



**European Research Council**

Established by the European Commission



**SAPIENZA**  
UNIVERSITÀ DI ROMA

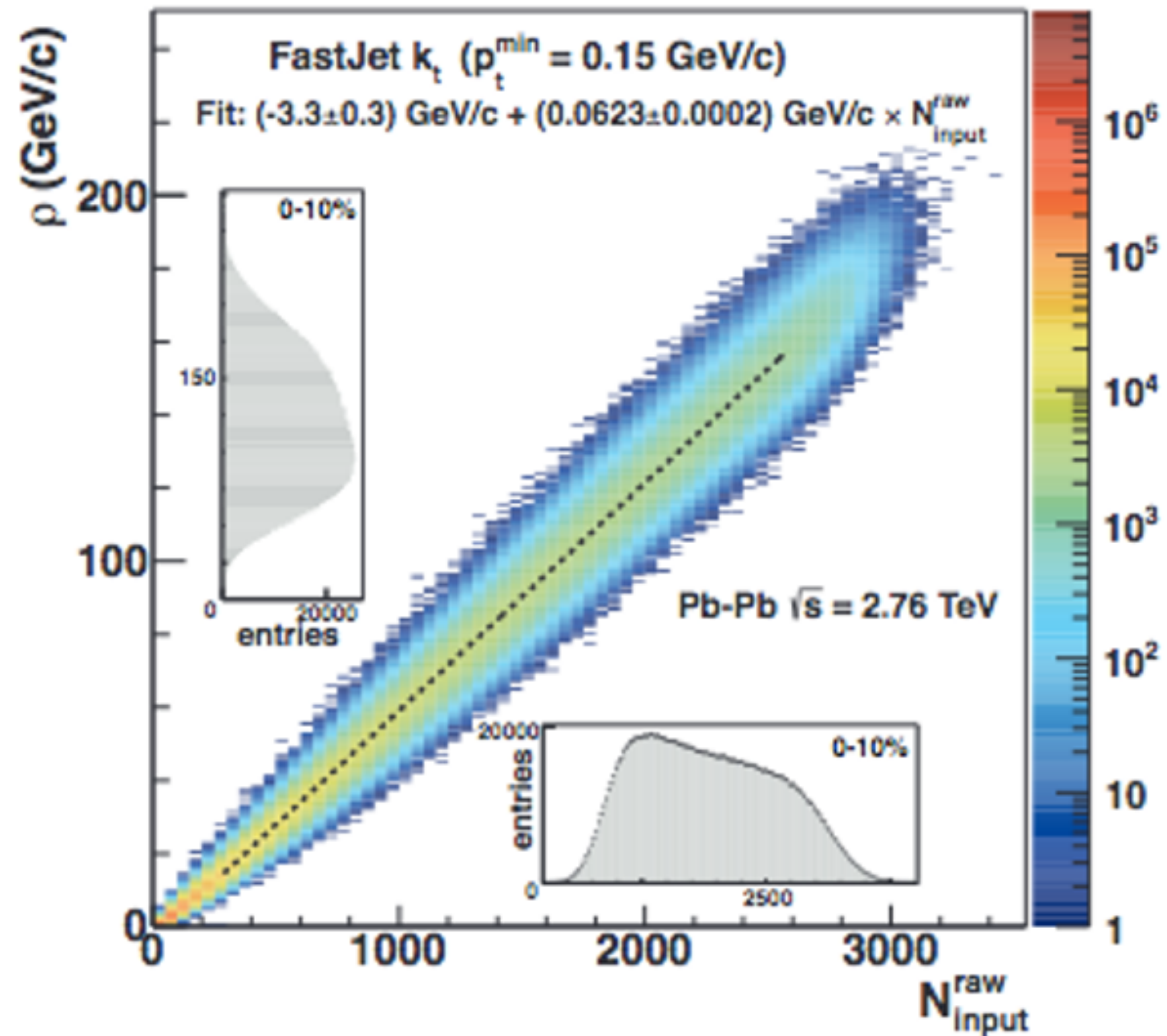
# **Lesson 4**

# **from the raw to a fully corrected jet**

# **measurement**

Leticia Cunqueiro Mendez

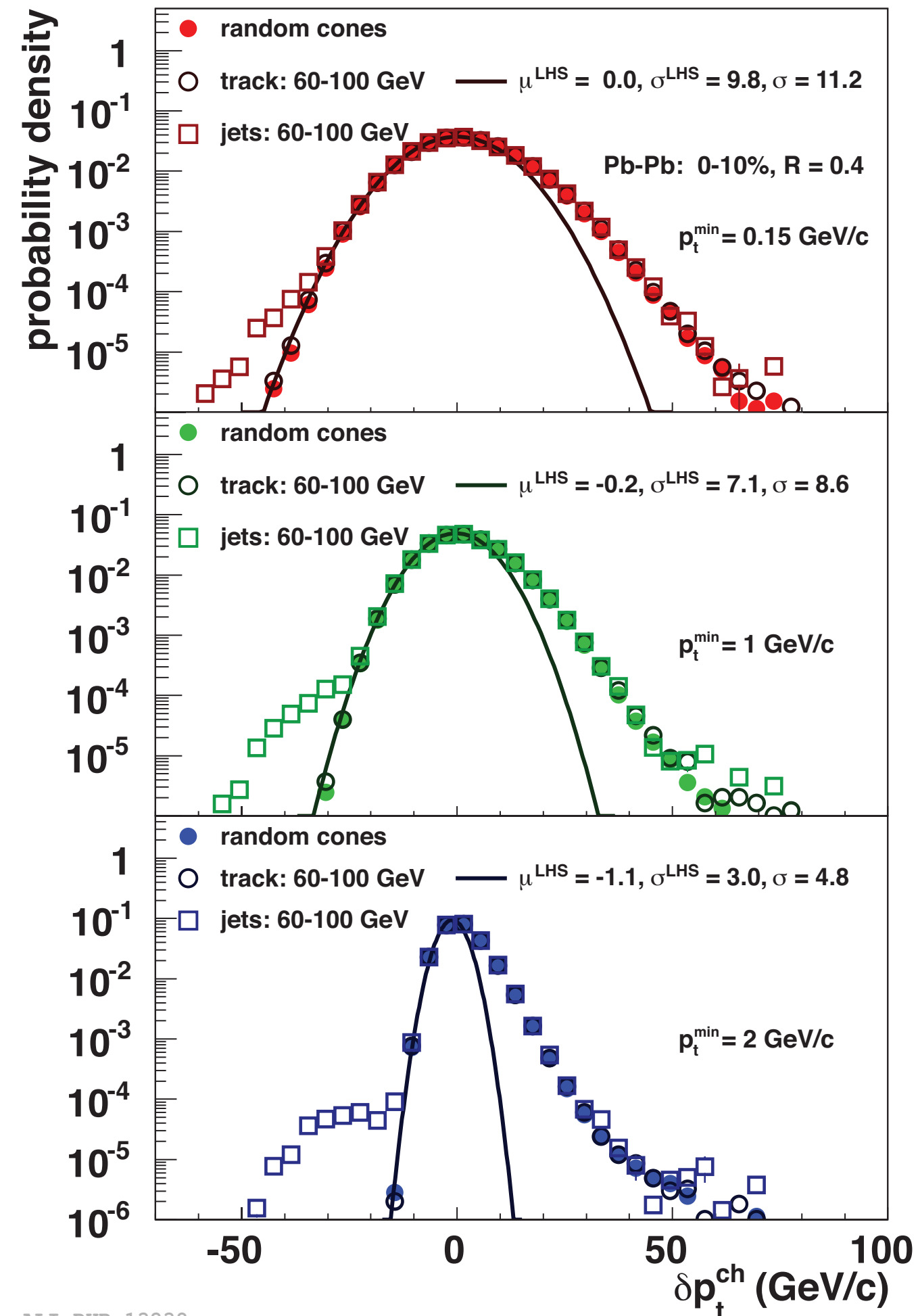
## Measurement of the underlying event



Underlying event  $\rho \approx 150$  GeV in central collisions

Remember UE contamination  $\propto$  jet area,  $150 * \pi * R^2 \sim \mathbf{75}$  GeV in a jet of  $R=0.4$

# Background response: region-to-region fluctuations



ALI-PUB-13238

[ALICE, JHEP03(2012)053]

$$\delta p_T = p_{T,\text{jet}}^{\text{reco}} - \rho \cdot A_{\text{jet}} - p_T^{\text{part,embed}}$$

We embed different probes into Pb-Pb events and estimate the background response through  $\delta p_T$

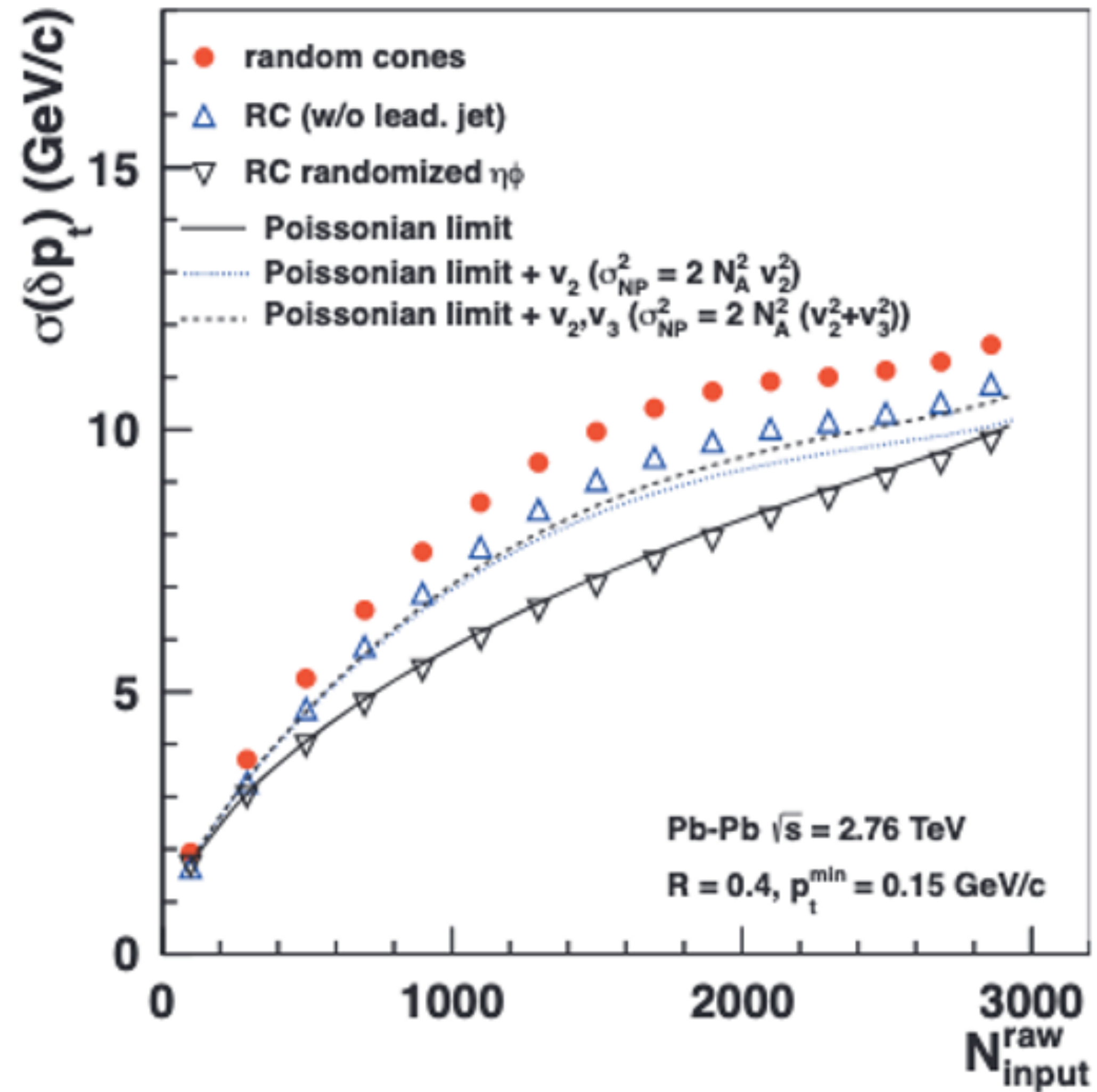
-Small dependence on the probe fragmentation pattern

-Small **back reaction** effects in the tails of the response due to **jet splitting** and **jet merging**

