



ALICE

Measurement of jet spectra in Pb-Pb collisions at $\sqrt{s_{NN}}=2.76$ TeV with ALICE

Salvatore Aiola,
on behalf of the ALICE collaboration
INFN Catania and LBNL

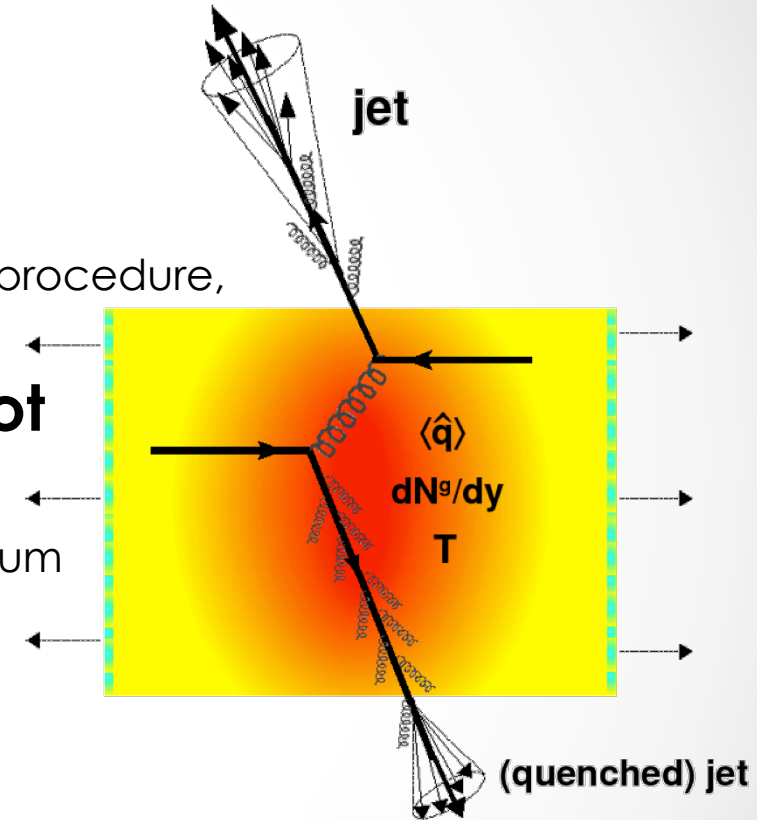


Outline

- Introduction
- Analysis overview
- Jet reconstruction
- Background estimation
- Unfolding
- Jet spectra
- Conclusions & Outlook

Jets in heavy-ion collisions

- Jet: **collimated spray of hadrons**
 - QCD branching of a high p_T parton
 - Subsequent hadronization of fragments
 - Experimentally grouped according to given procedure, *jet algorithm*
- Jets can be used to **probe the hot QCD medium**
 - Observable properties modified by the medium
 - p_T distribution
 - ...and many more



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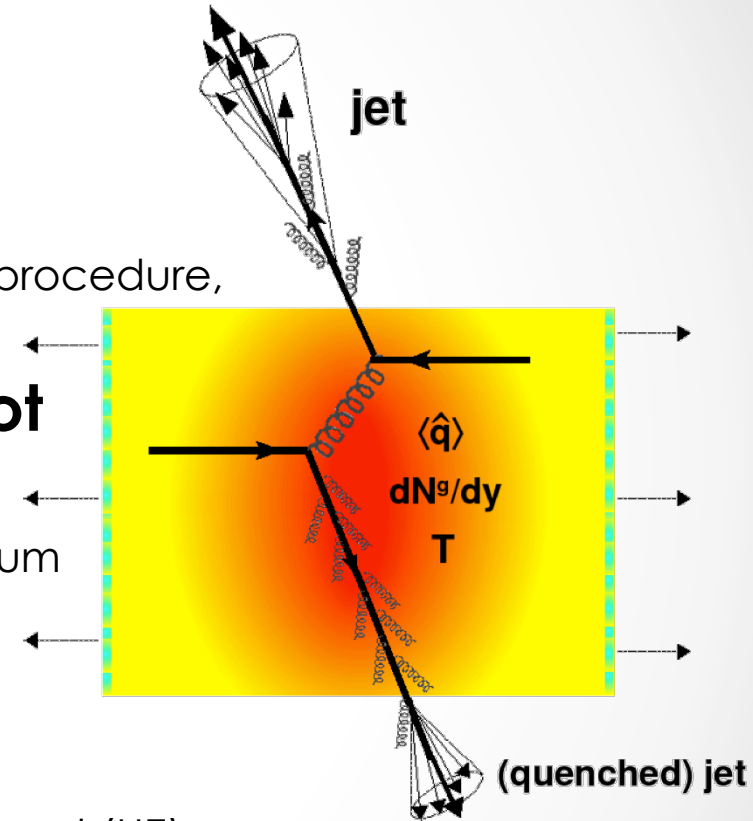
- **Experimentally challenging**

- Huge background given by the underlying event (UE)

1. **Average background**

2. **Combinatorial jets**

3. **p_T smearing due to background fluctuations**



Jet-by-jet and event-by-event correction

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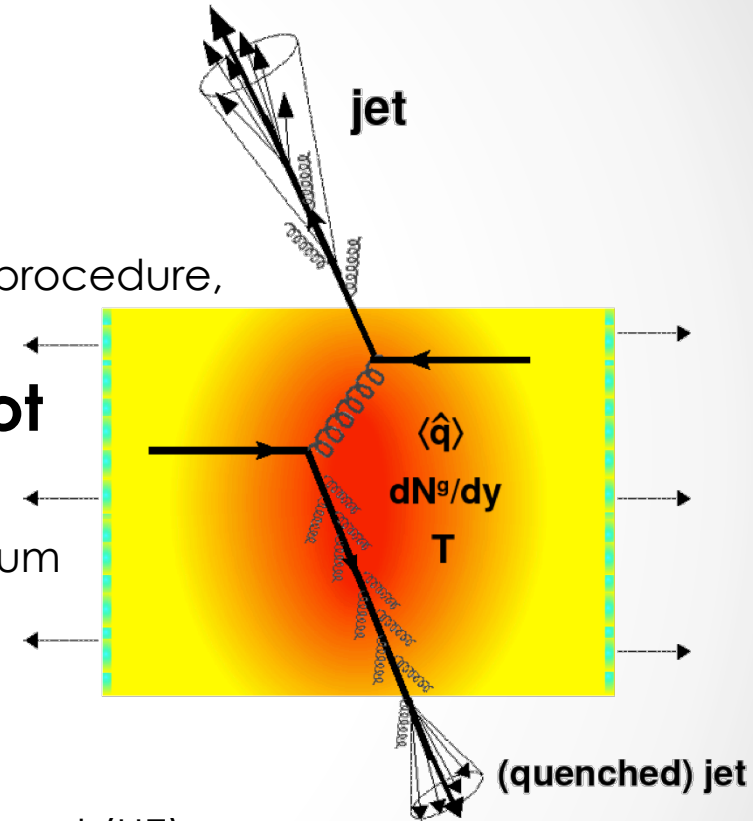
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Efficiently removed
by a leading hadron
 p_T requirement



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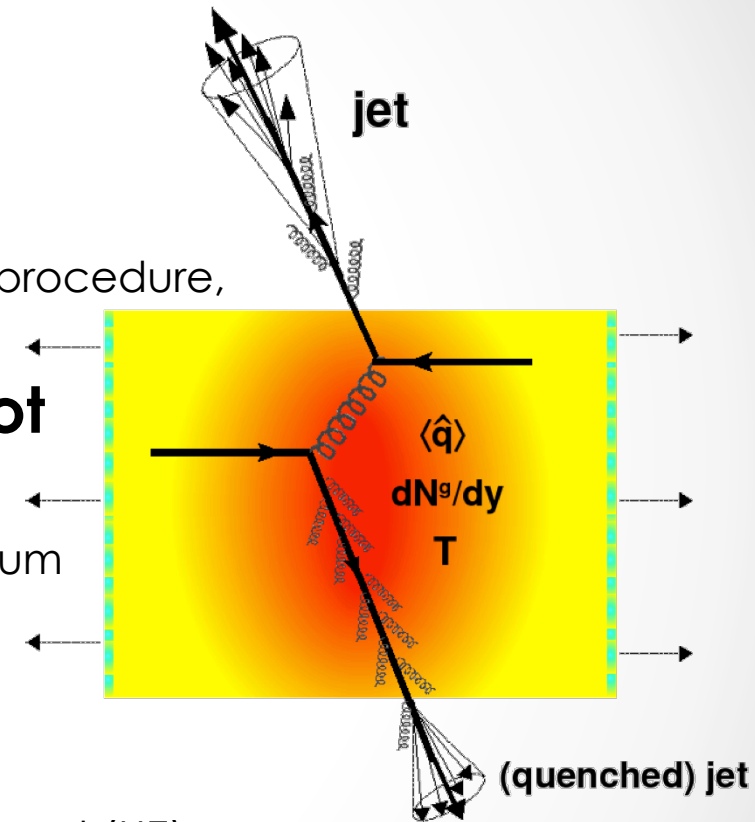
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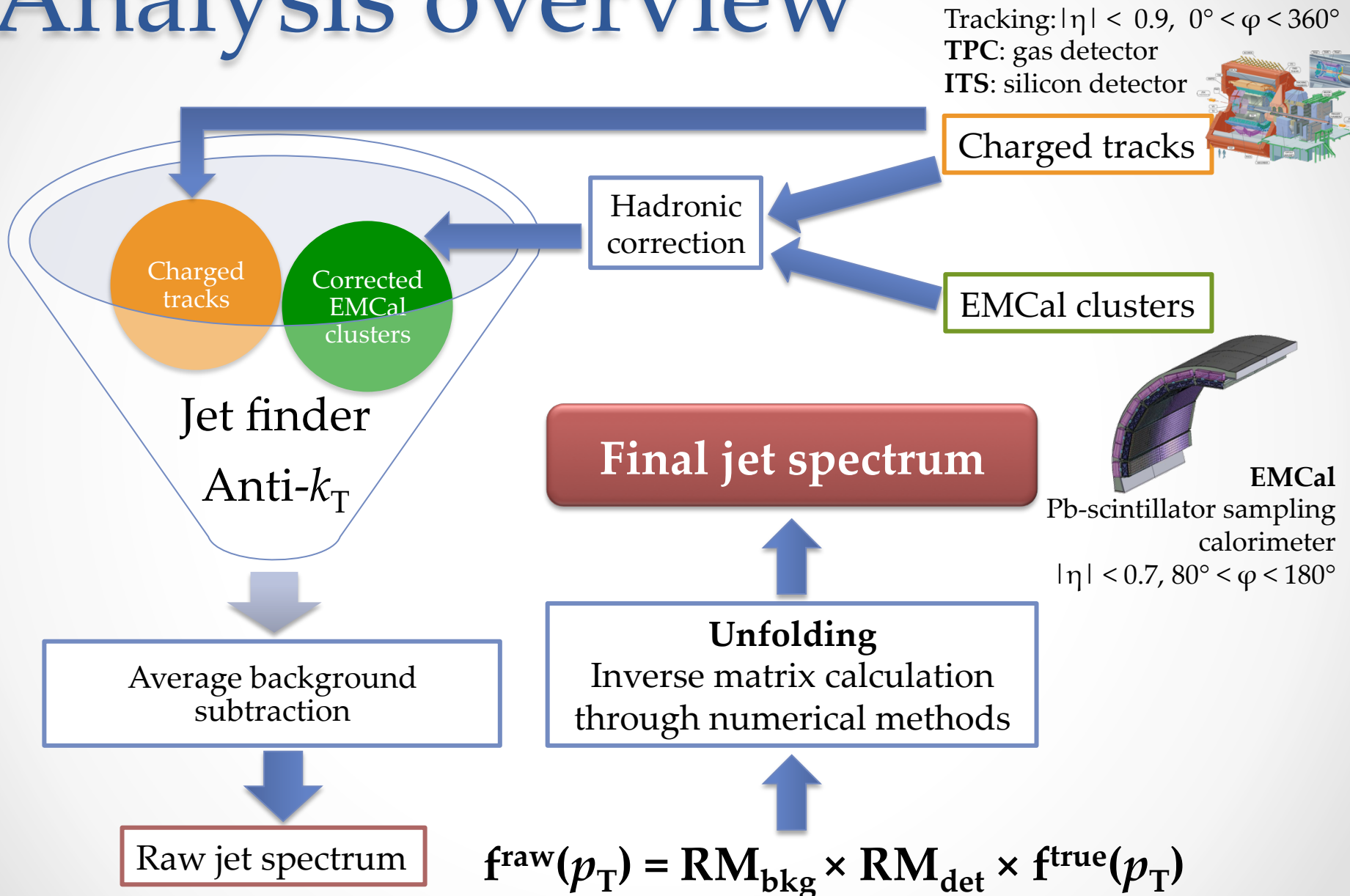
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3. **p_T smearing due to background fluctuations**



