

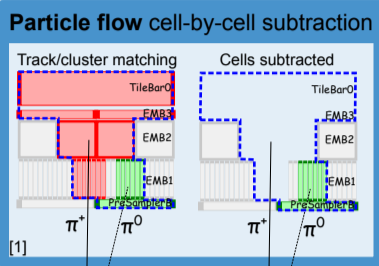
IN SITU MEASUREMENTS OF THE ATLAS JET ENERGY RESOLUTION USING 13 TeV PP DATA

LHCC Poster Session - CERN, 27 February 2019

JER Parametrization

$$\frac{\sigma_{p_T}}{p_T} = \frac{N}{p_T} \oplus \frac{S}{\sqrt{p_T}} \oplus C$$

- N: noise term** → pile-up and electronics noise
- S: stochastic term** → sampling calorimeter
- C: constant term** → passive material



The abundance of jets in proton-proton collisions makes it very important to understand the typical range of energy values a given jet can deposit in the detector. This quantifies the Jet Energy Resolution (JER) and is studied *in situ* by two analyses which are then inputs to a **statistical combination**. The first analysis measures the effect of **pile-up and detector noise** on the JER, the second measures the JER of anti- k_t $R = 0.4$ jets with p_T up to 1.5 TeV by using **well-balanced dijet systems**. This poster focuses on the JER measurement of **particle flow jets** that are reconstructed by combining information from topological clusters and tracks.

Noise Term Measurement

$$N_{\text{full}} = N_{\text{pile-up}} \oplus N_{\mu=0}$$

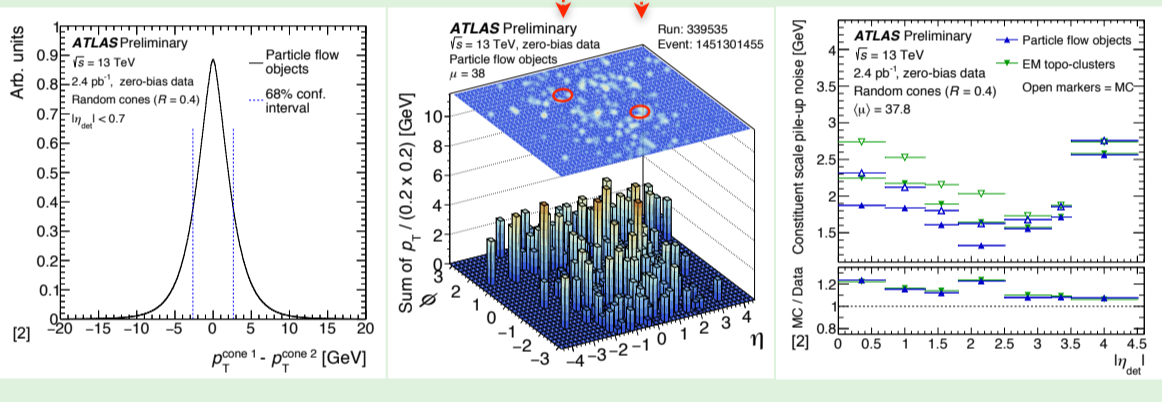
Noise from electronics

Extracted from JER of MC with no pile-up

Noise from pile-up

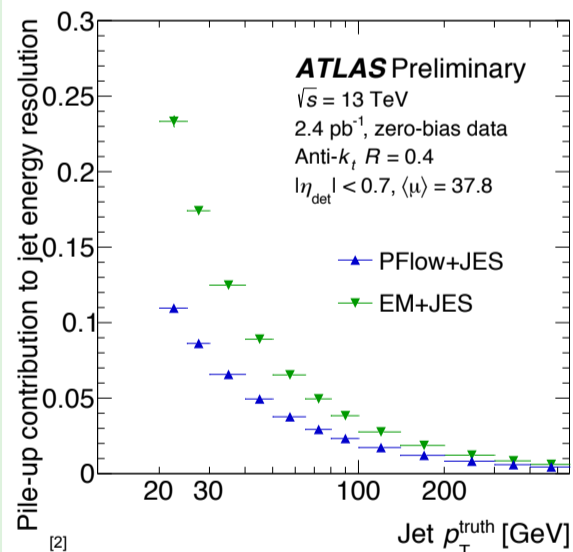
Random cones method:

- Sum energies within randomly distributed and non-overlapping pairs of cones ($R = 0.4$) in **zero-bias data**
- Width of momentum difference of two cones gives **noise at constituent scale**



Extraction of N_{full}

The random cones noise is scaled to JES in bins of particle-level jet p_T . The **pile-up noise term** is extracted from a fit to the pile-up resolution and added in quadrature to the noise from electronics.



Dijet Direct Balance

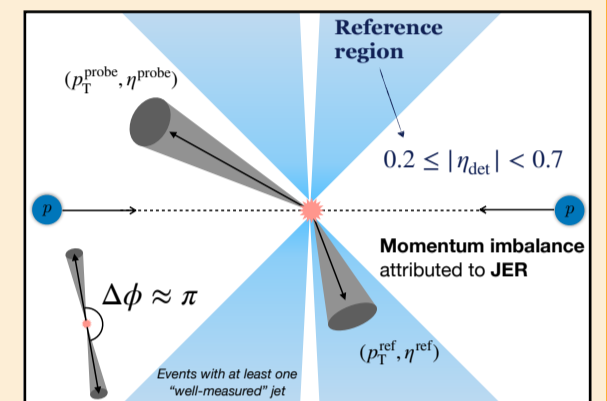
Resolution of probe jet studied from **momentum balance** with well-measured reference jet

$$A = \frac{p_T^{\text{ref}} - p_T^{\text{probe}}}{(p_T^{\text{ref}} + p_T^{\text{probe}})/2}$$

Studied across η with the **central reference method**:

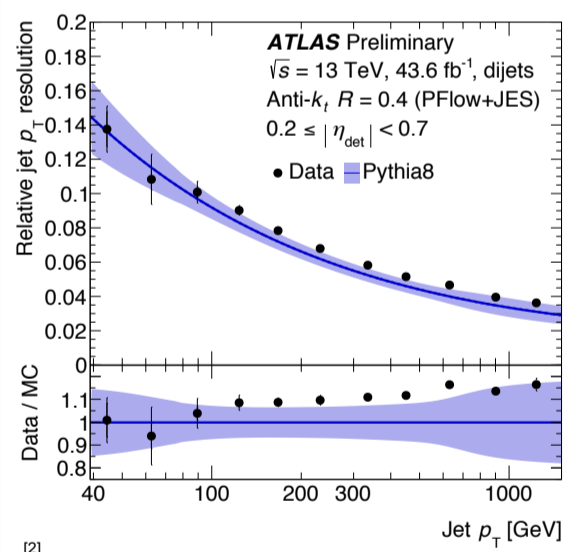
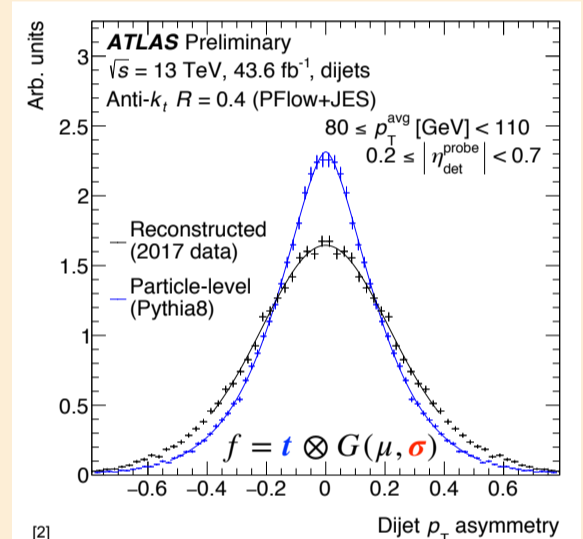
$$\left(\frac{\sigma_{p_T}}{p_T}\right)^{\text{ref}} = \frac{\sigma_A^{\text{det}}}{\sqrt{2}}, \text{ for } 0.2 \leq |\eta_{\text{det}}^{\text{probe}}| < 0.7$$

