

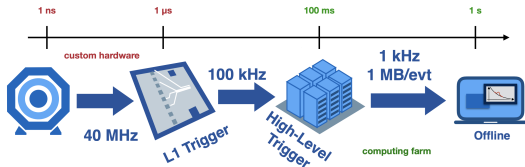
ATLAS jet trigger system: overview and performance at the beginning of Run 3

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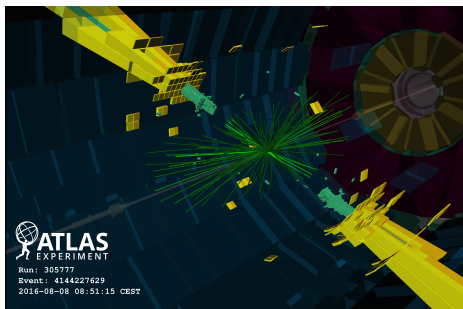
ATLAS trigger system - general overview

- If all events were saved - ~ 1 PB of data in 10 seconds of data taking
 - not feasible \rightarrow use trigger system to record only **interesting events**
- ATLAS has two level trigger system
 - **Level-1 (L1) trigger:**
 - hardware based
 - reduces 40 MHz to 100 kHz, decision made within $2.5 \mu\text{s}$
 - high- p_T muons, electrons, photons, jets, large E_T^{miss}
 - **High Level Trigger (HLT):**
 - more sophisticated algorithms (tracking, PFlow, jet reconstruction), higher latency ($\mathcal{O}(100)$ ms)
 - reduces 100 kHz down to 3 kHz



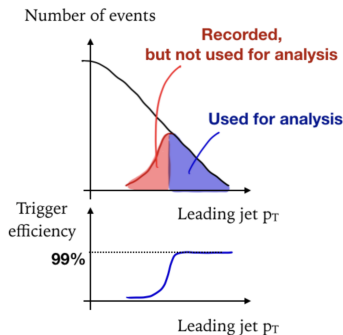
Jets in ATLAS

- **Collimated bursts of hadrons** produced by the hadronisation of quarks and gluons
- Reconstructed by using calorimeter clusters and charged particle tracks as inputs
 - jets are formed by clustering inputs together (different strategies)
 - **small- R jets** $R = 0.4$, **large- R jets** $R = 1.0$
- Associated with many interesting Standard Model (SM) and Beyond Standard Model (BSM) processes
 - the **most common** signature **at the LHC** \rightarrow powerful jet triggers essential for all physics purposes



Trigger efficiency

- Way to assess the performance of trigger is to study **trigger efficiency**
- Computed with respect to the reference trigger which is chosen to be fully efficient in the studied region
- The **sharper the turn-on** (rising slope), the less unused data \rightarrow **better** performing trigger



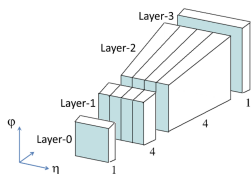
Source: [1]

Level-1 Trigger

ATLAS jet trigger in Run 3

L1 calorimeter trigger (L1Calo)

- Jet trigger at Level-1 based on the information from the calorimeter
- **Upgrade** of the system between Runs 2 and 3
 - Runs 1 and 2 - trigger towers 0.1×0.1 in $\eta - \phi$
 - Run 3 finer-granularity input from LAr - SuperCells (10 in one trigger tower)
 - trigger towers distributed to **Feature Extractor (FEX)** processor
- FEX has electromagnetic (eFEX), jet (jFEX) and global (gFEX) feature extractors
 - jFEX and gFEX used for jet triggers



Schema of 0.1×0.1 trigger tower formed by ten SuperCells. [7]

ATLAS jet trigger in Run 3

L1 jet trigger

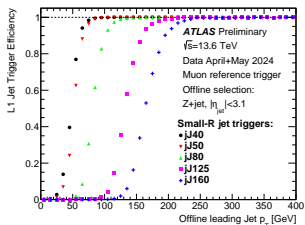
- **Jet Feature Extractor (jFEX)**

- reconstructs small-R jets, sliding-window algorithm
- similar performance to L1Calo from Run 2 in single jets
- higher efficiency in multi-jet triggers due to better resolution of close-by jets

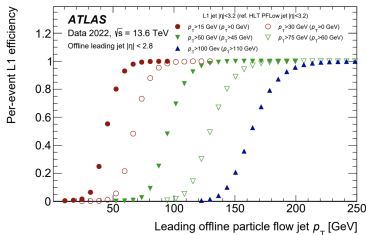
- **Global Feature Extractor (gFEX)**

- coarser granularity compared to jFEX \rightarrow large-R jets

- jJ40 \rightarrow jFEX jet, gLJ80 \rightarrow gFEX \dots , number is a jet p_T



L1Calo single jet trigger efficiencies for the Phase-I system (Run 3). [4]



