

ALICE b-jet tagging in Run2 5 TeV pPb collisions with the SV method

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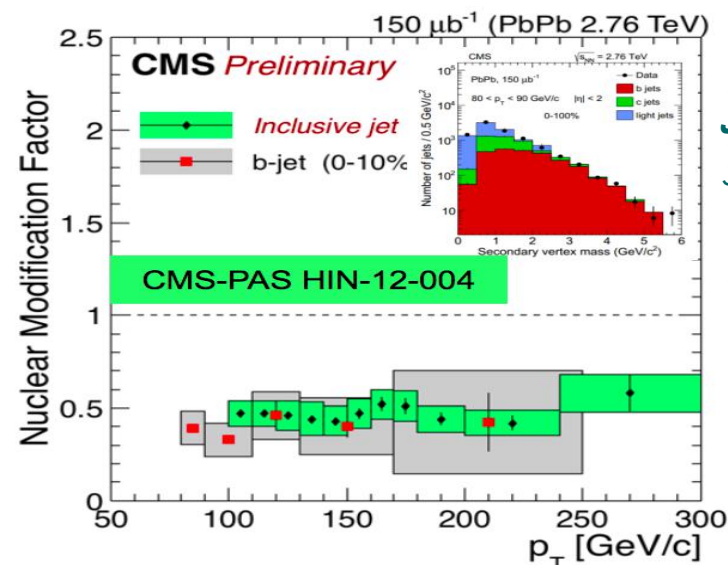


ALICE



goals and datasets

- **pp:**
 - pQCD benchmark and baseline for nuclear modification
 - Color charge vs. mass/flavor effects?
- **p-A:**
 - CNM effects?
Baseline for modification in hot medium
- **A-A:**
 - mass ordering?
 - Low/intermediate p_T - unique
 - contribution of gluon-splitting to direct b quark production?
 - Radiative or collisional energy loss?
- **Experimental data:**
 - pp 2017 data at 5 TeV (IP method) ~600M evts
 - p-Pb 2016 data at 5 TeV (IP and SV methods) ~900M Minimum Bias evts
 - ITS+TPC tracks, $p_T > 0.15$
 - Anti-kT jets, $R=0.4$, $|\eta| < 0.5$



extracting the b-jet cross section

A. Jet Reconstruction

B. b-jet selection

b-jet: presence of a b-hadron inside a cone with given R centered on the jet axis

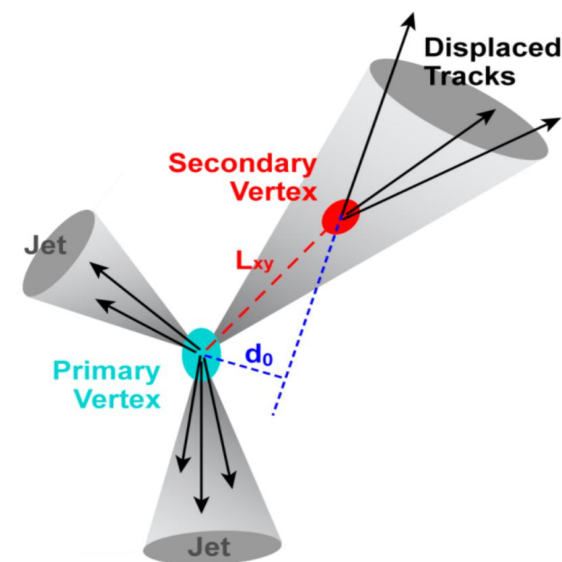
1. *Impact parameter significance method*

based on the closest approach to the primary vertex of tracks inside a jet

Hadi Hassan (Linus Feldkamp, Min Jung Kweon, Minjung Kim)

2. *Displaced secondary vertex method*

secondary vertex reconstruction and evaluation of its distance from the primary vertex - *Ashik Ikbal Sheikh, Filip Křížek, Artem Isakov, R.V. (Elena Bruna, Lukás Kramárik, Gyulnara Eyyubova)*



C. Statistically remove non-b jets from tagged sample

D. Unfolding

E. Efficiency correction

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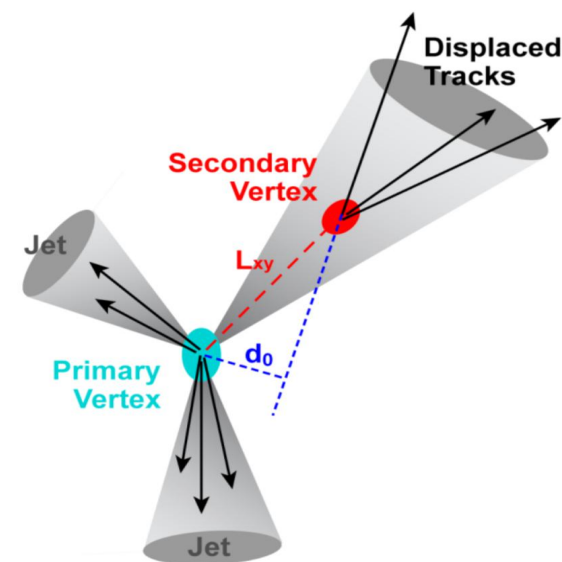
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spectrum extraction

- Extract SV tagging **efficiency** from Monte Carlo

$$\epsilon_{b,c,udsg}(p_{T,\text{ch. jet}}^{\text{det.}}) = \frac{N_{b,c,udsg}^{\text{tagged}}(p_{T,\text{ch. jet}}^{\text{det.}})}{N_{b,c,udsg}^{\text{gen.}}(p_{T,\text{ch. jet}}^{\text{det.}})}$$

**==> Reliable MC
is essential!**

- Extract **purity** from template fit and MC

$$P(p_{T,\text{ch. jet}}^{\text{det.}}) = \frac{N_{\text{b-jets}}(p_{T,\text{ch. jet}}^{\text{det.}})}{N_{\text{all jets}}(p_{T,\text{ch. jet}}^{\text{det.}})}$$

- Correct tagged inclusive raw spectrum:

$$\frac{1}{N} \frac{dN_{\text{measured,b}}}{dp_T} = \frac{1}{N} \frac{1}{\epsilon_b} \cdot P \cdot \frac{dN_{\text{tagged}}}{dp_T}$$

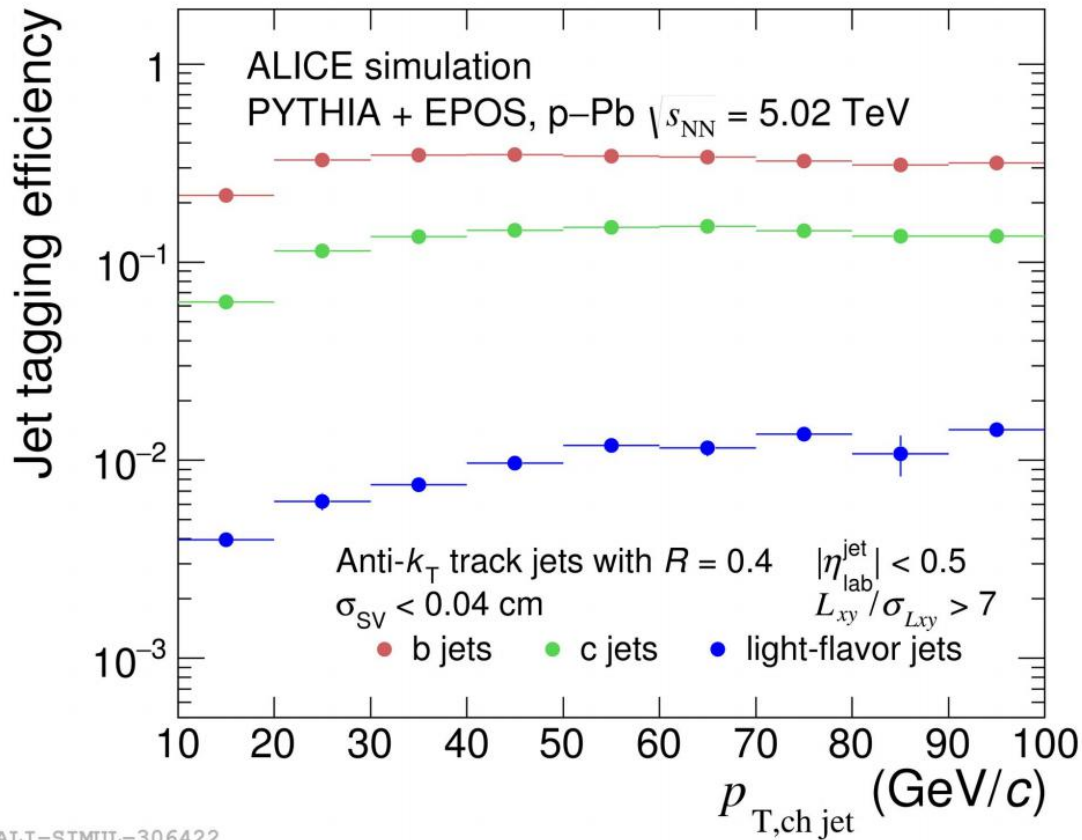
- Unfolding and final correction step $p_{T,\text{ch. jet}}^{\text{det.}} \rightarrow p_{T,\text{ch. jet}}$

$$\frac{d\sigma_{\text{b-jet}}}{dp_T} = \frac{1}{L_{\text{int}}} \cdot \text{Unfolded} \left(\frac{dN_{\text{measured,b}}}{dp_T} \right)$$

analysis

- Data extraction (Run2 p-Pb, LHC16{q,t}):
- Efficiency and purity corrections
 - Efficiency: LHC17h6{a,b,c,d,e,f}2
 - Purity #1 "POWbc": real inclusive jets and POWHEG c,b spectra
 - Purity #2: "data-driven": template fits from LHC17h6_2 simulations
==> a combined "hybrid" method
- Unfolding (SVD & Bayesian, binned)
 - Matrix based on LHC17h6_2, outliers removed ($p_T < p_{T\text{hard}} \times 4$)
 - PYTHIA hard processes + EPOS underlying event
- Systematics
 - Tracking & jet reconstruction related
 - b-tagging related