$$\left(\frac{\sigma_{p_T}}{p_T}\right)^{\text{probe}} = \sigma_A^{\text{det}} \ominus \left(\frac{\sigma_{p_T}}{p_T}\right)^{\text{ref}}, \text{ else}$$



Fit Convolution

The **particle-level asymmetry** (ISR, jet out-of-cone radiation) is modelled with an *ad-hoc* function f . Fitting the convolution with a **Gaussian** provides a measure of the detector JER σ .



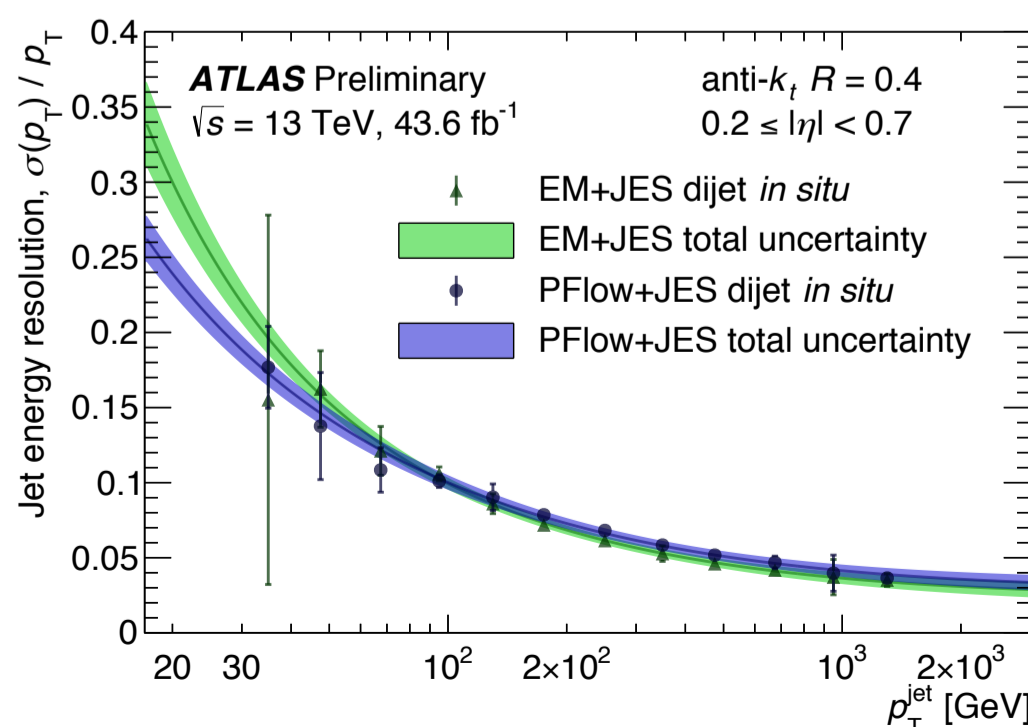
Comparison to MC

The Monte Carlo JER predictions are extracted from $\sigma(p_T^{\text{reco}}/p_T^{\text{truth}})$ and compared to the dijet results. After the combination, the average JER in MC is smeared to match data.

JER Combination

Combined fit

Dijet results fitted with N, S, C parametrization with fixed noise term from the random cones measurement. Particle flow jets show lower JER and smaller uncertainties at low p_T , which is driven by the noise term.



Uncertainties

The main systematic uncertainties arise from method non-closure, jet energy scale uncertainty, and the modelling of dijet events. The noise term uncertainties are important at low p_T .

