

Peter Loch University of Arizona

Tucson, Arizona-USA



Plots for this session

THE UNIVERSITY

OF ARIZONA.

Most if not all plots shown in this session are meant as examples and for illustration purposes

Educational showcases to highlight certain features of energy scales and calorimeter response

They do not represent the up-to-date estimates for ATLAS jet reconstruction performance

In general much better than the (old) results shown here!

Not many new plots can be shown in public yet!

The performance plots shown are published

Reflection of state-of-art at a given moment in time

No experimental collision data available at that time!



Summary Of Jet Inputs

Experiment and simulation

Calorimeter towers

2-dim signal objects from all cells or only cells surviving noise suppression (topological towers in ATLAS)

Calorimeter clusters

3-dim signal objects with implied noise suppression (topological clusters in ATLAS)

Tracks

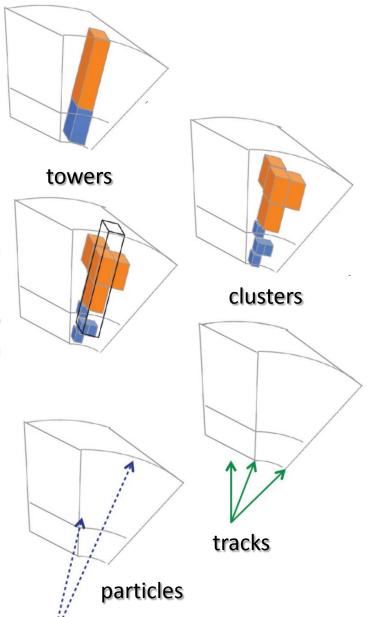
The University . of Arizona.

> Reconstructed inner detector tracks – only charged particles with $pT > pT_{threshold} = 500$ MeV – 1 GeV (typically)

Simulation only

Generated stable particles

Typically $\tau_{lab} > 10$ ps to be a signal source



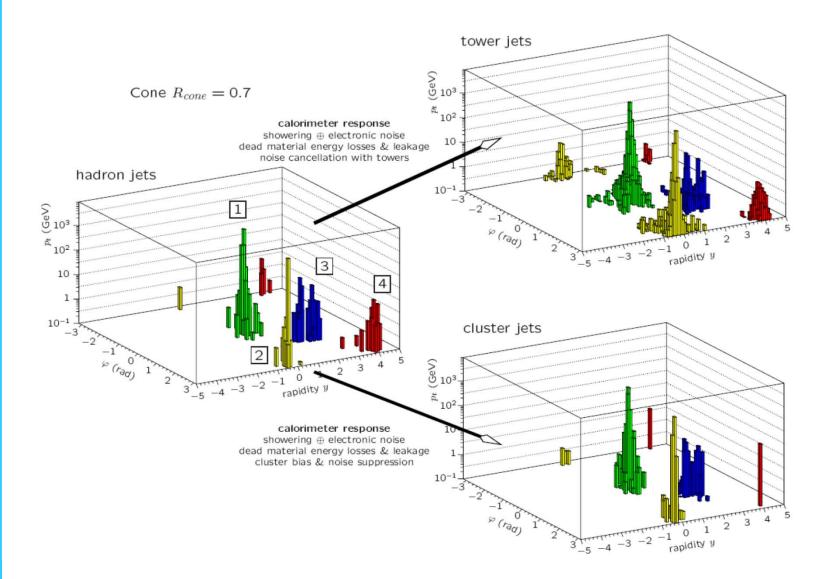




4

Image Of Jets In Calorimeter

P. Loch U of Arizona April 15, 2010





S.D. Ellis, J. Huston, K. Hatakeyama, P. Loch, M. Toennesmann, Prog.Part.Nucl.Phys.60:484-551,2008

Calorimeter jet response

THE UNIVERSITY

CF ARIZONA.

Electromagnetic energy scale

Available for all signal definitions

No attempt to compensate or correct signal for limited calorimeter acceptance

Global hadronic energy scale

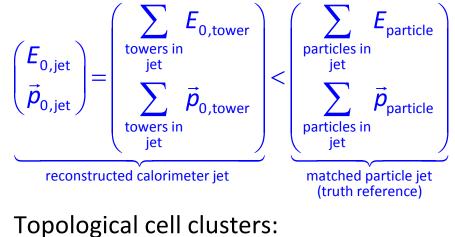
All signal definitions, but specific calibrations for each definition

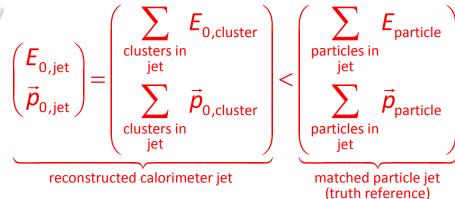
Calibrations normalized to reconstruct full true jet energy in "golden regions" of calorimeter

Local hadronic energy scale

Topological clusters only No jet context – calibration insufficient to recover calorimeter acceptance limitations – no corrections for total loss in dead material and Note at any scale: magnetic field charged particles losses)

Unbiased and noise-suppressed towers:





$$m_{\rm jet} = \sqrt{E_{\rm jet}^2 - \vec{p}_{\rm jet}^2} > 0$$
 for $N_{\rm towers}$, $N_{\rm clusters} > 1$



Calorimeter jet response

THE UNIVERSITY

. OF ARIZONA.