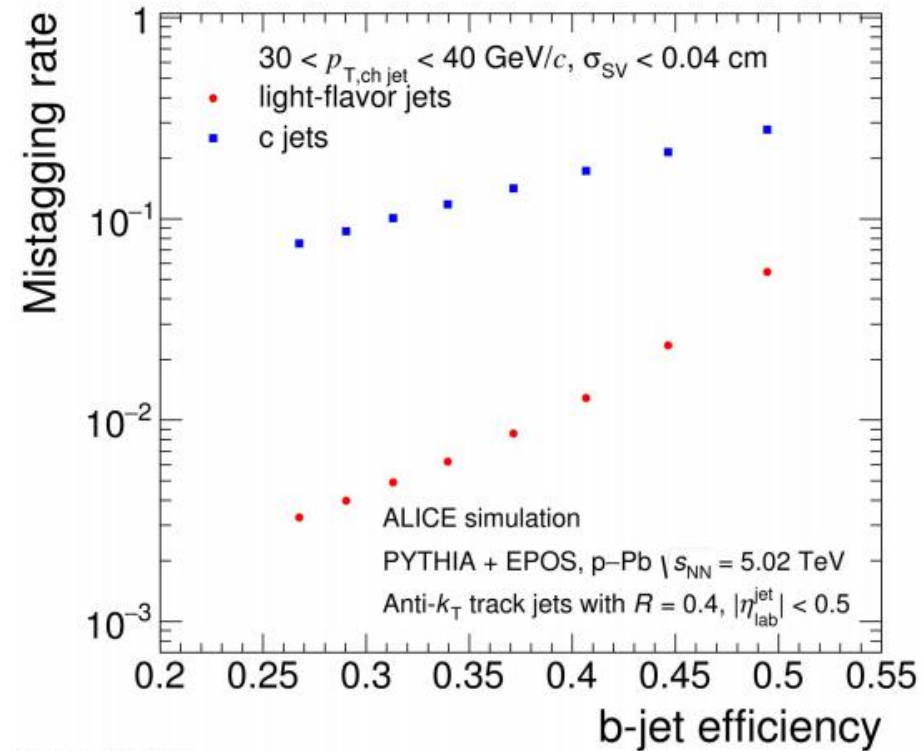
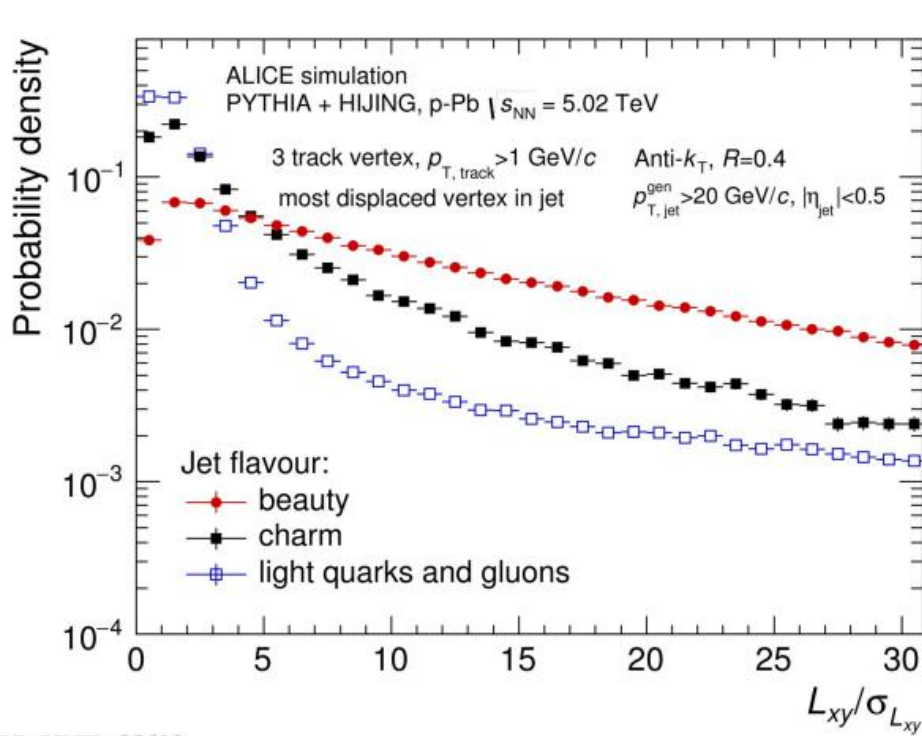
tagging efficiencies vs. p_T



■ Tagging cuts

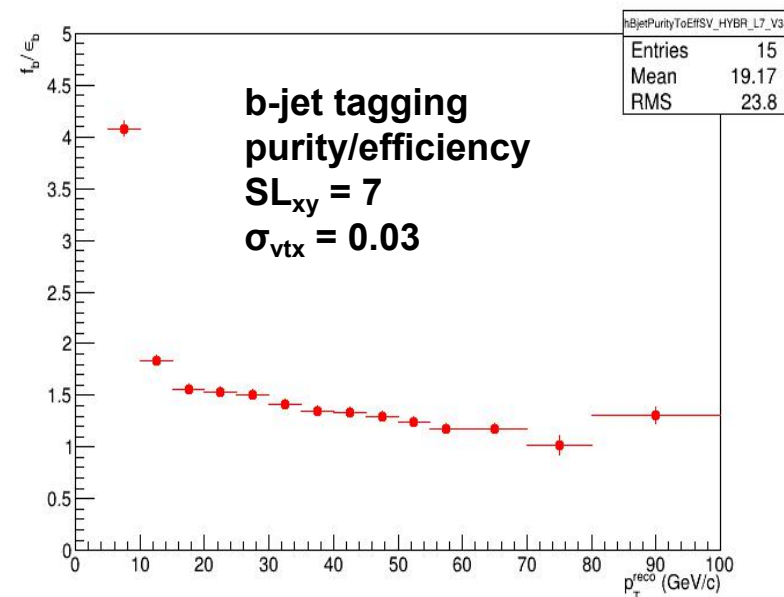
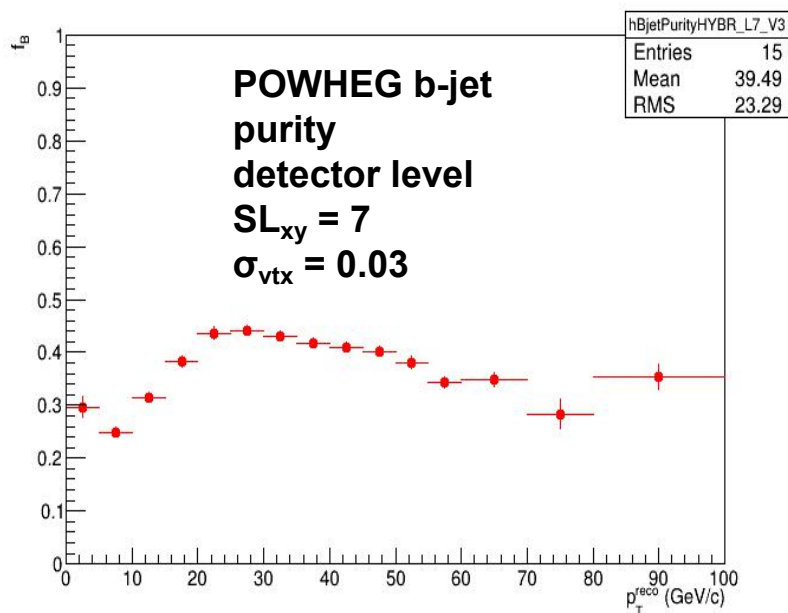
- Dispersion of reconstructed secondary vertex σ_{vtx}
- Significance of primary-secondary vertex distance $SL_{xy} = L_{xy}/\sigma_{Lxy}$

tagging performance



- Evolutions of efficiencies and mistagging rates with SL_{xy}
 - Left: efficiency vs. SL_{xy} , no sigvtx cut applied
 - Right: efficiency vs. mistagging rates for different SL_{xy} values

purity & tagging correction, POWHEG



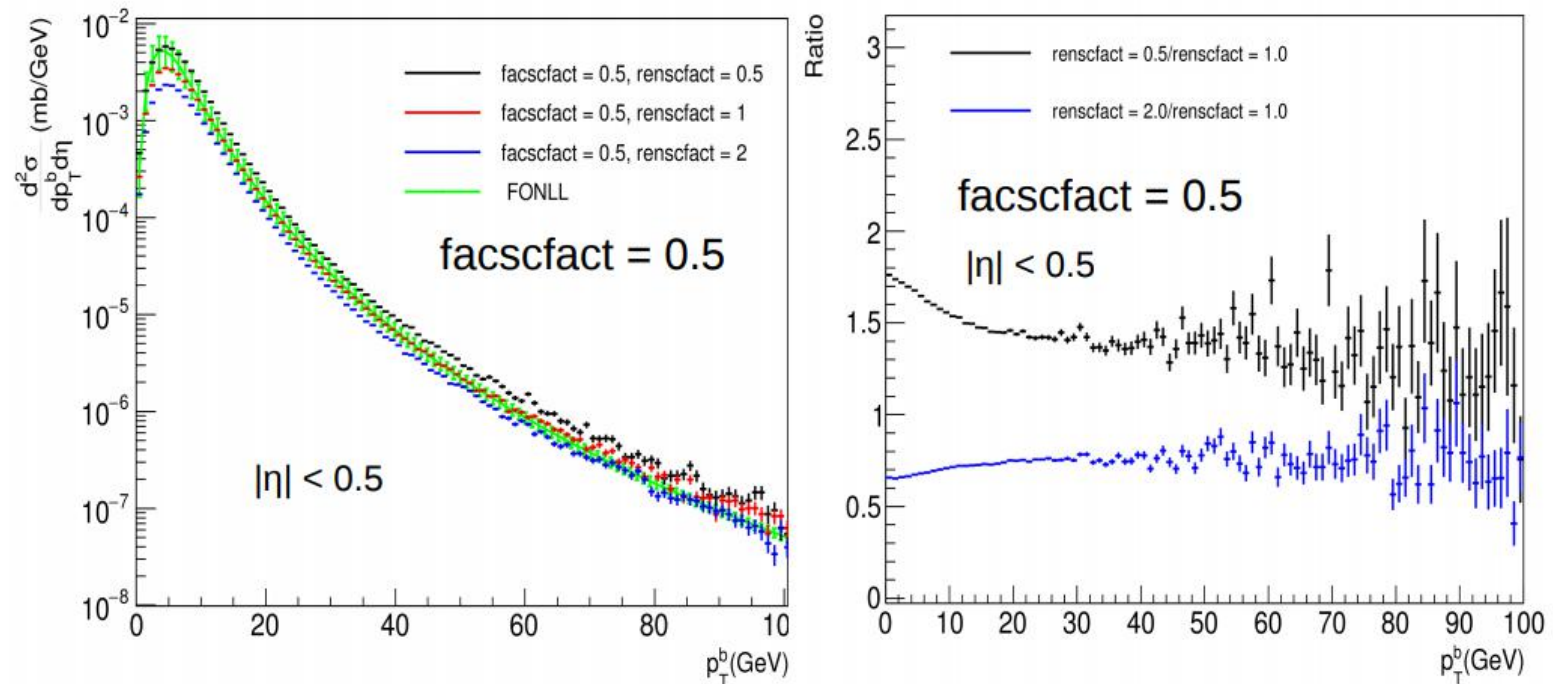
■ Purity obtained as:

