

Jet trigger algorithms with GPUs for the ATLAS Upgrade

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The Jet trigger challenges at the LHC:

Jet:

- Collimated spray of particles initiated by quarks or gluons
- Important for QCD studies, SM physics and search for new physics (SUSY,...)

Trigger system:

- 40 MHz event rate
- Volume of data/s \sim PB/s
- Trigger systems designed to select interesting physics events, for storage and offline analysis

	Run 2	LS2 - Phase-I	Run 3
	2015-18		2020-22
Centre of mass Energy \sqrt{s} (TeV)	13-14		14
Luminosity ($\text{cm}^{-2}\text{s}^{-1}$)	1×10^{34}		2×10^{34}
Bunch spacing (ns)	25		25
Number of interaction/event	~ 27		$\sim 55 - 80$
Total Integrated luminosity (fb^{-1})	~ 100		~ 300

ATLAS Jet trigger for Run 3:

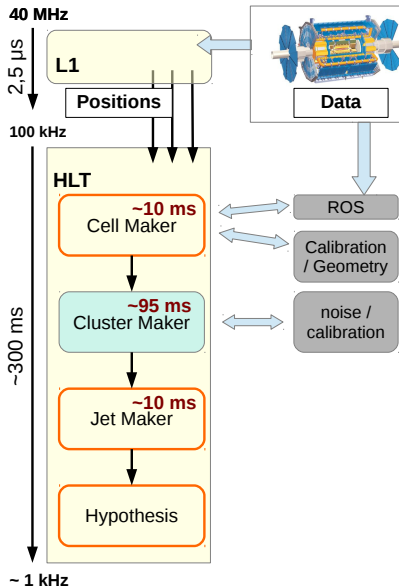
Jet energy deposits are aggregated by the reconstruction algorithms and calibrated to provide the jet momentum measurement.

In Run 3:

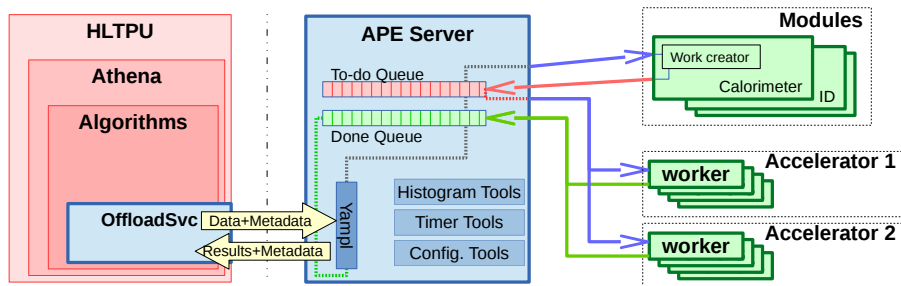
- Up to 80 collisions per event
- Advanced algorithms needed to better suppress pile-up
- More computation required

Evaluate the use of GPU's on the HLT

- Demonstrator prototype under implementation
 - ▶ L1's responsibility: calorimeter trigger chains
- Needs re-design of the algorithms: single instruction - multiple data paradigm



Demonstrator architecture:



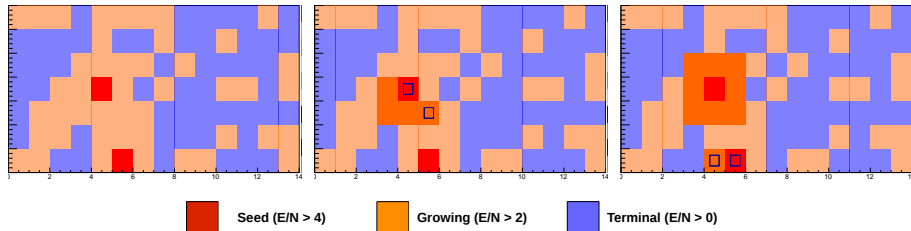
Client side

- One HLT processing unit per core
- Athena framework
 - ▶ Executes chains of algorithms
 - ▶ Provides data and monitoring services
 - ▶ Caches responses

