Redefining
Performance:
New Techniques
for ATLAS Jet
Calibration

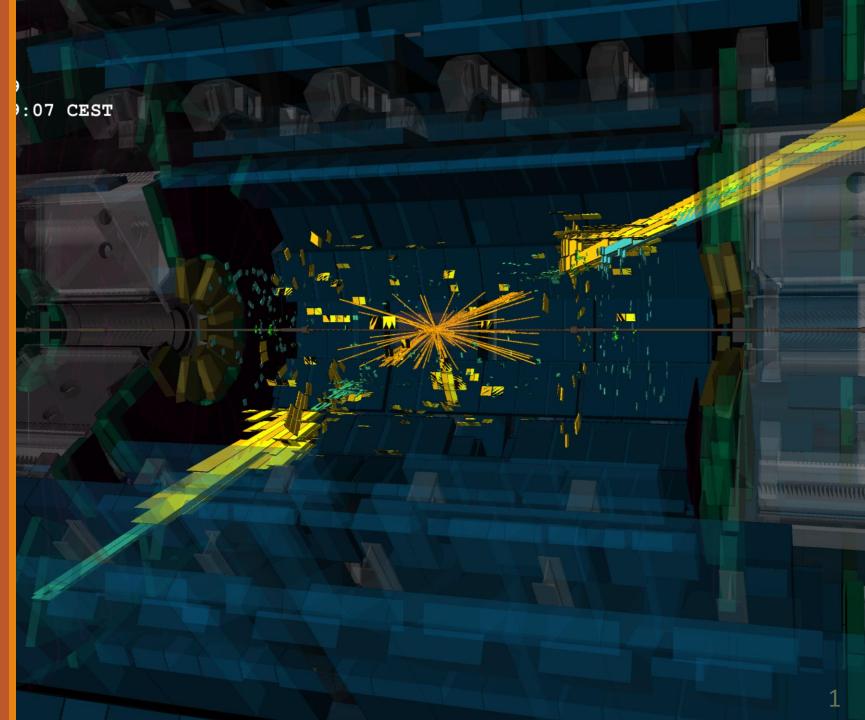
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LPNHE/CNRS
On behalf of the ATLAS
Collaboration

HEP2023

January 12th, 2023





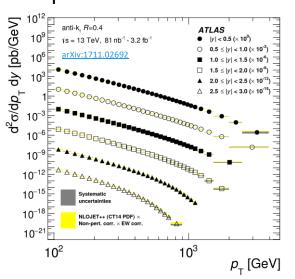
Importance of Jets at LHC

LHC: Particle accelerator at CERN, 27 km circumference

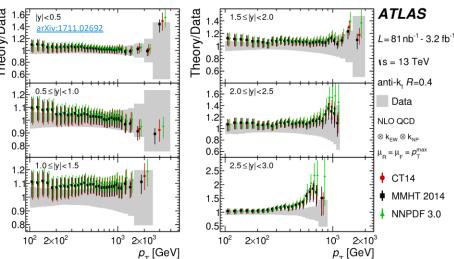
→ currently produce proton-proton collisions at a center-of-mass energy of 13.6 TeV

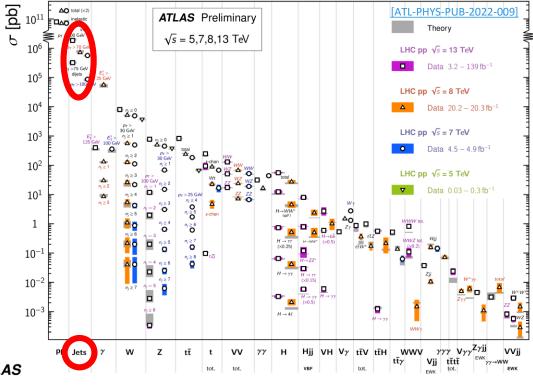
Quarks and gluons, resulting in "jets" in the detector, are by far the dominant production at LHC

Jet production cross-section



Data-theory tension





Jets are used both in:

Standard Model Production Cross Section Measurements

- precision measurement of standard model processes
- beyond the standard model searches
 - → Precision is crucial



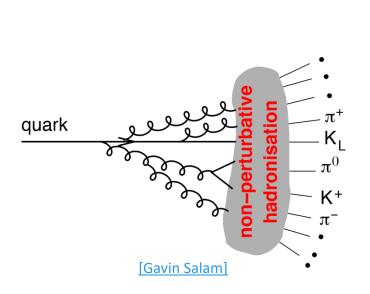
Jet Production and Detection in ATLAS

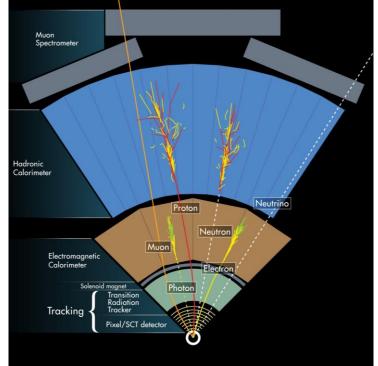
ATLAS: multi-purpose particle detector

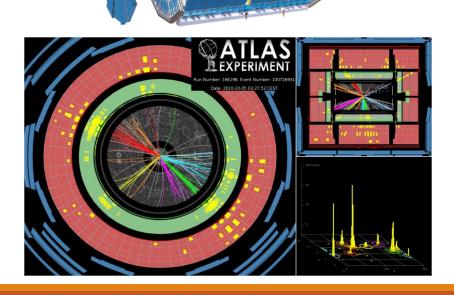
Jet: collimated spray of hadrons produced by the hadronization of a quark or a gluon