- Going to detector level: POWHEG spectrum * detector matrix
- Using the good old formulae

$$f_b(p_T^{\det}) = \frac{N_b^{\text{tagged}}(p_T^{\det})}{N_{\text{inclusive}}^{\text{tagged}}(p_T^{\det})}$$

$$N_b^{\text{tagged}}(p_T^{\det}) = N_{\text{inclusive}}^{\text{tagged}}(p_T^{\det}) - N_b^{\text{Powheg}}(p_T^{\det}) \cdot \epsilon_b(p_T^{\det}) - N_c^{\text{Powheg}}(p_T^{\det}) \cdot \epsilon_c(p_T^{\det}) - \left(N_{\text{inclusive}}(p_T^{\det}) - N_c^{\text{Powheg}}(p_T^{\det}) - N_b^{\text{Powheg}}(p_T^{\det}) \right) \cdot \epsilon_{lf}(p_T^{\det})$$

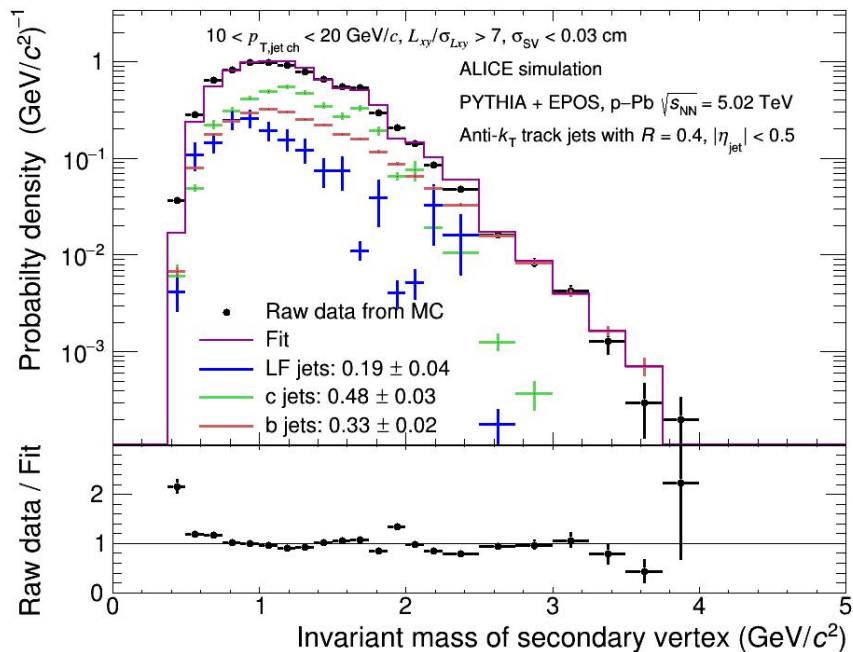
uncertainty in POWHEG



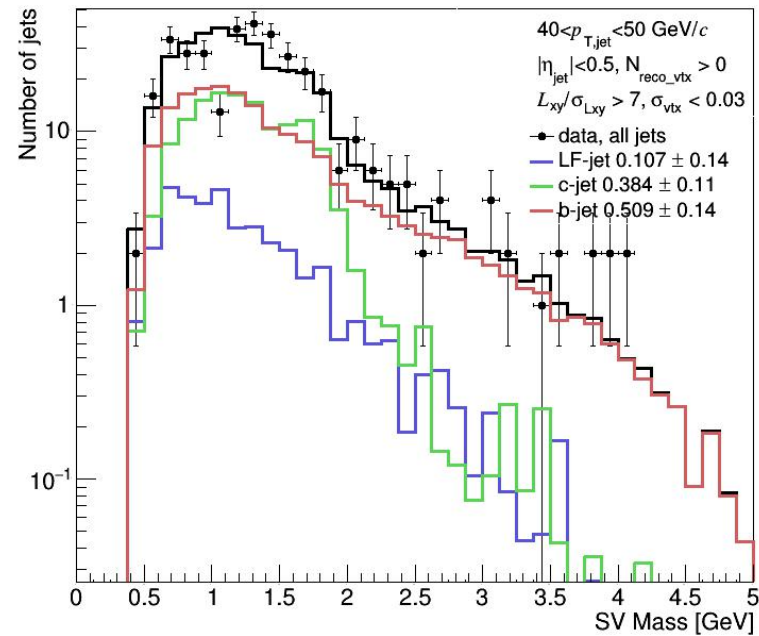
- Several variations (defaults in bold) and cross-variations:
 - $m_b=4.5$, **4.75**, 5.0 GeV ; $m_c=1.3$, **1.5**, 1.7 GeV
 - factorization scale = 0.5, **1.0**, 2.0 ; renormalization scale = 0.5, **1.0**, 2.0
 - Translates to a factor ~ 2 uncertainty on the corrected spectrum (later)
 - See more: backup slides and https://twiki.cern.ch/twiki/pub/ALICE/BtagSecVtx/PowhegSystematicsBeauty_ashik.pdf

purity, data-driven - template fits

Example: Minuit, lowest p_T bin



Example: RooFit, higher p_T bin



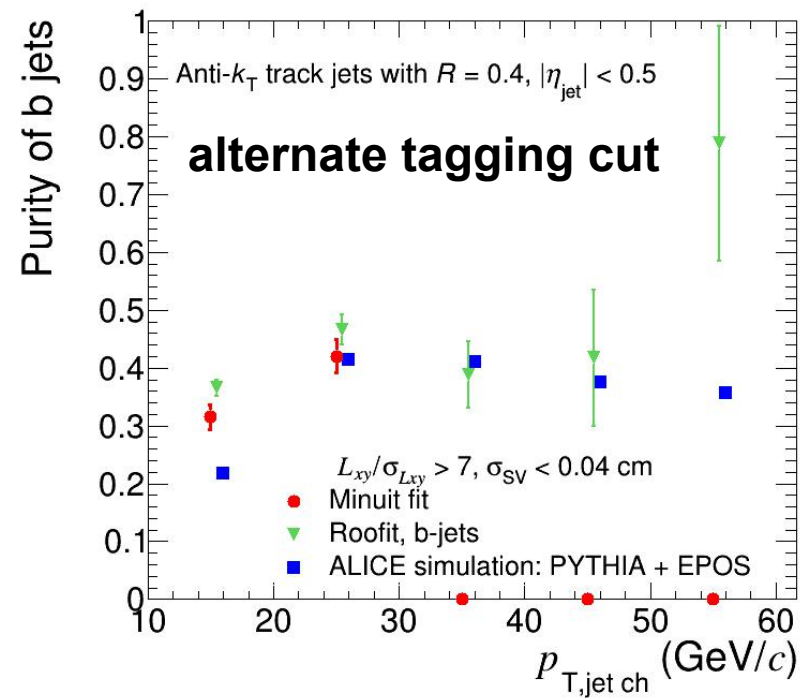
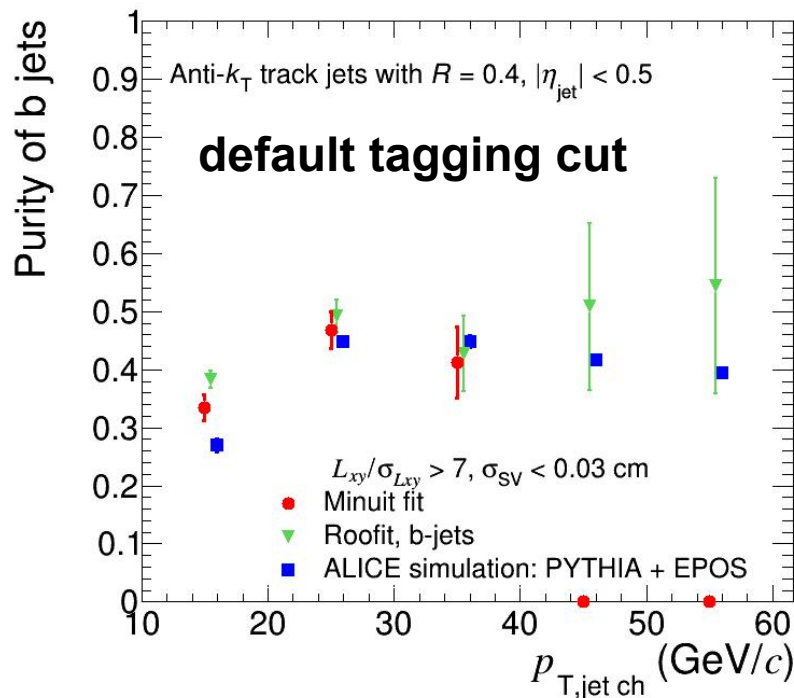
■ Minuit vs. RooFit on measured data

- RooFit: template errors ignored - can be a problem
- Minuit: correct treatment of errors;

$$F(n) = \sum_{i=1}^{n_{\text{bins}}} \frac{(Data_i - B_i * p_B - C_i * p_C - LF_i * p_{LF})^2}{\sigma_{Data_i}^2 + (\sigma_{B_i} * p_B)^2 + (\sigma_{C_i} * p_C)^2 + (\sigma_{LF_i} * p_{LF})^2}$$