Statistical
correction based on
 δp_T distributions

Analysis overview



Jet reconstruction

- Inputs to the jet finder
 - Assumed to be massless
 - **Charged tracks** with $p_T > 150 \text{ MeV}/c$
 - **EMCal clusters** with $E_T > 300 \text{ MeV}/c$ after charged particle correction
- Jet reconstructed using FastJet* package
 - **Infrared-** and **Collinear-Safe** algorithms
 - Good for comparison with theory
 - **Radii** $R = 0.2, 0.3$
 - **Area cut** $A > 0.6 * \pi R^2$ removes extremes
 - Fiducial cut selects jets **fully contained in the EMCal** acceptance

Sequential recombination

$$d_{ij} = \min(k_{Ti}^{2p}, k_{Tj}^{2p}) \frac{\Delta_{ij}^2}{R^2} \quad d_{iB} = k_{Ti}^{2p}$$

$p = 1$ **k_T algorithm**

Clustering starts from the low p_T particles

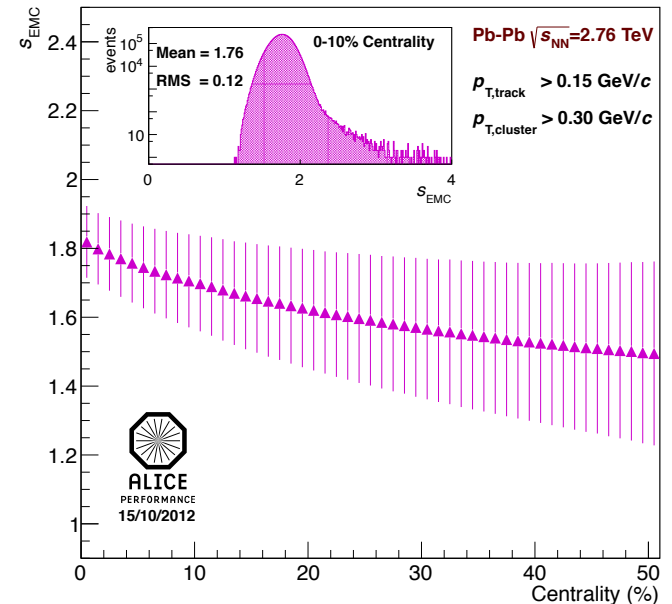
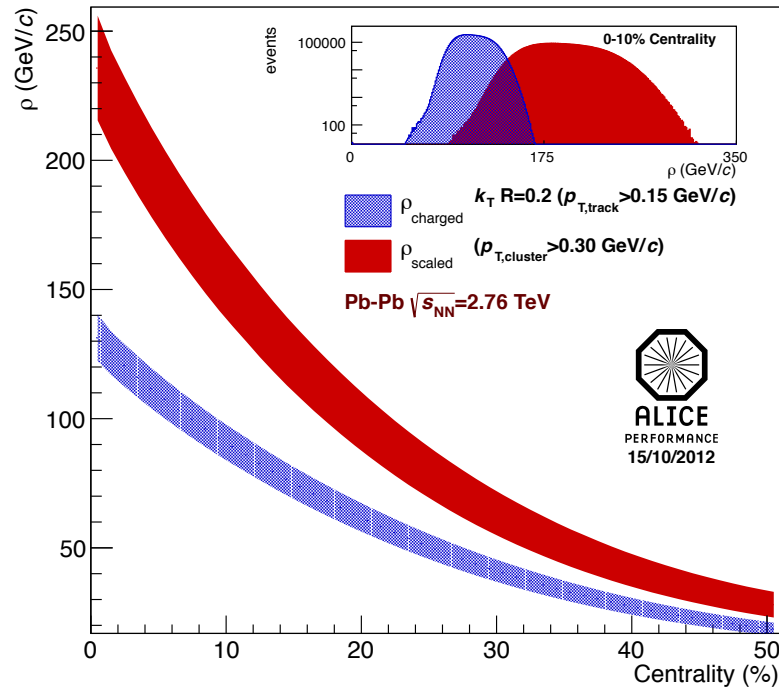
Used for q calculation

$p = -1$ **Anti- k_T algorithm**

Clustering starts from the highest p_T particle, regular cone-like jet shapes

Used for signal jets

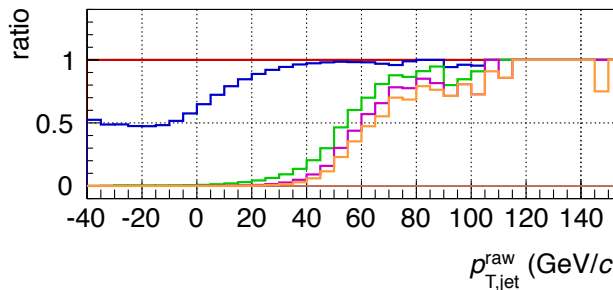
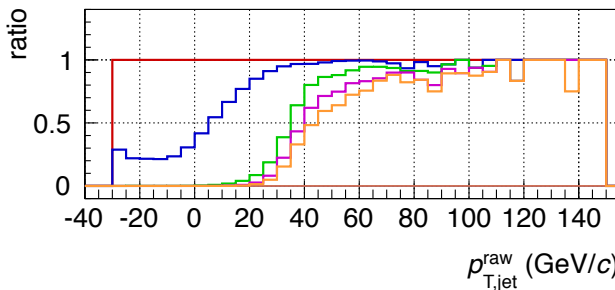
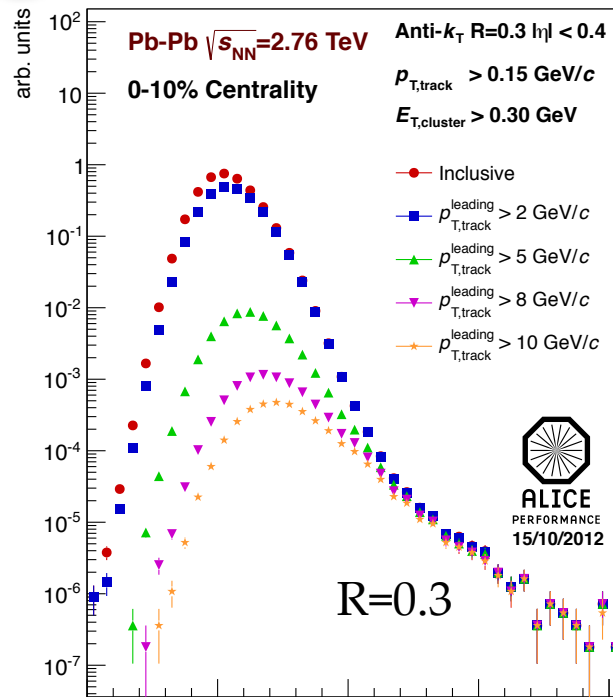
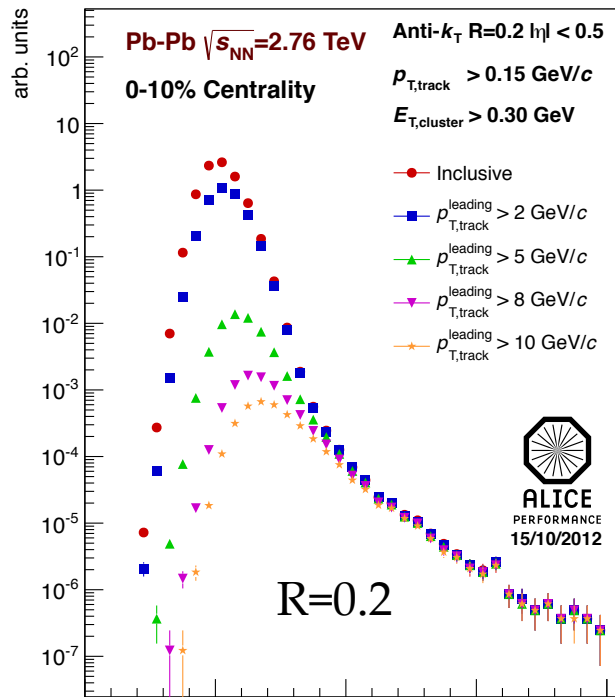
1. Average background



Charged+neutral momentum density / charged only momentum density

- Calculated on an **event-by-event basis**
- The median of the density of k_T charged jets gives $\rho_{\text{ch}} = \text{median} \left\{ \left[\frac{p_{Tj}^{k_T}}{A_j^{k_T}} \right] \right\}$
- A parametrization of the scale s_{EMC} factor as a function of centrality is used to obtain $\rho_{\text{ch+ne}} = \rho_{\text{ch}} \times s_{EMC}$

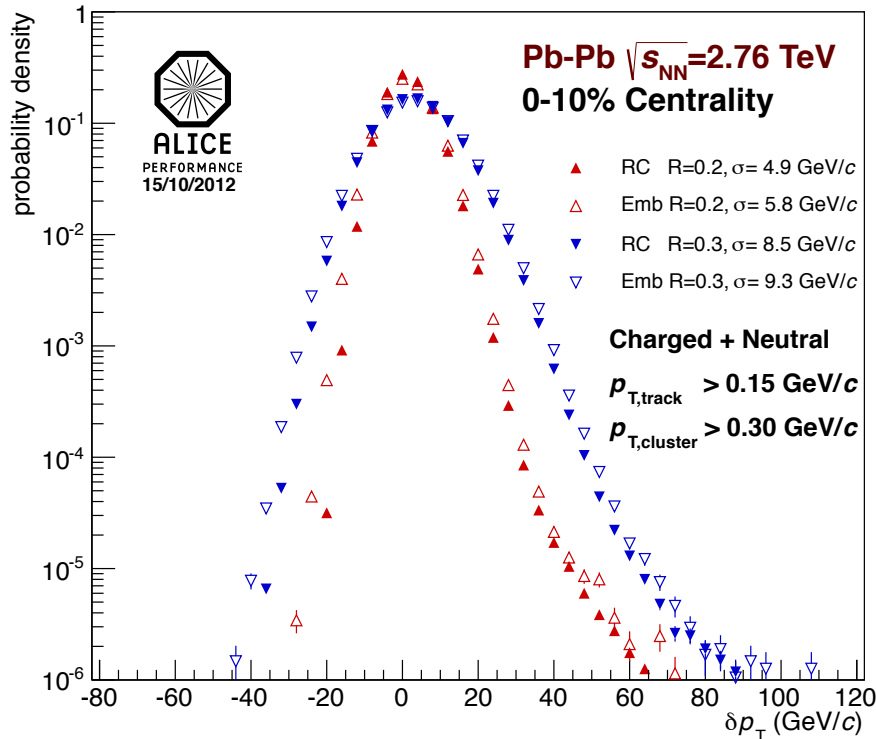
2. Leading hadron trigger



5 GeV/c is our default choice for R=0.2 jet spectra

- **Combinatorial jets:** jet finder clusters together also soft particles from the UE
- Efficiently remove the fake jets by **requiring a high p_T constituent**
- Bigger effect when changing the p_T threshold from 0 \rightarrow 5 GeV/c than 5 \rightarrow 10 GeV/c
- Bias still effective up to 100 GeV/c