→ Produce tracks in the trackers and energy deposits in the calorimeters, that are clustered

together to obtain the properties of the initial quark or gluon









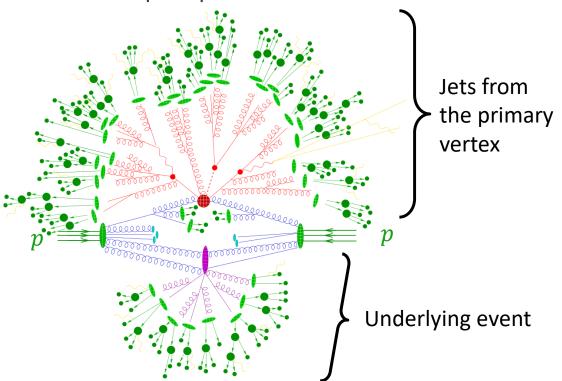
Pileup

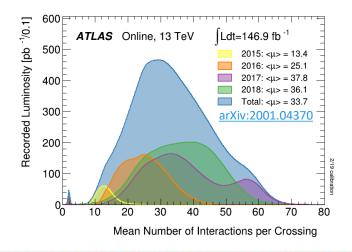
Pileup: all the effects of other interactions other than the hard scatter

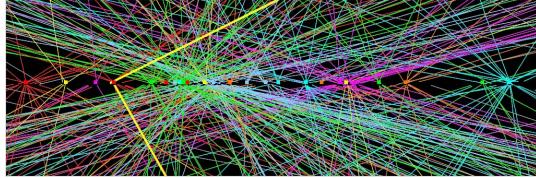
→ Challenge for the jet reconstruction and calibration

In-time pileup: underlying event, other collisions in the same bunch crossing

Out-of-time pileup: collisions in other bunch crossings









Jet Reconstruction in ATLAS used during Run 2

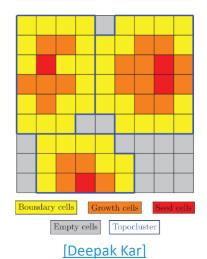
Step 1: Building particle-like **clusters** from the energy deposits and the tracks

Topological clusters

arXiv:1603.02934

Connect group of calorimeter cells

- 1) Seed: cell with $|E| > 4\sigma$
- 2) Growth: adjecent cells with $|E| > 2\sigma$ I separation
- 3) Additional layer
- 4) Splitting algorithm (local maxima)

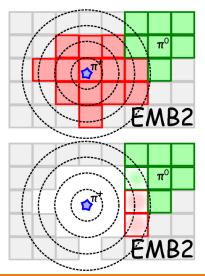


Particle flow objects (PFOs)

arXiv:1703.10485

Combine track and cluster information

- \rightarrow Better resolution at low p_T , pileup separation
- 1) Select good tracks
- 2) Subtract them from the topo-clusters



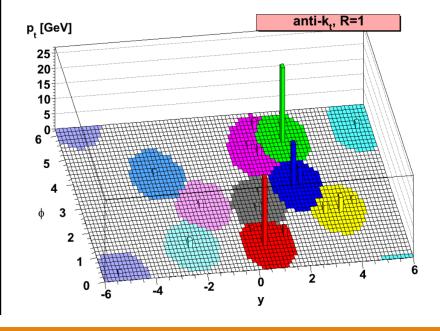
Step 2: Building jets from clusters/PFOs

Anti- k_t algorithm

arXiv:0802.1189

Sequentially gather nearby constituents

Small radius jets: parameter R=0.4 Large radius jets: parameter R=1.0





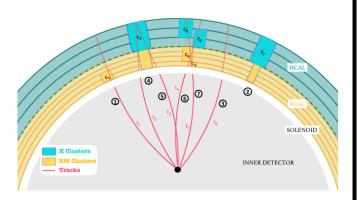
Unified Flow Objects

Track-CaloClusters (TCC)

[ATL-PHYS-PUB-2017-015]

Use tracks to split up large clusters based on the energy flow and their direction.

 \rightarrow Improvement of the mass resolution at high- p_T

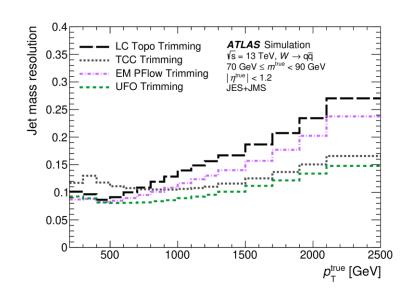


Unified Flow Objects (UFOs)

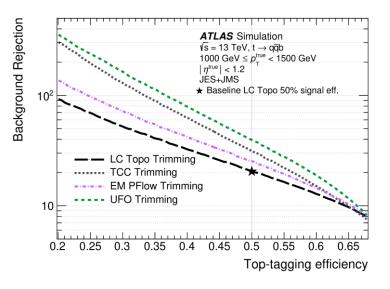
<u>arXiv:2009.04986</u> [ATL-PHYS-PUB-2022-038]

Apply TCC technique to split PFOs

- \rightarrow Low p_T : good pileup subtraction from PFOs
- \rightarrow High p_T : large clusters are split



- \rightarrow Best mass resolution across p_T
- \rightarrow Best background rejection for a simple top-tagger at both low and high p_T



Results shown here for large radius jets (R = 0.1); results for small radius jets in <u>backup</u>.



Small-R Jet Energy Scale Calibration used in ATLAS during Run 2

Pileup corrections

Absolute MC-based calibration

Global sequential calibration

Residual in situ calibration

- 1) Correction based on the **pileup density** and the **jet area**
- 2) Remove residual pileup dependency

Energy and direction are corrected to particle-level scale based on MC simulations.