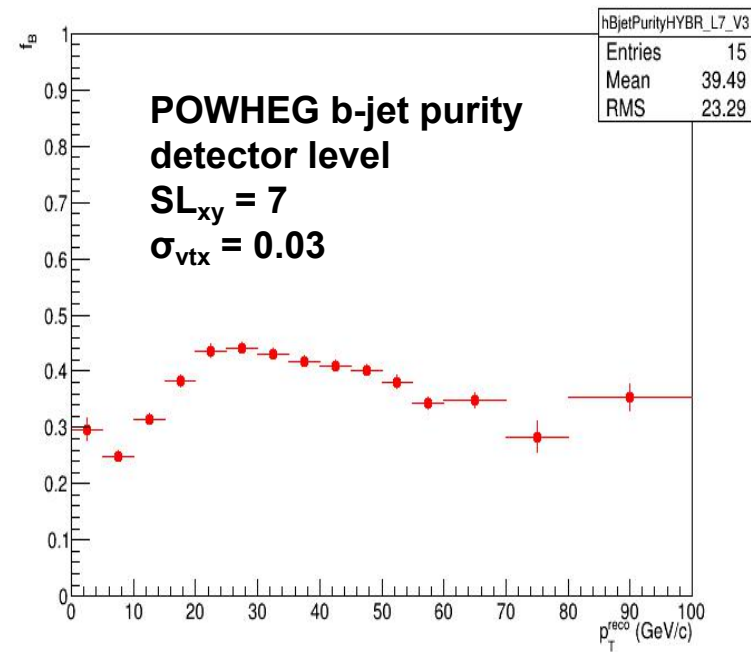
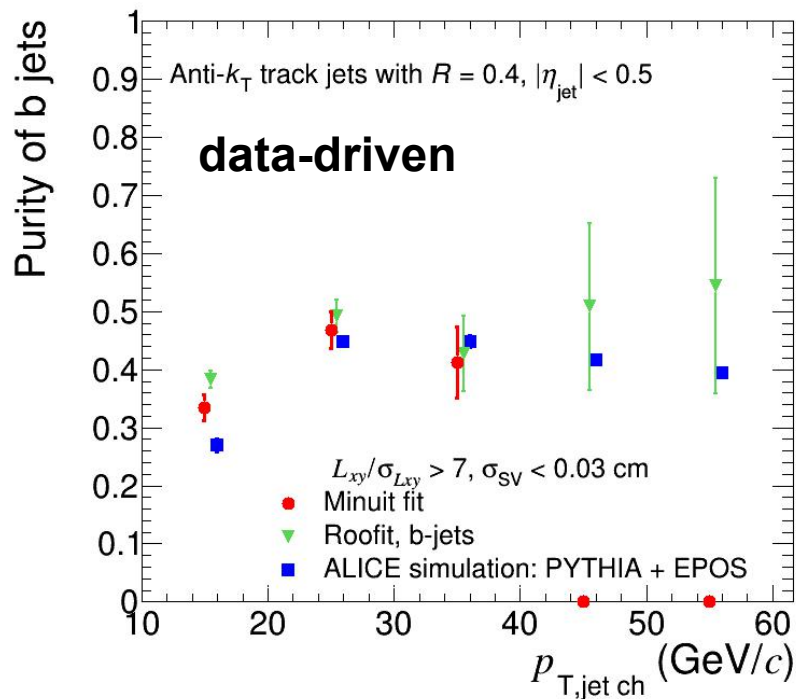
but: convergence problems at higher p_T

purity, data-driven - p_T -dependence



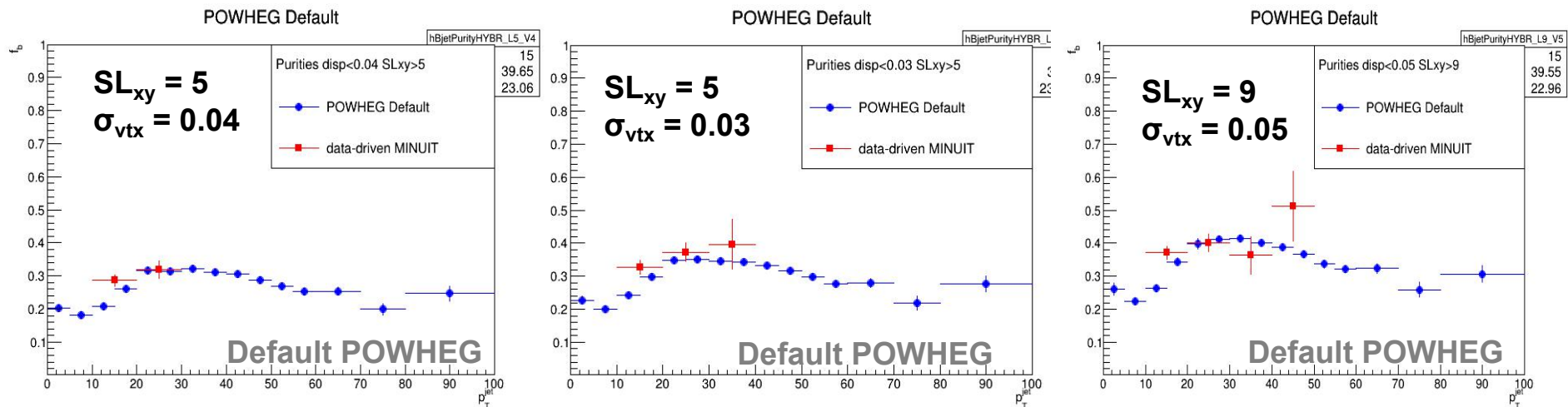
- Minuit convergence problems already from 30 GeV/c in some cases, above 40 in most cases
 - Note: merging the bins did not help
- RooFit different at low- p_T (and we trust it less than Minuit)
- *But:* At higher p_T -bins, RooFit and Minuit always match
 - Perhaps less effect of template errors because of wider distributions

purity - comparison of methods

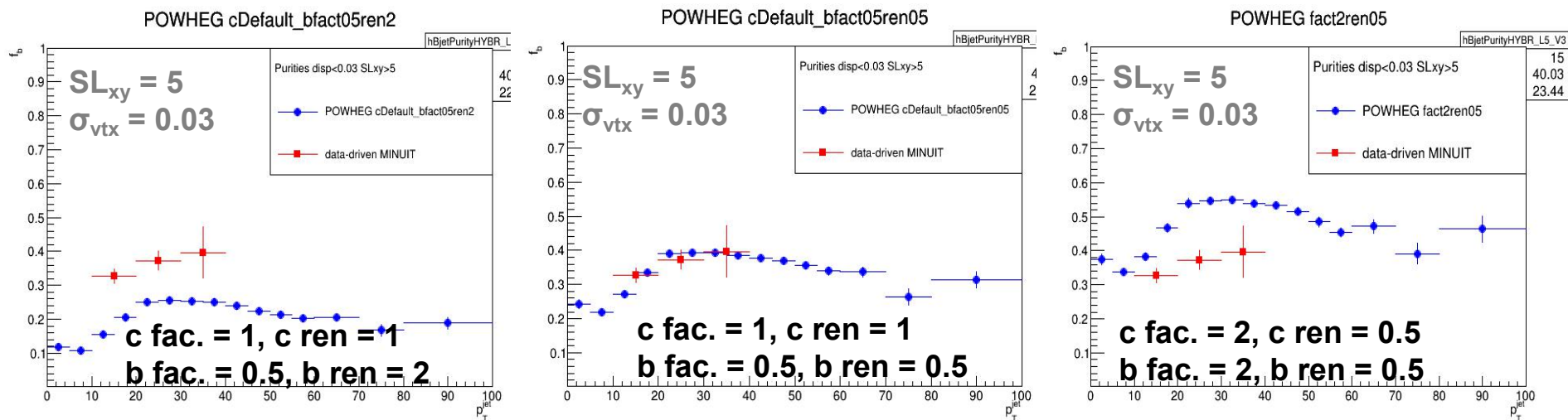


- Good news: very good consistency with the POWbc method
 - POWbc and data-driven MC template closure: good match
 - POWbc and data-driven with real data consistent within errors
- Strategy: data-driven constraints to be used to constrain purities from POWHEG

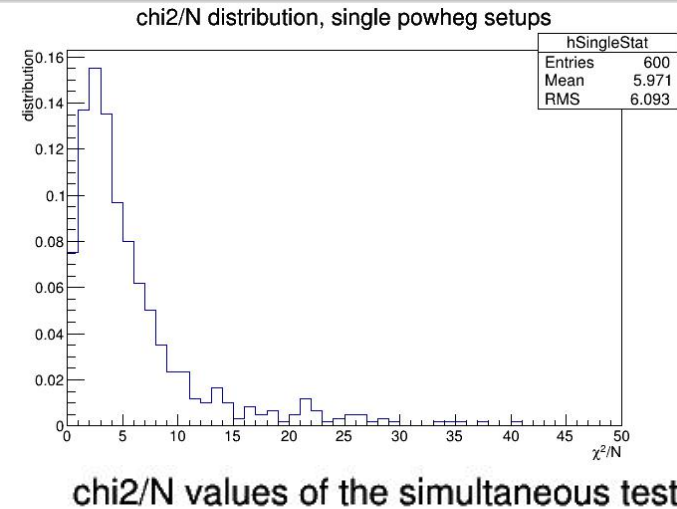
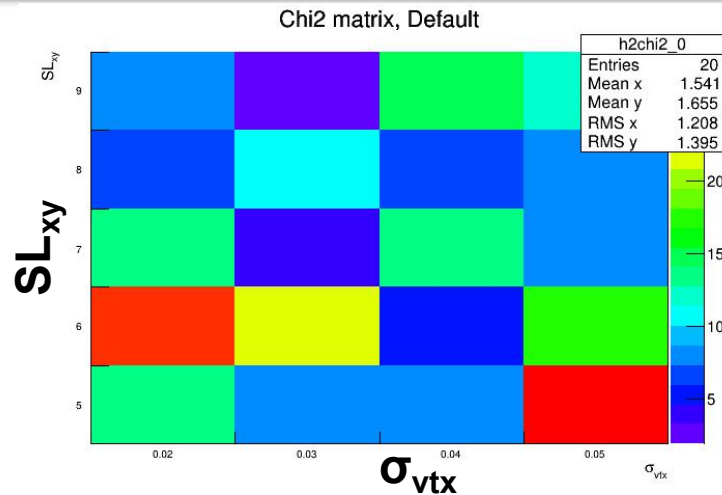
purity comparison examples



- Default POWHEG describes data regardless of tagging cut
- Scale variations cause big differences