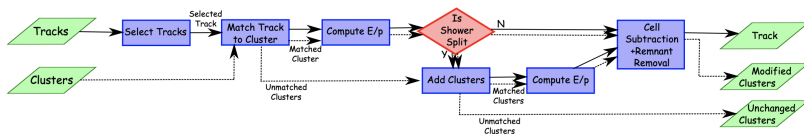
Efficiency of the legacy L1 single jet triggers. [7]

High-Level Trigger

Jet reconstruction for the HLT

Particle Flow jets

- Tracks are extrapolated to calorimeter and are matched to clusters
- Computing E/p
 - deposited energy $\langle E_{dep} \rangle = p^{trk} \langle E_{ref}^{clus} / p_{ref}^{trk} \rangle$
- Cell energy subtraction
- Positive energy topo-clusters surviving the energy subtraction are used for anti- k_t jet clustering algorithm ($R = 0.4$)

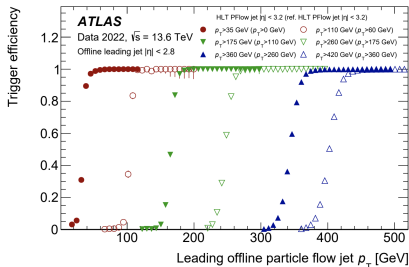


Source: [6]

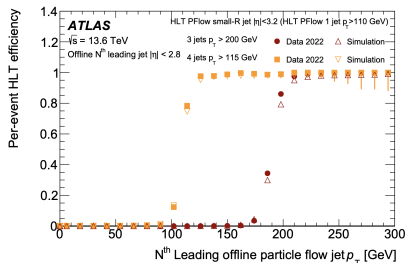
ATLAS jet trigger in Run 3

HLT jet trigger

- Triggers based on **Particle Flow Objects (PFO)** thanks to great tracking capability
 - Run 2 triggers used information only from topo-clusters
- Full scan tracking run after preselection stage to reduce CPU costs
- PFO - matching tracks to topo-clusters based on extrapolated track and cluster location
- Calibration → directly controls the shape of trigger efficiency turn on curves



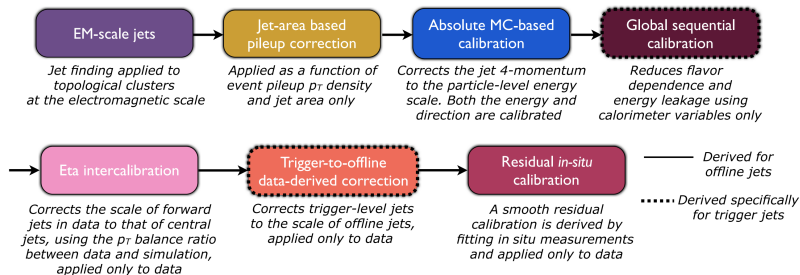
Single jet HLT efficiencies for jets in the central region. [7]



Multi-jet trigger efficiency at the HLT. [7]

HLT jet calibrations

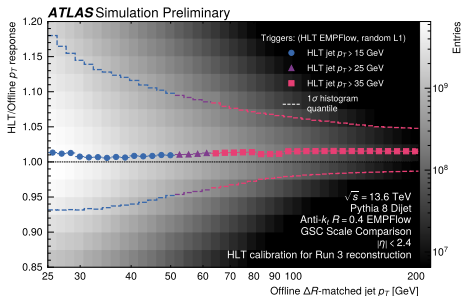
- **Response** of the detector for **HLT jets** with **different** p_T is different compared to **offline jets** at EM-scale → need for the **calibration**
 - ideally we want to get the same response as for offline jets
 - we want distribution HLT/offline $\mu \approx 1$ with small σ



Steps of the HLT jet energy scale calibration [5]

Performance of the trigger-level jet calibrations

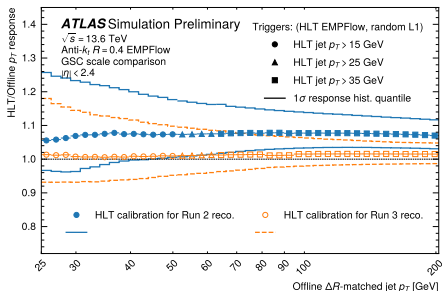
- HLT jets are corrected with an offline jet calibration configured for Run 3 reconstruction
- Good trigger-level calibration crucial for well-performing trigger
- Dijet Monte-Carlo HLT/offline response
 - **distribution** of **HLT/offline response** for different p_T bins is computed
 - each bin is fitted Gaussian and **mean** (full points in the graph) is estimated
 - boundaries of **1σ quantile** are drawn (dashed lines)