-Minimum constituent  $p_T$  cut-off reduces fluctuations

Note that different experiments have different effective  $p_T$  cutoffs

# Components of the background fluctuations



Randomized event or poissonian limit:  
the number of particles is kept but their  
 $\eta, \phi$  coordinates are randomly redistributed

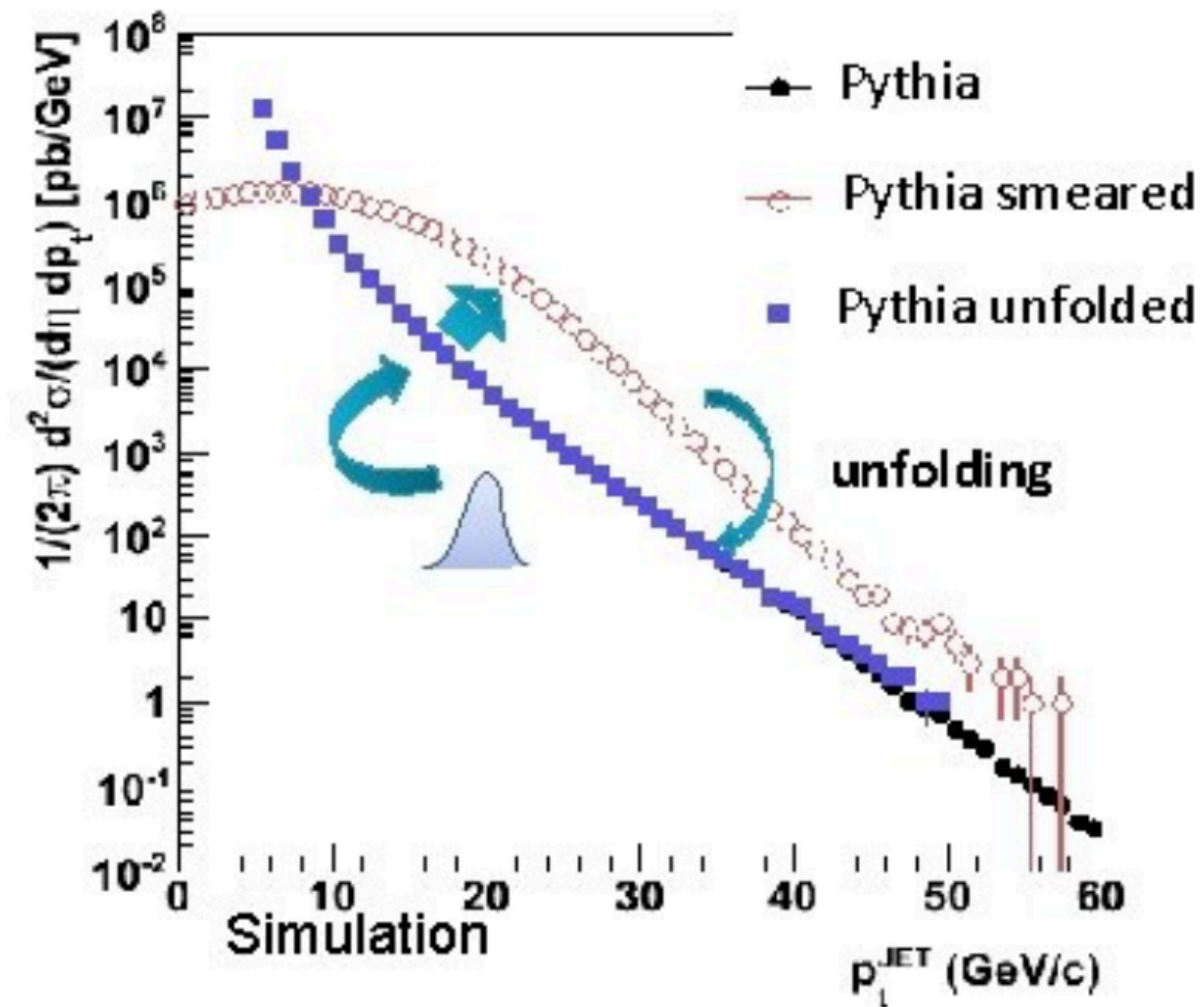
When azimuthal anisotropies are simulated in the event,  
naturally fluctuations grow

## What does a $\delta p_T$ of 11 GeV mean?

$$\frac{dN^{\text{Meas}}}{dp_T} = \frac{dN^{\text{true}}}{dp_T} \otimes f^{\text{Resol}}(\delta\rho)$$

$\delta p_T$  distribution:  
 'smearing' of jet spectrum  
 due to background fluctuations

Large effect on yields  
 Need to unfold



Note: the impact of the smearing depends on the underlying distribution, if you have a flat distribution for instance, smearing does not modify the truth

With a power-law, smearing goes in one direction preferentially 5

# Fake jets

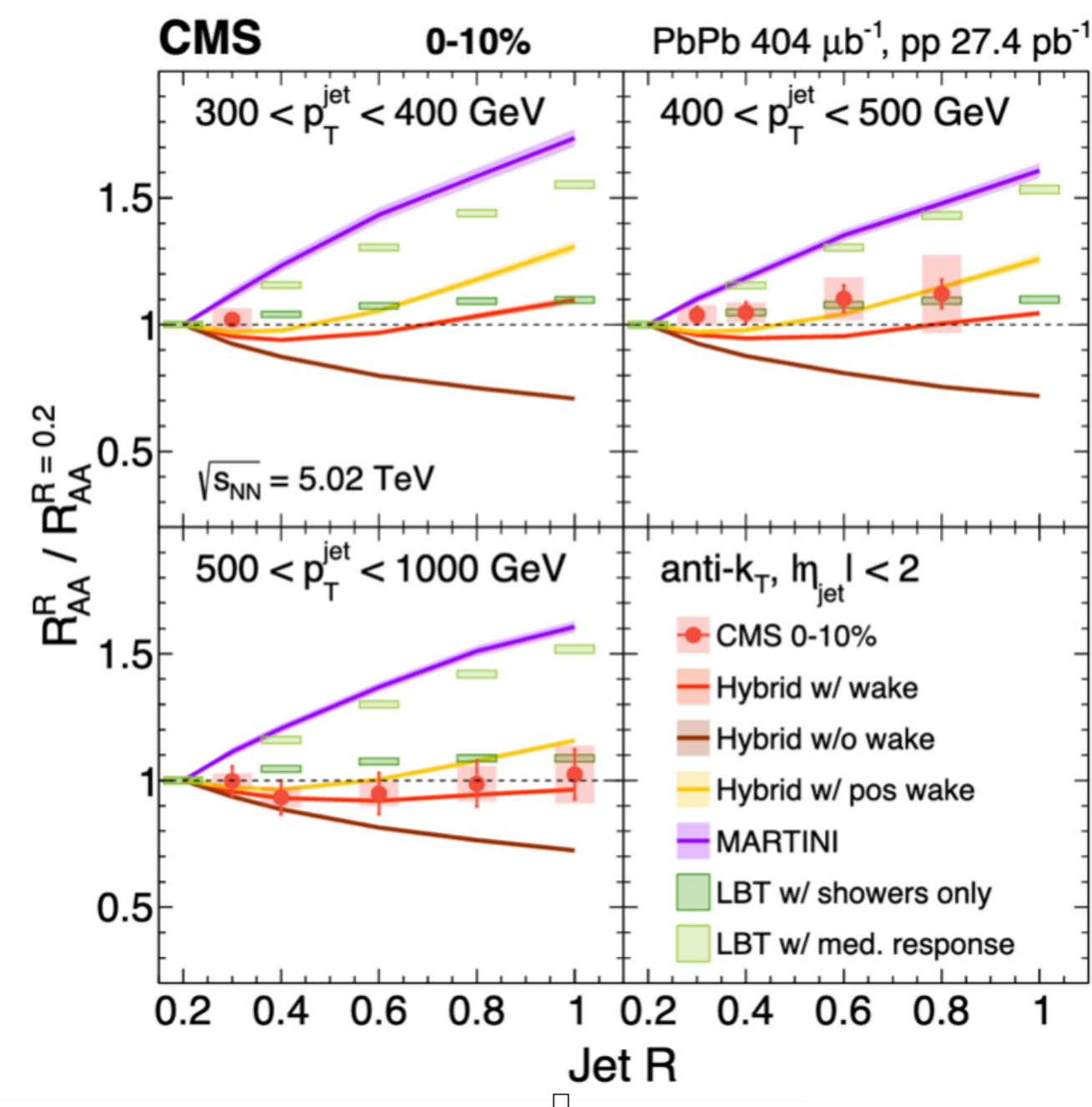
It also means that it is likely to reconstruct jets that are not correlated to any signal

Those purely bkg jets are called **fake jets**

**Fake jets** need to be suppressed before unfolding, cause unfolding conserves counts.

This is done typically:

a) by considering very high  $p_T$  jets (recall CMS results for large-R jets in lesson 2.)