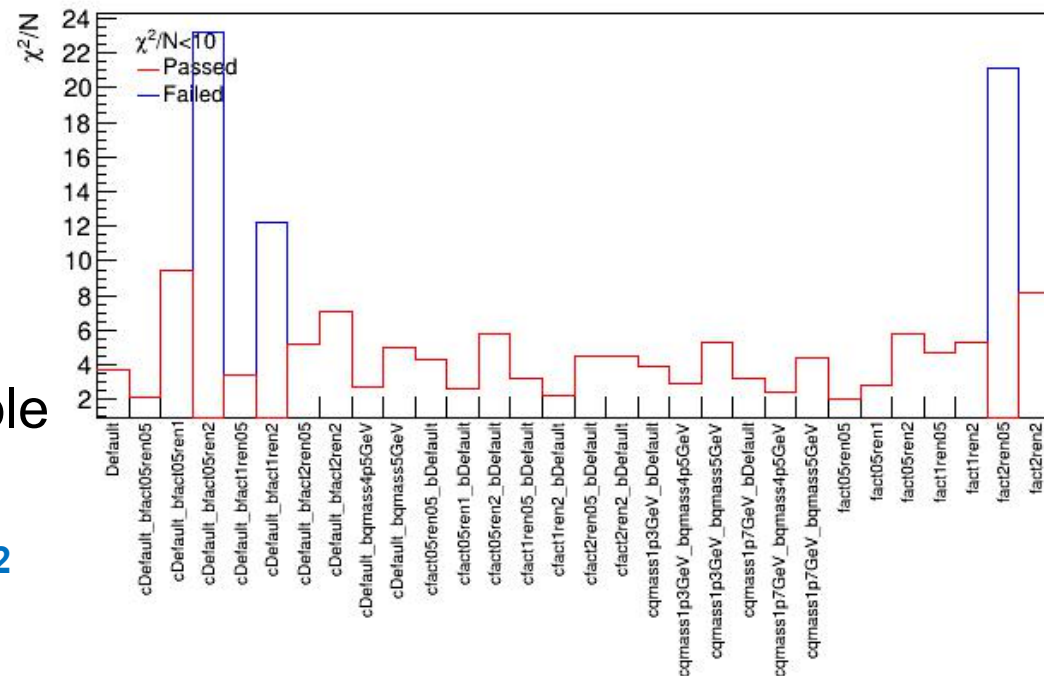


hybrid method: statistical exclusion



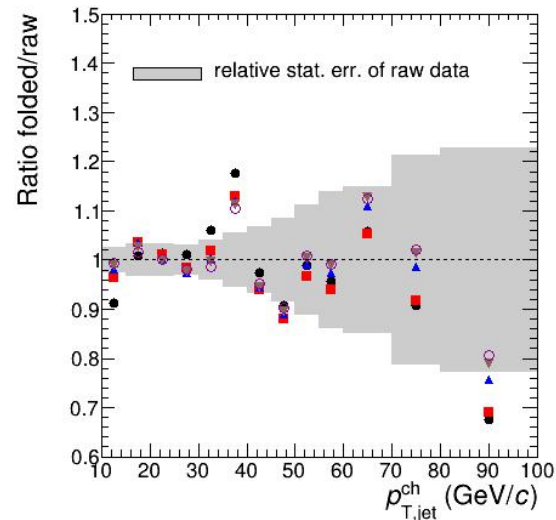
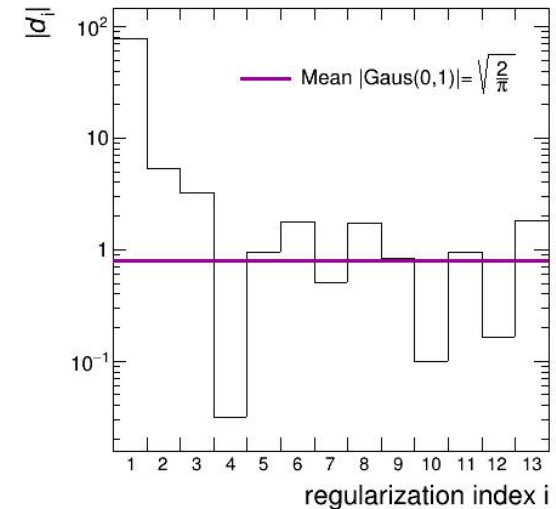
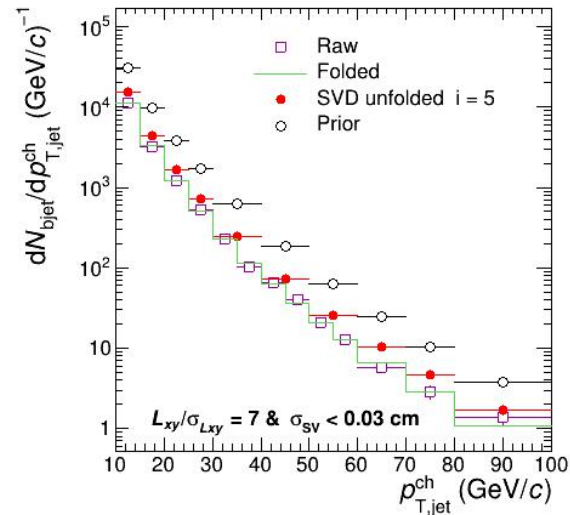
For each POWHEG setting:

- Compute χ^2 for each tagging cut
- sum them up
- Divide by sum of N_{points}
- Keep statistically acceptable settings only ($\chi^2/N < 10$)
 - c fac=1 c ren=1 b fac=0.5 b ren=2
 - c fac=1 c ren=1 b fac=1 b ren=2
 - c fac=2 c ren=2 b fac=2 b ren=2

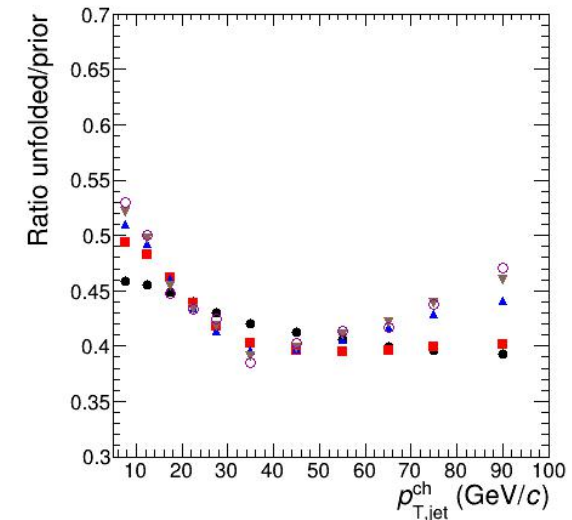


unfolding - SVD

- Generally good
 - Folded/raw ~ 1
 - Uncertainties below fluctuations
 - Convergent iteration (unfolded/prior)
 - $k\text{SVD} \geq 4$
- Some oscillation
 - $\sim 2\sigma$; plan to take care by rebinning or cropping the response matrix



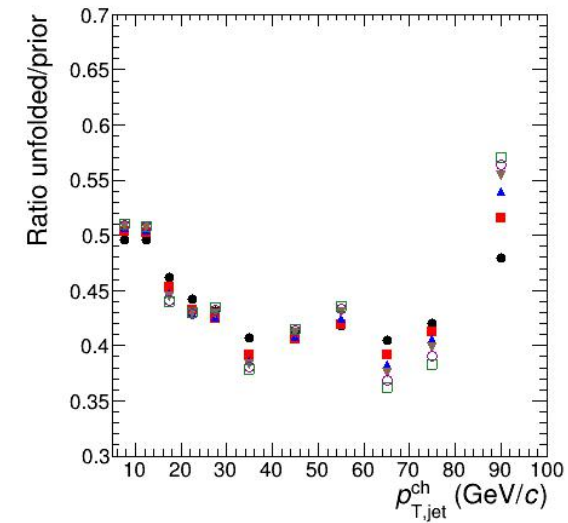
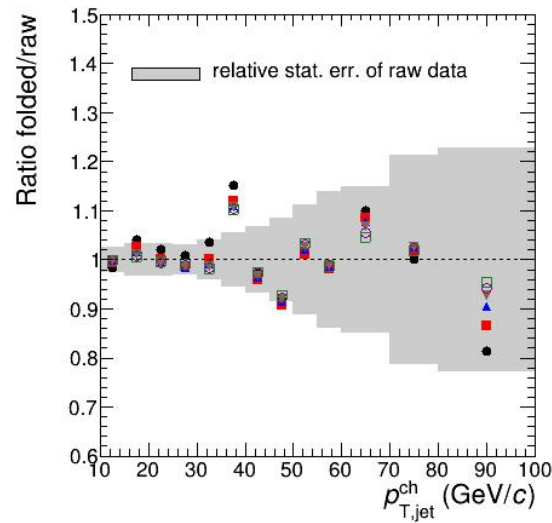
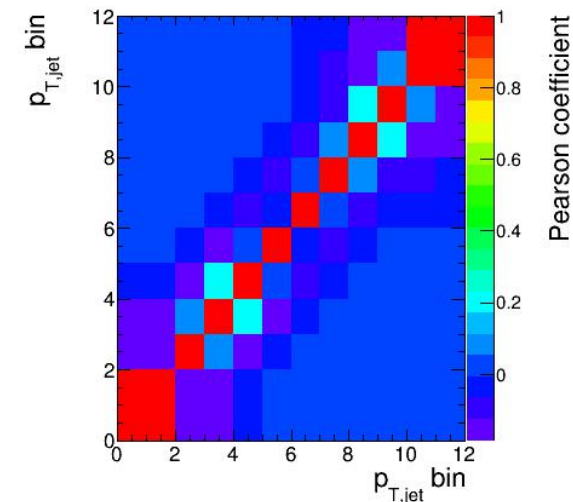
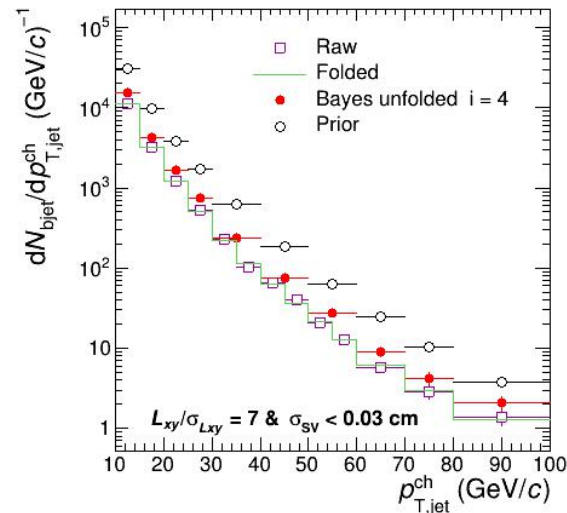
● $i = 2$ ■ $i = 3$ ▲ $i = 4$ ▼ $i = 5$ ○ $i = 6$



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unfolding - bayesian

- Generally good
 - Folded/raw ~ 1
 - Uncertainties below fluctuations
 - Convergent iteration
 - $k_{\text{Bayes}} \geq 4$
- Some oscillation
 - $\sim 2\sigma$; plan to take care by rebinning or cropping the response matrix