HLT jet calibration performance for 2023 [4]

Comparison of the HLT jet calibrations during 2022 and 2023

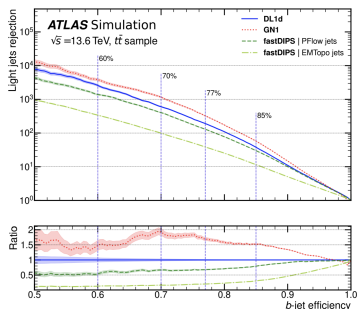
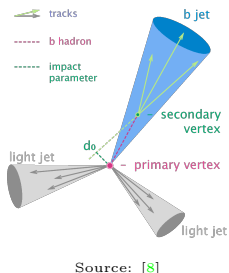
- At the **beginning of Run 3** HLT used Run 2 and offline used Run 3 **calibrations**
 - HLT/offline response 5-7% above unity
- **2023** - improvement by applying the **same calibration** to the HLT and offline reconstruction
 - narrower 1σ bands and mean response closer to 1
 - this can help achieving sharper trigger efficiency turn on



Comparison between HLT jet calibration using Run 2 (blue) and Run 3 (orange) reconstruction [4]

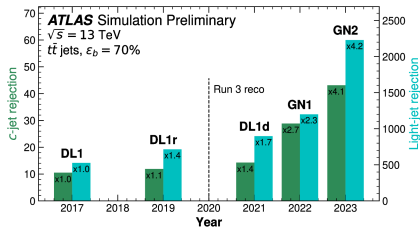
b -jet triggers

- b -jet triggers crucial for precision measurements and searches
- b -tagging requires **precise tracking** and dedicated approach
 - inputs are reconstructed tracks and jets, and primary vertex position
- Identification done in two steps
 - fast b -tagging is run on tracks inside super-Regions of Interest
 - final b -tagging algorithm
 - low level presel.: fastDIPS \rightarrow fastGN2; higher level: DL1d \rightarrow GN1 \rightarrow GN2

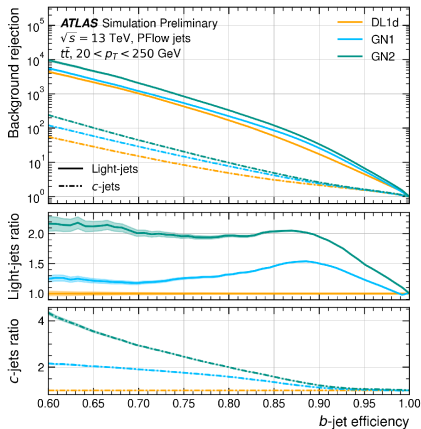


Comparison of b -jet taggers used in Run 3

- Continuous improvement of b -tagging algorithms used in ATLAS



The c -jet and light-jet rejection of different ATLAS flavour tagging algorithms in $t\bar{t}$ Monte-Carlo samples. [2]



The c -jet and light-jet rejections as a function of the b -jet tagging efficiency for jets in a $t\bar{t}$ sample. [2]

Conclusion

- Efficient jet trigger system helps targeting many interesting SM and BSM processes
- Several improvements to the jet trigger system with respect to Run 2
 - L1: gFEX and jFEX
 - HLT: PFlow jets
- Improvements to the trigger-level jet calibration
- Better b -tagging capability thanks to the graph neural networks GN1, and currently GN2
- b -tagging preselections (fastDIPS, fastGN2) introduced in Run 3 allowed for better b -tagging triggering

References

- [1] AMERL, Max: The ATLAS Jet Trigger in Run-3, URL https://indico.jlab.org/event/459/contributions/11409/attachments/9379/13601/CHEP2023Slides_MAmerl_TheATLASJetTriggerInRun3.pdf (CHEP2023)
- [2] ATLAS COLLABORATION : *Jet Flavour Tagging With GN1 and DL1d. Generator dependence, Run 2 and Run 3 data agreement studies.* – URL <https://atlas.web.cern.ch/Atlas/GROUPS/PHYSICS/PLOTS/FTAG-2023-01/>
- [3] ATLAS COLLABORATION : *Particle-hunting at the energy frontier.* – URL <https://atlas.cern/updates/briefing/particle-hunting-energy-frontier>
- [4] ATLAS COLLABORATION : *Public Jet Trigger Plots for Collision Data.* – URL https://twiki.cern.ch/twiki/bin/view/AtlasPublic/JetTriggerPublicResults#2022_pp_data
- [5] ATLAS COLLABORATION : Jet energy scale measurements and their systematic uncertainties in proton-proton collisions at $\sqrt{s} = 13$ TeV with the ATLAS detector. In: *Physical Review D* 96 (2017), Nr. 7. – URL <http://dx.doi.org/10.1103/PhysRevD.96.072002>
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