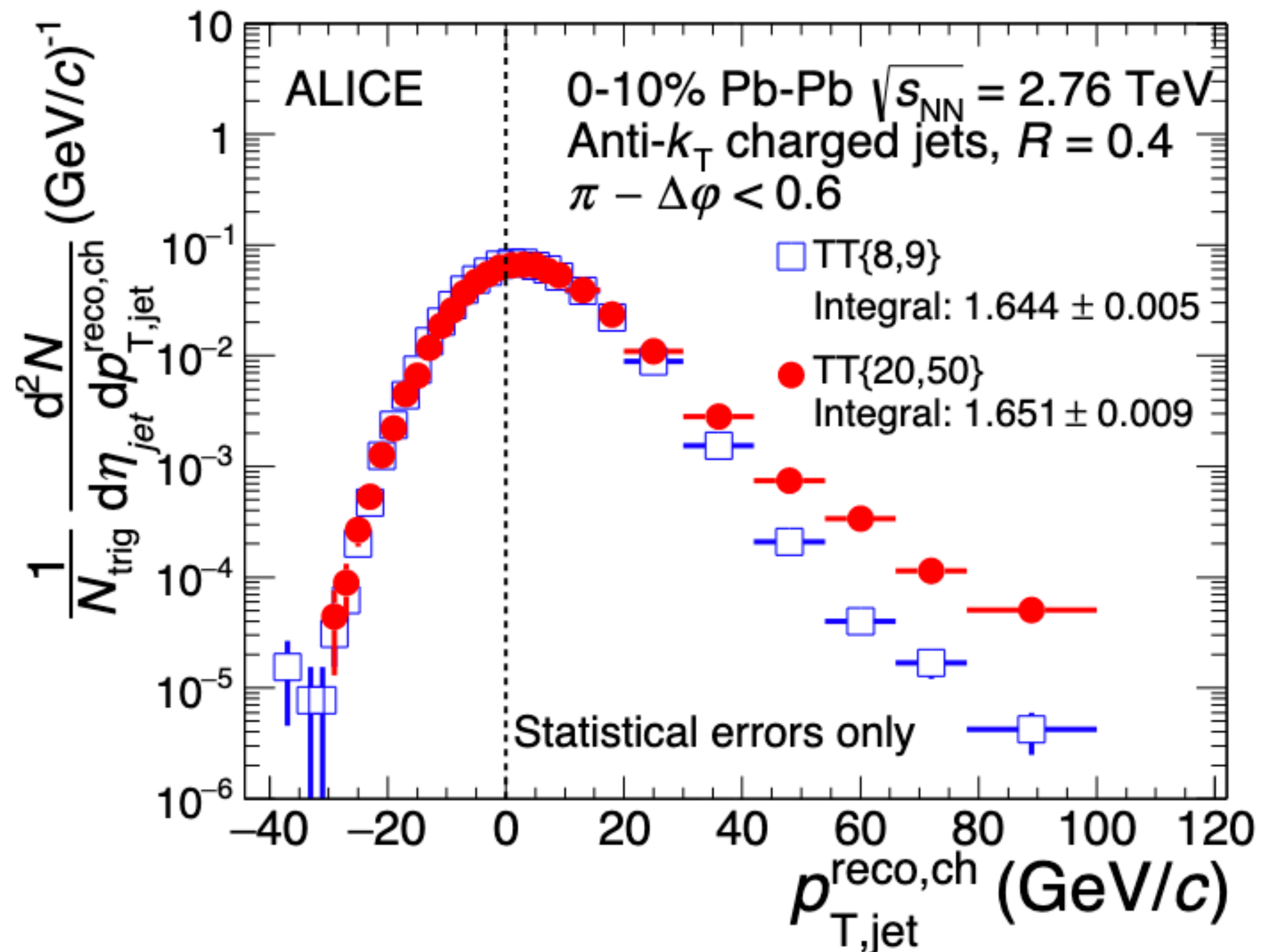


R=1 jets are only reported in the very high  $p_T$  bin

# Faket jets

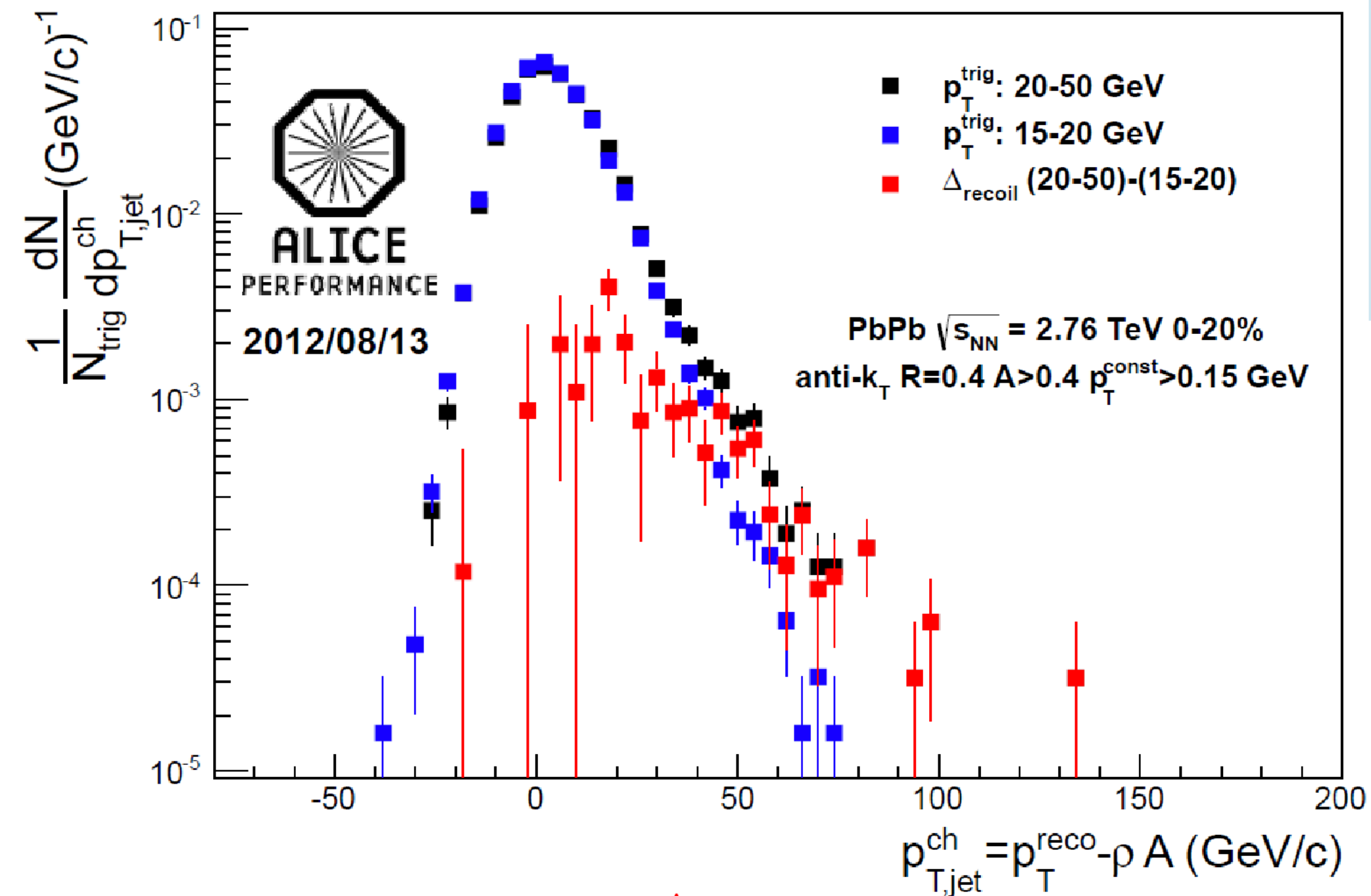
$$\Delta_{\text{recoil}} = \frac{1}{N_{\text{trig}}} \frac{d^2 N_{\text{jet}}}{dp_{T,\text{jet}}^{\text{ch}} d\eta_{\text{jet}}} \Big|_{TT_{\text{sig}}} - c_{\text{ref}} \cdot \frac{1}{N_{\text{trig}}} \frac{d^2 N_{\text{jet}}}{dp_{T,\text{jet}}^{\text{ch}} d\eta_{\text{jet}}} \Big|_{TT_{\text{ref}}}$$

b) by applying data-driven coincidence measurements



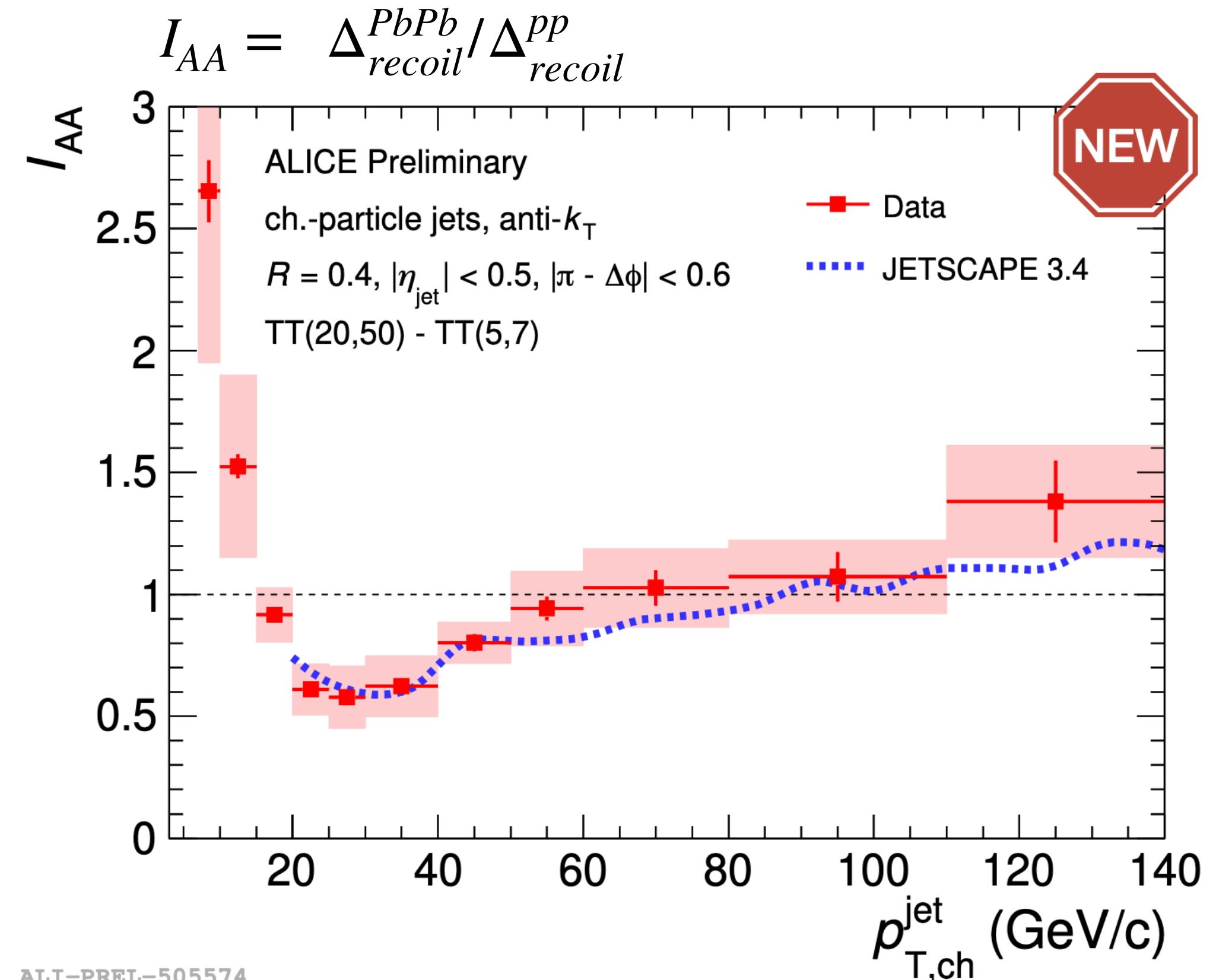
# Faket jets

b) by applying data-driven coincidence measurements



$$\Delta_{recoil} = \frac{1}{N_{trig}} \frac{d^2 N_{jet}}{dp_{T,jet}^{ch} d\eta_{jet}} \Big|_{TT_{sig}} - c_{ref} \cdot \frac{1}{N_{trig}} \frac{d^2 N_{jet}}{dp_{T,jet}^{ch} d\eta_{jet}} \Big|_{TT_{ref}}$$

$\Delta_{recoil}$  is a fake-free distribution that can be unfolded down to zero transverse momentum

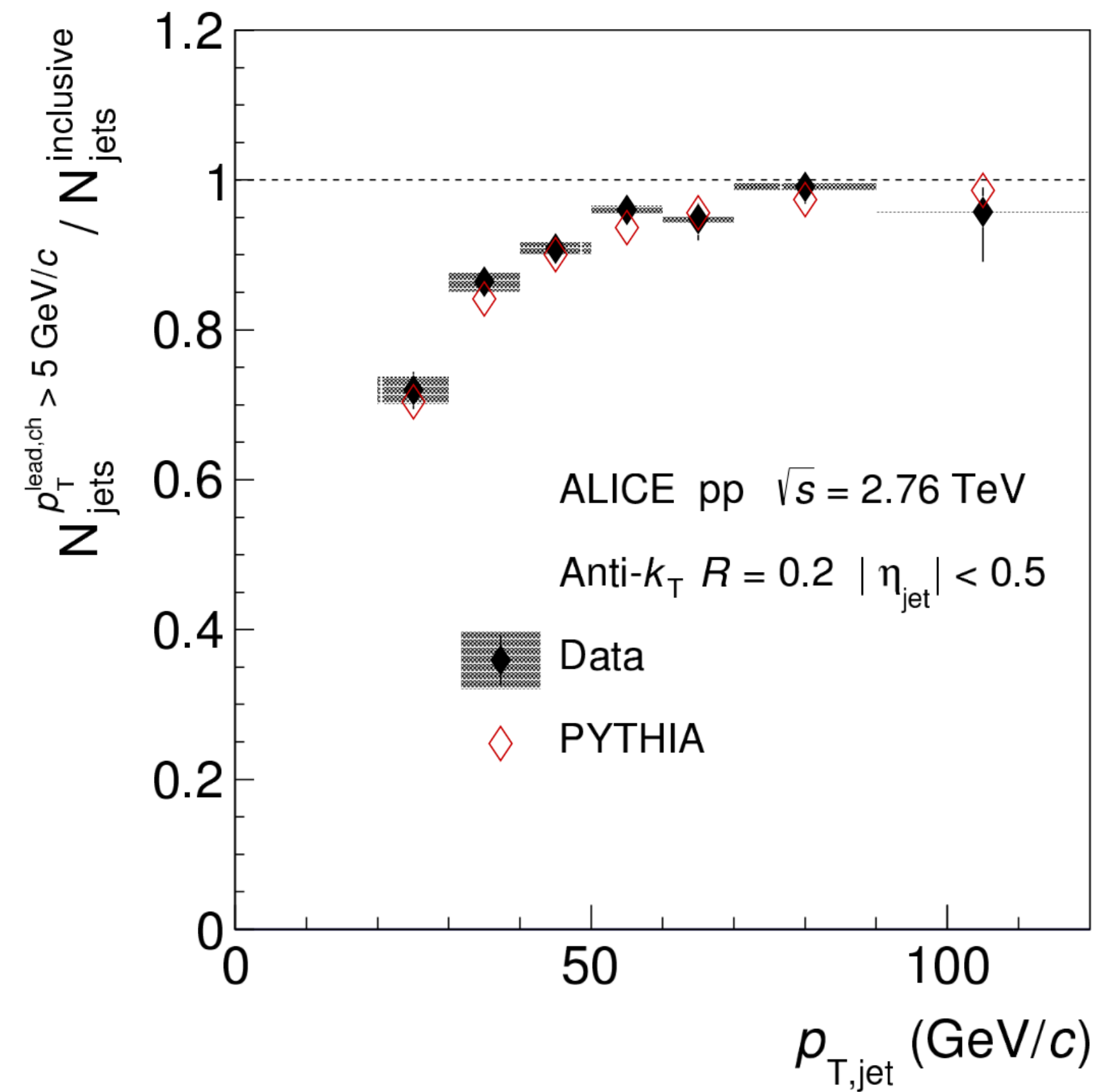


ALI-PREL-505574



# Fake jets

c) by requiring hard fragmentation (whatch out, it introduces a potential bias!)