Available for all signal definitions

No attempt to compensate or correct signal for limited calorimeter acceptance

Global hadronic energy scale

All signal definitions, but specific calibrations for each definition

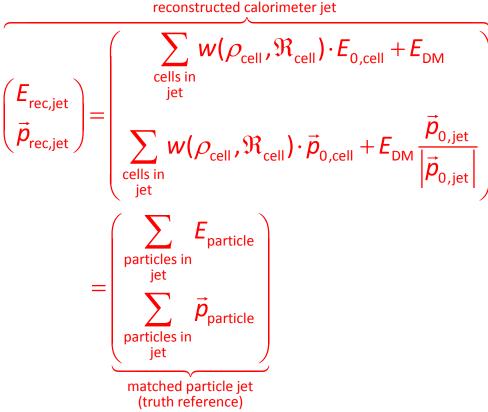
Calibrations normalized to reconstruct full true jet energy in "golden regions" of calorimeter

Local hadronic energy scale

Topological clusters only No jet context – calibration insufficient to recover calorimeter acceptance limitations – no corrections for total loss in dead material and magnetic field charged particles losses)

Cell based calibration for all calorimeter

signals and jets in "golden spot":



(cells are extracted from unbiased or noise suppressed

towers or topological clusters forming the jet)



Calorimeter jet response

THE UNIVERSITY

. OF ARIZONA.

Electromagnetic energy scale

Available for all signal definitions

No attempt to compensate or correct signal for limited calorimeter acceptance

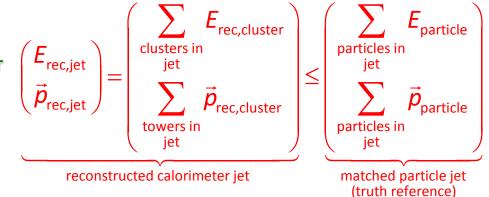
Global hadronic energy scale

- All signal definitions, but specific calibrations for each definition
- Calibrations normalized to reconstruct full true jet energy in "golden regions" of calorimeter

Local hadronic energy scale

Topological clusters only No jet context – calibration insufficient to recover calorimeter acceptance limitations – no corrections for total loss in dead material and magnetic field charged particles losses)

Locally calibrated clusters only:



THE UNIVERSITY . OF ARIZONA.

Final Jet Energy Scale (JES)

- **Final jet calibration**
 - All corrections applied

Best estimate of true (particle) jet energy

- Flat response as function of pT
- Uniform response across whole calorimeter
- **Relative energy resolution**
 - Depends on the calorimeter jet response calibration applies compensation corrections
- Resolution improvements by including jet signal features
 - Requires corrections sensitive to measurable jet variables
 - Can use signals from other detectors

Determination with simulations

- Measure residual deviations of the calorimeter jet response from truth jet energy
 - Derive corrections from the calorimeter response at a given scale as function of pT (linearity) and pseudorapidity (uniformity) for all particle jets
- Use numerical inversion to parameterize corrections
 - Conversion from truth variable dependence of response to reconstructed variable response



From simulations

. OF ARIZONA.

THE UNIVERSITY

Compare calorimeter response with particle jet energy as function of the particle jet energy

All jets, all regions, full kinematic coverage

Residual deviation from linearity

Depend on calorimeter energy scale – large for electromagnetic energy scale and local calibration due to missing jet level corrections

Small for global calibration due to jet energy normalization

Corrections can be extracted from residuals

A bit tricky – need to use numerical inversion (see later)

From experiment

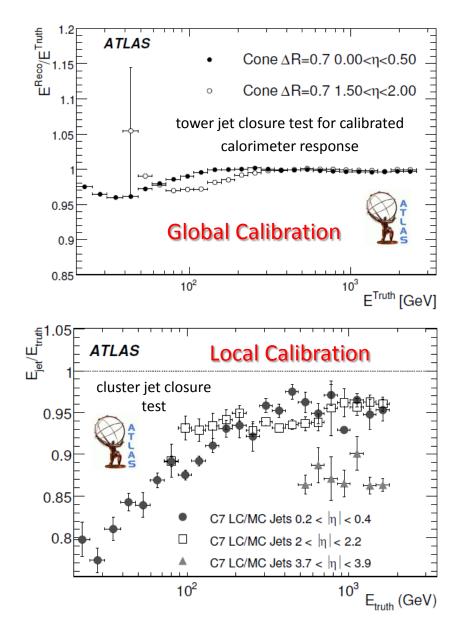
Validate and extract calibrations from collision data

W boson mass in hadronic decay is jet energy scale reference

pT balance of electromagnetic signal (Z boson, photon) and jet

Note change of reference scale

In-situ channels provide interaction (parton) level truth reference!