Response:

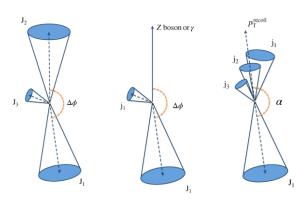
$$\mathcal{R} = \frac{E^{detector}}{E^{particle}}$$

The scale depends on visible **features of the jets**, like the **quark/gluon nature**.

The resolution can be improved by removing these dependencies thanks to MC simulations.

MC does not perfectly represent data

→ In situ corrections needed using different event topologies.

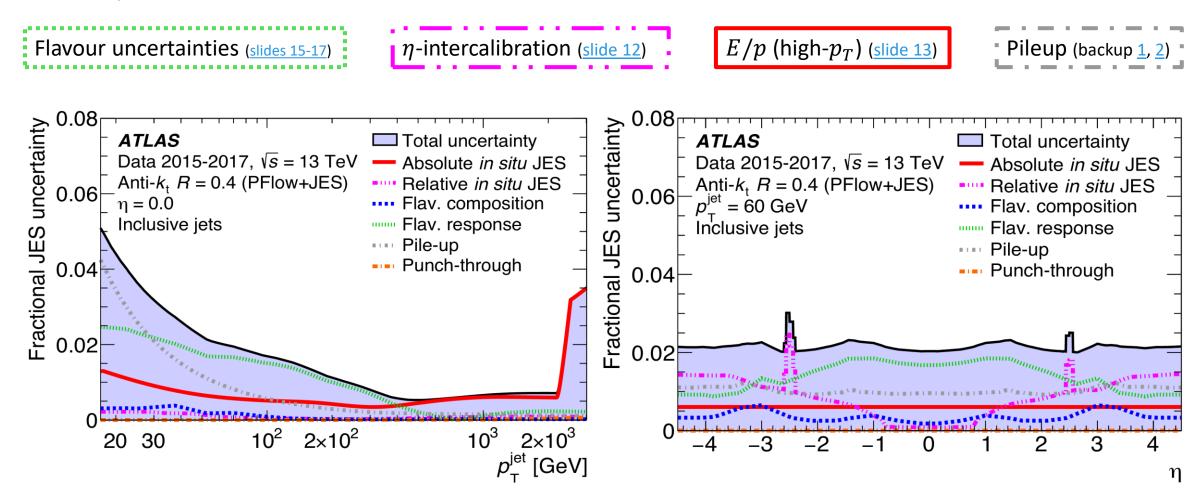


Backup: small-R JES calibration more in depth



The Uncertainties on the Jet Energy Scale

Improvements of some dominant uncertainties:





Global Neural Network Calibration

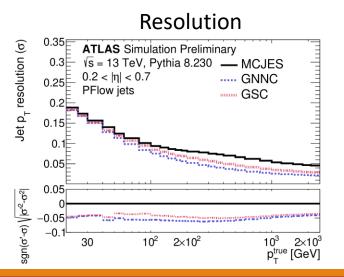
The Global Sequential Calibration (**GSC**) sequentially corrects for the response dependency on six variable so does not exploit correlations.

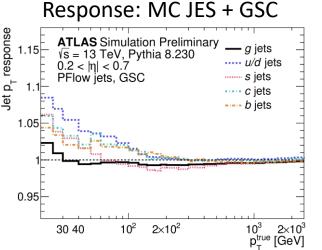
The Global Neural Network Calibration (GNNC) uses a **Deep Neural Network (DNN)** trained to predict the p_T response. It can use more correlated variables. Architecture and inputs in backup.

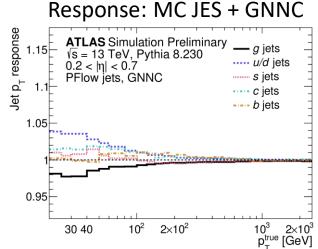
- \rightarrow The Jet Energy Resolution (JER) is improved by 15-25~% on average compared to the GSC.
- → The responses of quark-initiated jets is closer to unity → Reduction of the flavour uncertainties.

 $(\rightarrow \text{ see } \underline{\text{slide } 15})$

The in situ studies allow to validate the improvements in the calibration chain (see backup).







 p_T^{ref}

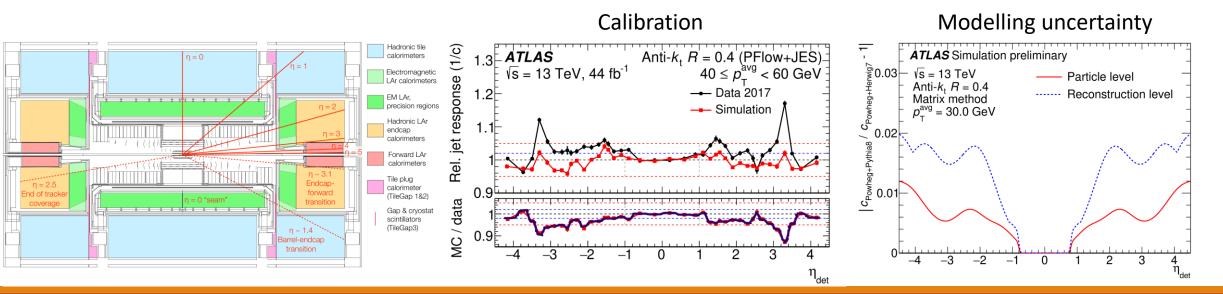
 p_T^{probe}

η -intercalibration Uncertainty Reduction

In situ η -intercalibration: make the calibration homogeneous across the whole η range using dijet events and transverse momentum conservation: $c = p_T^{ref}/p_T^{probe} \approx 1$.