Require that your jets have a constituent with  $p_T > 5$  GeV

This suppresses fakes but also potentially softer quenched jets!

# Fake jets

d) by reducing bkg fluctuations using for instance ML techniques

Area-based is an unbiased method, knows nothing about the jet structure

$$p_{Tsub} = p_{Traw} - \rho Area$$

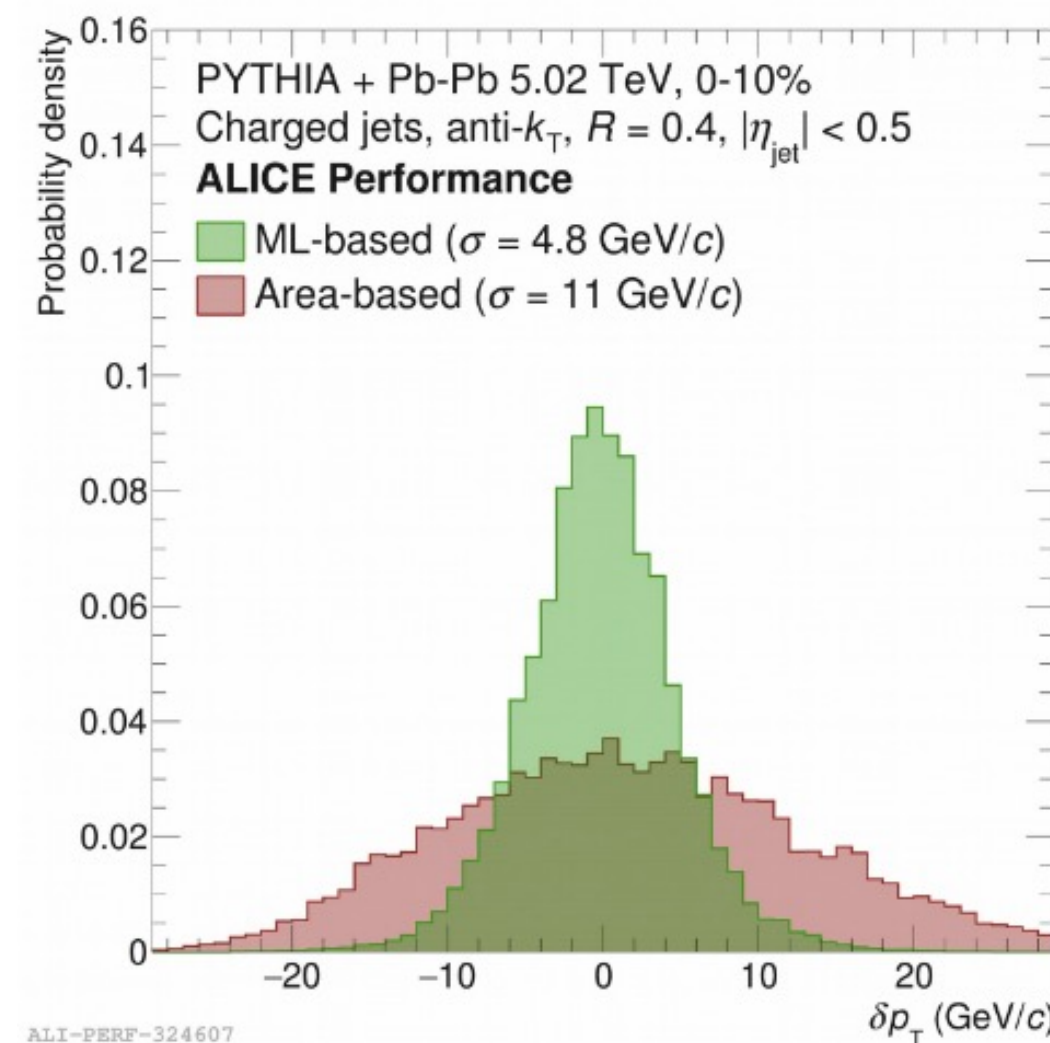
New ML approach: a regression model to subtract background. Input parameters: area-based subtracted jet  $p_T$ , angularity, number of constituents,  $p_T$  of the 8 leading jet constituents

Background fluctuations suppressed:

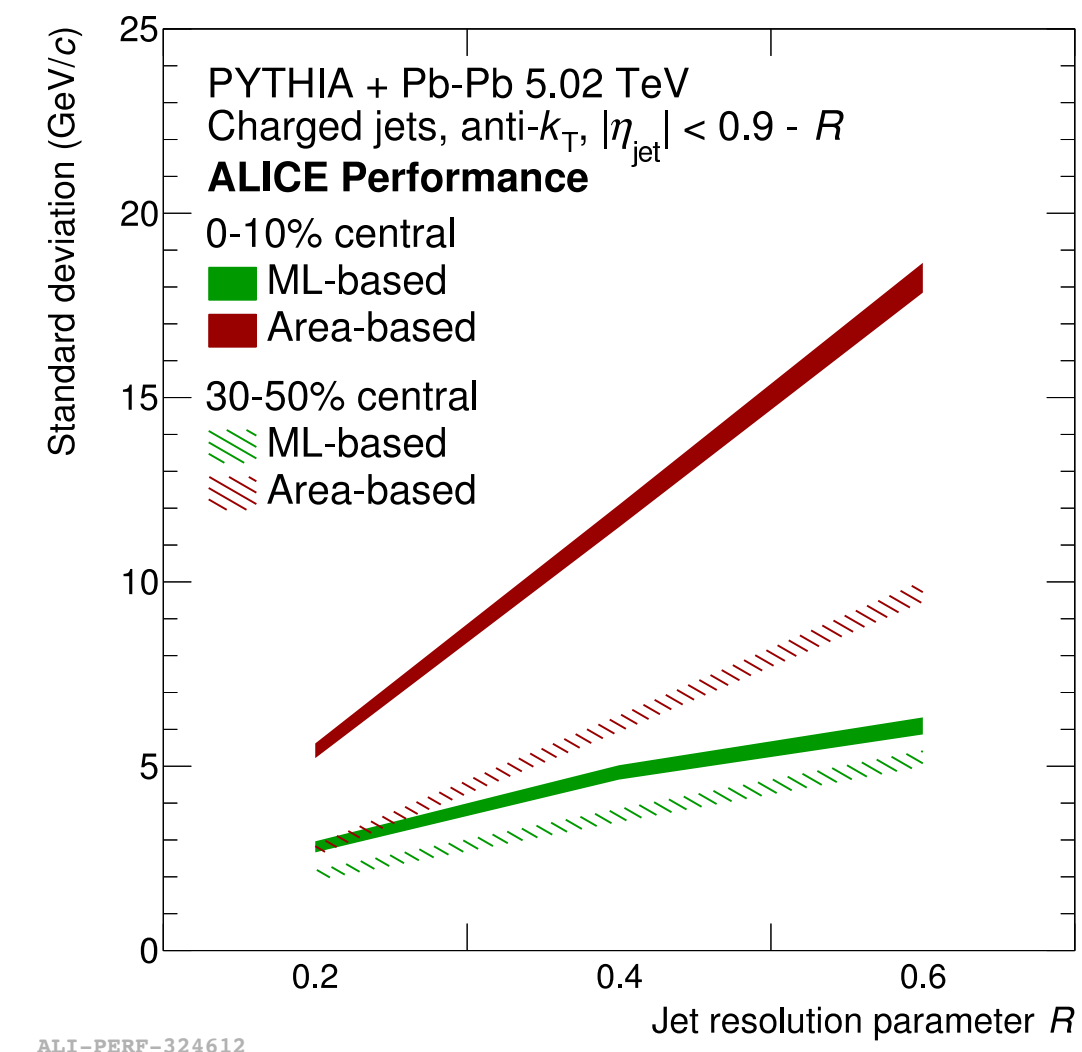
- Combinatorial jets reduced
- Smaller unfolding correction
- Extend to low jet  $p_T$  and large  $R$

Caveat: background subtraction might become model dependent

**Background fluctuations**



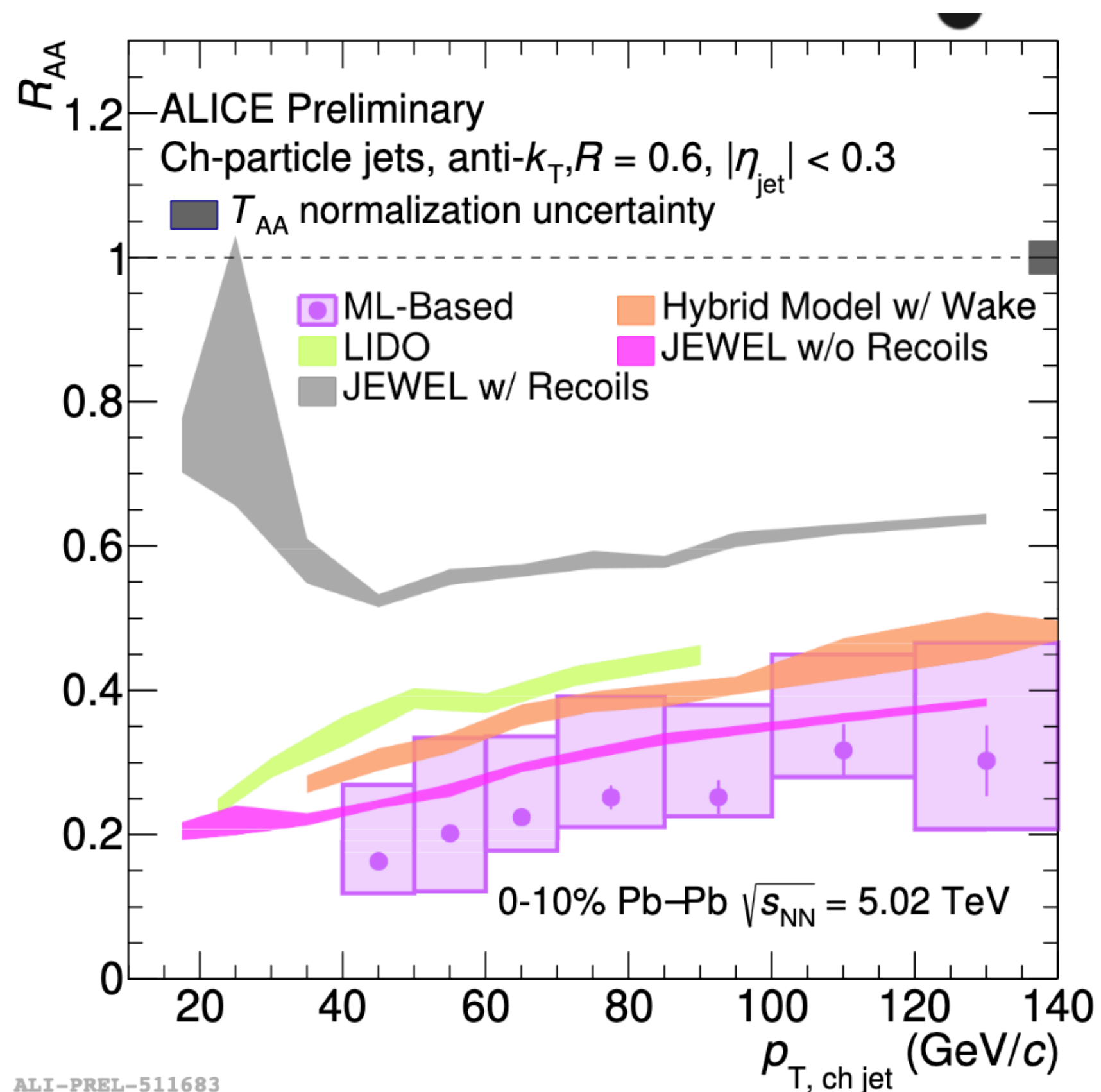
**Width of the fluctuations vs jet  $R$**