9

Signal Uniformity

Simulations

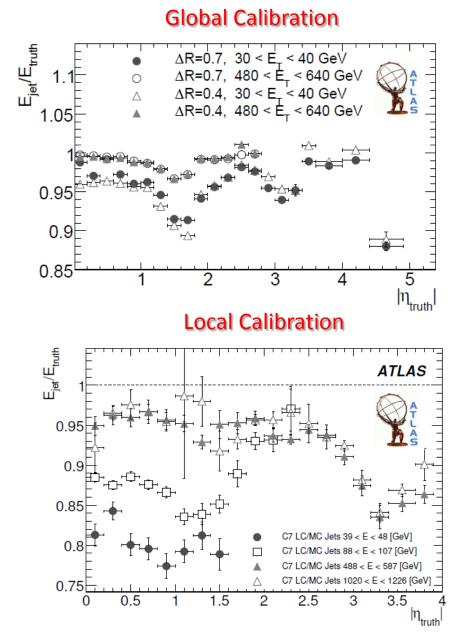
The University . of Arizona.

> Compare calorimeter response with particle jet energy as function of the jet direction All jets in full kinematic range Residual non-uniformities expected in cracks Only jets in "golden regions" used for calibration

From experiment

Di-jet pT balance

Balance pT of well calibrated jet in "golden region" with jet in other calorimeter regions Can also use photon pT balance with jets outside of "golden region"



10



ATLAS plots from **arXiv:0901.0512** [hep-ex]

Relative Jet Energy Resolution

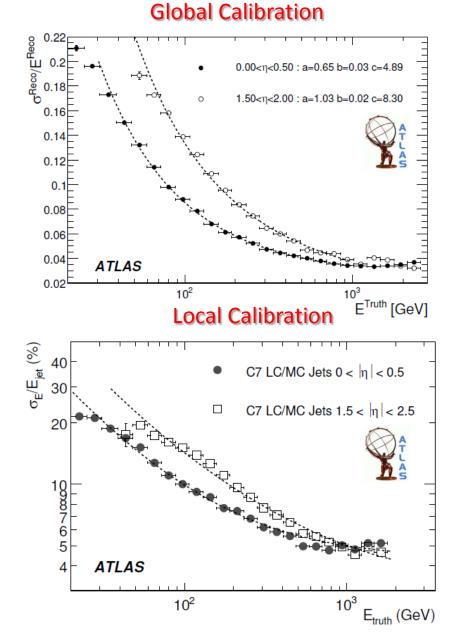
Simulations

Measure fluctuations of calorimeter jet energy as function of truth jet energy All jets in full kinematic range and in various regions of pseudo-rapidity

From experiment

Di-jet final states

Measure relative fluctuations of jet energies in back-to-back (pT) balanced di-jets





Golden rule of calorimetric energy measurement

The fully calibrated calorimeter signal is most probably the true jet (or particle) energy

Interpretation holds only for symmetrically distributed fluctuations – mean value is identical to average value

The resolution of the measurement is given by the characteristics of the signal fluctuations

Can only be strictly and correctly understood in case of Gaussian response distributions We need a normally distributed response!

Problem for all calibration techniques

Residual deviations from expected jet reconstruction performance must be measure as function of true quantities

Only then is the fluctuation of the response $R = E_{reco}/E_{true}$ really Gaussian after calibration

But need to apply corrections to measured jets

Need parameterization as function of reconstructed quantities

Simple re-binning does not maintain the Gaussian characteristics of the fluctuations – hard to control error!

Use numerical inversion to transfer the calibrations from true to measured parameters



Maintains Gaussian character

THE UNIVERSITY

. OF ARIZONA.

Toy model

The University • OF Arizona •

Generate flat jet energy spectrum

Uniform energy distribution for E_{jet} in $[E_{min}, E_{max}]$ Smear true jet energy with

Gaussian

Assume perfect average calibration

Width of distribution follows calorimetric energy resolution function

Calculate the response

In bins of E_{true} and in bins of $E_{smear} = E_{reco}$ Repeat exercise with steeply falling energy spectrum Calibrated response:

$$\langle E_{\rm smear} \rangle = \langle E_{\rm reco} \rangle = \langle E_{\rm true} \rangle$$

Calorimeter resolution function (no noise):

$$\frac{\sigma_{E}}{E} = \sqrt{\frac{a^{2}}{E_{true}} + c^{2}}$$

Smeared energy:

 $E_{\text{smear}} = E_{\text{true}} + r \cdot \sigma_{E}$

r is a random number following the Gaussian PDF:

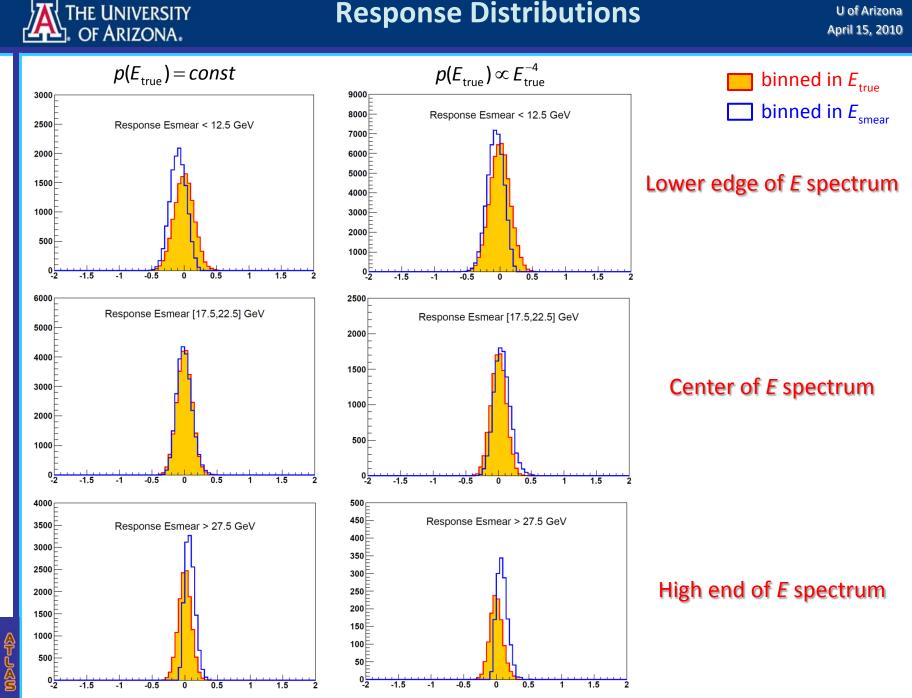
$$g(r) = \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2}r^2\right]$$

i.e. distributed around 0 with a width of 1 Response fluctuations:

$$R = \frac{E_{\text{smear}} - E_{\text{true}}}{E_{\text{true}}} \text{ with } \langle R \rangle = 0$$