Modelling differences affect:

- the dijet balance (extra radiations like a third jet, out-of-cone radiations)
- the detector response (through the jet composition and substructure)
 - → Possible double-counting of detector effects with the flavour uncertainty.
 - → Solution: separating physics effects from detector effects
 - → Factor 2 reduction of the modelling uncertainty below 40 GeV, which was dominant

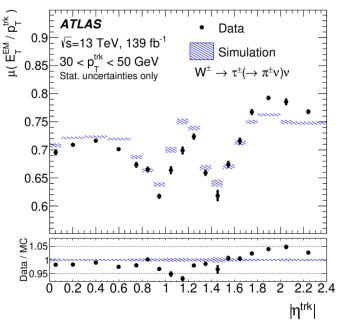


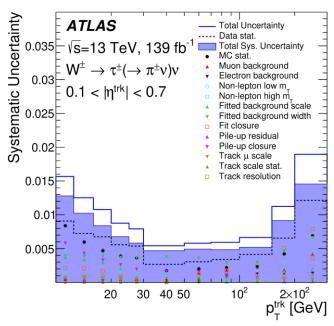


E/p Calibration Using $W^{\pm} \to \tau^{\pm} (\to \pi^{\pm} \nu_{\tau}) \nu_{\tau}$ Events

 $W^{\pm} \to \tau^{\pm} (\to \pi^{\pm} \nu_{\tau}) \nu_{\tau}$ events provide **highly-energetic isolated charged pions**

Gaussian fit of the $E^{calorimeter}/p_T^{tracker}$ distribution in data and MC





- \rightarrow Calorimeter response obtained in a very wide range (10 < p_T < 2000 GeV and $|\eta|$ < 2.4)
- ightarrow Excellent precision ightarrow High p_T uncertainties improvement
- → Improvement over the combined test-beam data previously used

More E/p results

b Quark Jet Energy Scale in $\gamma+$ jet Balance

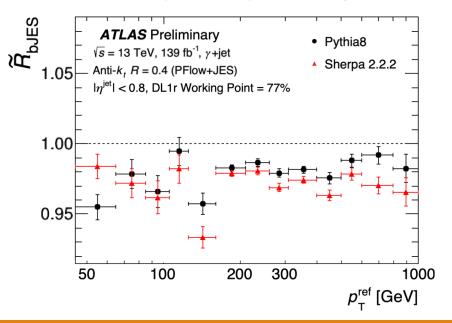
 γ +jet balance: exploit the transverse momentum balance between a jet and a reference photon.

The MC-to-data response ratio is obtained for two selections: inclusive, and b-tagged jets.

$$ilde{\mathcal{R}}_{bJES} = rac{\mathcal{R}_{b-tagged}^{MC}/\mathcal{R}_{b-tagged}^{data}}{\mathcal{R}_{inclusive}^{MC}/\mathcal{R}_{inclusive}^{data}}$$

The energy scale of b-tagged jets with respect to that of inclusive jets is determined in with a precision of about 1.5 to 2 % on a broad phase-space range.





- → Will improve measurements of:
- Top mass
- Bottom Yukawa ($VH, H \rightarrow b\overline{b}$)

<u>b-JES for different b-tagger</u> <u>working points in backup</u>



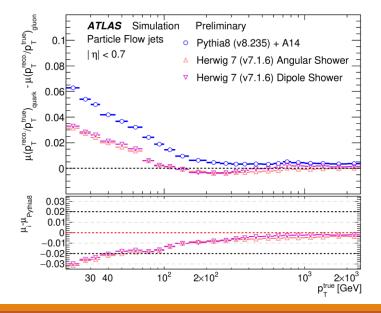
The Flavour Uncertainty

Different MC generators, different hadronization models

- → Different fragmentation pattern and hadron content
 - → Different detector responses
 - → Flavour uncertainty

Origin: non-compensating nature of the calorimeters. Response of π^0 and π^\pm are very different. Response of pions, kaons, baryons are different too.

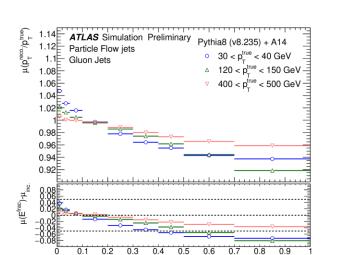
→ Important response difference observed between quark and gluon jets



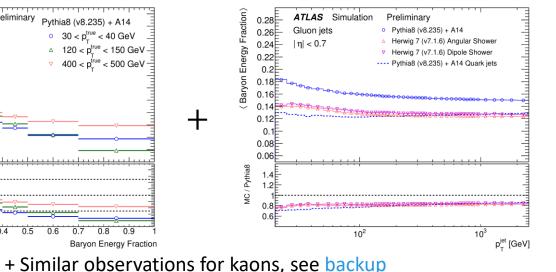


Response of Gluon Jets vs. Baryon Energy Fraction

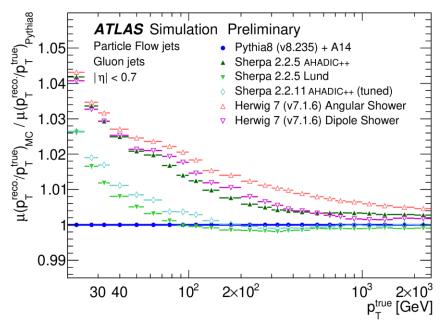
Dependency of the response on the baryon energy fraction



Different baryon energy fraction between Lund string-based and
cluster-based **hadronization models**