Method paper: *Haake, Loizides Phys.Rev. C99 (2019) no.6, 064904*

# Fake jets

d) by reducing bkg fluctuations using for instance ML techniques

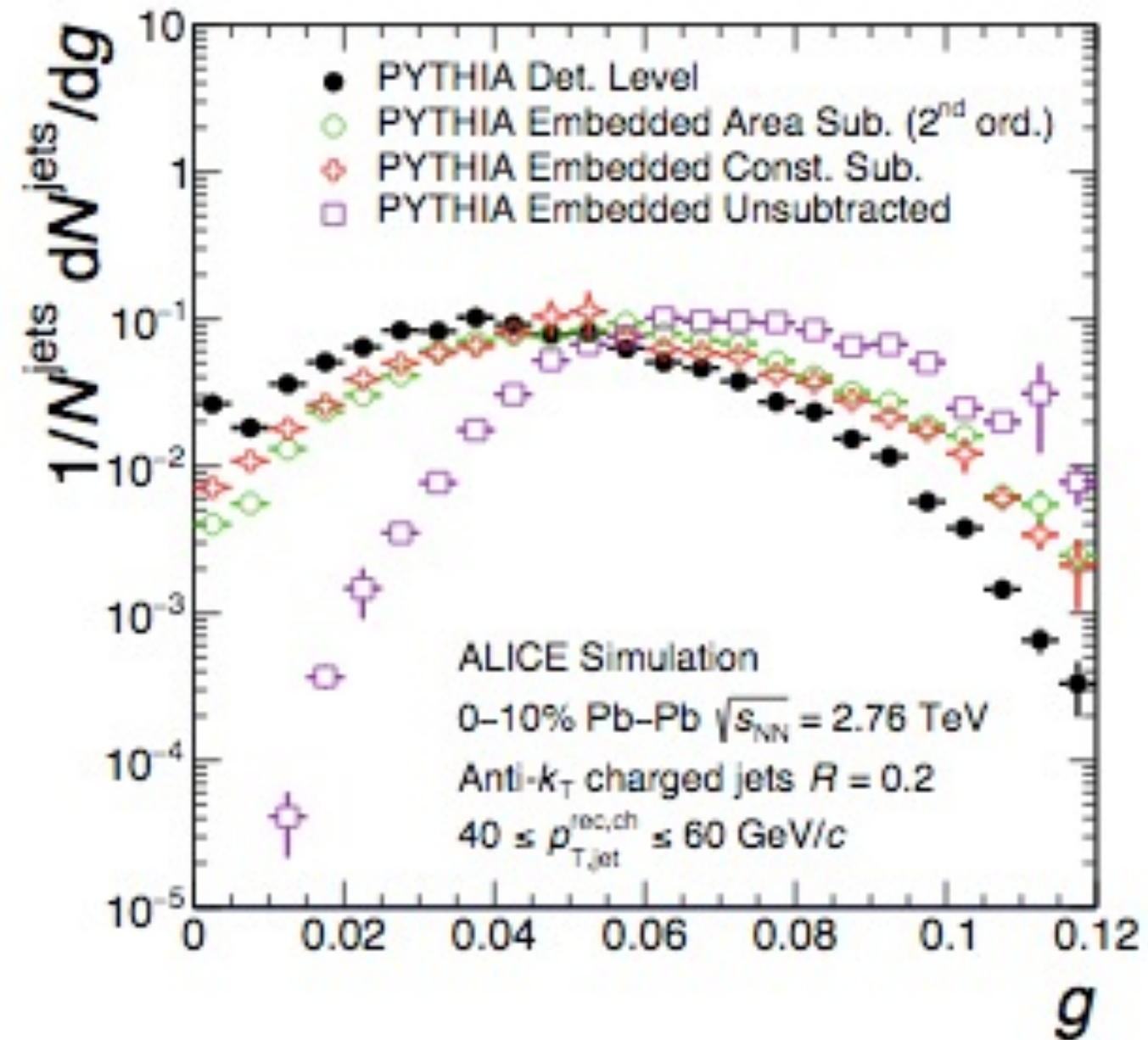


ALI-PREL-511683

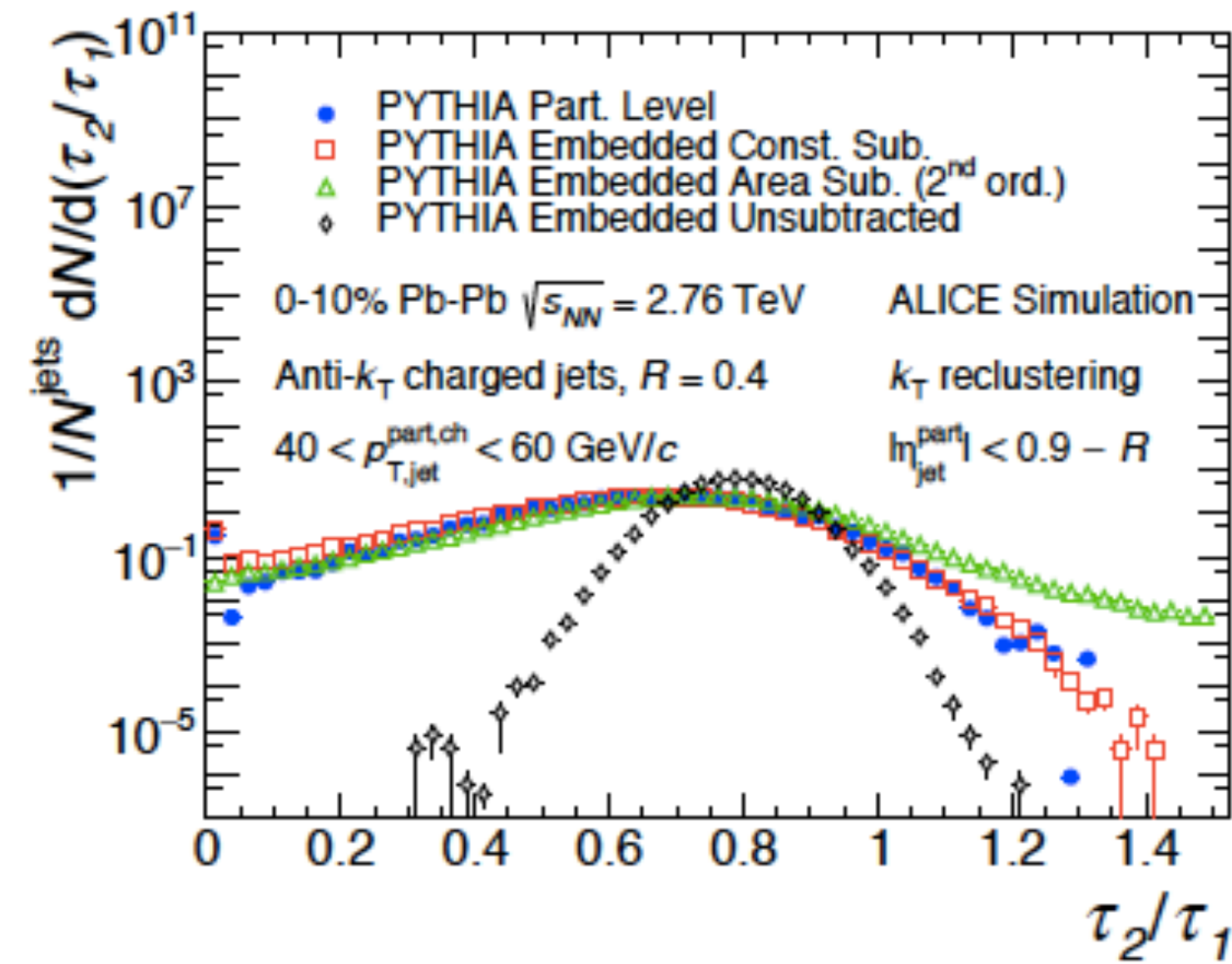
(recall lesson 2. and ALICE results for large-R jets at unprecedently low  $p_T$ )  
 Uncertainty dominated by systematic uncertainty on training model dependency

# Background at the level of jet substructure

Subtraction Performance angularity



Subtraction performance 2-subjettiness



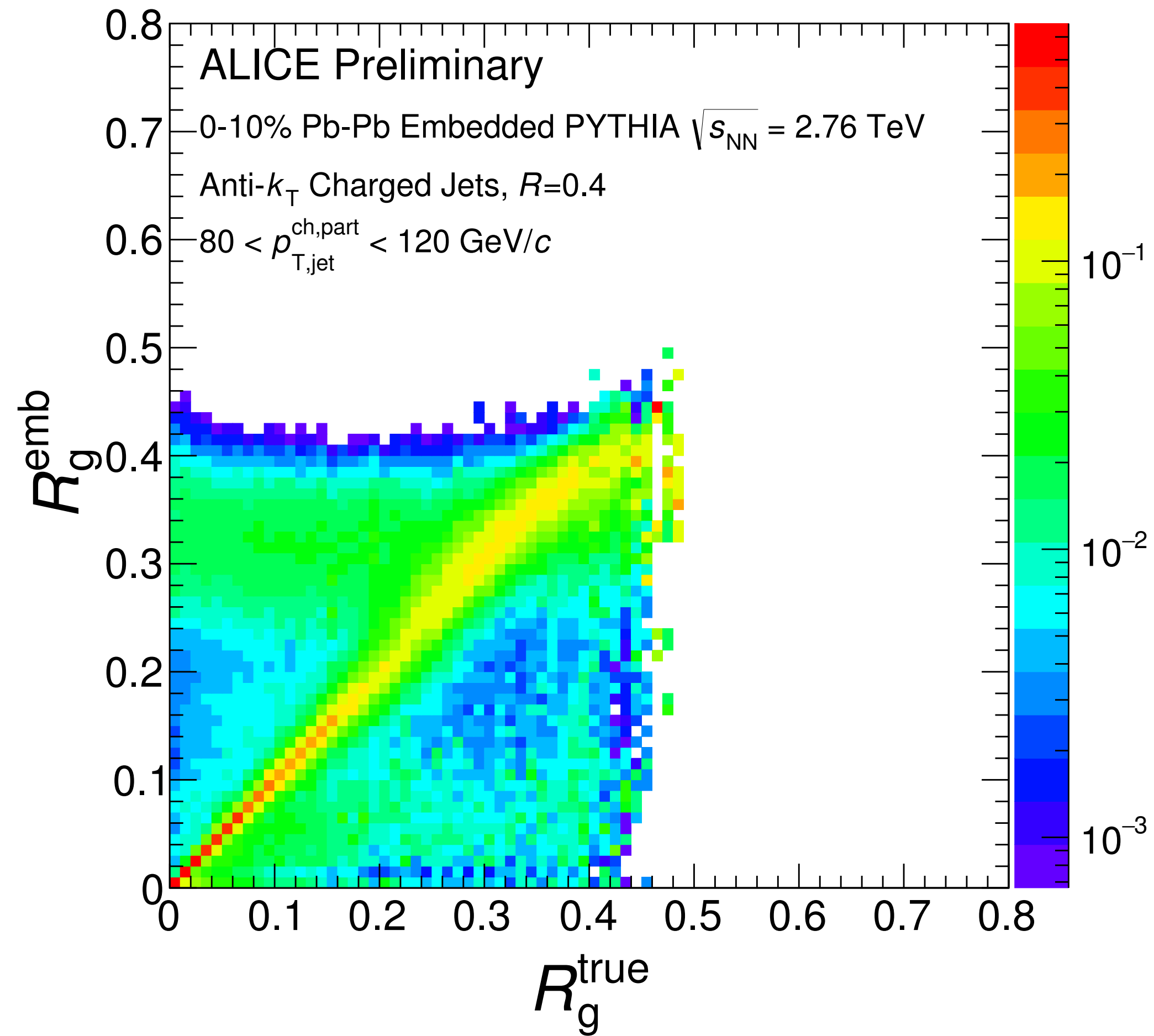
Pedestal background contamination subtracted using Area-derivatives [1] and Constituent Subtraction [2] methods.