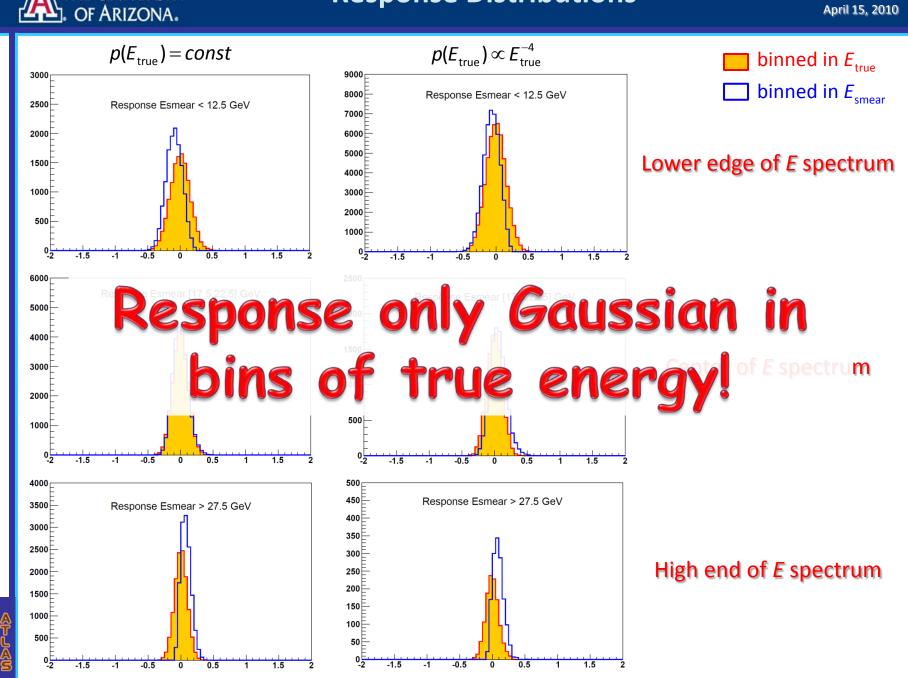


Response Distributions



14

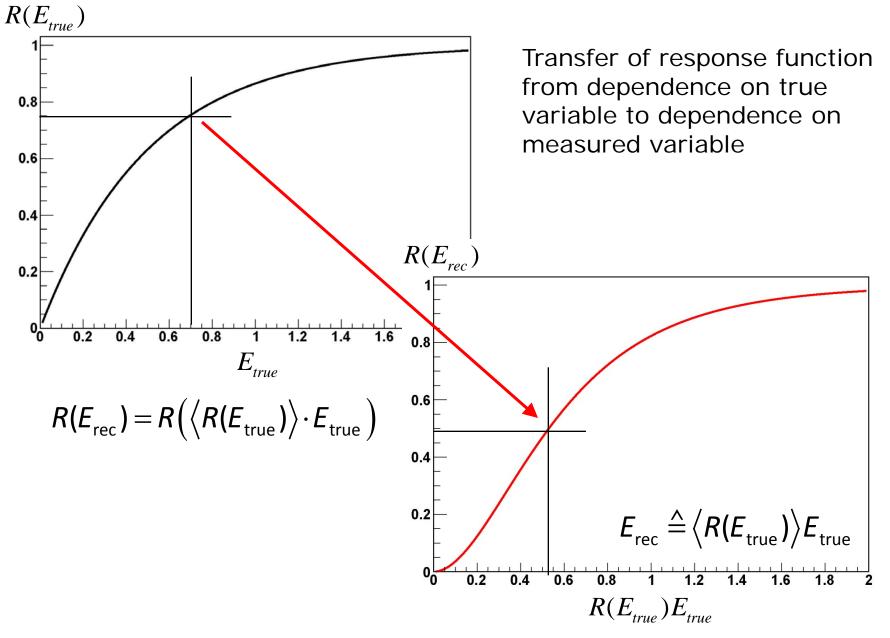
Response Distributions



THE UNIVERSITY







Often simple functions

THE UNIVERSITY

. OF ARIZONA.

Address residual energy (pT) and direction dependence of calorimeter jet response

> Determine response functions *R* in bins of true jet pT and reconstructed pseudo-rapidity $\eta_{\text{rec,jet}}$ Apply numerical inversion to determine calibration functions in reconstructed variable space ($p_{\text{T,rec,jet}}$, $\eta_{\text{rec,jet}}$)

Use calibration functions to get jet energy scale

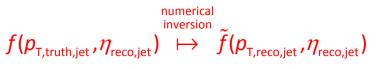
Technique can be applied to locally or globally calibrated jet response, with likely different calibration functions

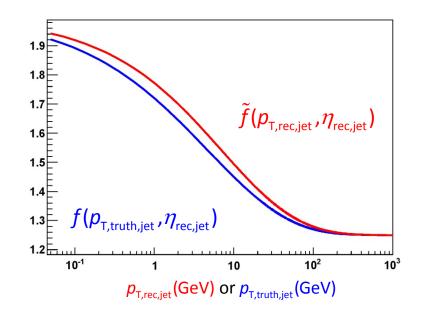
$$f(p_{T,truth,jet}, \eta_{reco,jet}) = R^{-1}(p_{T,truth,jet}, \eta_{reco,jet})$$

with $\eta_{reco,jet} \simeq \eta_{truth,jet}$ and

$$R^{-1}(p_{\mathrm{T,truth,jet}},\eta_{\mathrm{reco,jet}}) = \left\langle \frac{E_{\mathrm{truth,jet}}}{E_{\mathrm{rec,jet}}} \right\rangle (p_{\mathrm{T,truth,jet}},\eta_{\mathrm{reco,jet}})$$

then apply numerical inversion







Often simple functions

THE UNIVERSITY

. OF ARIZONA.

Address residual energy (pT) and direction dependence of calorimeter jet response

> Determine response functions R in bins of true jet pT and reconstructed pseudo-rapidity $\eta_{\rm rec, jet}$ Apply numerical inversion to determine calibration functions in reconstructed variable space (p_{T,rec,jet}, $\eta_{\rm rec,iet}$)

Use calibration functions to get jet energy scale

> Technique can be applied to locally or globally calibrated jet response, with likely different calibration functions

global calibration: $\left(\begin{array}{c} E_{\text{calib,jet}} \\ \vec{p}_{\text{calib,jet}} \end{array}
ight)$

$$ilde{f}(extsf{p}_{ extsf{T}, extsf{reco}, extsf{jet}}, extsf{\eta}_{ extsf{reco}, extsf{jet}}) \cdot$$

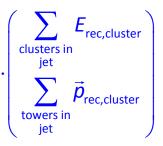
$$\sum_{\substack{\text{cells in}\\ \text{jet}}} W(\rho_{\text{cell}}, \Re_{\text{cell}}) \cdot E_{0,\text{cell}} + E_{\text{DM}}$$

$$\sum_{\substack{\text{cells in}\\ \text{jet}}} W(\rho_{\text{cell}}, \Re_{\text{cell}}) \cdot \vec{p}_{0,\text{cell}} + E_{\text{DM}} \frac{\vec{p}_{0,\text{jet}}}{|\vec{p}_{0,\text{jet}}|}$$

$$\sum_{\substack{\text{cells in}\\ \text{jet}}} W(\rho_{\text{cell}}, \Re_{\text{cell}}) \cdot \vec{p}_{0,\text{cell}} + E_{\text{DM}} \frac{\vec{p}_{0,\text{jet}}}{|\vec{p}_{0,\text{jet}}|}$$

local calibration:

$$\begin{pmatrix} E_{\text{calib,jet}} \\ \vec{p}_{\text{calib,jet}} \end{pmatrix} = \tilde{f}'(p_{\text{T,reco,jet}}, \eta_{\text{reco,jet}})$$







Why not use direct relation between reconstructed and true energy?

