



Real Time Analysis in ATLAS Trigger with Machine Learning

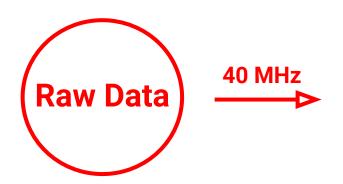
Claire Antel, <u>Leon Bozianu</u>, Denis Damazio, Anna Sfyrla, Steven Schramm

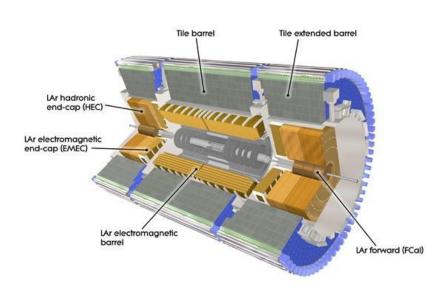


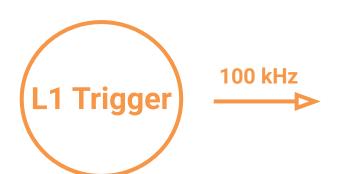




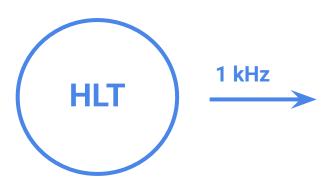
ATLAS Trigger







Hardware based (FPGA/ASIC).
Calorimeter and muon systems only.
Latency 2.5µs



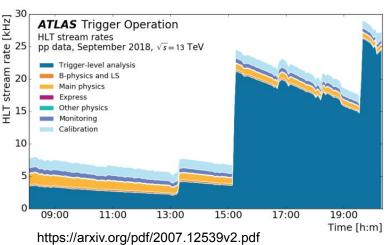
Software based (ATHENA).

Seeded by L1.

40k CPU cores.

Full detector output.

Average execution time 300ms.

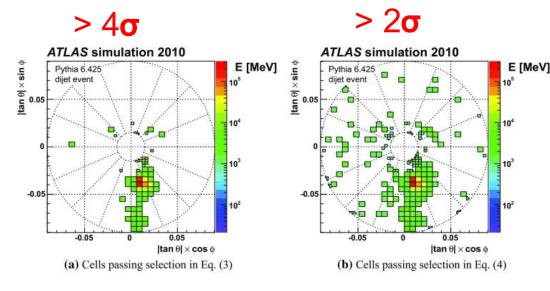


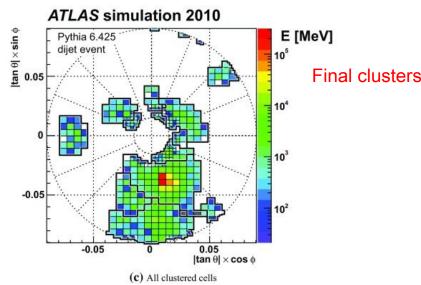


What is TopoClustering?

Cluster Formation:

- Cluster Seeding
 - Cell signal significance > 4σ.
 - Initialise highly significant "seed" cells.
- Cluster Growth
 - Cell signal significance > 2σ.
 - Neighbours of seed cells extend cluster.
- Cluster Boundary Cells
 - Cell signal significance > 0.
 - Add all neighbouring cells to existing cluster.

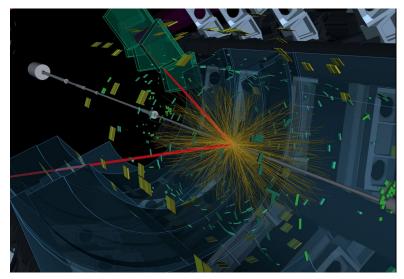


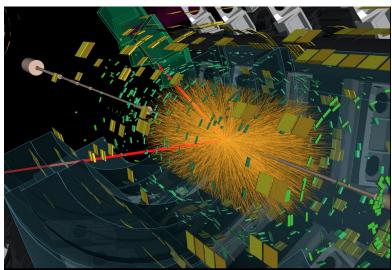




Problem Statement

- → Topoclustering is one of the most resource intensive algorithms in use in HLT.
- → Crucial role in **jet** and **MET** reconstruction.
- → Far worse pile-up conditions in HL-LHC.
- → We pursue faster solutions with (at least) similar performance.





50 vs 200 p-p collisions per bunch crossing.