Response difference between the two types of hadronization models



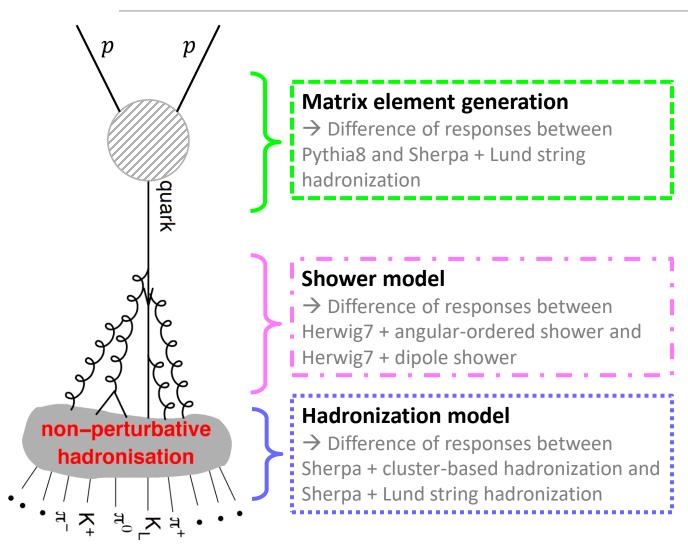
Fragmentation pattern + hadron content → need to be tuned to experimental measurements

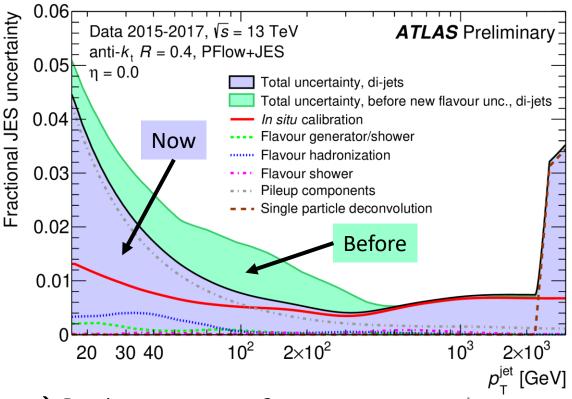
Reweighting the events based on the baryon and kaon energy fractions reduces the differences (backup)

Reduction in the dependence on the generator configuration would result in a significant improvement of the flavour uncertainties.



Flavour Uncertainty Decomposition





- → Previous treatment: 2 components: sample composition + response difference of gluon-initiated jets between Pythia8 and Herwig7
- → New treatment: 3 components
- → Improvement of the overall flavour uncertainty



Conclusion

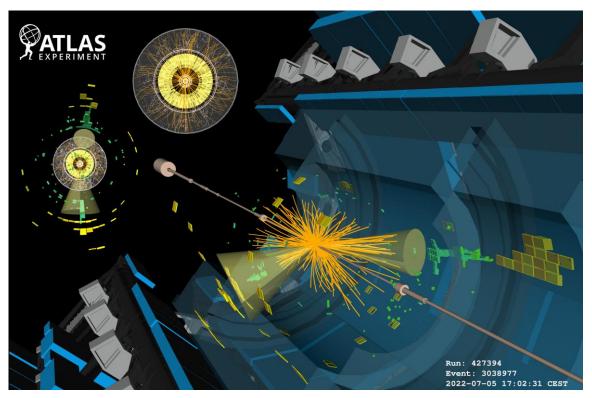
Hadronic objects reconstruction and calibration are important for precision measurements and BSM exploration at the LHC (and beyond)

Lots of improvements at every step of the definition and calibration of jets, thanks to the exploration of new techniques

→ Also see the use of ML/AI techniques for hadronic object performance (see Reina Camacho's talk)

Run 3 just started: great opportunity to exploit these developments and continue refining our strategies

→ Lot of work ongoing in that direction, stay tuned!



Run 3 dijet event



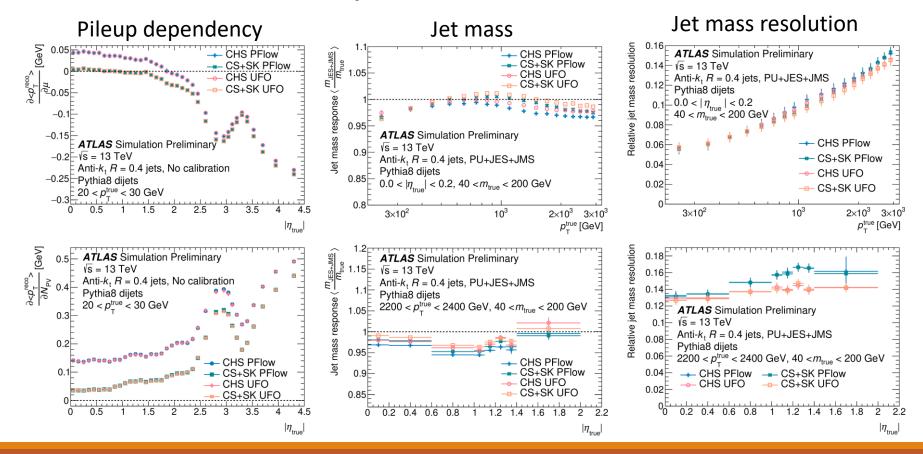
Backup



UFOs and CS+SK for small-R jets

CS+SK reduces the pileup dependency (see next slide).

- CS+SK UFO → Better calibration of the jet mass and better mass resolution.
 - → Consistent with PFlow jets on other observables.