3. Background fluctuations



- Region-by-region background fluctuations are estimated through **random cones** (solid symbols) and **single particle embedding** (open symbols)
- Fluctuations limit **jet energy resolution**
 - **Smearing** of the p_T raw jet spectrum

Random cones

1. Throw a random direction in the (η, ϕ) plane
2. Sum up the momenta of all particles in the cone of radius R
3. Calculate

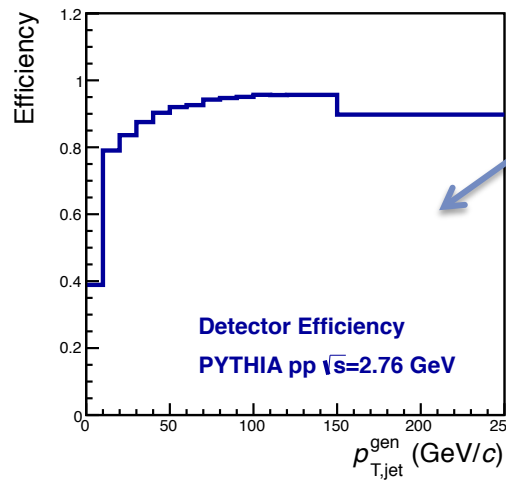
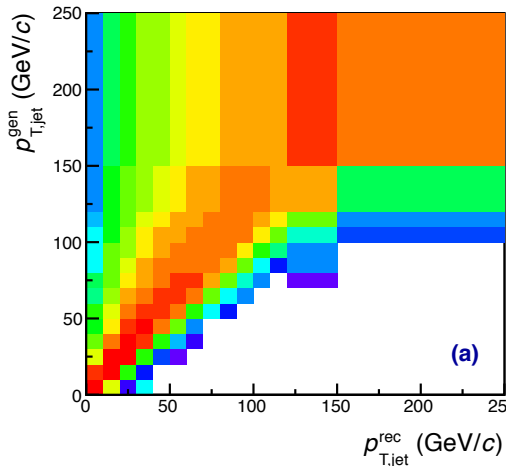
$$\delta p_T^{RC} = p_T^{RC} - \pi R^2 \rho$$

Single particle embedding

1. Embed a high p_T particle
2. Run the Anti- k_T jet finder
3. Pick up the jet which contain the embedded particle and calculate

$$\delta p_T^{emb} = p_T^{jet} - p_T^{probe} - A^{jet} \rho$$

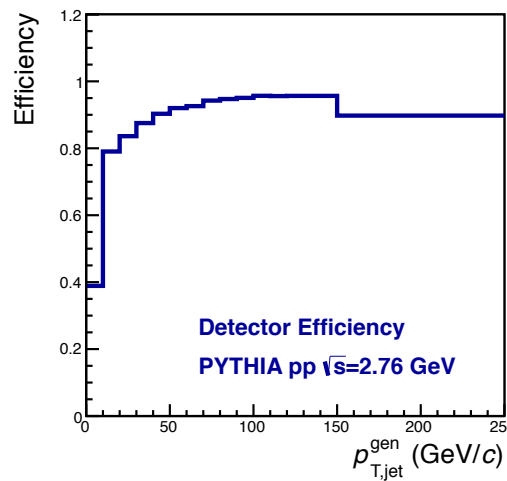
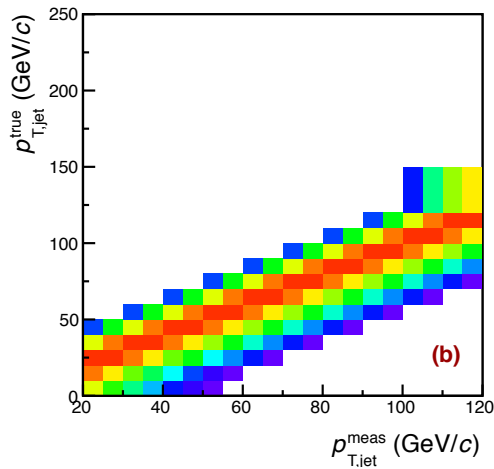
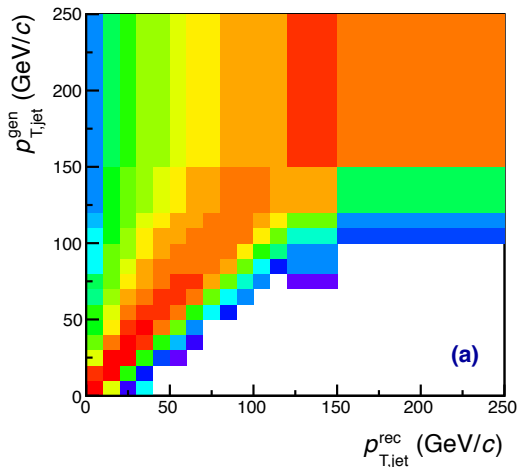
Response matrices



- **Detector effects**

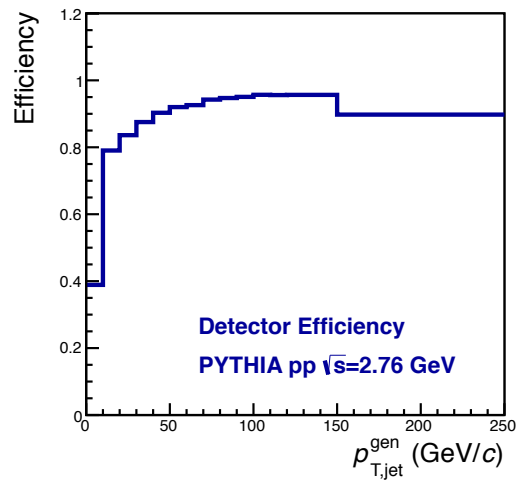
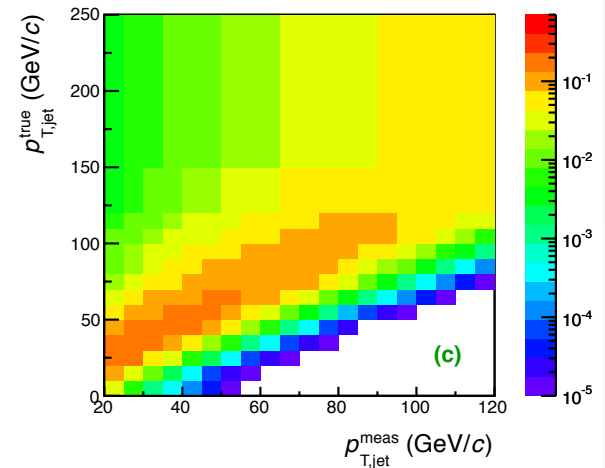
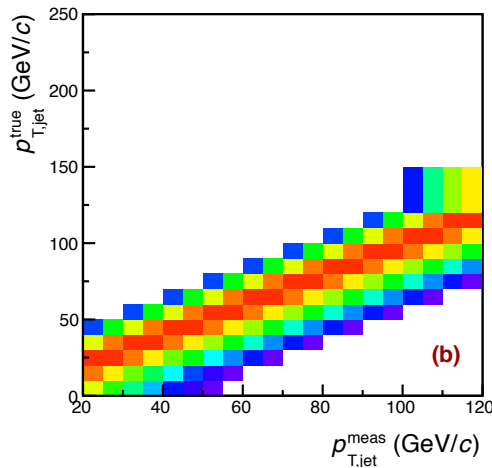
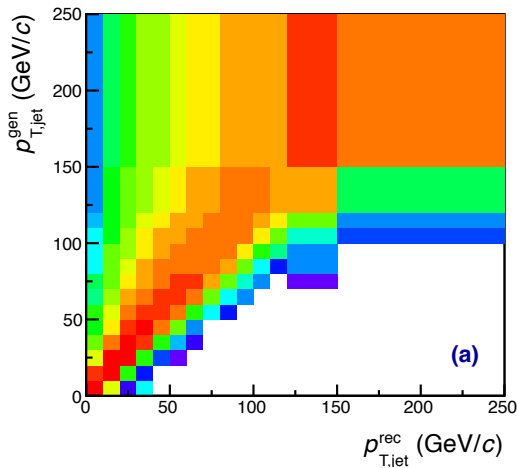
- PYTHIA+GEANT3 simulation
 - Assumes jet detector response is the same as in **vacuum fragmentation**
- Shift **jet energy scale** $\sim 20\text{-}25\%$
 - Unreconstructed neutrons and K_L^0
 - Tracking inefficiency
- **Jet energy resolution** $\sim 18\%$
- **Jet reconstruction efficiency** $\sim 95\%$ at 80 GeV/c

Response matrices



- **Background fluctuations**
 - δp_T distribution (RC)
 - $R = 0.2 \rightarrow \sigma = 4.9$ GeV/c
 - $R = 0.3 \rightarrow \sigma = 8.5$ GeV/c

Response matrices



Anti- k_T R=0.2

$$p_{T,track} > 0.15 \text{ GeV}/c$$

$$E_{T,cluster} > 0.30 \text{ GeV}$$

$$p_{T,track}^{leading} > 5 \text{ GeV}/c$$

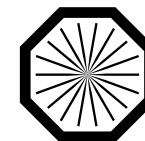
(a) RM_{det} Detector response matrix

(b) RM_{bkg} Background fluctuation matrix

$$(c) RM_{tot} = RM_{bkg} \times RM_{det}$$

Pb-Pb $\sqrt{s_{NN}}=2.76$ TeV

0-10% Centrality



ALICE
PERFORMANCE
15/10/2012

Why unfolding

- **Remove detector and analysis specific features** from the observables to be compared with theory
- Another **simpler but less elegant approach**: use the response matrix to include these distortions in the theoretical predictions
- **Advantages of unfolding**
 - Comparison between experiments is possible
 - Final results are more “fundamental” and easier to read

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 - **χ^2 minimization**

$$\chi^2 = \sum_{\text{refolded}} \left(\frac{y_{\text{refolded}} - y_{\text{measured}}}{\sigma_{\text{measured}}} \right)^2$$

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- **Bayesian**

- Uses the Bayes' theorem to iteratively update the estimators' probabilities, starting from a “prior” guess