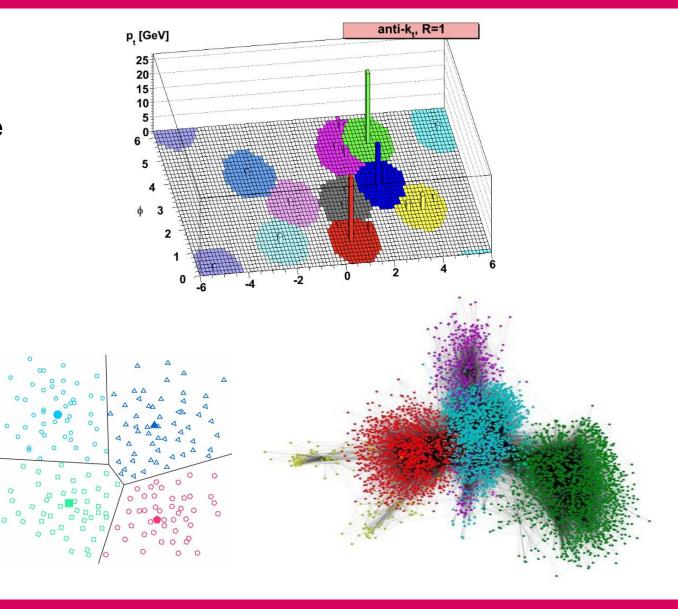
Work already underway to migrate current Topocluster to GPUs, and starting up on FPGA at present.



Proposal Solution

Any solution should address:

- Permit arbitrary cluster topology, shape and size.
- Detector geometry barrel/end-cap,
 EM/HAD calorimeter.
- Adjacent calorimeter layers granularity and irregularity.
- Robust to noise and cluster overlap.
- Variable number of clusters to find.
- Speed of execution.

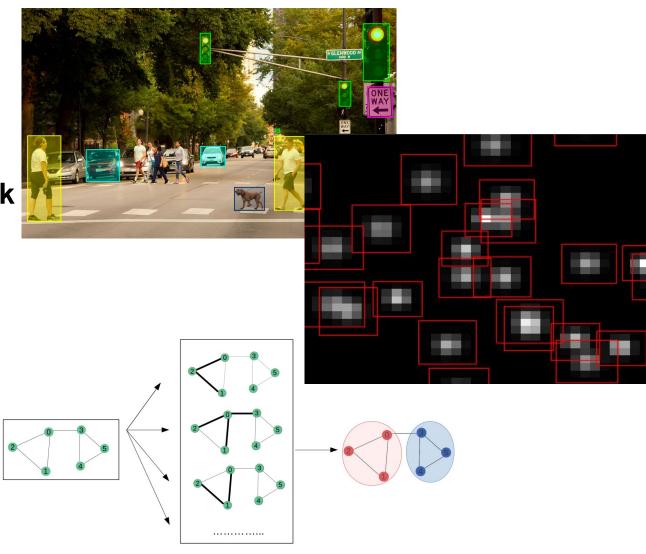




Proposal Solution

Two-step CNN + GNN

- Transform detector response into eta-phi projection of calorimeter.
- Use lightweight Region Proposal Network to identify Regions of Interest (ROIs).
- 3. Pass ROIs to **GNN** in original calorimeter coordinates.
- Classify nodes in the graph as noise or members of a cluster.
- (Incorporate splitter decisions, cluster classification and energy regression.)

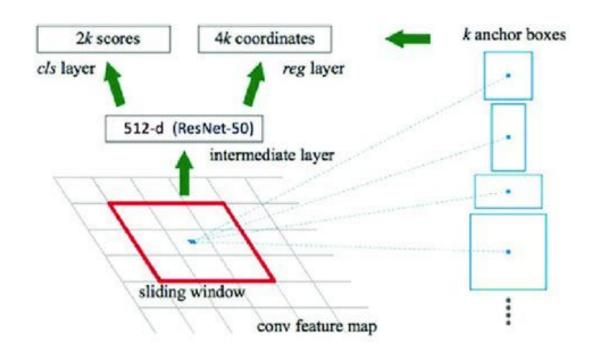


https://arxiv.org/abs/1603.02934

ML Proposal

Benefits of the approach

- Eta-phi projection simplifies problem of irregular detector geometry for CNN.
- Use of CNN to reduce search area from full calorimeter readout.
- GNN preserves detector geometry, preserving causal relationships between cells.
- Permit clusters of arbitrary shape and topology.
- Speed of GNN node classification within Rol.
- Potential cluster regression using graph level prediction.





Thanks for listening



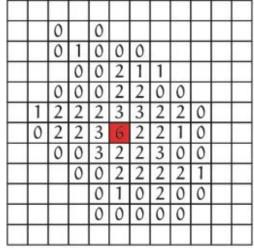
Backup



TopoCluster Algorithm

TopoCluster Algorithm:

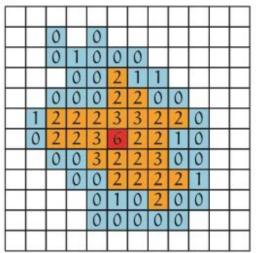
- Cluster formation is then later refined to jets and physics objects
- Base clusters from cell energy significance
 - Seed clusters
 - Grow protoclusters
 - Merge protoclusters
 - Splitting algorithm into clusters
- Calculation of cluster kinematics/moments
- Local hadronic cell weighting calibration