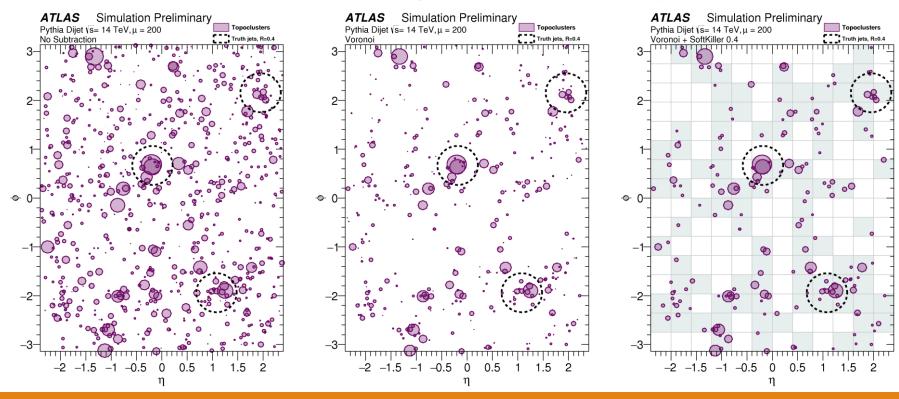


CS+SK

Some pileup can be removed prior to jet reconstruction

 \rightarrow Constituent subtraction (CS): assign an area to each cluster in (η, ϕ) (region where points are closer to that cluster than any other cluster), and subtract the expected pileup for that area.

→ Soft killer (SK): divide the event into a grid and remove the squares below a threshold.





Small-R JES Calibration more in depth

Back to small-R
JES calibration

Pileup corrections

Absolute MC-based calibration

Global sequential calibration

Residual in situ calibration

1) Area-based correction with pile-up density ρ and jet area A.

2) Local MC-based correction. Characterization:

- In-time pileup: N_{PV}
- Out-of-time pileup: μ

$$p_T^{corr} = p_T^{reco} - \rho A$$

$$-\alpha (N_{PV} - 1) - \beta \mu$$

$$0.8 \quad \text{A7LAS Simulation}$$

$$0.6 \quad \text{NS} = 13 \, \text{TeV}, \text{ Pythia8 dijet}$$

$$0.4 \quad \text{Anti-k}, R = 0.4 \, \text{ (PFlow)}$$

$$0.2 \quad \text{O.2}$$

$$0.4 \quad \text{O.2}$$

$$0.4 \quad \text{O.2}$$

$$0.4 \quad \text{O.3}$$

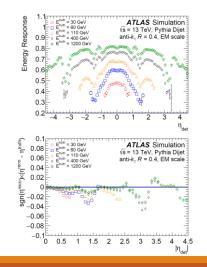
$$0.5 \quad \text{Anti-k}, R = 0.4 \, \text{ (priow)}$$

$$0.6 \quad \text{Anti-k}, R = 0.4 \, \text{ (priow)}$$

$$0.8 \quad \text{O.5} \quad \text{O.5} \quad \text{I.5} \quad \text{I.5$$

1) **Energy** is calibrated to the true jet energy from MC using a smooth multiplicative function in bins of η .

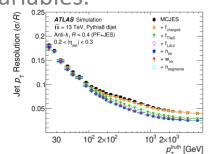
2) **Direction** η receives an additive correction.



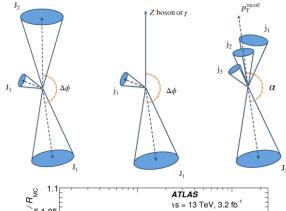
The scale depends on visible features of the jets, like the quark/gluon nature or the energy in dead material.

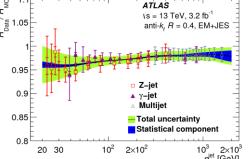
The resolution can be improved by removing these dependencies thanks to MC simulations.

This is done sequentially for 6 variables.



MC does not perfectly represent data → In situ corrections needed using different event topologies.







Improved Pileup Energy Density Determination

Including the

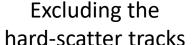
Pile-up energy density $\langle \rho \rangle$ determination:

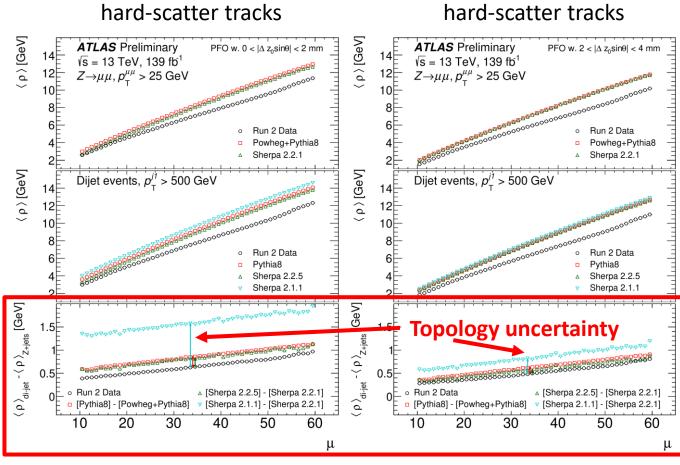
$$\langle \rho \rangle = \underset{i \in \{jets\}}{\text{median}} \left(\frac{p_{T,i}}{A_i} \right)$$

for two different selections: $Z o \mu \mu$ and dijet events.