Server side

- Independent from Athena
 - Efficient accelerator resource management
 - Can exploit several technologies
- Specific modules and services implemented for the calorimeter data/tasks offloading and processing

Topological Clustering algorithm:



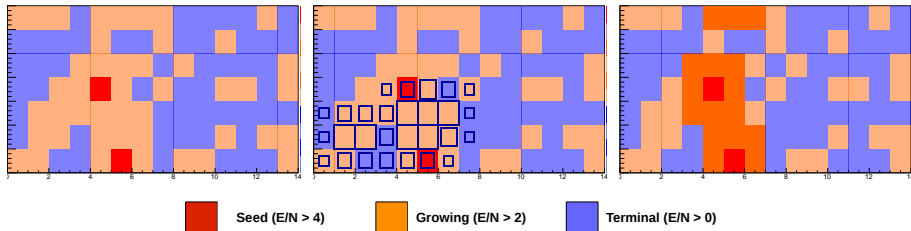
Group cells into clusters according to their Energy/Noise level.

Cluster Growing:

- Each SEED initiates a clusters
- For each SEED (and later GROWING) cell:
 - ▶ Find all neighbours (3D) and add them to clusters:
 - ★ If a neighbour is a SEED or GROWING in another cluster → merge them
 - ★ If a neighbour is a GROWING cells → add it to next iteration list
- Terminal cells do not propagate the cluster

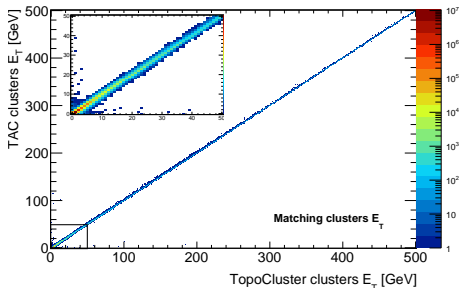
Topological Automaton Clustering algorithm:

- Designed to exploit the parallelism of the GPU



- Based on a Cellular Automaton
 - Propagate of flags on a grid of elements
- Assign unique flag (energy ordered) to each **SEED** cell
- Process all cells in each iteration (until no flag changes)
 - Compare flag to all neighbours
 - Get the largest flag
- Implementation:
 - Maximize parallelism: data organized in cell-neighbour pairs
 - Only pairs with at least a **SEED** or **GROWING** cell
 - Each GPU thread process a pair

Results and next steps



First results very promising:

- Clusters energy correctly reconstructed
- $\sim 1\%$ of `TERMINAL` cells associated to different clusters
 - ▶ Expected due to small differences on GPU algorithm

• Processing time (di-jet events):

- ▶ Cluster Growing CPU ~ 22 ms \rightarrow GPU ~ 8 ms ($\sim 2.75X$ faster)
- ▶ Data conversion overhead ~ 12 ms
- ▶ Optimization ongoing

Next steps

- Cluster Splitting parallelisation just started (Braga)
- Finish optimization and evaluation of the demonstrator prototype
 - ▶ Next 2-3 months
- ATLAS will take a decision on whether to use GPU's for Phase I HLT Upgrade based on these results

Thank you!