Same simulation data input Has been used in some experiments

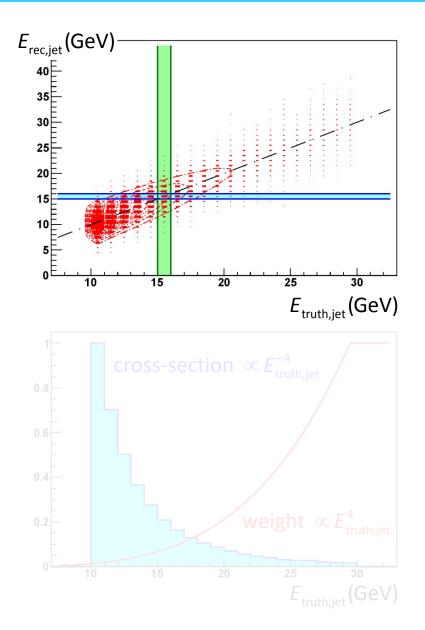
Dependence on truth energy spectrum

Need to make sure calibration sample is uniform in truth energy

Alternatively, unfold driving truth energy spectrum

Residual non-gaussian behaviour of truth energy distribution

Error on reconstructed energy hard to understand Could still use response distribution → same issues as discussed on previous slide!





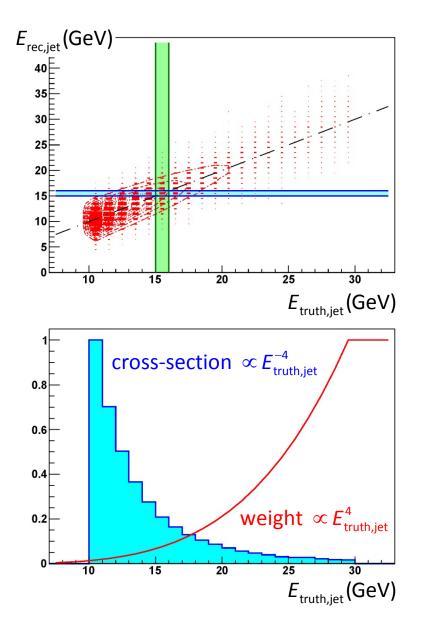
The University . Of Arizona.

Why not use direct relation between reconstructed and true energy?

- Same simulation data input
 - Has been used in some experiments
- Dependence on truth energy spectrum
 - Need to make sure calibration sample is uniform in truth energy
 - Alternatively, unfold driving truth energy spectrum

Residual non-gaussian behaviour of truth energy distribution

Error on reconstructed energy hard to understand Could still use response distribution → same issues as discussed on previous slide!



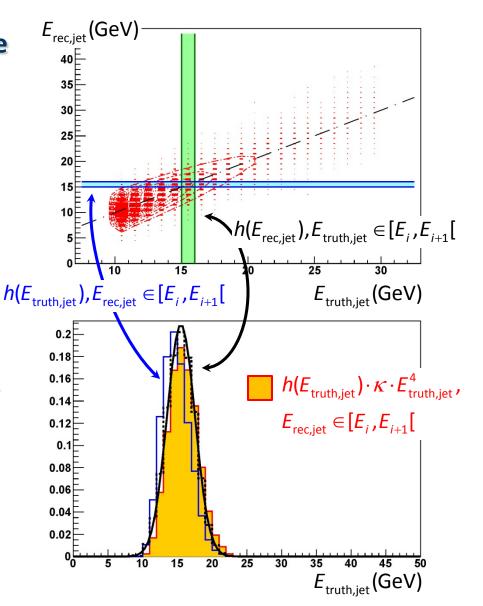


Why not use direct relation between reconstructed and true energy?

THE UNIVERSITY

. OF ARIZONA.

- Same simulation data input
 - Has been used in some experiments
- Dependence on truth energy spectrum
 - Need to make sure calibration sample is uniform in truth energy
 - Alternatively, unfold driving truth energy spectrum
- Residual non-gaussian behaviour of truth energy distribution
 - Error on reconstructed energy hard to understand Could still use response distribution → same issues as
 - distribution → same issues as discussed on previous slide!