- \rightarrow Previous determination was including the hard-scatter tracks, shifting $\langle \rho \rangle$ to higher values.
 - → Reduction in the dependency on hard scattering topologies
 - → Improvement on topology uncertainties by a factor of 3
- → Updated Sherpa Muti-Parton Interaction (MPI) results in a factor of 4 improvement
- → Total improvement of nearly a factor 7 of the topology uncertainty

Back to the JES uncertainties





Improved Pileup Residual Correction

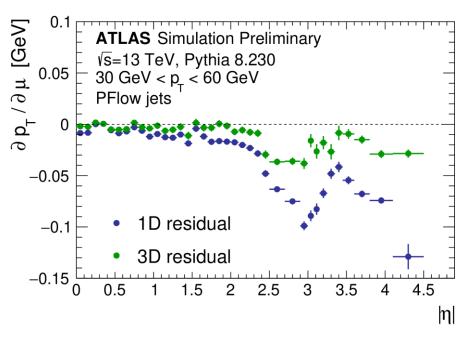
1D residual pileup correction: $p_T^{corrected} = p_T - \alpha(N_{PV} - 1) - \beta \mu$

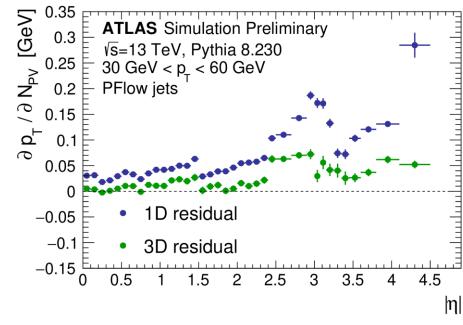
Back to the JES uncertainties

3D residual correction: $p_T^{corrected} = p_T - \Delta p_T(N_{PV}, \mu, \eta, p_T)$

The 3D residual correction takes into account correlations between N_{PV} , μ and jet properties.

→ Reduction of the pileup dependency above 30 GeV





GNNC architecture

3 hidden layers with swish activation functions (<u>arXiv:1710.05941</u>), each with 100 nodes, and a single-node output layer with linear activation.

Leaky Gaussian kernel loss function (arXiv:1910.03773), where the target is the jet p_T response.

Back to the GNNC



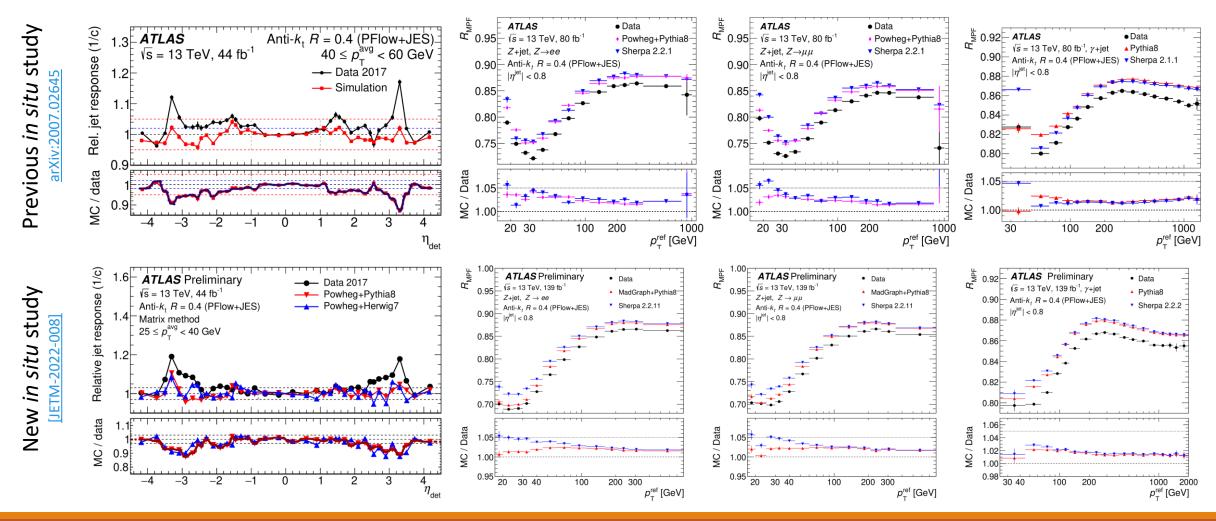
GNNC inputs

Back to the GNNC

Calorimeter	$f_{\text{LAr,0}-3}$	The fraction of energy measured in the 0th-3rd layer of the EM Lar calorimeter
	$f_{\rm TILE,0-2}$	The fraction of energy measured in the 0th-2nd layer of the hadronic tile calorimeter
	<i>f</i> _{HEC,0−3}	The fraction of energy measured in the 0th-3rd layer of the hadronic end cap calorimeter
	$f_{\text{FCAL},0-2}$	The fraction of energy measured in the 0th-3rd layer of the forward calorimeter
	N _{90 %}	The minimum number of clusters containing $90\ \%$ of the jet energy
Jet kinematics	$p_T^{ m JES}$	The jet p_T after the MC JES calibration
	$\eta_{ ext{det}}$	The detector η
Tracking	^w track	The average p_T -weighted transverse distance in the (η,ϕ) plane between the jet axis and all tracks of $p_T>1$ GeV ghost-associated with the jet
	$N_{\rm track}$	The number of tracks with $p_T>1~{ m GeV}$ ghost-associated with the jet
	$f_{\rm charged}$	The fraction of the jet p_{T} measured from ghost-associated tracks
Muon segments	N _{segments}	The number of muon track segments ghost-associated with the jet
Pileup	μ	The average number of interactions per bunch crossing
	N_{PV}	The number of reconstructed primary vertices