Bayes' theorem

$$P(C_i | E) = \frac{P(E | C_i)P(C_i)}{\sum_{l=1}^{n_c} P(E | C_l)P(C_l)}$$

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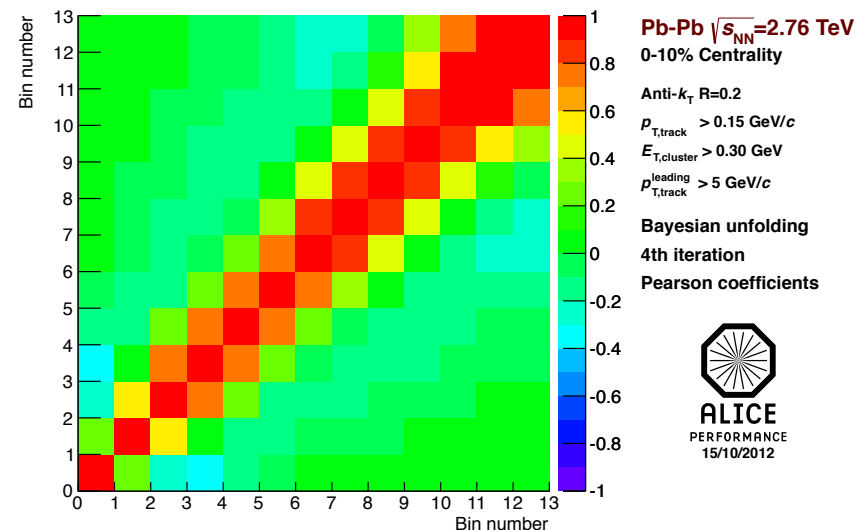
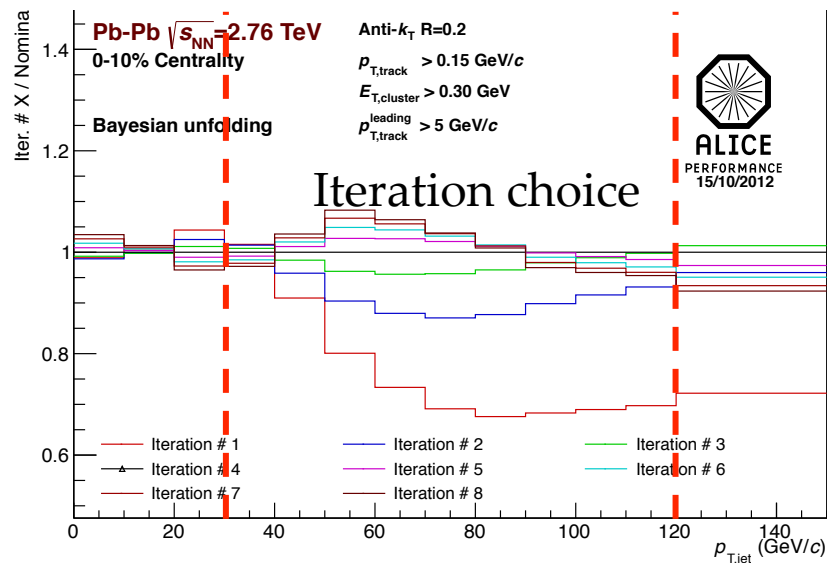
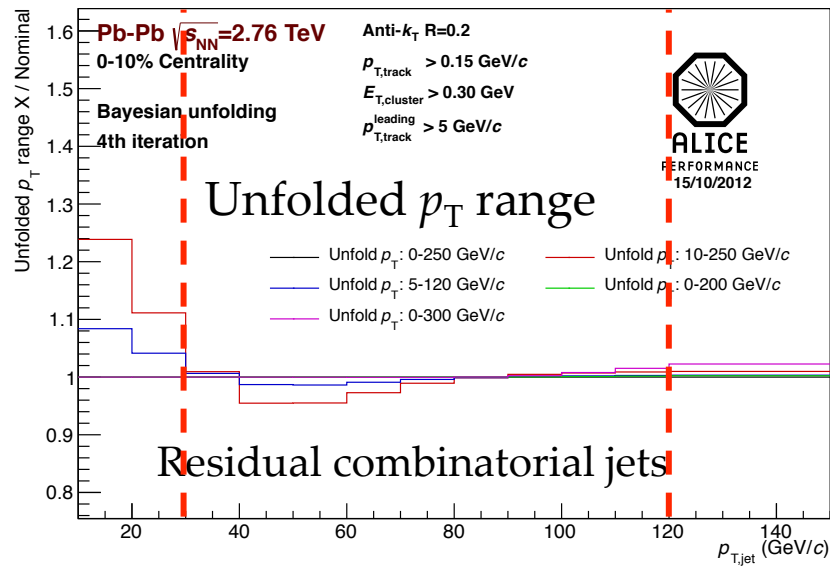
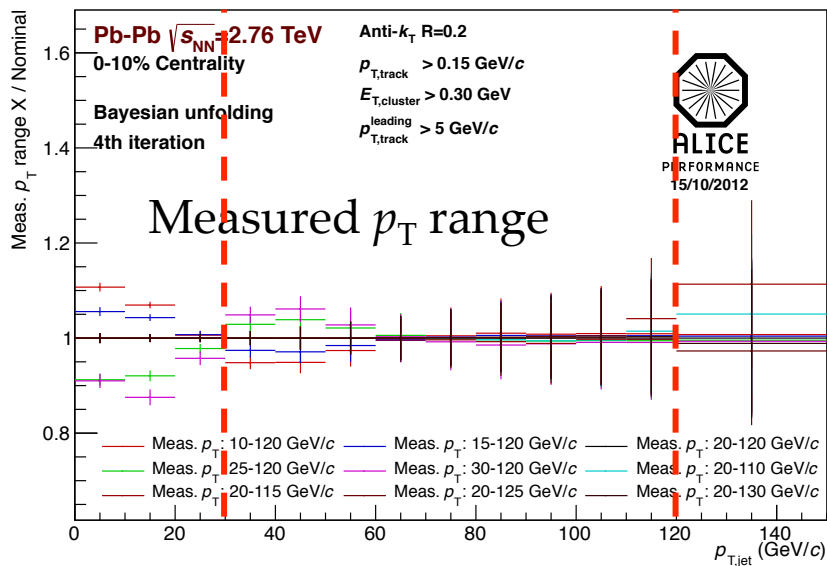
$$\chi^2 = \sum_{\text{refolded}} \left(\frac{y_{\text{refolded}} - y_{\text{measured}}}{\sigma_{\text{measured}}} \right)^2 + \beta \sum_{\text{unfolded}} \left(\frac{d^2 \log y_{\text{unfolded}}}{d \log p_T^2} \right)$$

- **Bayesian**

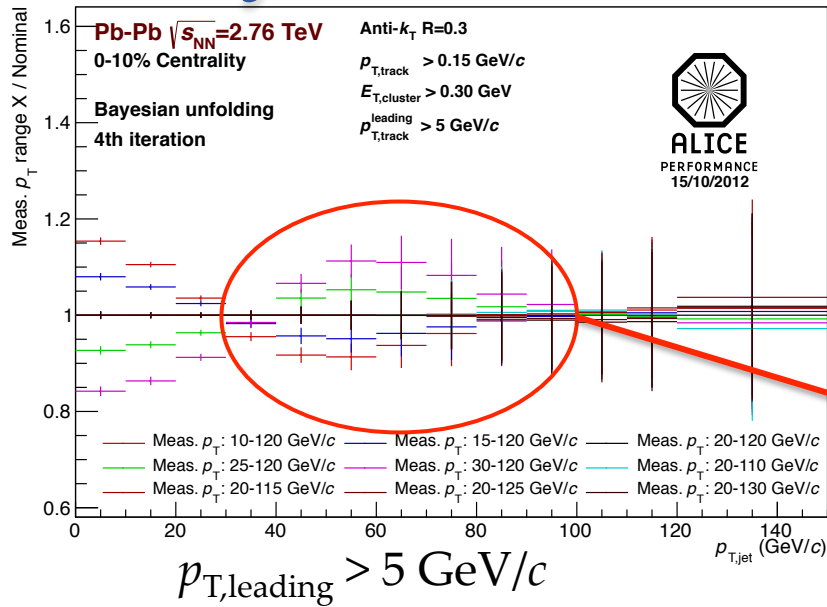
- Uses the Bayes’ theorem to iteratively update the estimators’ probabilities, starting from a “prior” guess (**finite # of iterations**)
 - Infinite iterations would give the best mathematical solution, but highly fluctuating, not useful for physics
- **Regularization parameter**: *a priori* knowledge about spectrum shape
 - Avoids wildly fluctuating results
 - Introduce a bias in the result

G. D’Agostini, Nuclear Instr. Meth., 362 (1995) 487-498
G. Cowan, A Survey of Unfolding Methods in Part. Phys., Proc. Adv. Stat. Tech. in Part. Phys., Durham (2002) ([link](#))