Strategy from simulations

THE UNIVERSITY

T. OF ARIZONA.

- Determine all calibrations with fixed conditions
 - Ideal detector model everything is aligned
 - Fixed (best) GEANT4 shower model from testbeam evaluations
 - Fixed calorimeter signal definition e.g., towers
 - Fixed jet definition like seeded cone with size 0.7
 - Fixed final state QCD di-jets preferred
- Study change in performance for changing conditions with ideal calibration applied
 - Detector misalignment and changes in material budgets
 - Different shower GEANT4 model
 - Different calorimeter signal definitions e.g., clusters
 - Different jet definitions e.g., kT, AntikT, different cone or cone sizes...
 - Different physics final state preferably more busy ones like SUSY, ttbar,...
- Use observed differences as systematic error estimates

Use of collision data

- Compare triggered final states with simulations
 - Level of comparison represents understanding of measurement systematic error (at least for standard final states)
- Use in-situ final states to validate calibration
 - Careful about biases and reference levels (see session 9)



Calibration functions determined with "perfect" detector description and one reference jet definition

Validate performance in perfect detector

Signal linearity & resolution

Quality of calibration for a real detector

A priori unknown real detector

Absolute and relative alignments, inactive material distributions Estimate effect of distorted (real) detector

Implement realistic assumptions for misalignment in simulations Small variations of inactive material thicknesses and locations But use "perfect" calibration for reconstruction

Change jet signals

Tower or clusters

E.g, change from reference calorimeter signal Different jet finder

E.g., use kT instead of cone

Different configuration

E.g., use narrow jets (cone size 0.4) instead of wide jets (0.7)



Response

THE UNIVERSITY

. OF ARIZONA.