Residual differences between the probe and the subtracted embedded probe are unfolded

[1] Soyez et al, Phys.Rev.Lett. 110 (2013) no.16, 162001

[2] Berta et al, JHEP 1908 (2019) 175

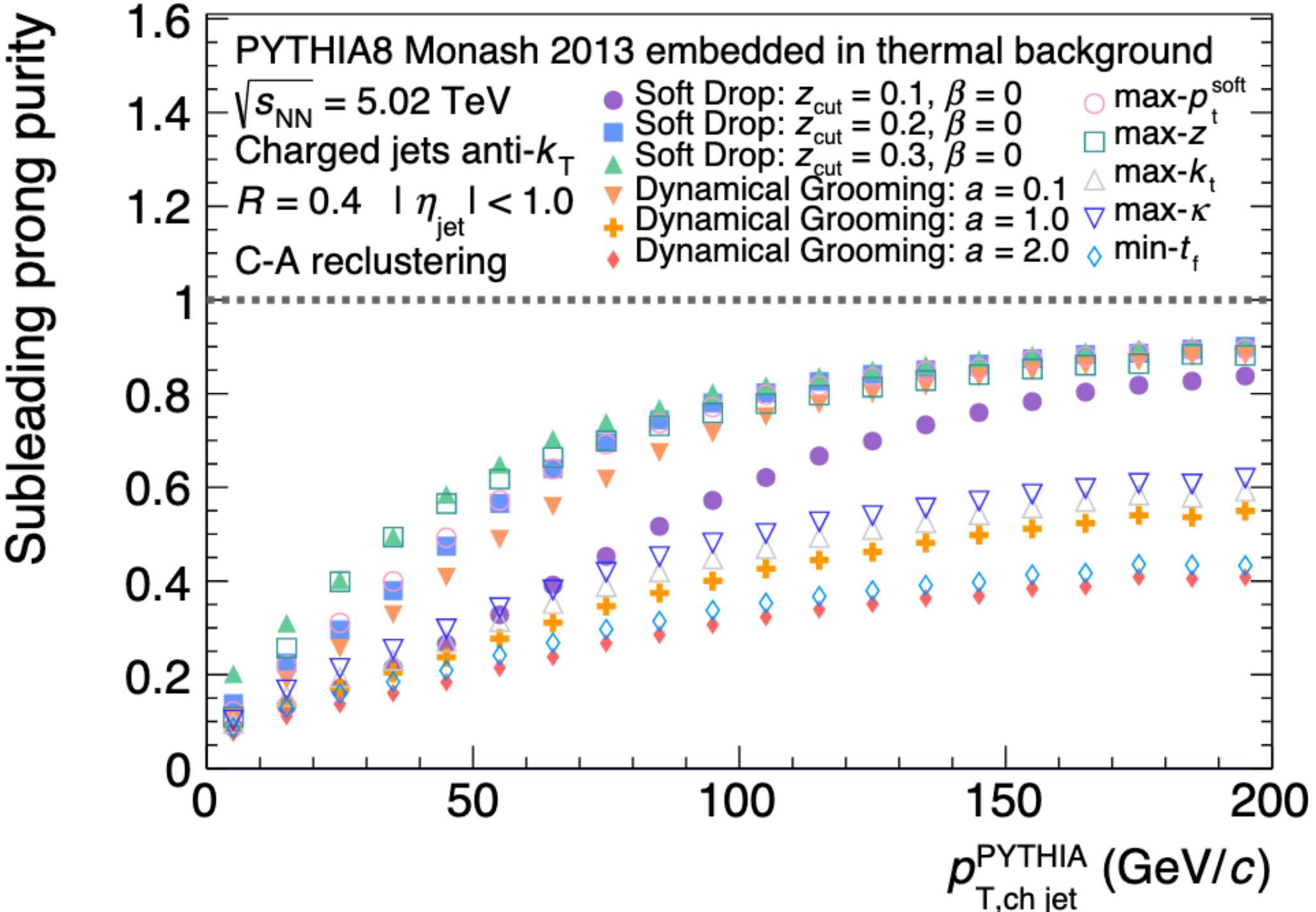
# The groomed substructure response



Off-diagonalities in the response due to fake prongs  
Render the problem not unfoldable

ALI-SIMUL-155665

# Fake splittings

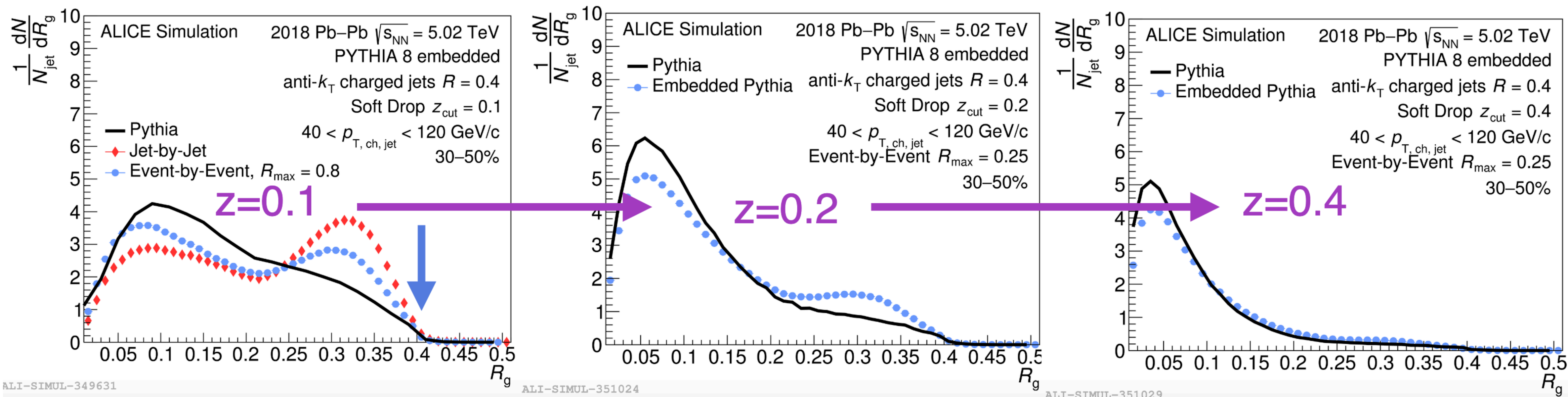


Subleading prong purity is the fraction of reconstructed subleading prongs that are connected to the true subleading prong

See big improvement when varying  $z_{cut}$  from 0.1 to 0.2 (harder prongs, more difficult that it's just bkg)

See worse performance for dynamical grooming

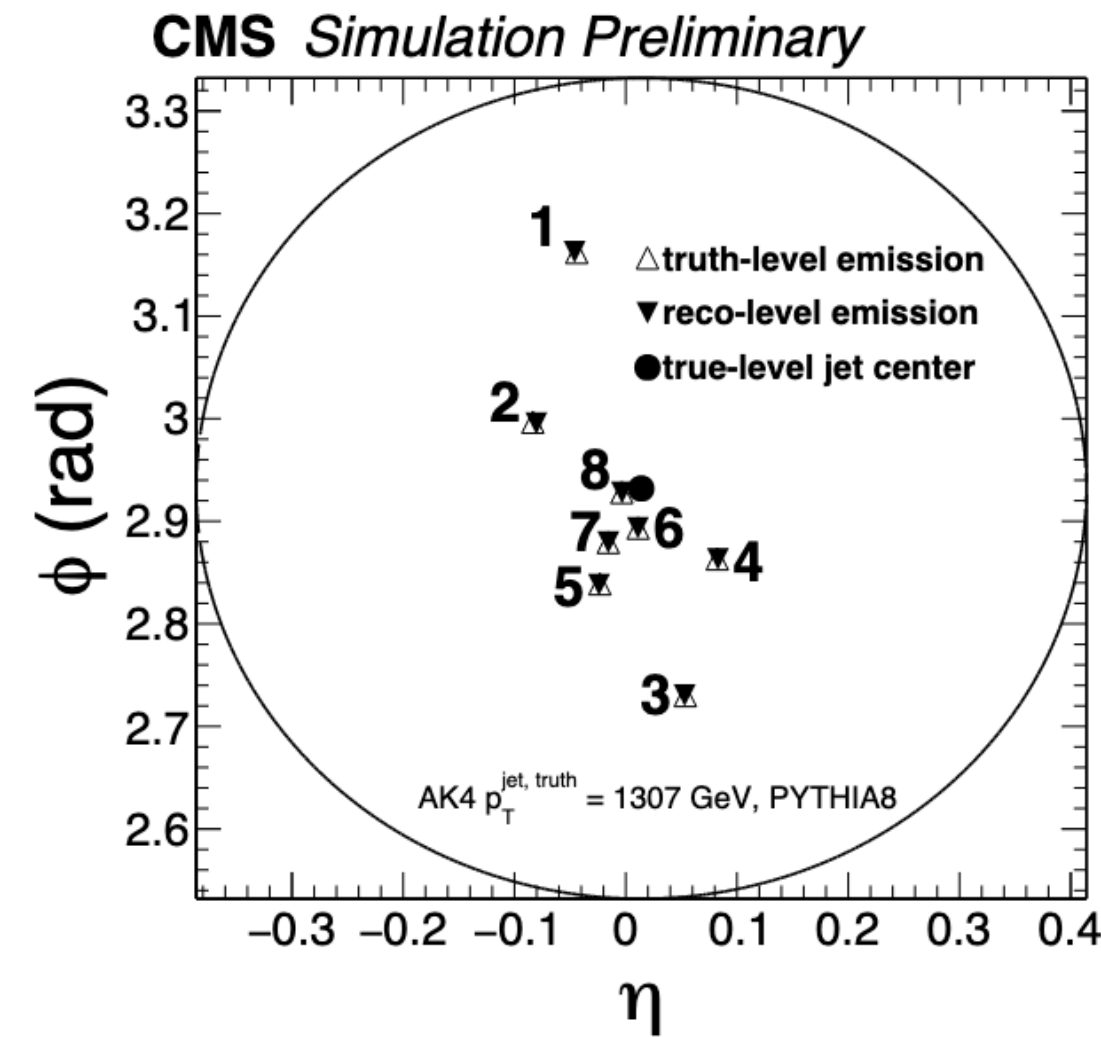
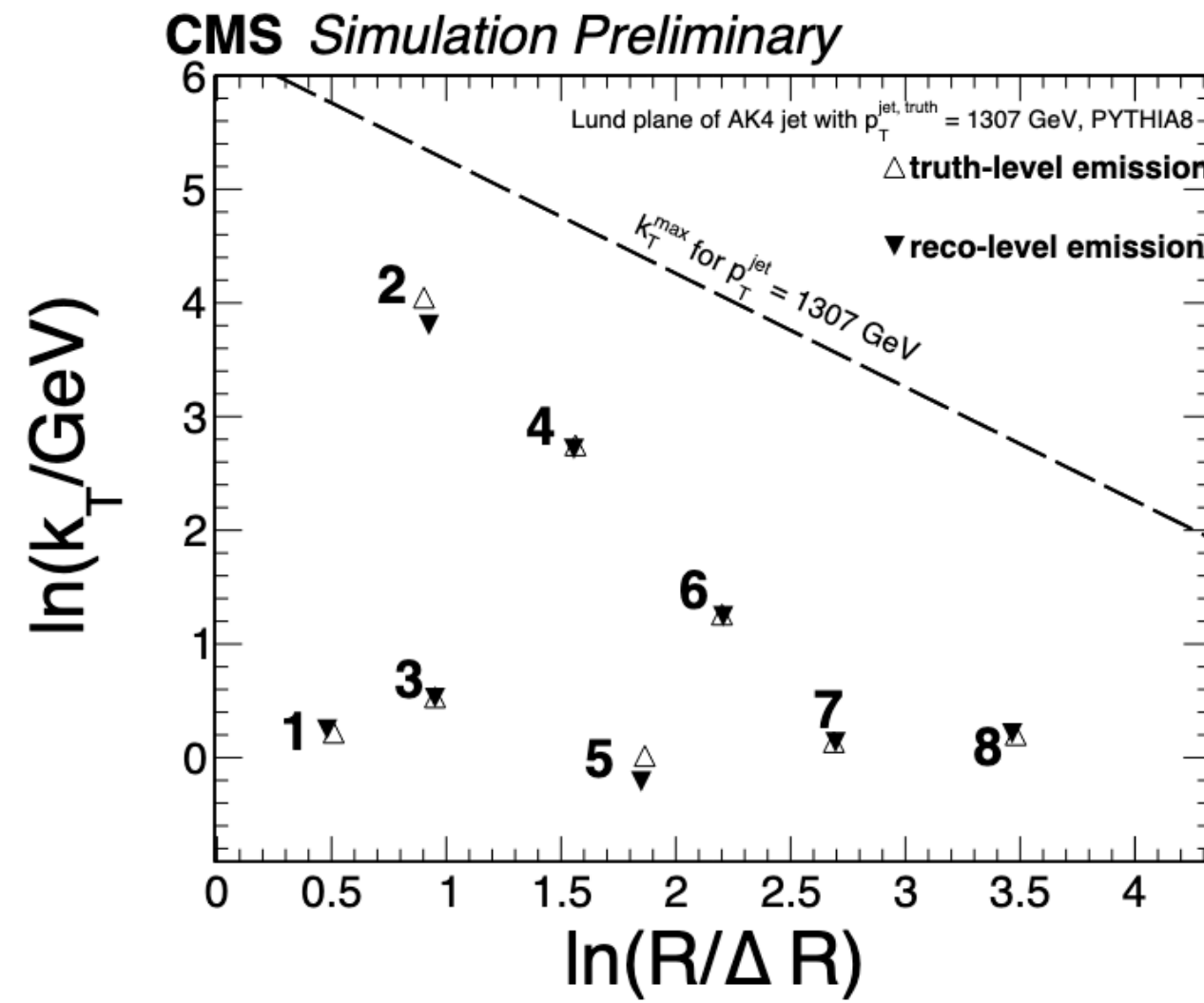
# Experimental strategies in HIC