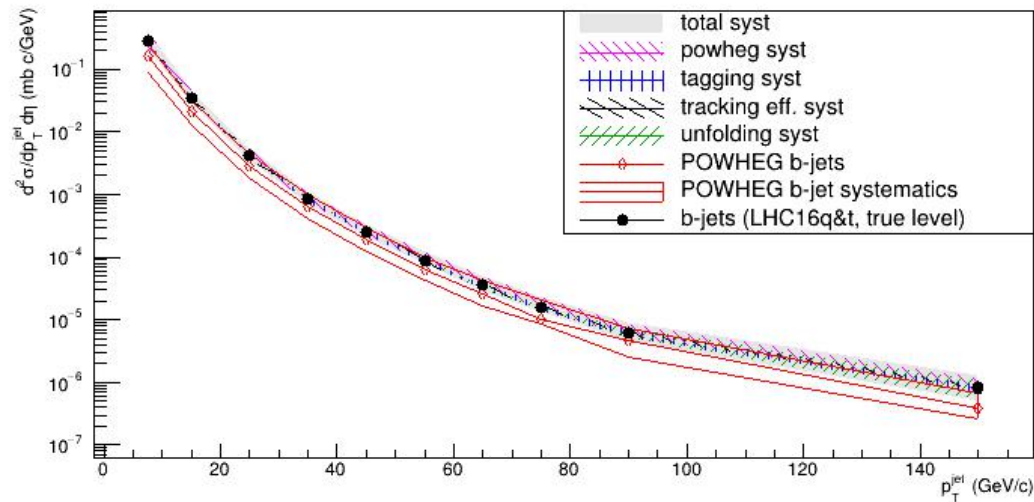


● $i=1$ ■ $i=2$ ▲ $i=3$ ▼ $i=4$ ○ $i=5$ □ $i=6$

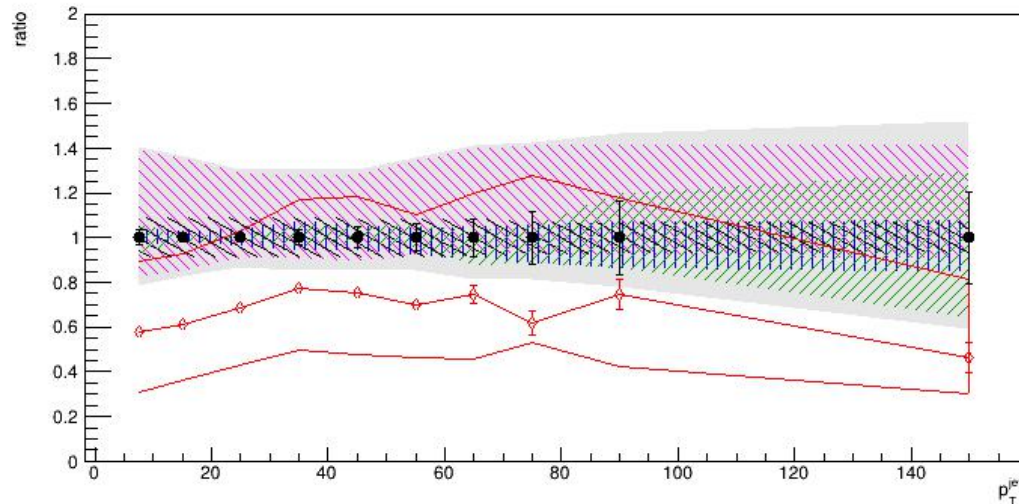
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spectrum with systematics

corrected b-jet spectrum, true level

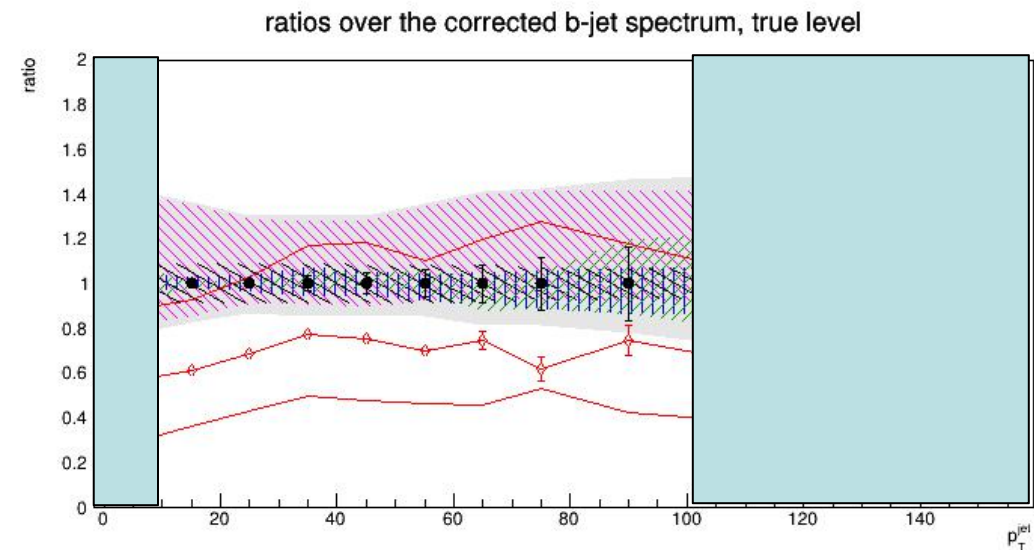
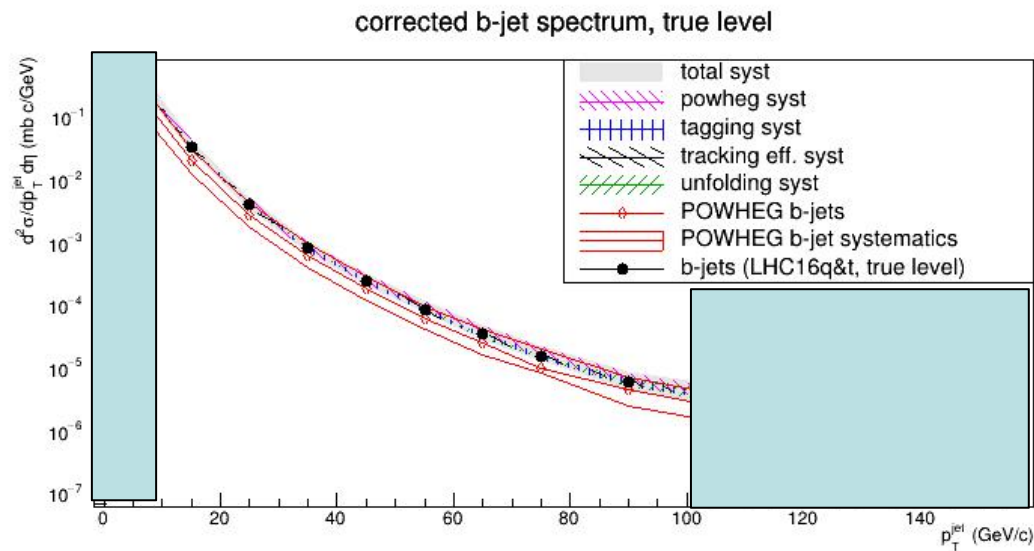


ratios over the corrected b-jet spectrum, true level



- Corrections with the Hybrid method
- Principal analysis: $SL_{xy} > 7$, $\sigma_{vtx} < 0.03$
- Dominant uncertainties:
 - hybrid purity
 - unfolding (including method, regularization and prior)
 - tracking
 - tagging
- Consistent with POWHEG within errors

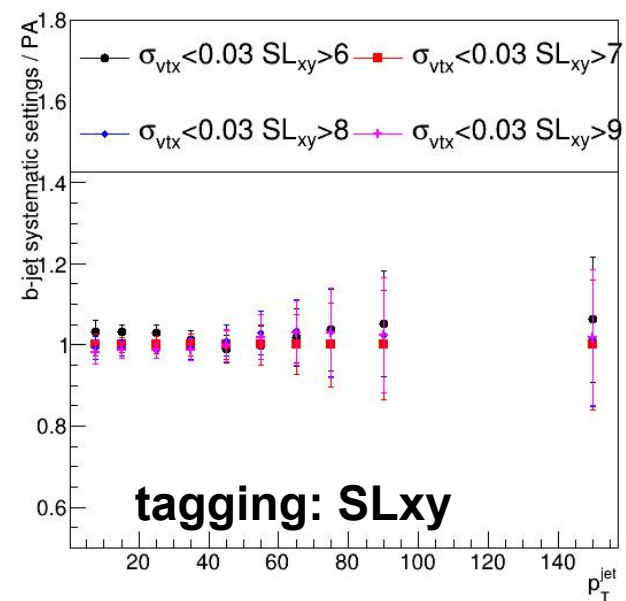
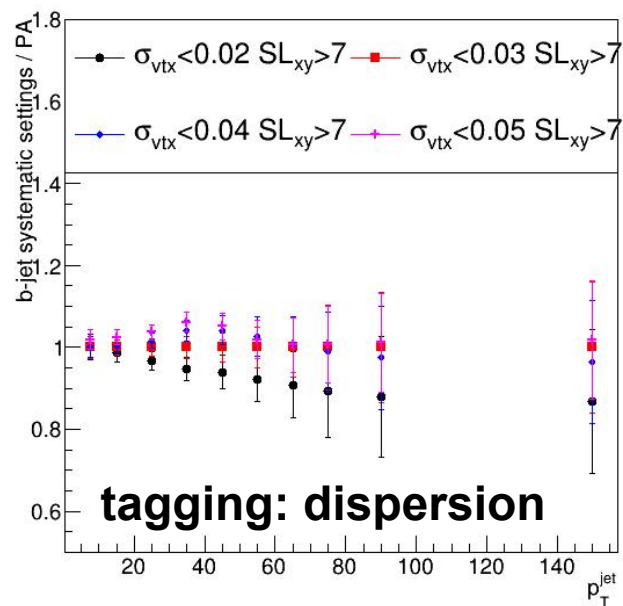
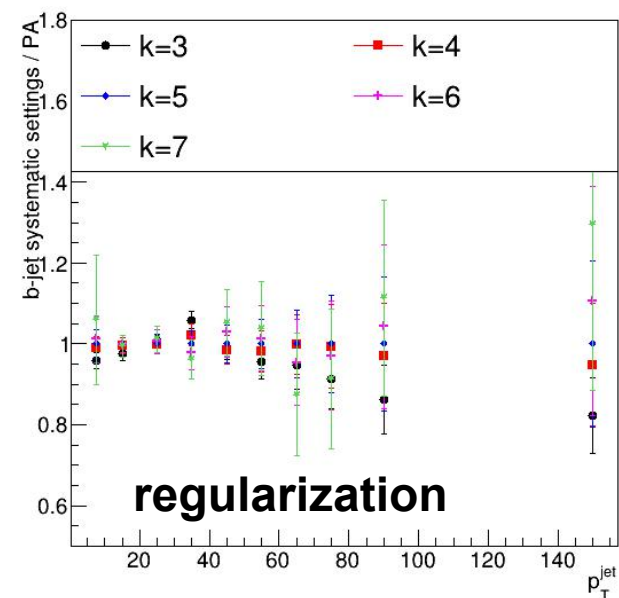
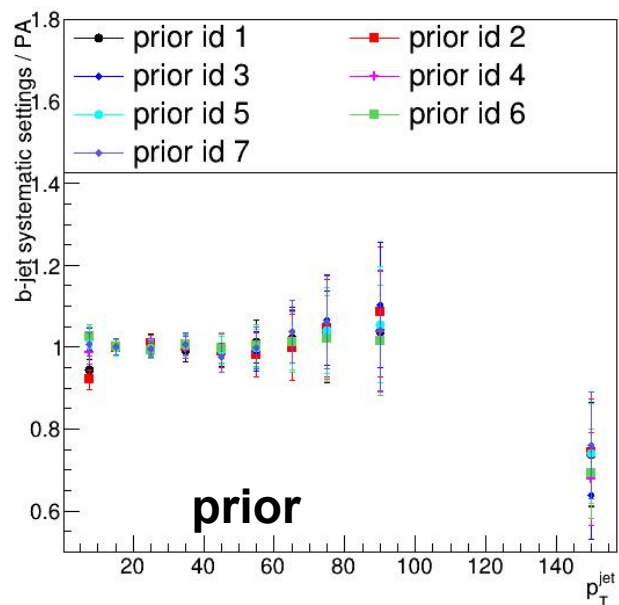
spectrum with systematics