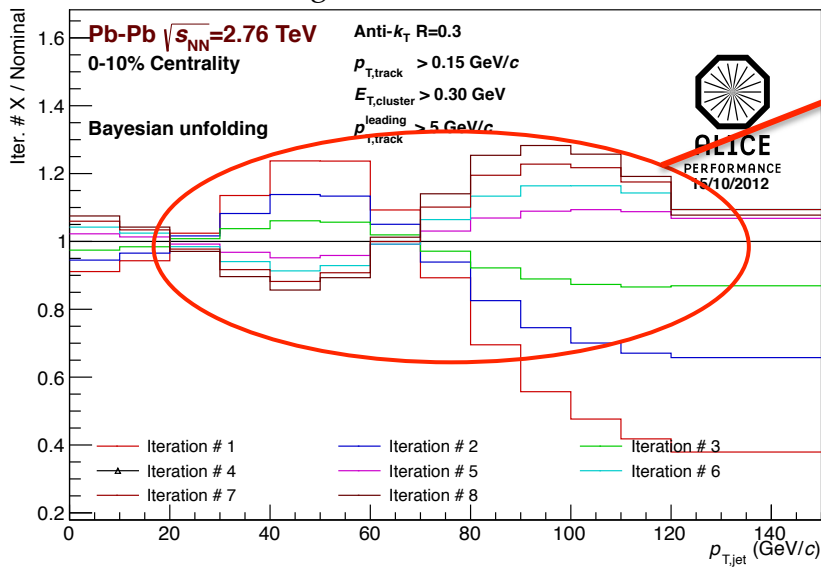
Bayesian unfolding results – R=0.2



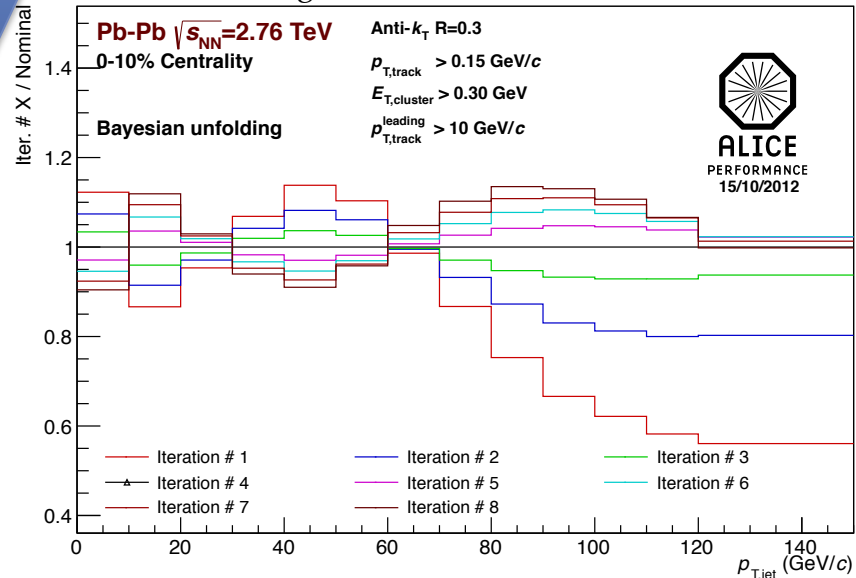
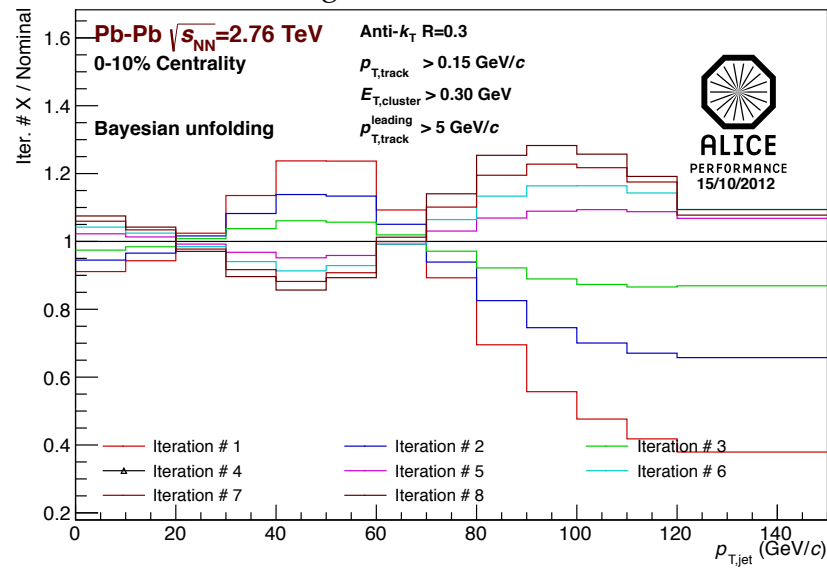
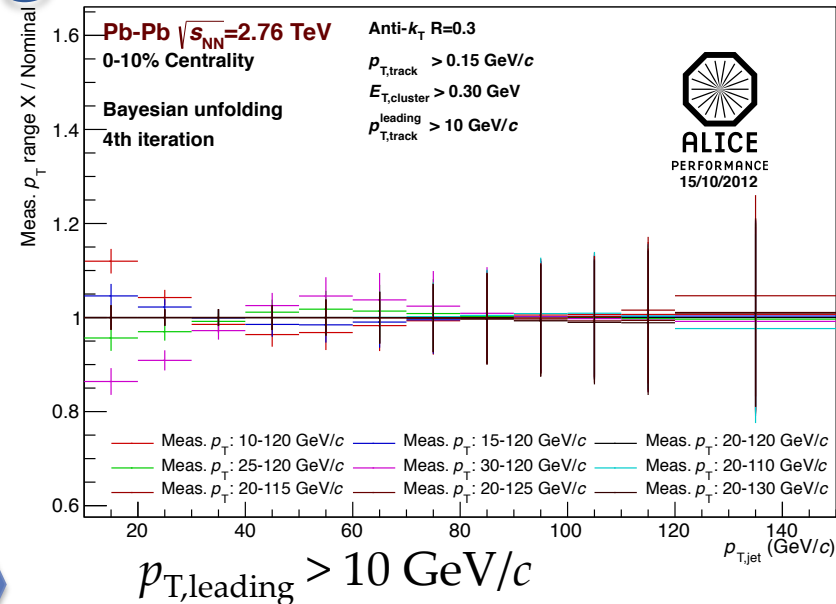
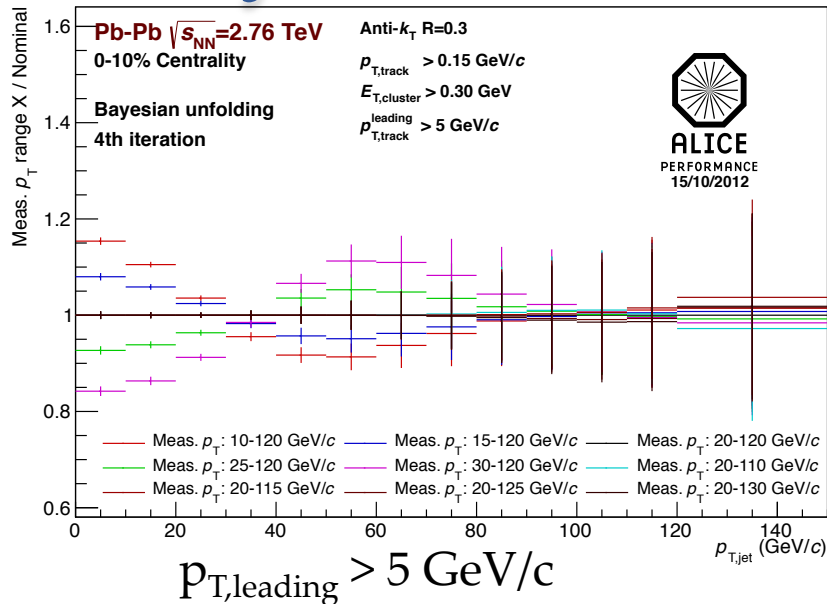
Bayesian unfolding results – $R=0.3$



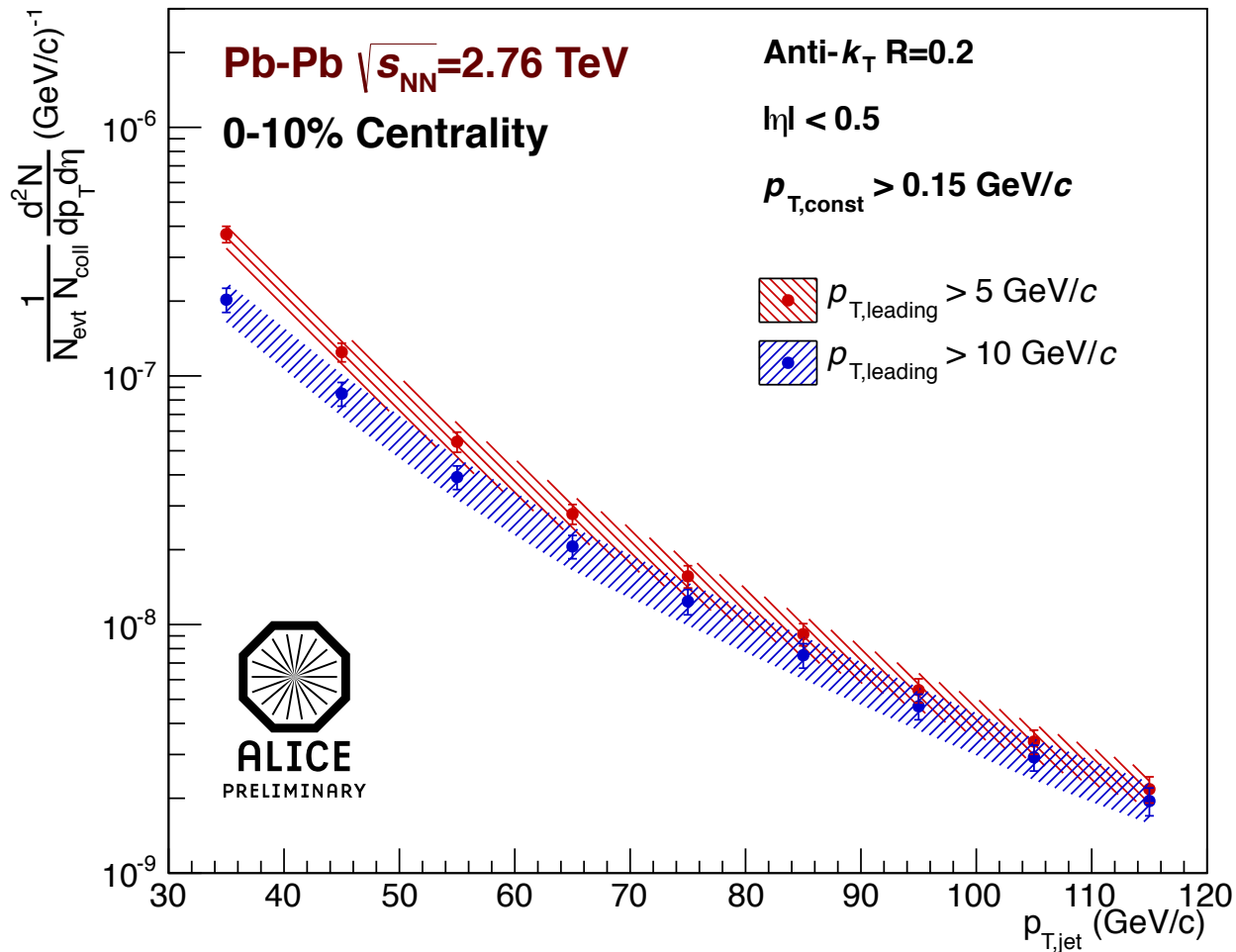
Unfolding instabilities...



Bayesian unfolding results – R=0.3

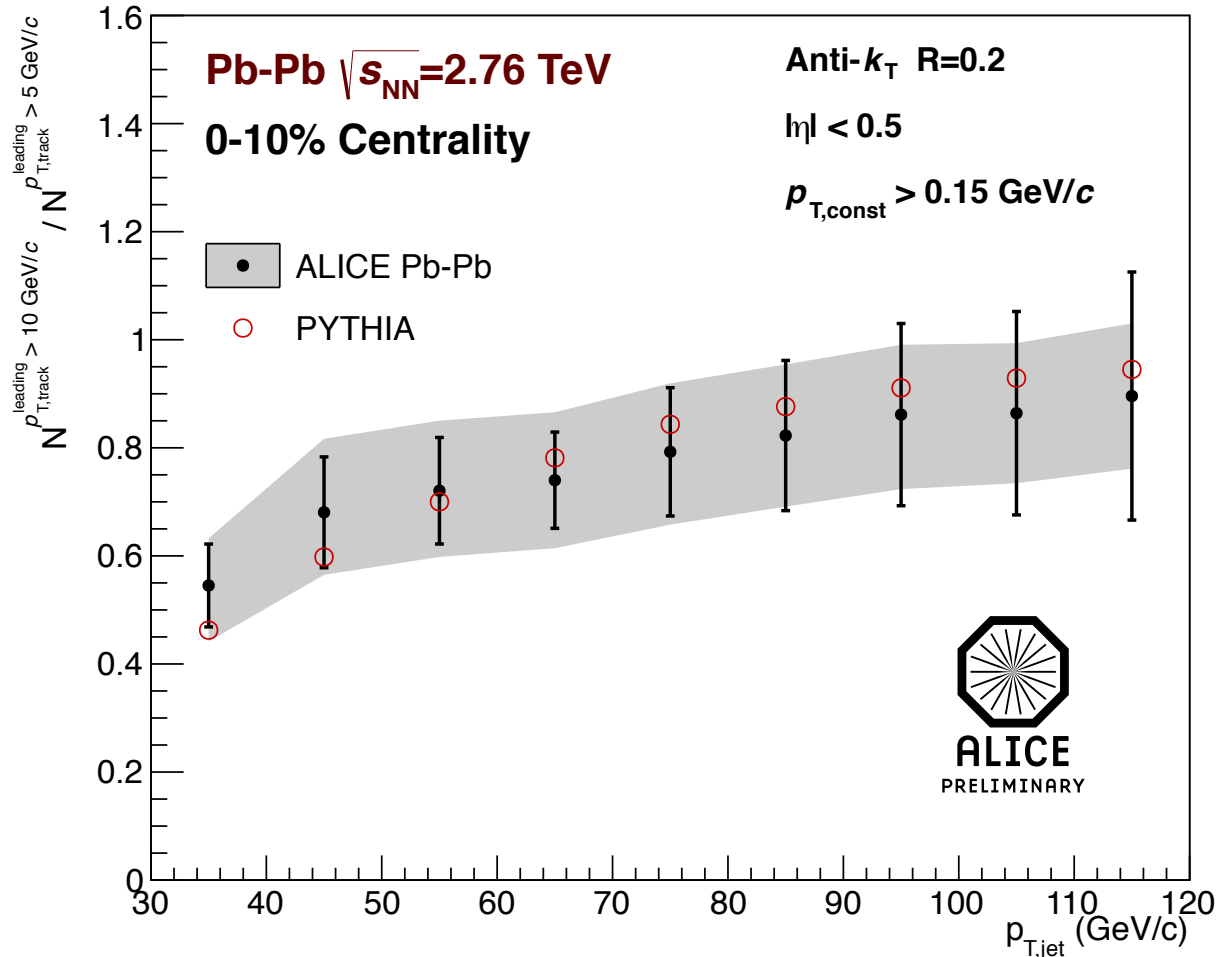


Unfolded biased spectra



- Fully corrected p_T jet spectra
 - $R=0.2$
 - $p_{T,const} > 0.15$ GeV/c
- Two different leading hadron triggers
 - trigger only on charged tracks to avoid difficulties in the theoretical predictions
 - $p_{T,track} > 5$ GeV/c
 - $p_{T,track} > 10$ GeV/c

Effect of the leading hadron requirement



- Statistics and systematics
 - Take into account correlated uncertainties
- Agreement with PYTHIA is consistent with vacuum-like fragmentation of the jet core