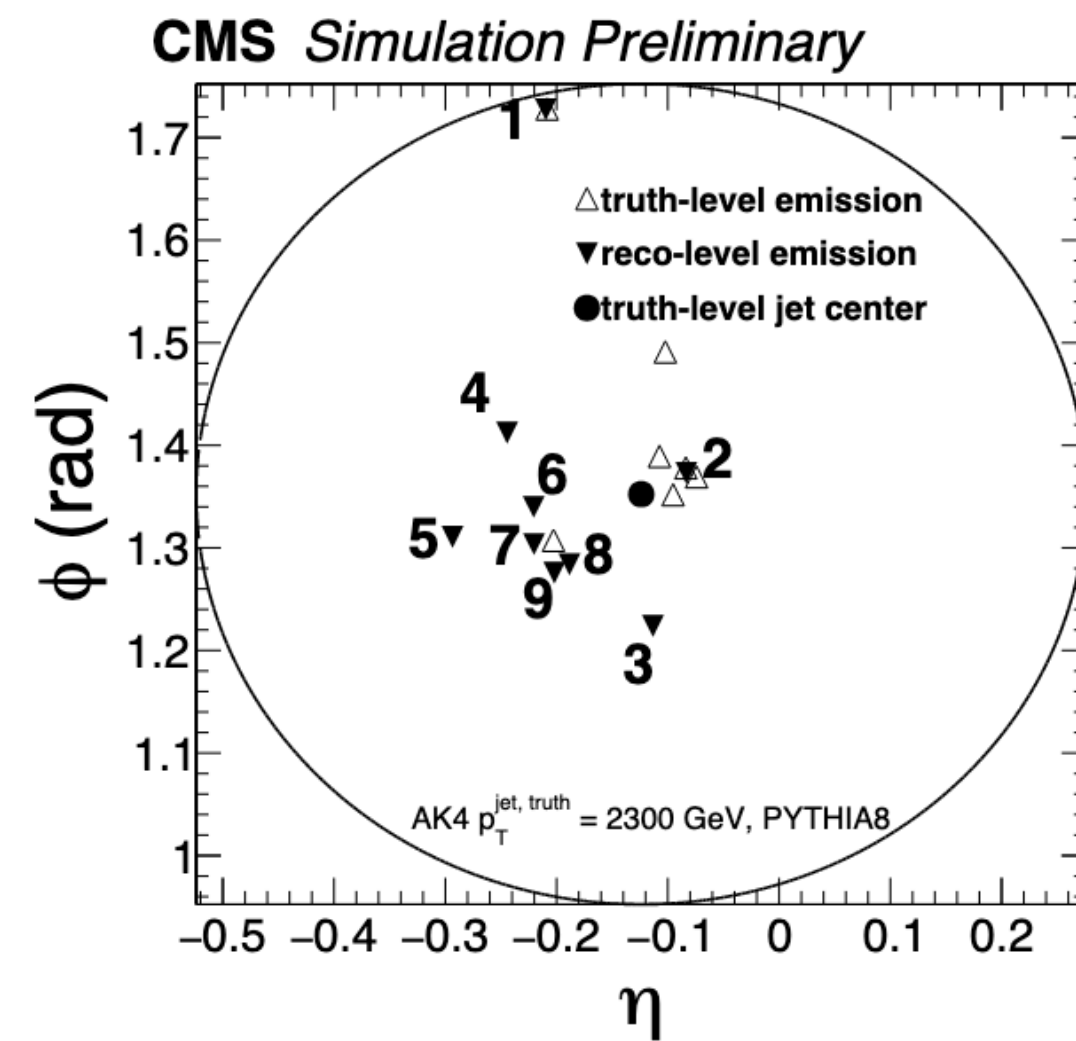
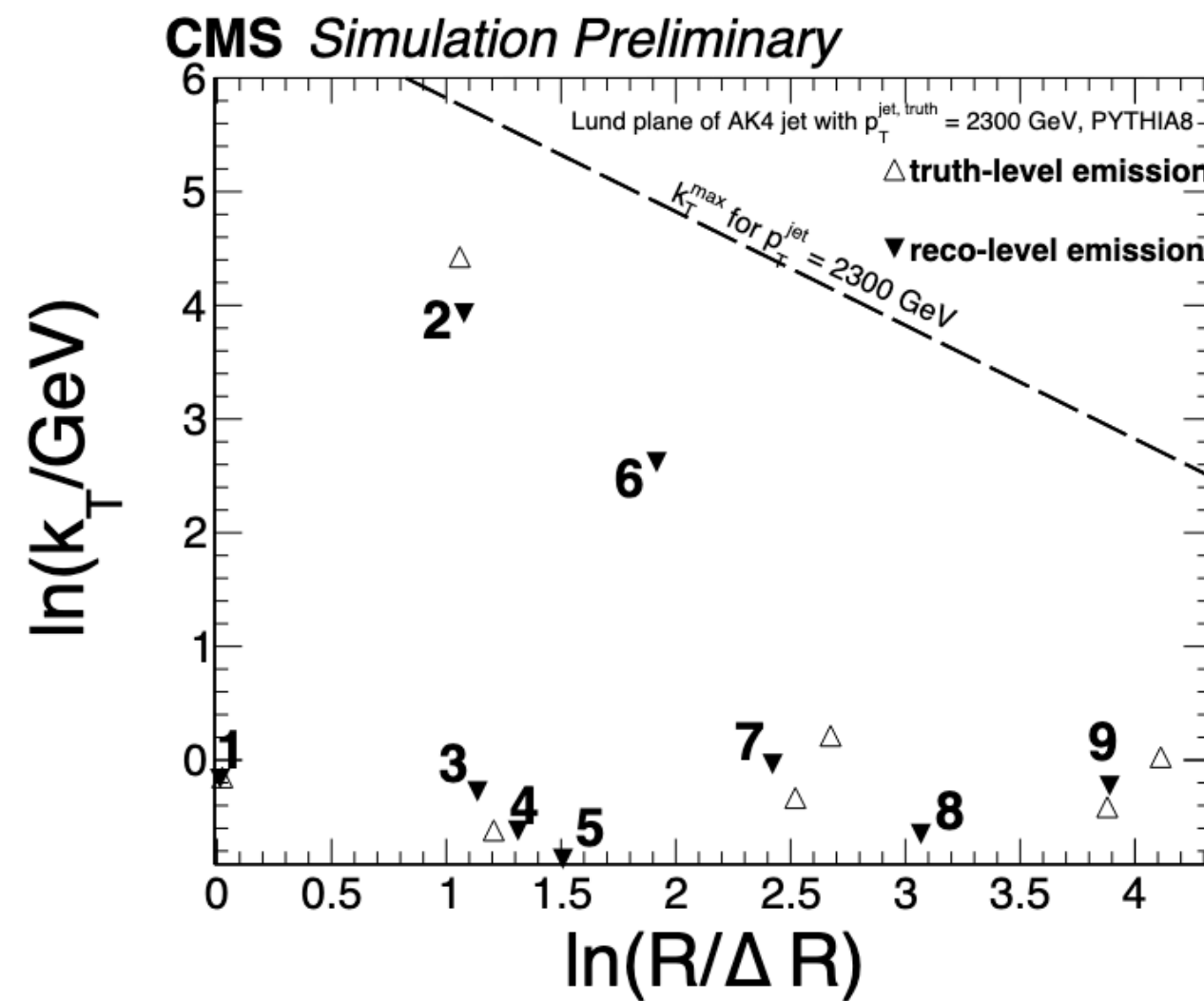
More aggressive grooming cuts reduce combinatorial prongs (but also naturally worsens statistics)

See [Laura Havener's talk at LHCP'20](#)

# Fake splittings and mismatches in pp



good mapping!



