): MPL()
  (conv2): MPL()
  (conv3): MPL()
  (conv4): MPL()
  (lin1): Linear(in_features=128, out_features=128, bias=True)
  (lin2): Linear(in_features=128, out_features=16, bias=True)
  (lin3): Linear(in_features=16, out_features=16, bias=True)
  (lin4): Linear(in_features=16, out_features=1, bias=True)
  (lin5): Linear(in_features=128, out_features=128, bias=True)
  (lin6): Linear(in_features=128, out_features=16, bias=True)
  (lin7): Linear(in_features=16, out_features=16, bias=True)
  (lin8): Linear(in_features=16, out_features=1, bias=True)
  (global_att_pool1): AttentionalAggregation(gate_nn=Sequential(
    (0): Linear(in_features=64, out_features=1, bias=True)
  ), nn=None)
  (global_att_pool2): AttentionalAggregation(gate_nn=Sequential(
    (0): Linear(in_features=64, out_features=1, bias=True)
  ), nn=None)
)
  
```

Results

- Trained with a reduced sample (~1M events) in 20 epochs (model converges in 10 or so) with all possible connections in place
- Different configuration tested for now, early stopping, and dropout layers are now in place.
- Re-training with displaced samples and with flat-pt prompt samples to avoid favoring higher p_T estimation.

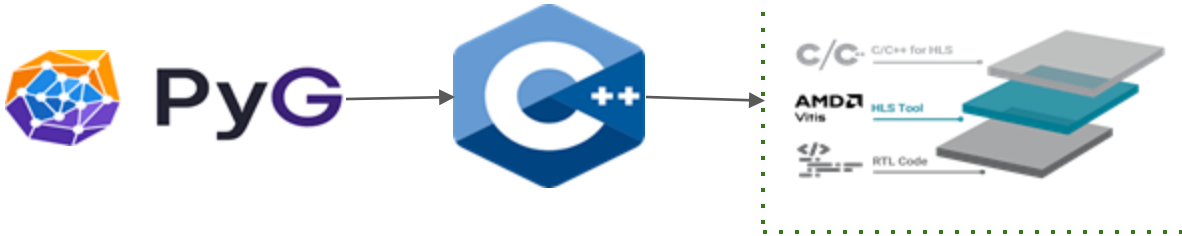


GNN

Status

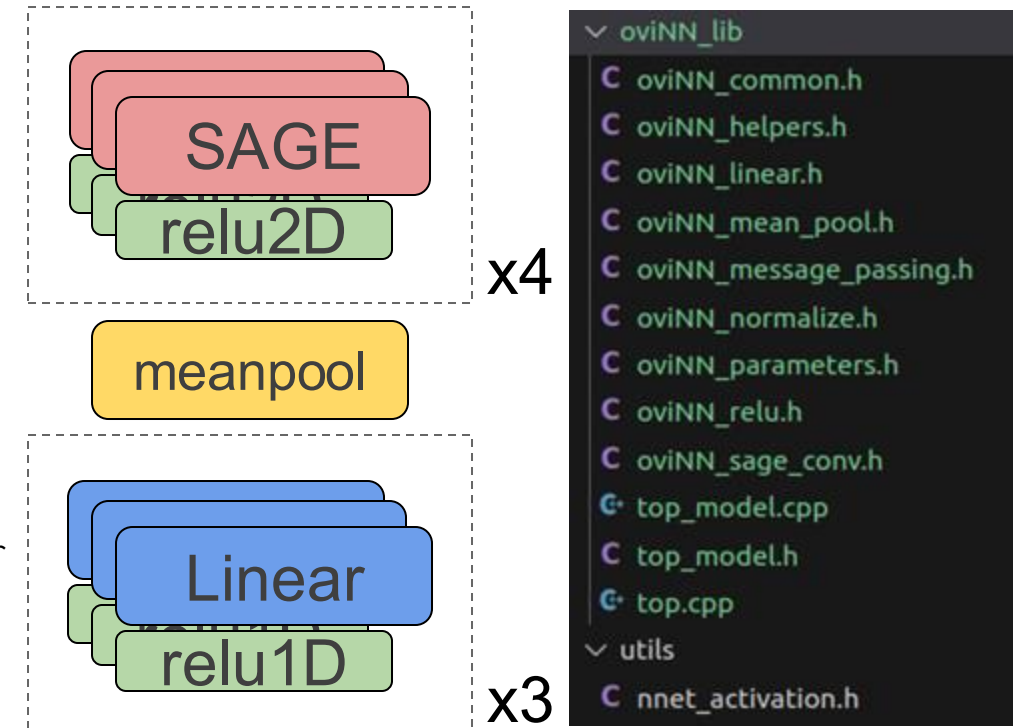
- Things achieved:
 - All modules defined
 - Message Passing (Mean & Sum agg)
 - Linear
 - Normalizer
 - ReLu
 - SAGE
 - Mean Pool
 - Behavior tested against SW Python/C++ implementation
- Ongoing work:
 - Complete synthesis
 - Implementation in versal aie cores

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- Python implementation and test
- C++ conversion and testing of Python modules
 - Unitary test for each layer/function to match behavior
- HLS translation
 - Static definition of input parameters
 - Based on the expected net size

- Development of own library for Oviedo NN acceleration
 - Python, C++ and HLS src/tests
- Designed following HLS4ML methodology
 - Template and parameter definition
 - Full partitioned (minimum latency)



Prospects for the future

- Explore alternative technologies and ideas which could not be otherwise investigated that could potentially lead to a significant breakthrough.
 - Both in the present architecture of the CMS experiment and beyond
- Using muons as probes for new physics, as they Project focuses on muons, but ideas can be ported elsewhere.
- If ideas are successful, we may want to have a small-scale system running in parallel to our future HL-LHC system to validate it (beyond the scope of the grant)

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Backup



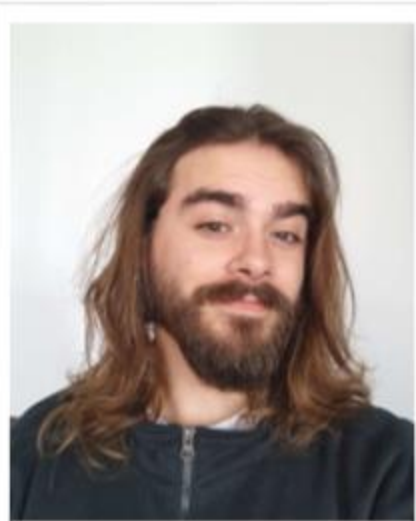
Who are we?

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Daniel Estrada

PhD. Student.
Physicists



Pelayo Leguina

PhD. Student.
Engineer



Santiago Folgueras



Andrea Cardini

Postdoc.
Physicist.



Javier Prado

PhD. Student.
Engineer