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- **Range: 10-100 GeV/c**

some systematics (visualization)

Ratios of systematic variations compared to principal analysis



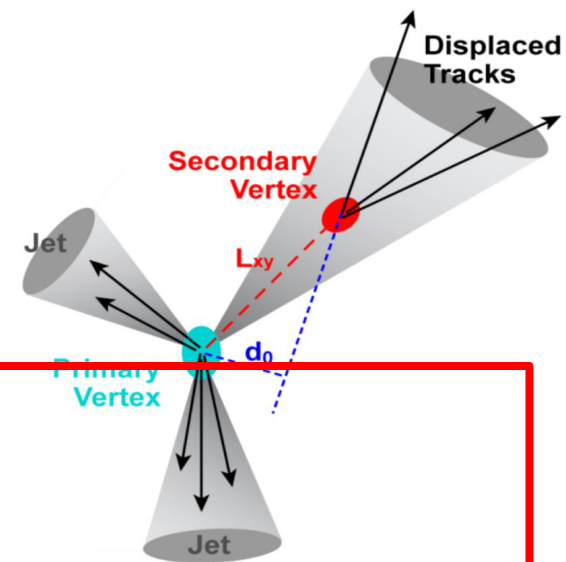
Status Summary

- We computed the 5 TeV pPb b-jet spectrum
 - New hybrid purities and efficiencies
 - Corrections are consistent
 - Most of the systematics are at hand
 - Detector matrix from EPOS+PYTHIA - the extent of background effect low p_T needs to be addressed
 - New unfolding method, slight oscillations - *crop matrix?*
 - Some minor (?) systematics needed:
 - unfolding: test with different binning
 - contamination of primary tracks by secondary tracks
 - track p_T smearing
- Next step: Preliminary for SQM

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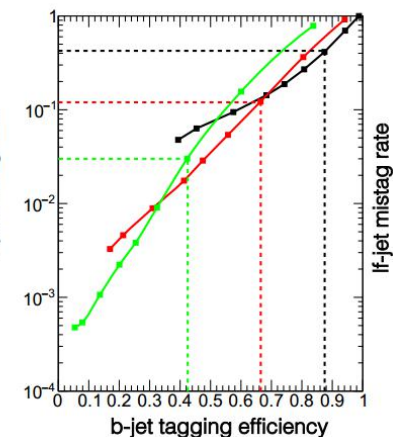
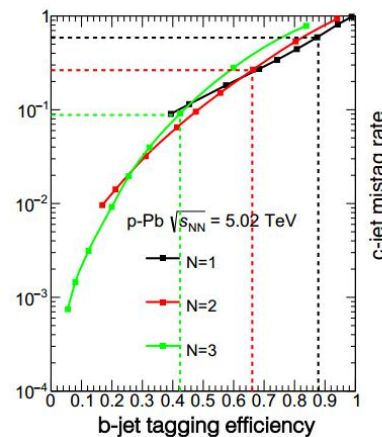
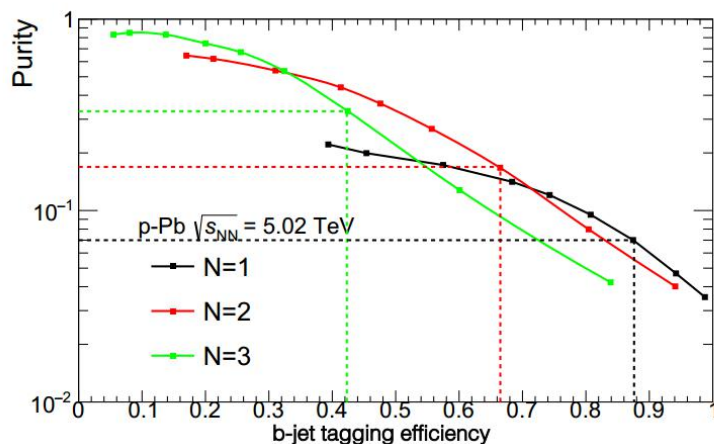
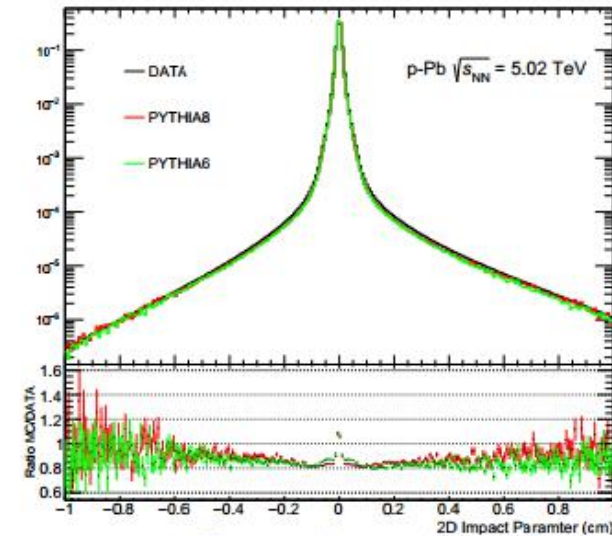
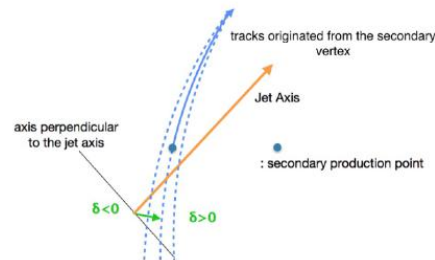
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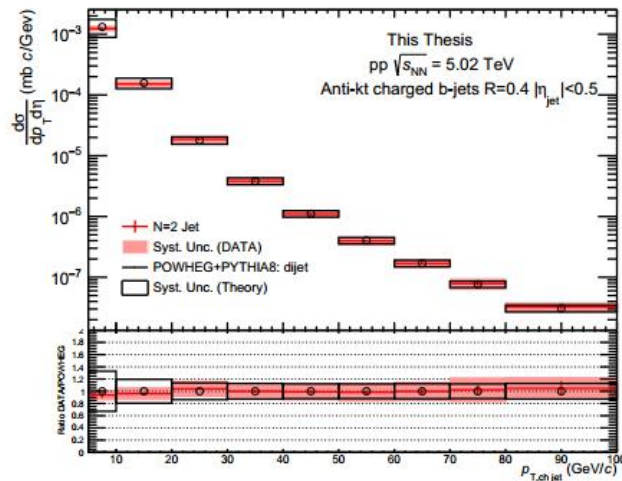
Impact parameter significance method

- Discriminator: $sd_{xy} = \delta \cdot d_{xy}$, where δ is the impact parameter sign: $sign(d_{xy}^{\vec{}} \cdot p_{jet}^{\vec{}})$.
- Track counting**
Based on #tracks fulfilling threshold
 - N=1 : high-efficiency ; N=2 ; N=3 : high purity
- Efficiency/purity curves:

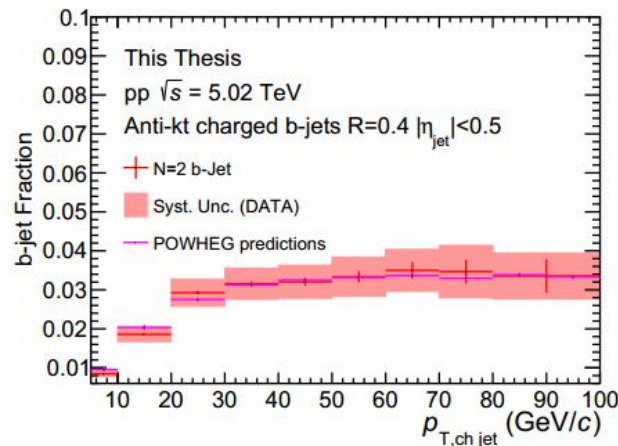
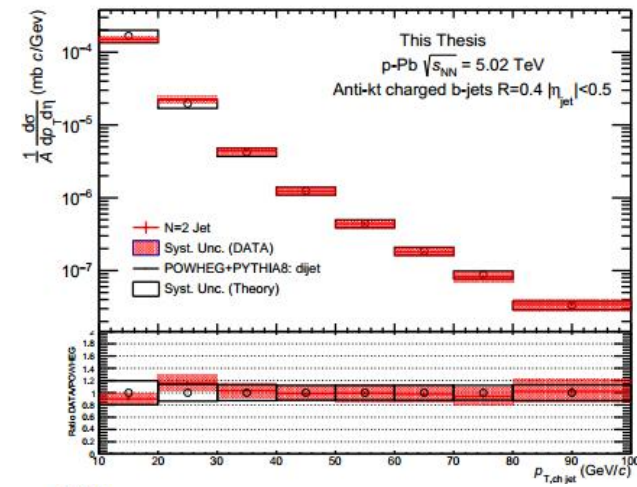