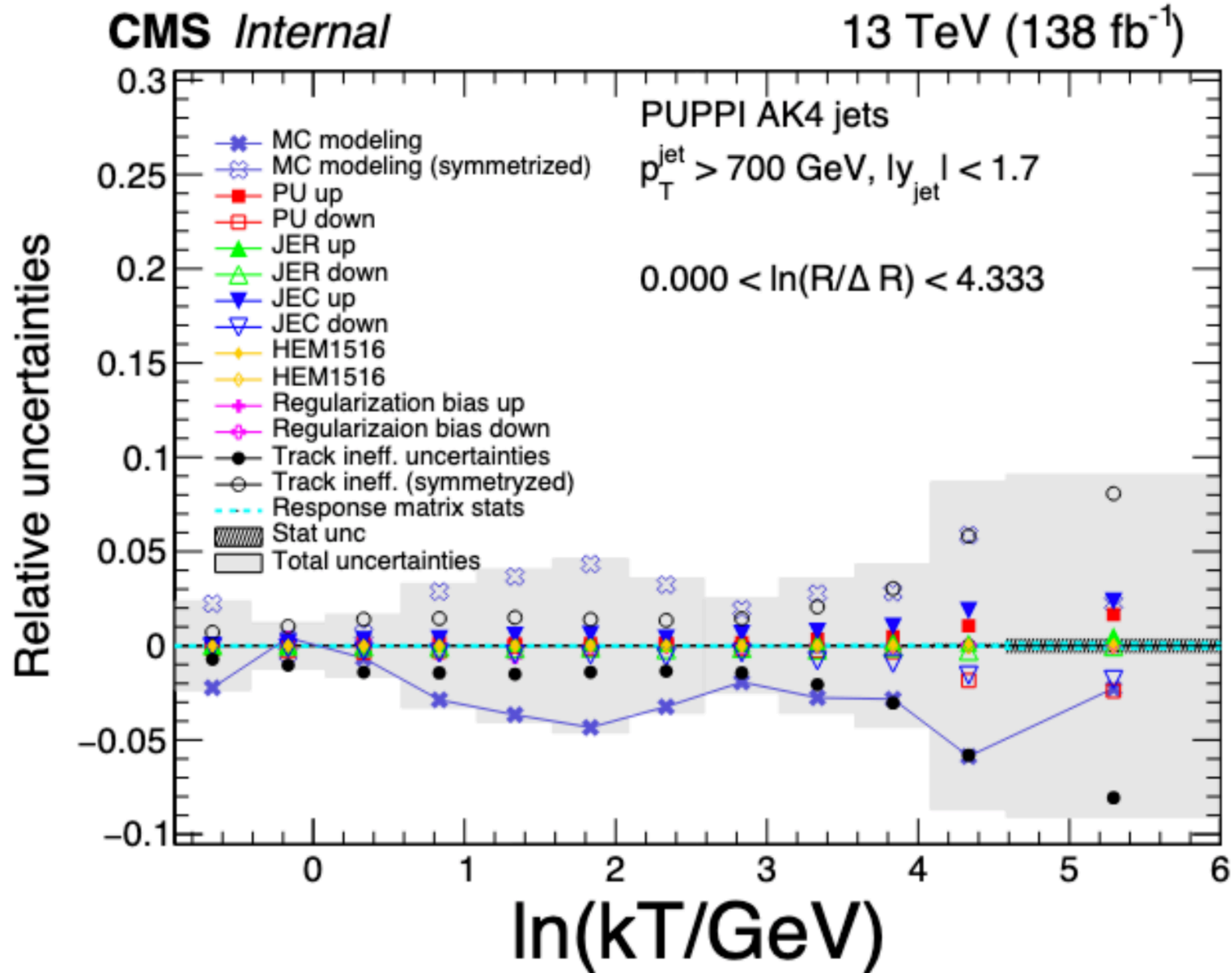
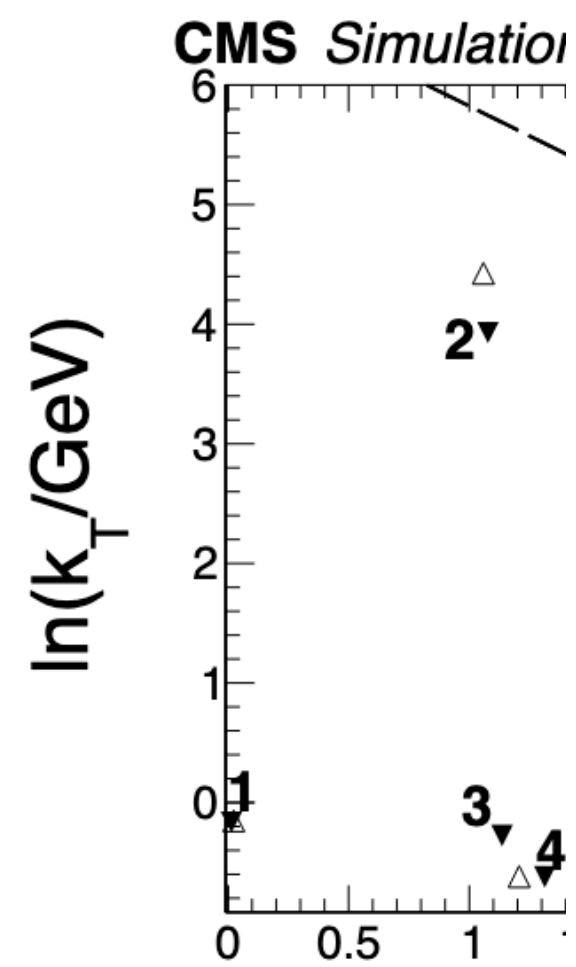
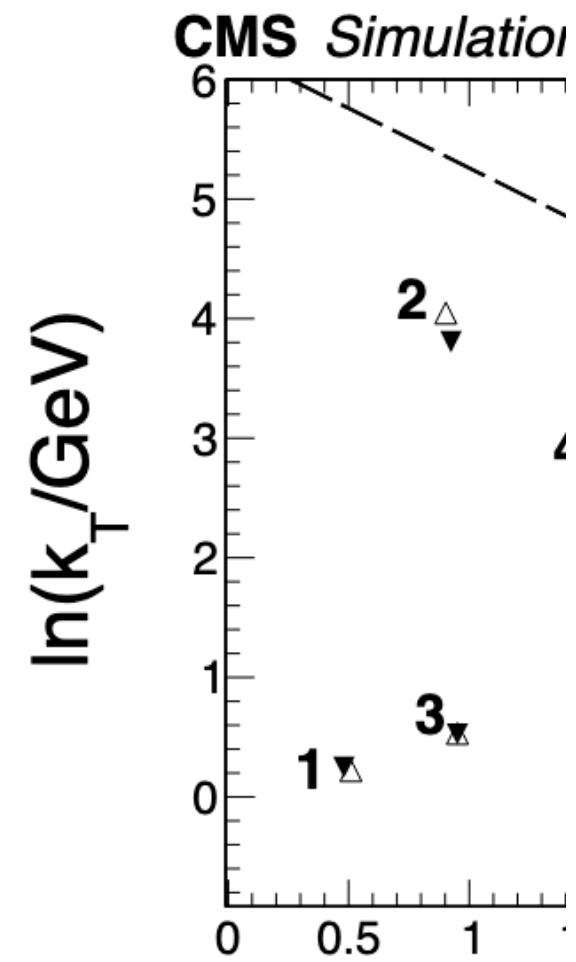
bad mapping!

all gone wrong since the beginning  
due to high- $k_T$  prongs being very symmetric in  $z$   
->easy to swap lead and subleading prong at det-level  
due to track.eff losses or pileup

from preliminary results of Lund plane in CMS by Cristian Baldenegro



# Fake splittings and mismatches in pp

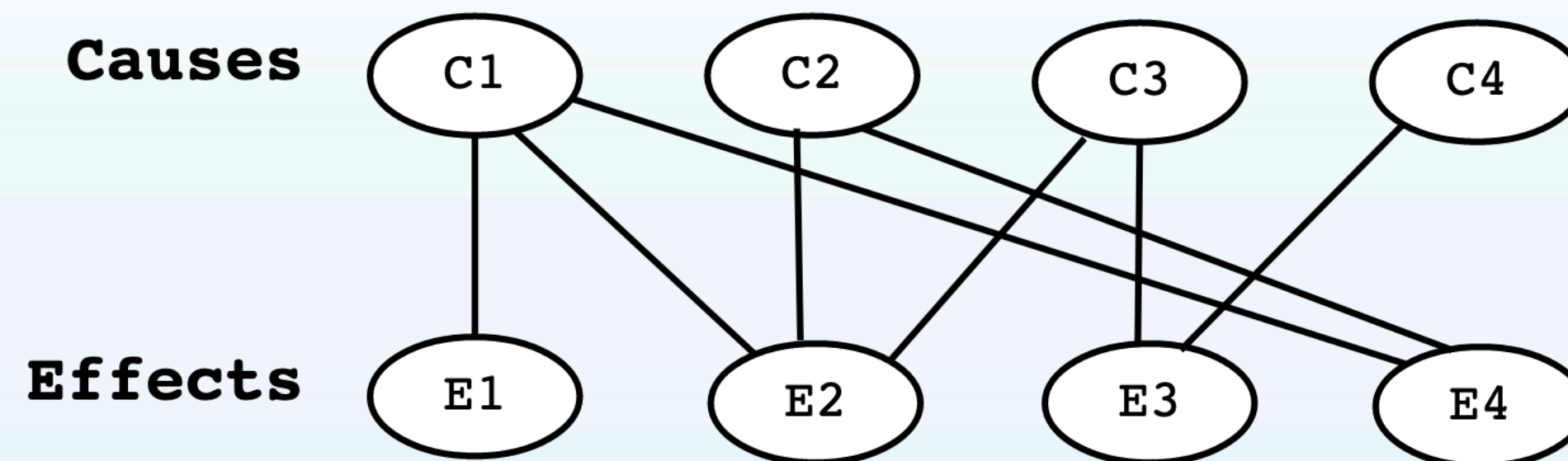


and this has consequences in terms of large sys. uncert. due to tracking ineff. uncertainty at high  $k_T$ , which is in principle the cleanest region for theory comparisons



## The unfolding

The same *apparent* cause might produce several, different effects



Given an observed effect, we are not sure about the exact cause that has produced it.

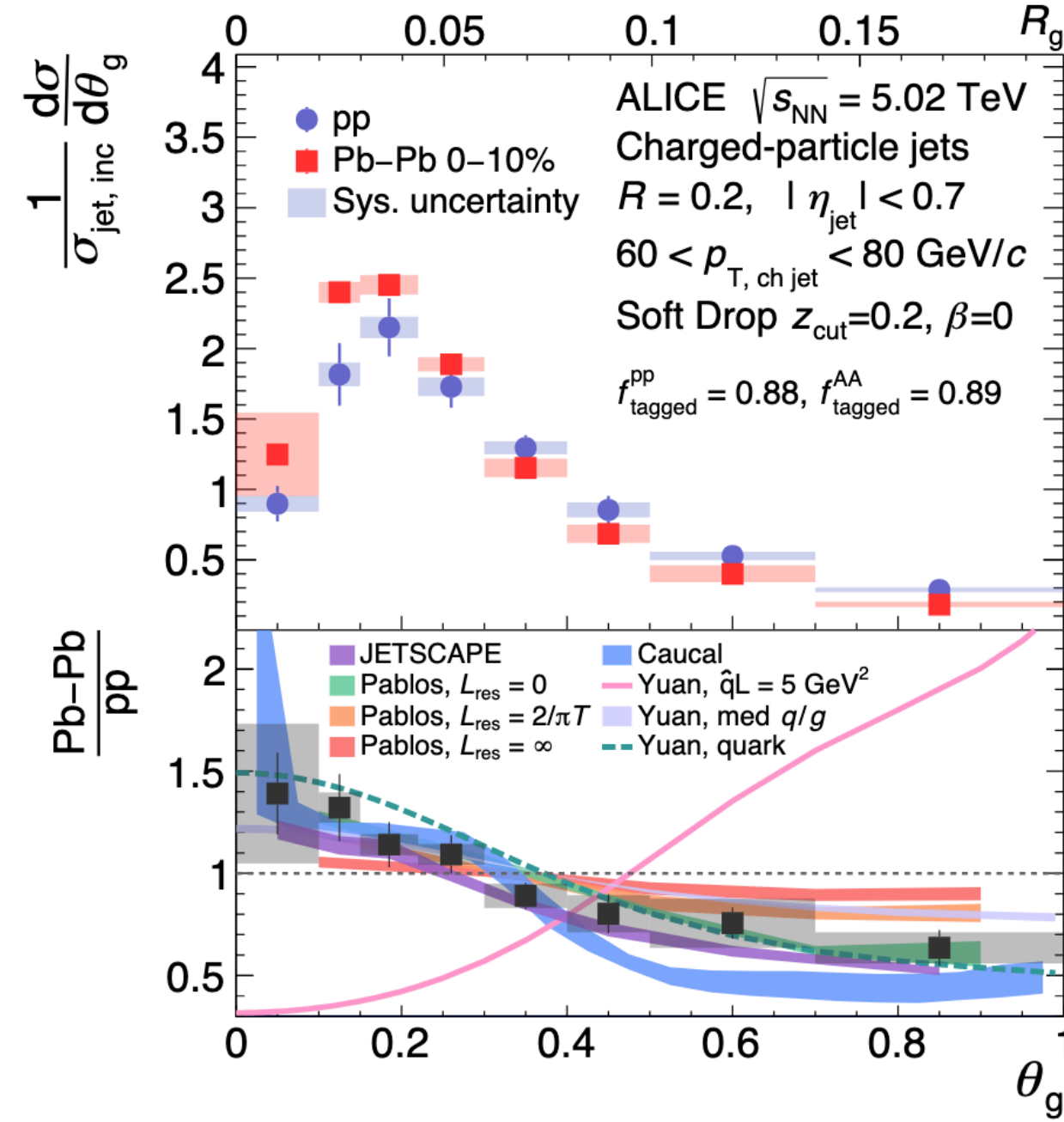
$P(i \rightarrow j)$  can be estimated by MC (response matrixes I've just shown you)

$\text{Measured}_i = \sum_j P(j \rightarrow i) \times \text{Truth}_j$  -> this is forward folding, clean numerical procedure

Truth has to be guessed, several methods available, have a look to the Bayesian method which is commonly used:

Bayesian method: [https://www.roma1.infn.it/~dagos/unf2\\_hh.pdf](https://www.roma1.infn.it/~dagos/unf2_hh.pdf)

# The final results with systematic uncertainties



Pb–Pb		Relative uncertainty (%)					
		Trk. eff.	Unfolding	Generator	Tagging	Bkgd. sub.	Total
$z_g$							
0–10%	$R = 0.2$	1–4%	1–4%	1–7%	1–2%	1–6%	4–10%
0–10%	$R = 0.4$	1–13%	1–4%	1–7%	2–26%	4–28%	9–41%
30–50%	$R = 0.4$	0–2%	0–5%	1–7%	1–6%	2–5%	5–9%
$\theta_g$							
0–10%	$R = 0.2$	1–8%	1–4%	1–5%	1–19%	1–14%	3–24%
30–50%	$R = 0.4$	3–6%	1–7%	1–5%	0–4%	2–15%	6–15%
30–50%	$R = 0.4$ $z_{\text{cut}} = 0.4$	4–11%	2–11%	1–5%	1–5%	1–13%	4–20%

**Table 1:** Summary of systematic uncertainties on the Pb–Pb measurements. The ranges correspond to the minimum and maximum systematic uncertainties obtained. All values correspond to  $z_{\text{cut}} = 0.2$  unless otherwise noted.

See dominant effects of tracking efficiencies and model dependency