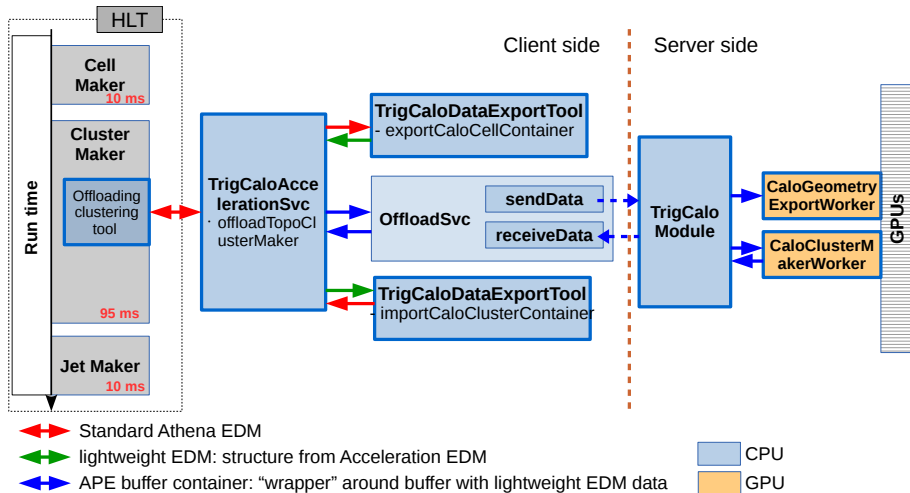
Acknowledgements



Run: 260272
Event: 5879
2015-04-05 10:46:41 CEST



Calorimeter demonstrator:



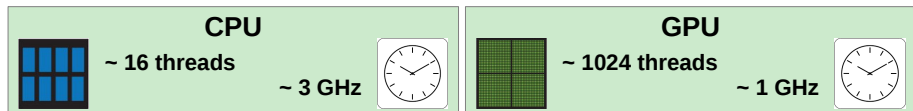
Trigger GPU demonstrator for the Phase I Upgrade:

Trigger farm hardware rolling replacement during shut-down

- New architectures need to be evaluated, parallel computing paradigm growing
- General-purpose computing on Graphics Processing Units (GPUs)

GPUs are massive parallel processing architectures

- single instruction multiple data



The Trigger GPU demonstrator comprises prototype of HLT algorithms

- Inner Detector tracking: (RAL)
- Calorimeter clustering: (LIP)
- Muon track finding: (Bologna, Rome)

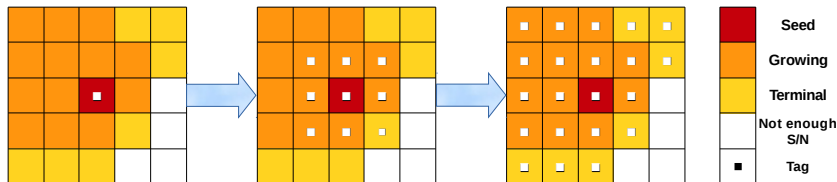
The goal is to assess the potential of GPUs for use in the HLT farm

- Main metric to be the throughput per unit cost

Topological Automaton Clustering:

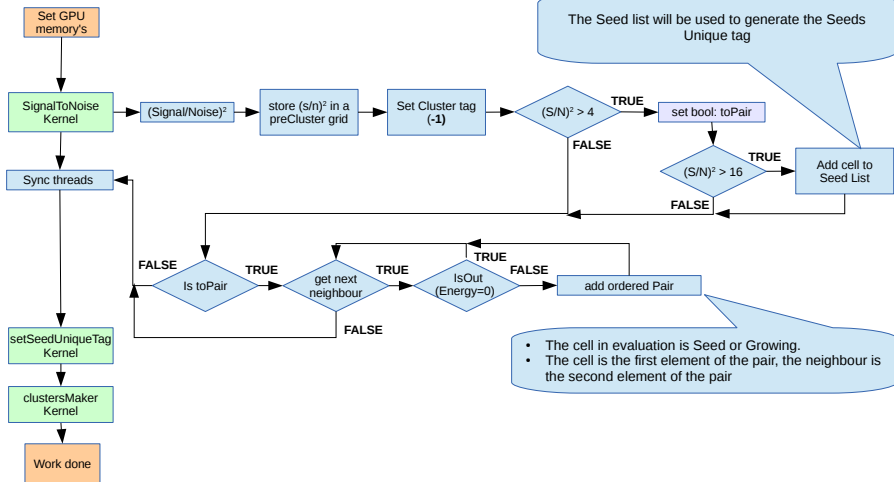
Based on a Cellular Automaton algorithm

- 1 Cell signal/noise rate
- 2 Fill a vector of cell pairs. (reduces branching due to irregular # neighbour)
 - ▶ Only SEED or GROWING cells generate pairs. (data reduction)
- 3 Orders the SEEDS (S/N), cluster tag is the SEED position in the list
- 4 Loop all cell pairs proposing the highest tag as new tag till all tags remain unchanged:
 - ▶ If no/same tag proposed: continues.
 - ▶ If has one proposed tag from a SEED/GROWING cell: accept tag.
 - ▶ If both cells propose a valid tag: accept the highest S/N tag. (merge clusters)

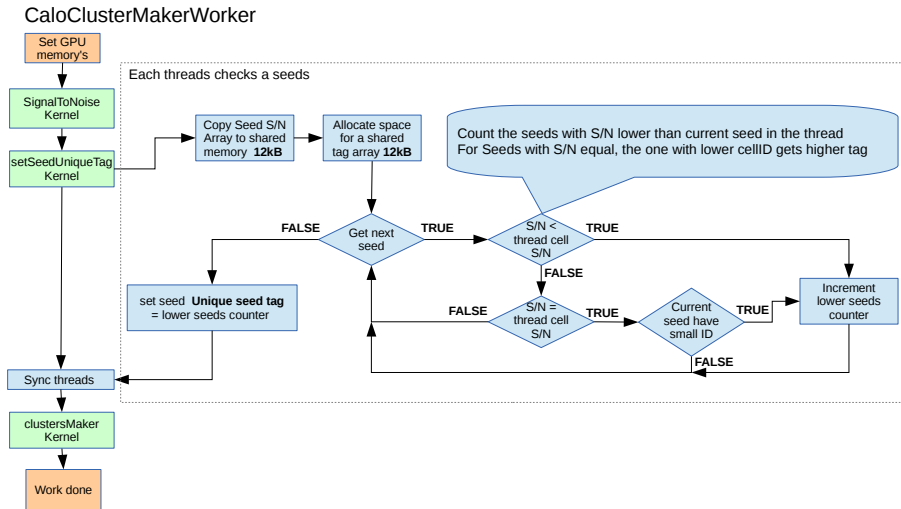


Topological Cluster alternatives: GPU CellRating

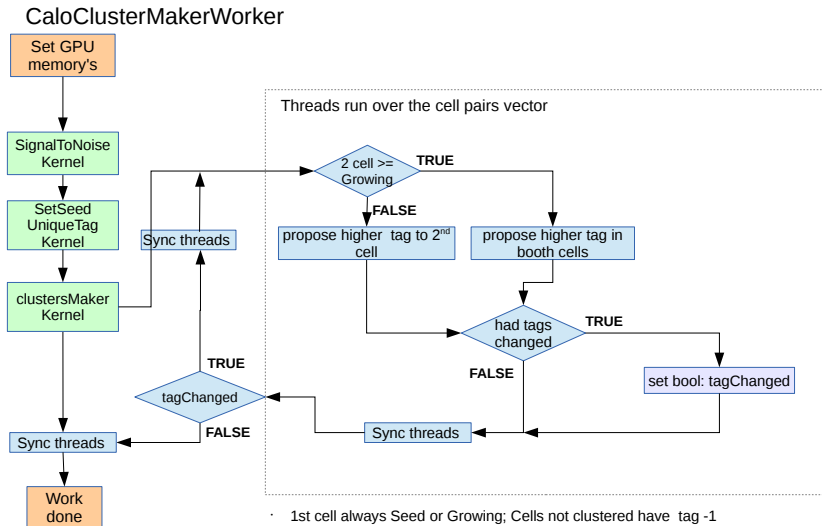
CaloClusterMakerWorker



Topological Cluster alternatives: GPU Set seed unique tag



Flux diagram of TAC:



Topological Cluster alternatives: CPU Clustering