Conclusions & Outlook

- **Background** in Pb-Pb central collisions has been studied in order to be able to extract a “clean” signal
- **Bayesian unfolding** used to correct spectra for **background fluctuations** and **detector-induced effects**
 - Studies are ongoing to unfold the **R=0.3 spectra**
- Effect of the **leading hadron requirement** studied for two different thresholds, 5 GeV/c and 10 GeV/c
 - Agreement with PYTHIA is **consistent with vacuum-like fragmentation** of the jet core

Backup

Systematic uncertainties

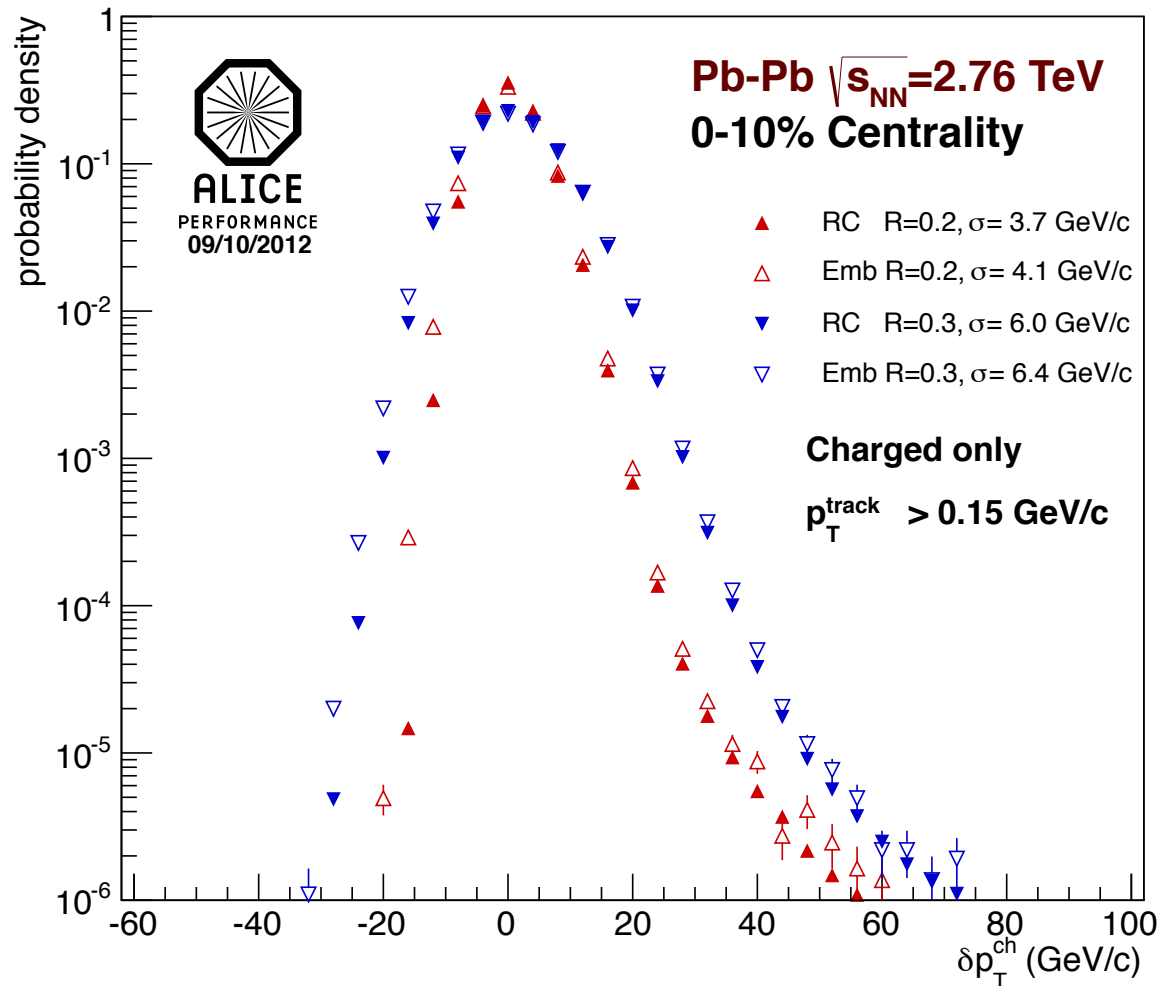
Uncertainty	% on jet spectrum
EMCal (resolution, energy scale, clusterizer, non-linearity)	5.3
Tracking Efficiency	10
Hadronic Correction	10
Scale Factor	2
δp_T RC vs. Embedding	5
Unfolding	14
Flow bias	< 1%
Total	19

Unfolding uncertainty	% on jet spectrum
Iteration	5
Measured p_T range	9
Unfolded p_T range	5
Prior choice	10
Total unfolding	14

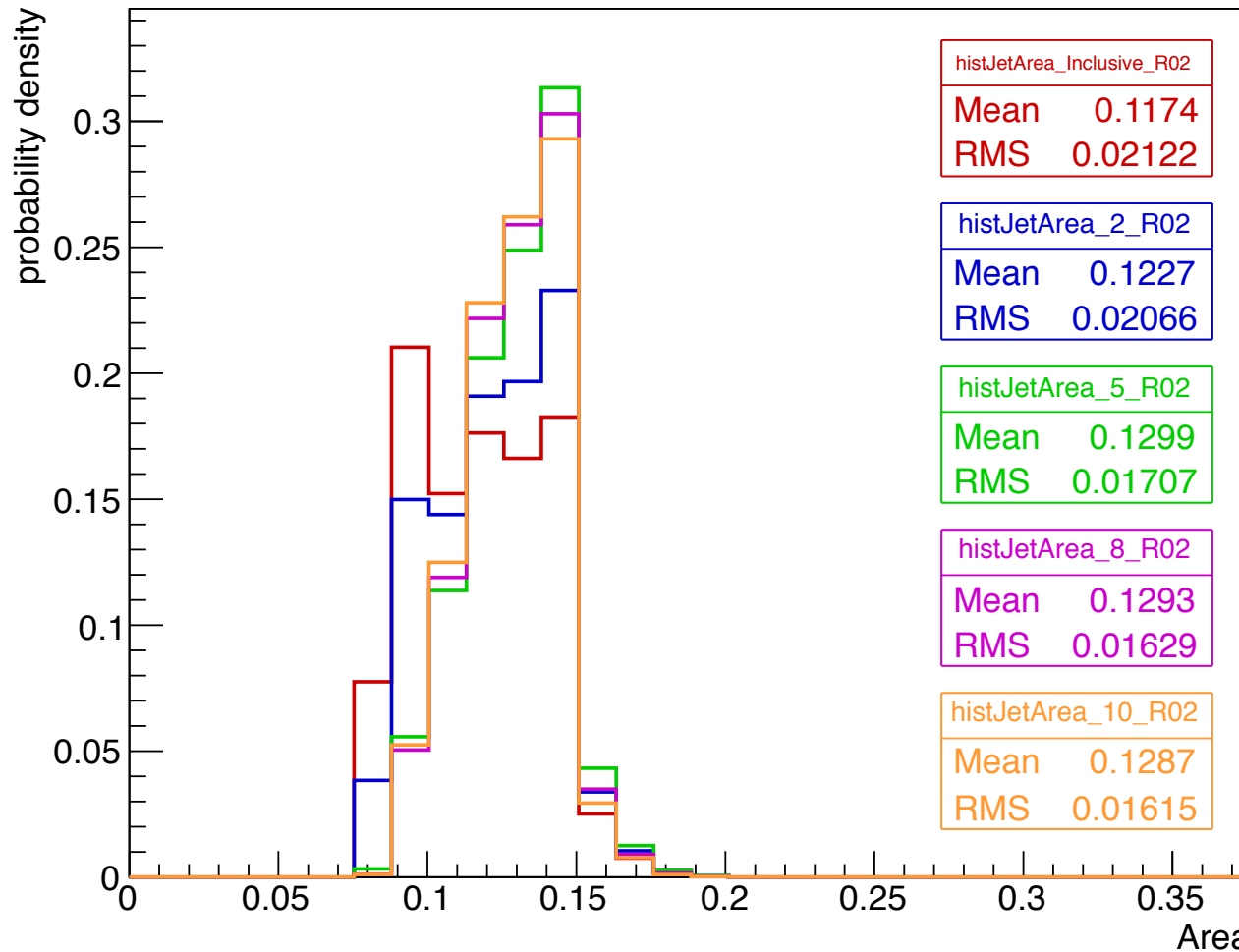
Some of the uncertainties are asymmetric and p_T dependent.
Here only the maximum value is shown.

Systematic uncertainties for 10 GeV/c bias are very similar

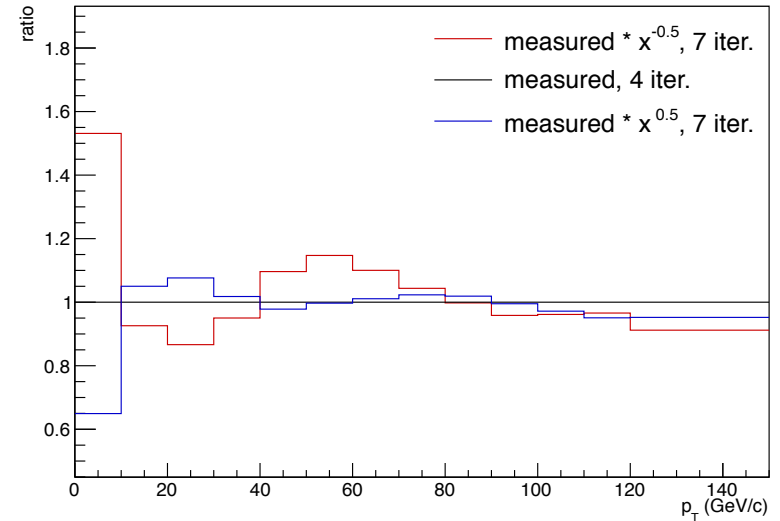
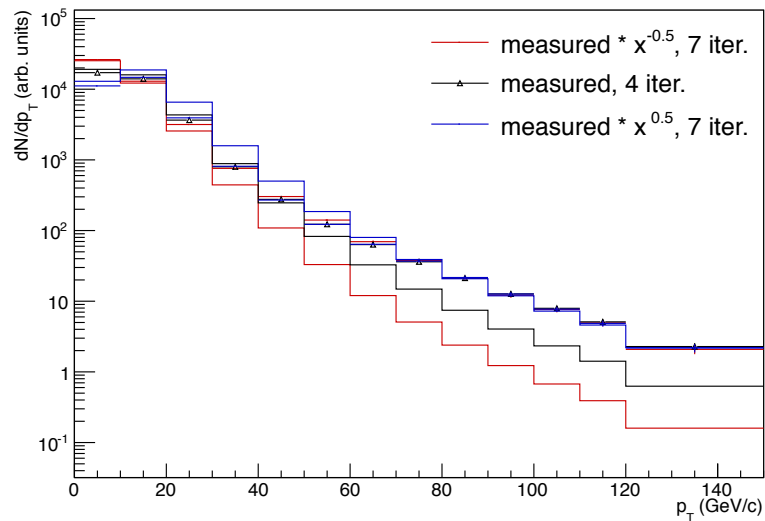
Charged δp_T