Validation of the MC calibration steps with the *In Situ* Techniques

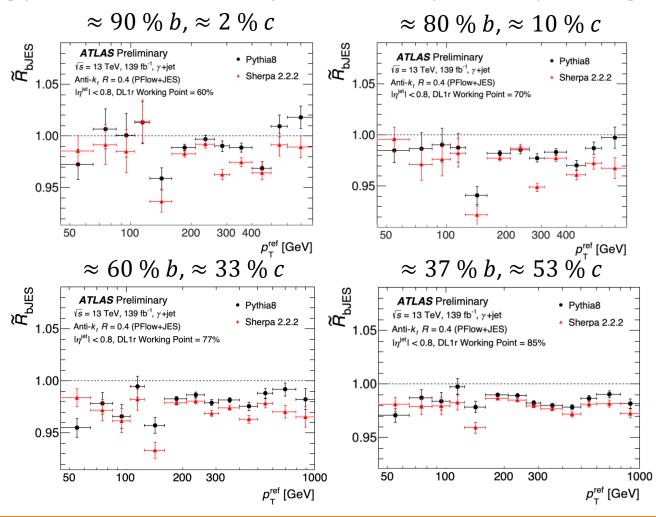
Improvements in the MC calibration steps are validated thanks to the in situ calibration techniques.





b Quark Jet Energy Scale for Different b-tagging Working Points

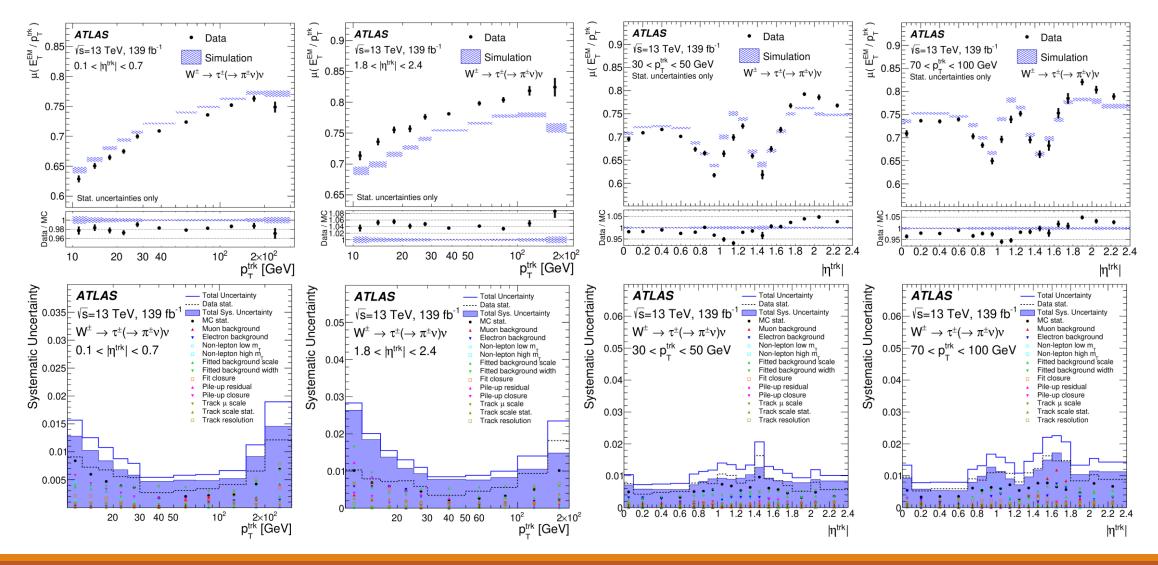
Multiple b-tagger working point for different compositions of b quarks, c quarks, light quarks and gluons.



Back to the *b*-JES

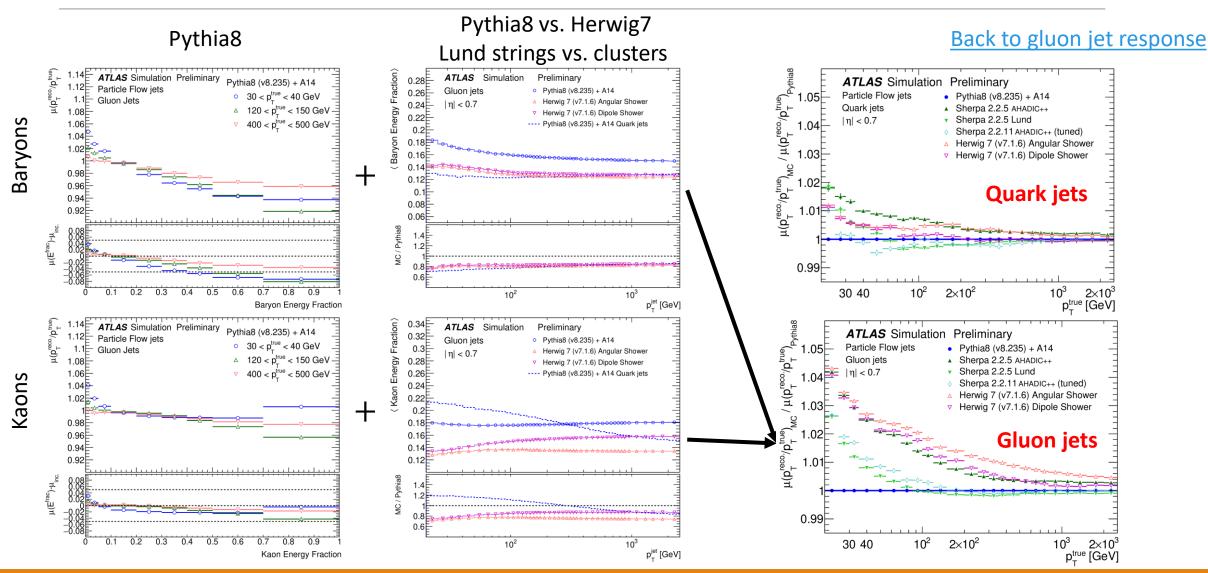
More E/p results

Back to E/p





Response of Gluon Jets vs. Baryon and Kaon Energy Fractions





Response After Reweighting of Hadron Content

Reweighting the events based on the baryon and kaon energy fractions reduces the differences.