0 1 0 0 0

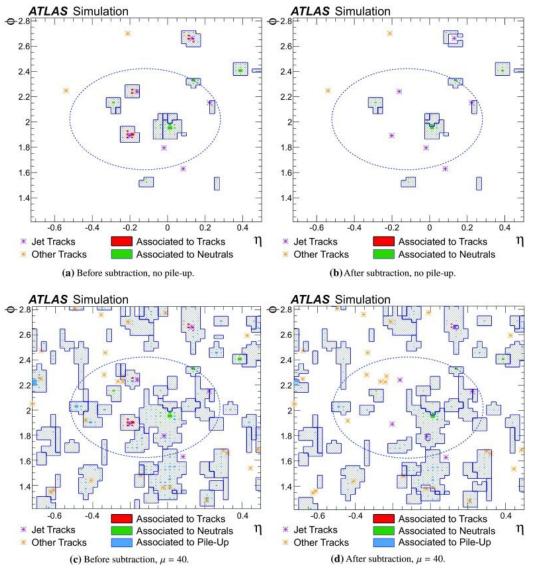
0 1 0 2 0 0



(c) Clustering of $|\epsilon_{cell}^{EM}| > 0$ cells.



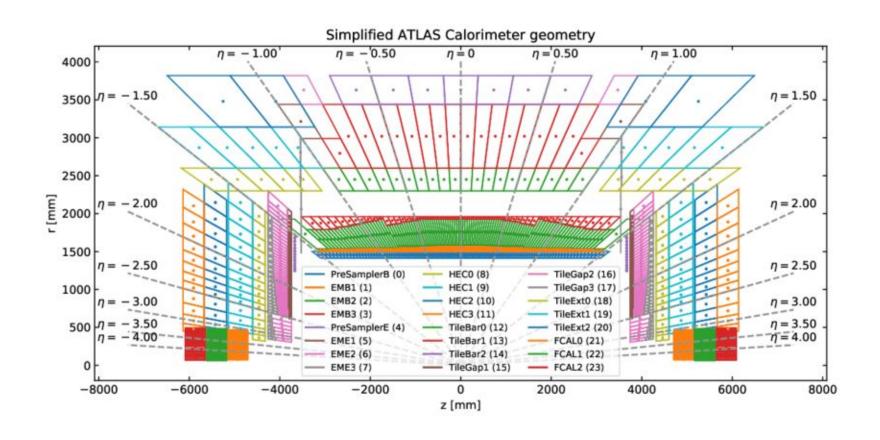
TopoCluster Algorithm



https://arxiv.org/abs/1703.10485

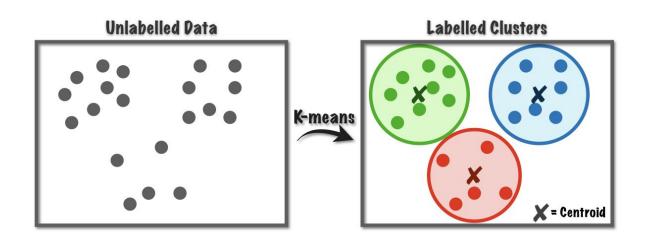


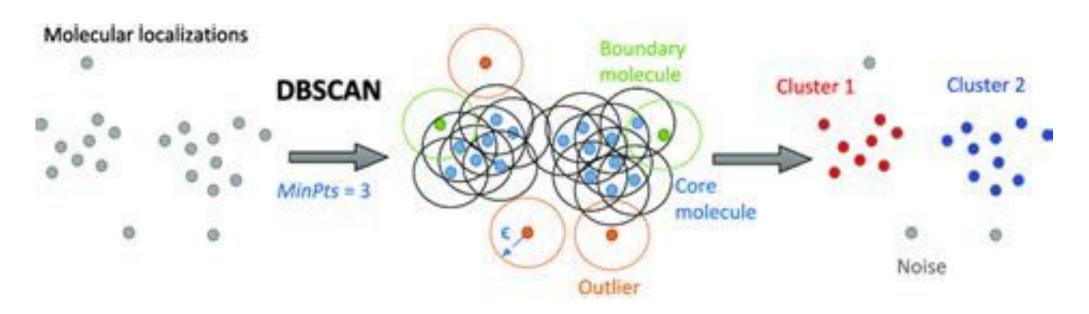
Calorimeter Granularity





Learning Paradigm





Project Objectives

- → Investigate ML clustering substitutes for TopoCluster algorithm in trigger
- → Deal with calorimeter granularity, complexity and irregularity
- → Benchmark tests performance + speed
- → Potential extension to parallel computing structure / GPUs
- → [Additional functionality cluster classification, energy/momentum regression]

https://arxiv.org/abs/1603.02934