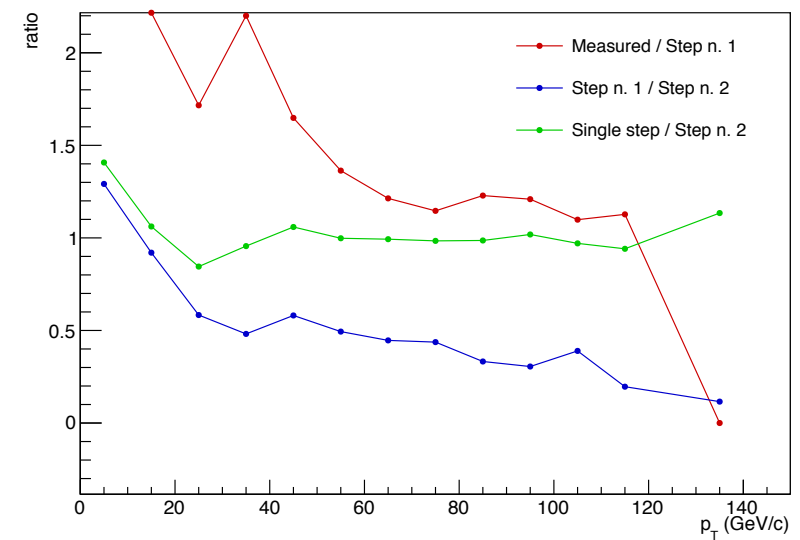
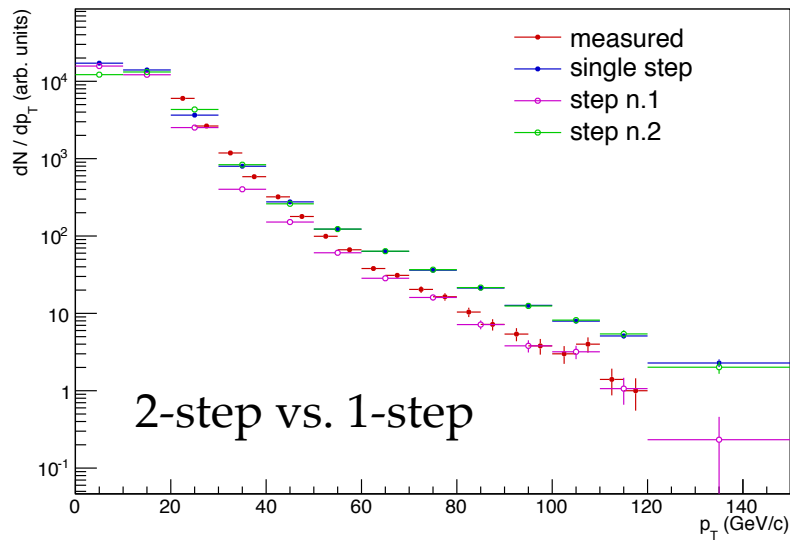
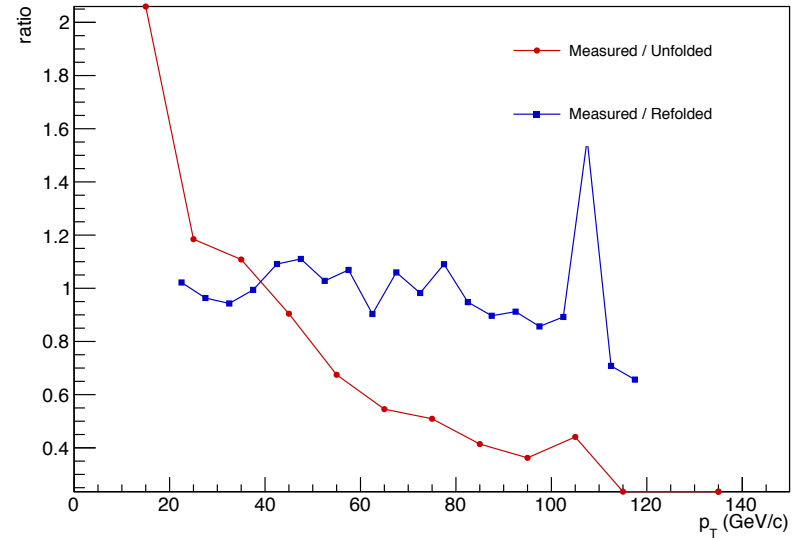
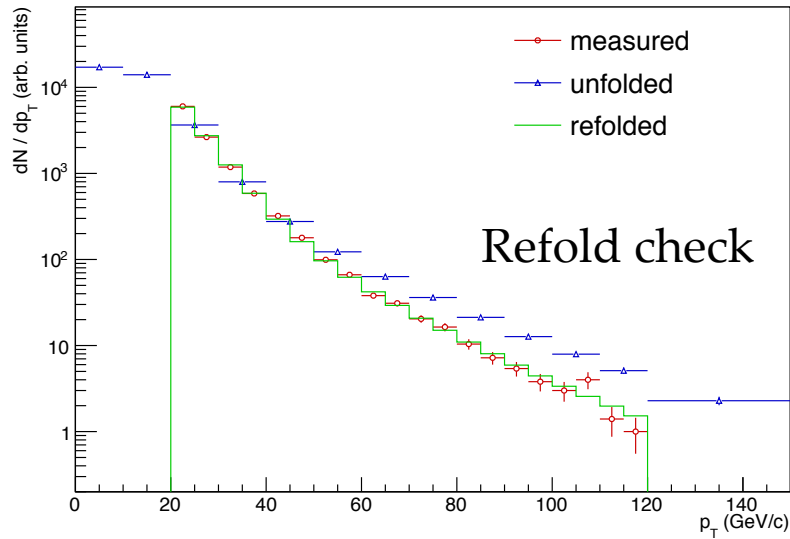
Jet area distribution



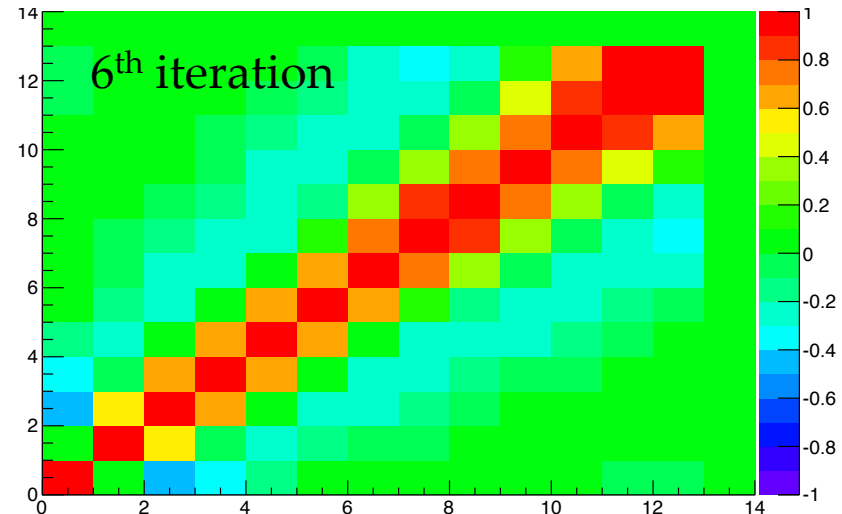
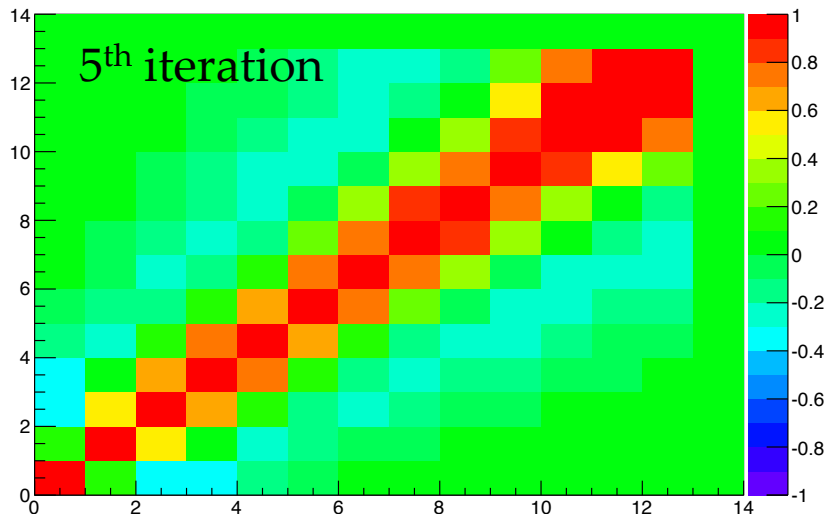
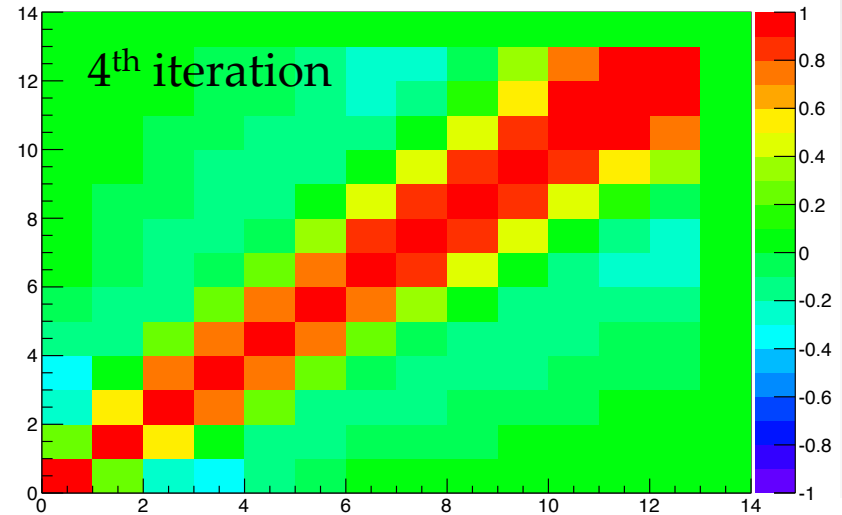
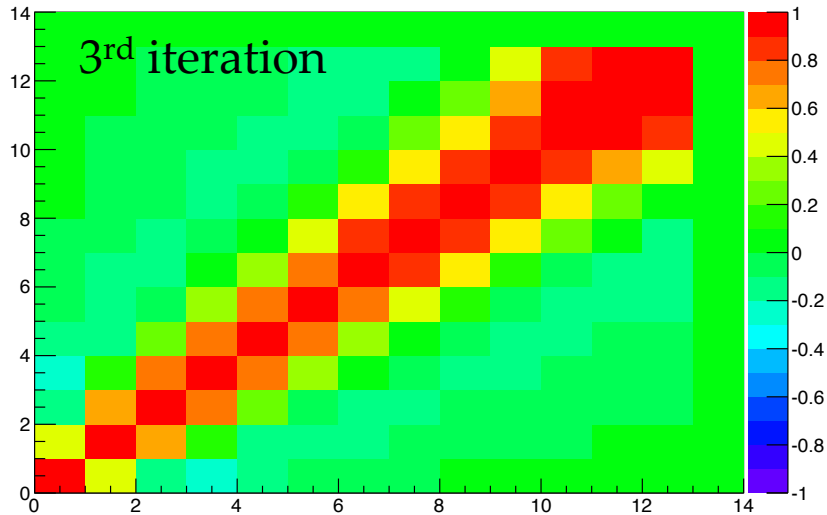
Unfolding prior choice