Linear within +/-1% after calibration applied for pT>100 GeV

Clear improvement compared to basic signal scale

Problems with low pT regime

ATLAS limit pT>20-40 GeV, depending on luminosity

May be resolution bias - under study

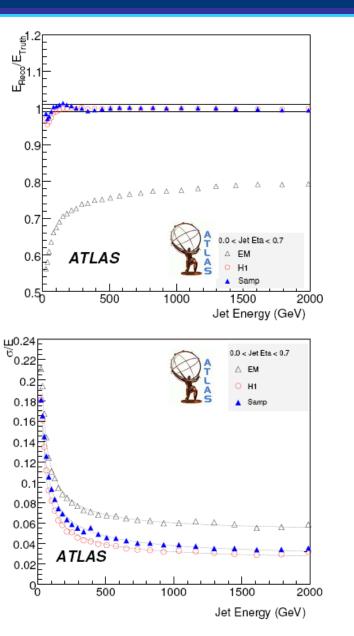
Resolution

Jet energy resolution clearly improved by calibration as well

Slight dependence on calibration strategy

Close to required performance

$$\frac{\sigma}{E} \approx \frac{65\%}{\sqrt{E}} \oplus 3\%$$



24





25

THE UNIVERSITY Signal Uniformity

Characterizes "real" detector jet response

Variation of response with direction

Changing inactive material distribution

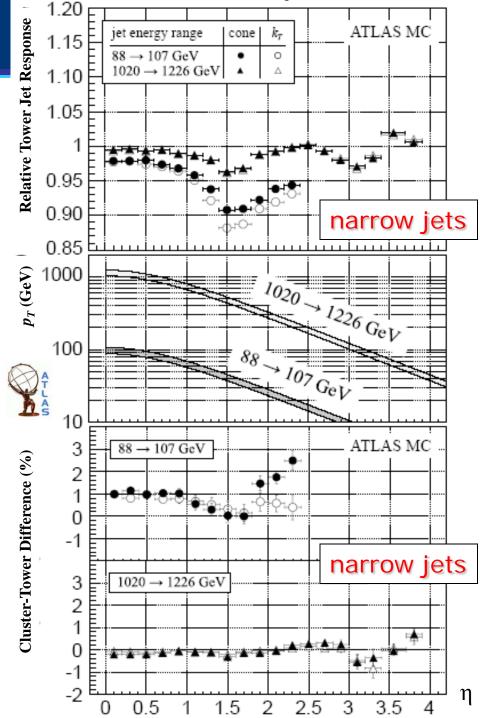
Cracks between calorimeter modules

Variations

No strong dependence on calorimeter signal definition

Towers/clusters

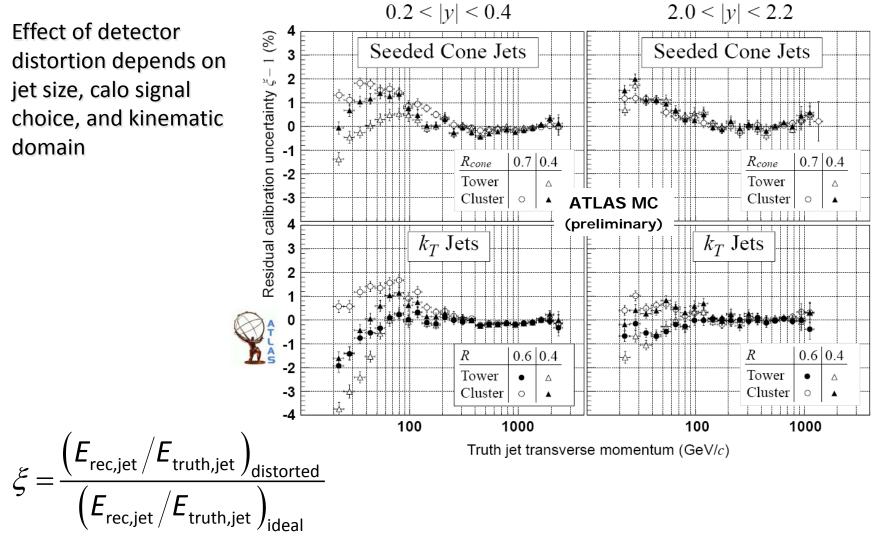
ATLAS cone jet performs better in crack region at low pT



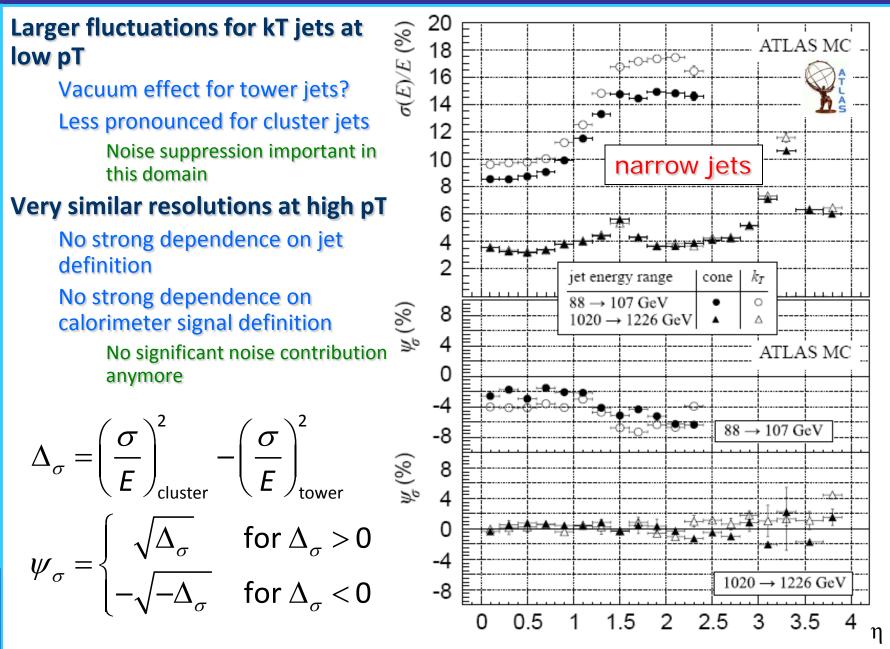


Estimated effect of a distorted detector

Effect of detector jet size, calo signal domain







27

THE UNIVERSITY

. OF ARIZONA.





Different Final States: Quark Jets

1.2 E^{Reco}/E^{Truth} E_T^{Reco}/E_T^{Truth} 0.00<η<0.50 Cone ∆R=0.4 39<E_T< 48 GeV 1.15 1.00<η<1.50 Cone ∆R=0.4 107<E_T< 130 GeV 0 Cone $\Delta R=0.4$ 336< E_T < 405 GeV 2.50<n<3.00 Φ ക 1.1 1.1 1.05 1.05 0.95 0.95 $t\overline{t} \rightarrow qqb$ $t\overline{t} \rightarrow qqb$ 0.9 0.9 ATLAS ATLAS 0.85<mark>L</mark> 0.85 0.5 2 2.5 3 3.5 4.5 1.5 10² 10^{3} E^{Truth} [GeV] nTruth E_T^{Reco}/E_T^{Truth} E^{Reco}/E^{Truth} 1.2 1.2 Cone $\Delta R = 0.4 \ 0.00 < \eta < 0.50$ Cone $\Delta R=0.4$ 39< E_T< 48 GeV 1.15 1.15 Cone ∆R = 0.4 1.00<η<1.50 Cone ∆R=0.4 107<E_T< 130 GeV 0 0 Cone $\Delta R=0.4$ 336< E_{τ} < 405 GeV Cone $\Delta R = 0.4 \ 2.50 < \eta < 3.00$ Φ Φ 1.1 1.1 **SUSY** 1.05 1.05 (high mulitplicity q jets) 0.95 0.95 SUSY 0.9 0.9 (high mulitplicity q jets) ATLAS ATLAS 0.85<mark>0</mark> 0.85 10² 0.5 1.5 2 2.5 3 3.5 4.5 10³ E^{Truth} [GeV] η^{Truth}