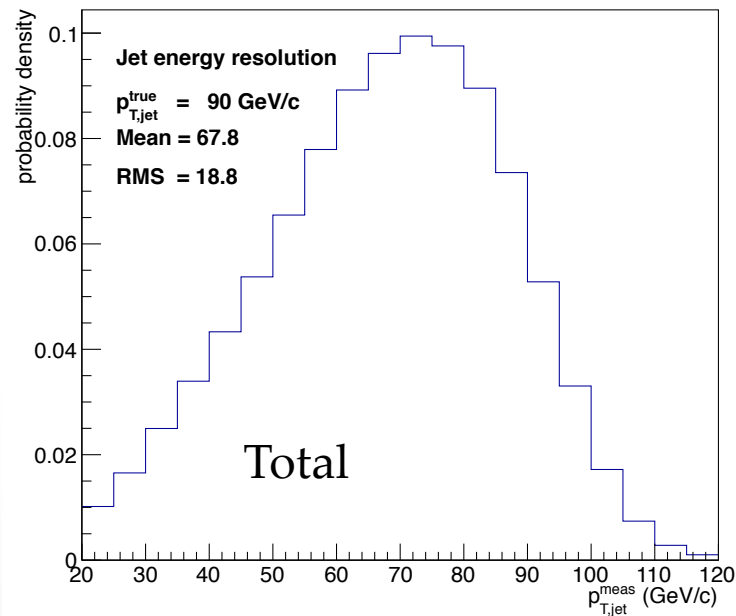
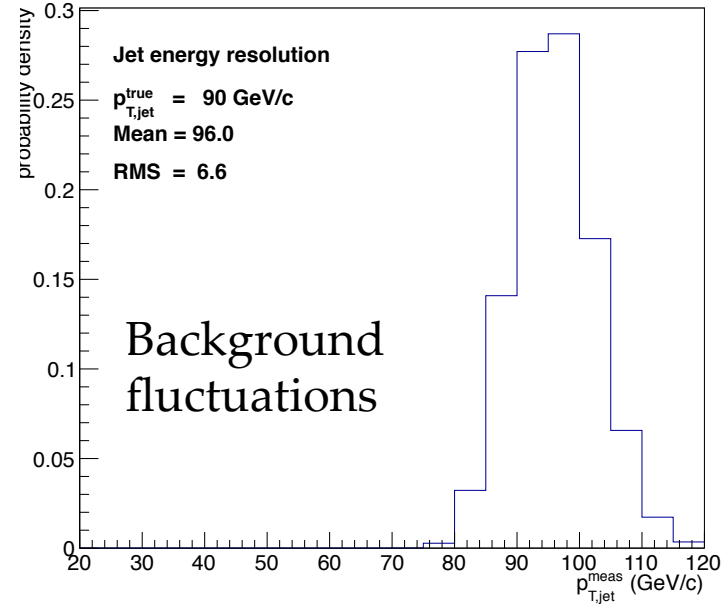
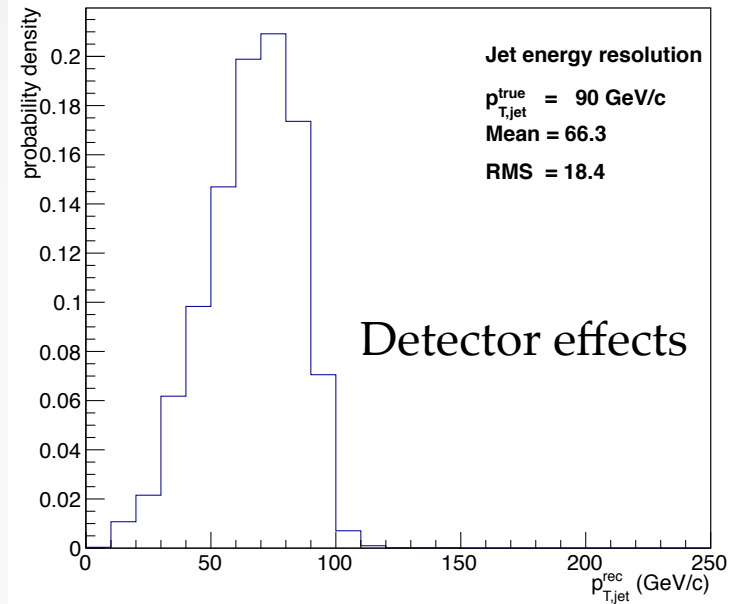
Refold, 2-step



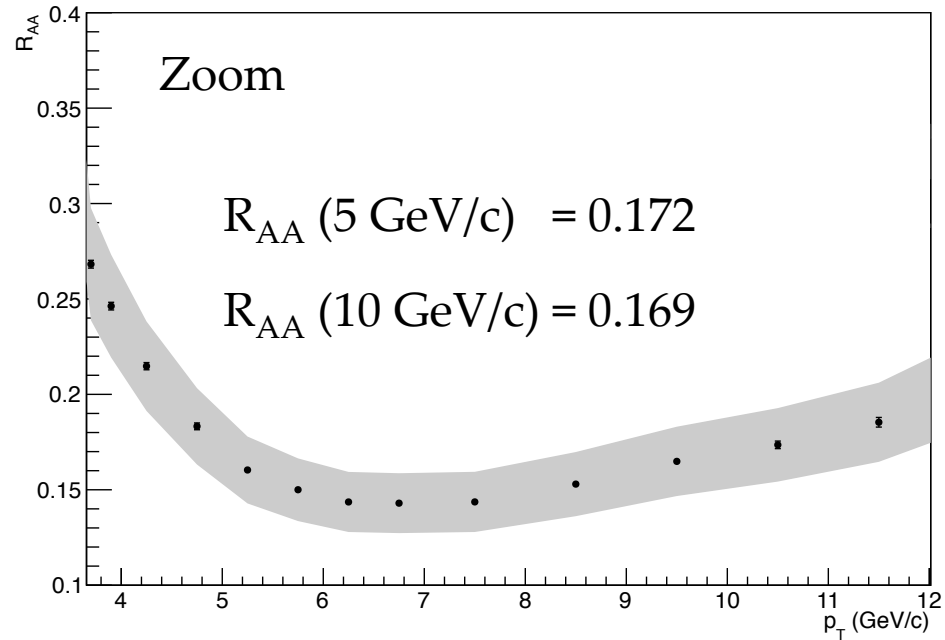
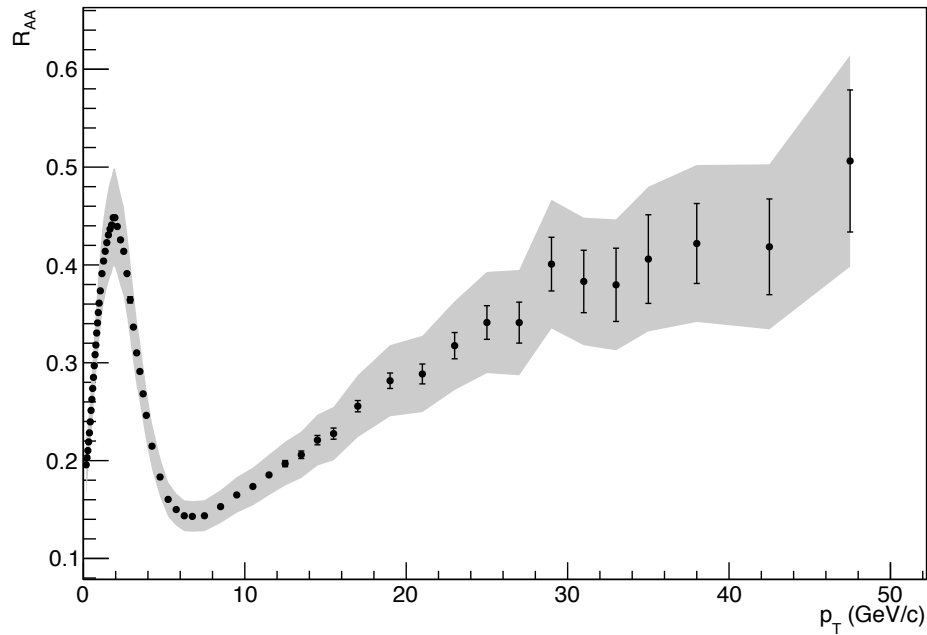
Pearson coefficients



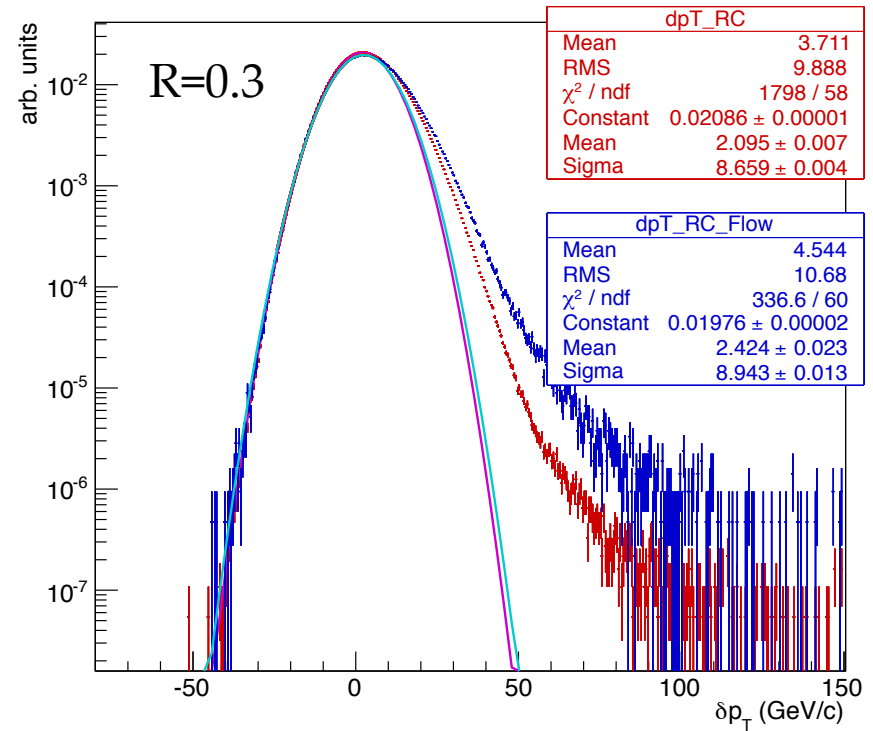
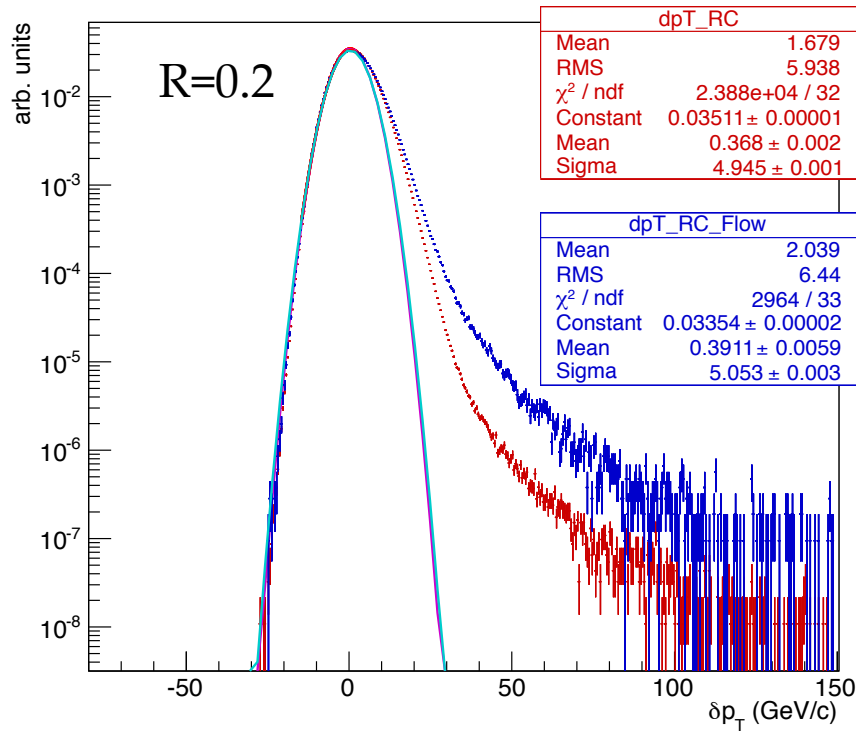
Jet energy resolution



Charged particles R_{AA}



Flow bias



Blue = random cones δp_T distributions obtained requiring a 5 GeV/c charged track in the calorimeter

Red = regular random cones δp_T distributions

The shift measured with the mean of the Gaussian fit on the LHS is **< 100 MeV/c for R=0.2** and **300 MeV/c for R=0.3**