- Working point: $d_{xy} > 0.008$ threshold

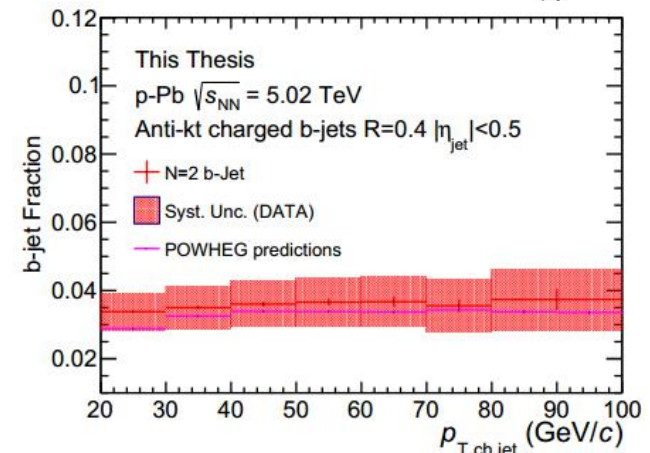
Production of b-jets (pp and pPb)



x-section

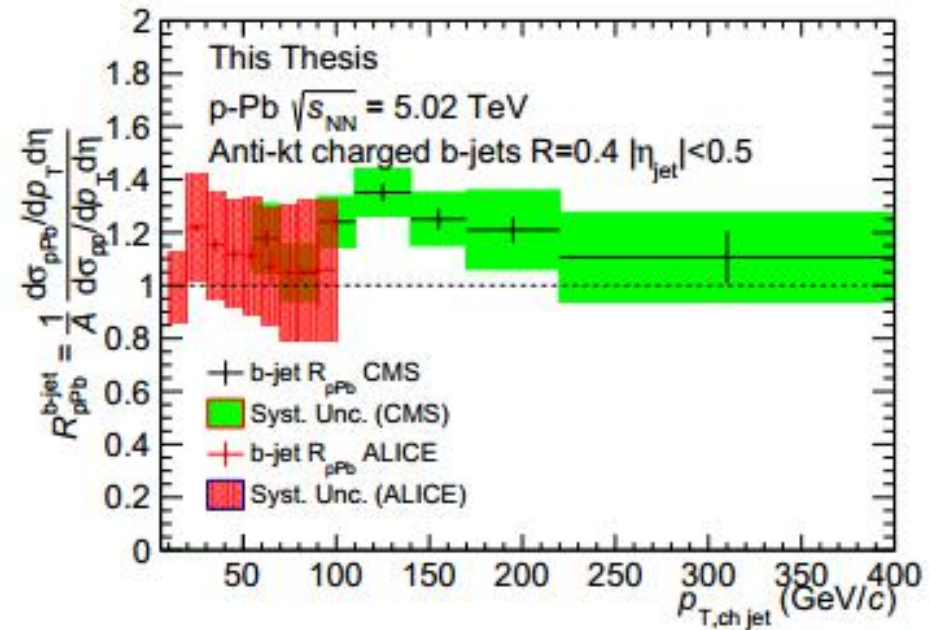
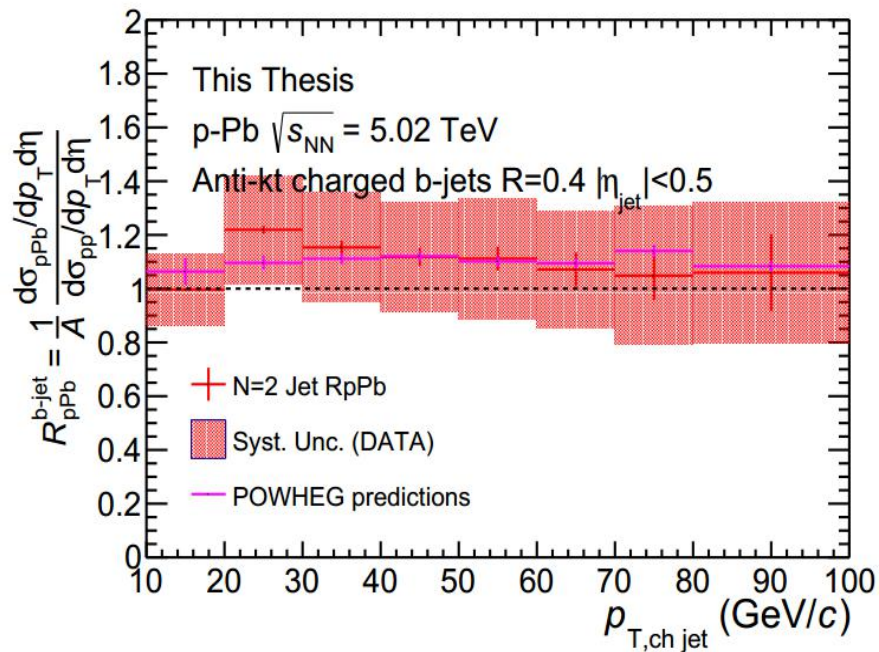


b-fraction



- Production is consistent within N=1, 2 (shown), 3
- Consistent with POWHEG w/ scale variation
- b-jet fraction drops at low- p_T in pPb

R_{pPb} of b-jets



- The R_{pPb} of b-jets is consistent with unity
- ...consistent with CMS measurements
- ...and with theory predictions within uncertainties
- The interaction of the b-jet with the cold nuclear matter has no effect on the b-jet within uncertainties.

Systematics and ToDo

Uncertainty source		p_T bins	
		10-20	20-30
p-Pb collisions	Unfolding algorithm	3.14%	
	Regularization parameter	2.03 %	
	Prior	2.09%	2.91%
	Unfolding range	1.40%	1.43%
	δp_T	0.12%	0.29%
	Tracking Efficiency	7.67%	10.60%
	Tagger working point	0.31%	0.24%
	V^0 rejection	0.20%	0.05%
	Normalization uncertainty	3.24%	
	Total	9.47%	11.7%
pp collisions	Unfolding algorithm	3.23%	
	Regularization parameter	3.3%	
	Prior	1.19%	0.19%
	Unfolding range	0%	
	Tracking Efficiency	9.3%	10.6%
	Tagger working point	0.13%	0.36%
	Normalization uncertainty	2.29%	
	Total	10.7%	11.7%

ToDo (Hadi)

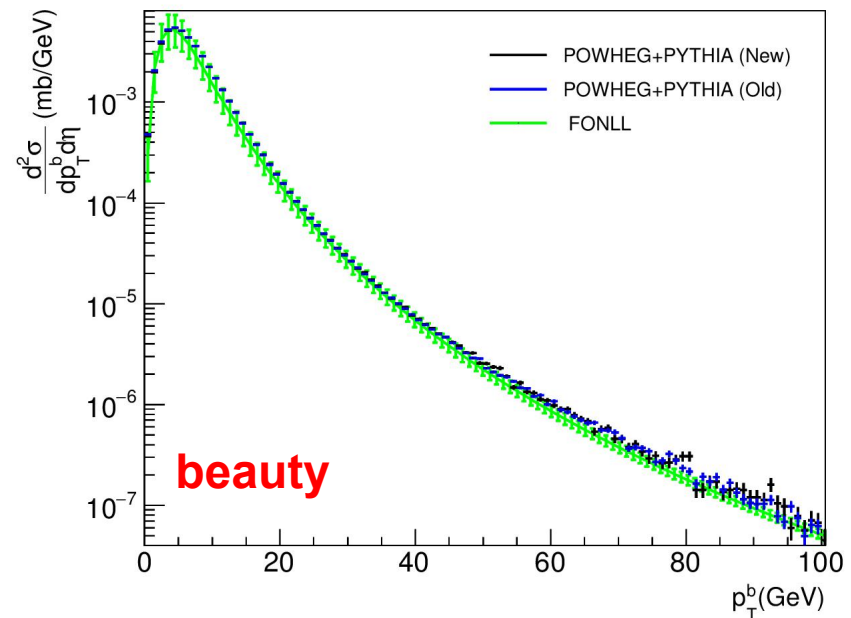
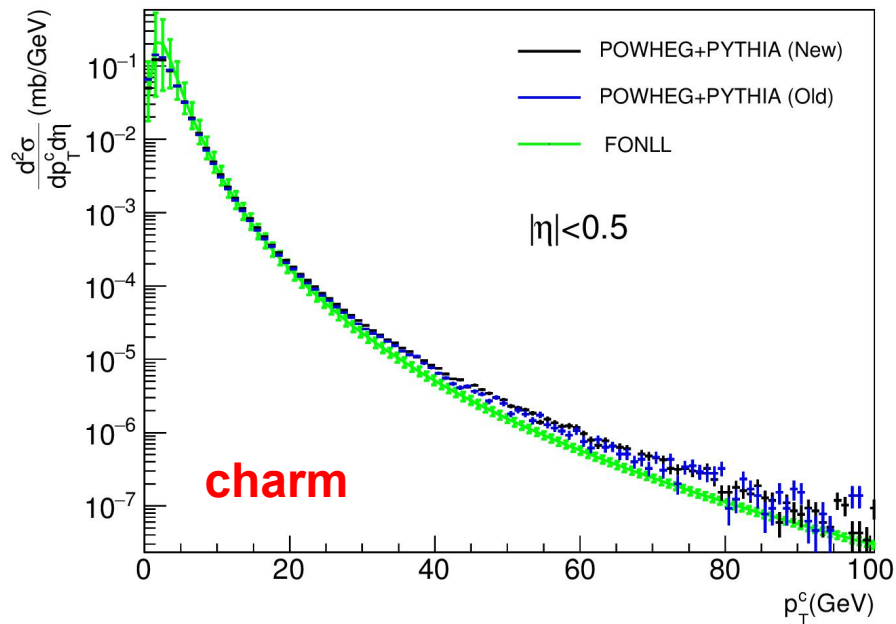
- Change the jet probability distribution, and use another discriminator for the templates used in the tagging efficiency determination.
- Use another distribution to fit the purity.
- Cancel the correlated uncertainties on the b-jet fraction and the R_{pPb}

15 March 1848
Hungarian
Revolution
against the
Habsburg rule



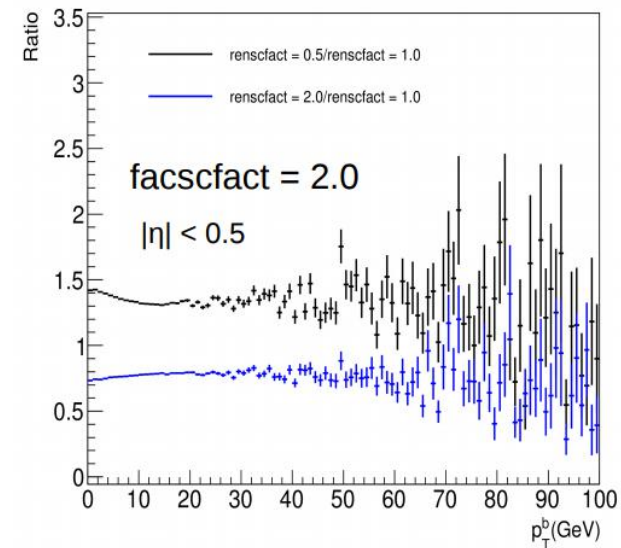
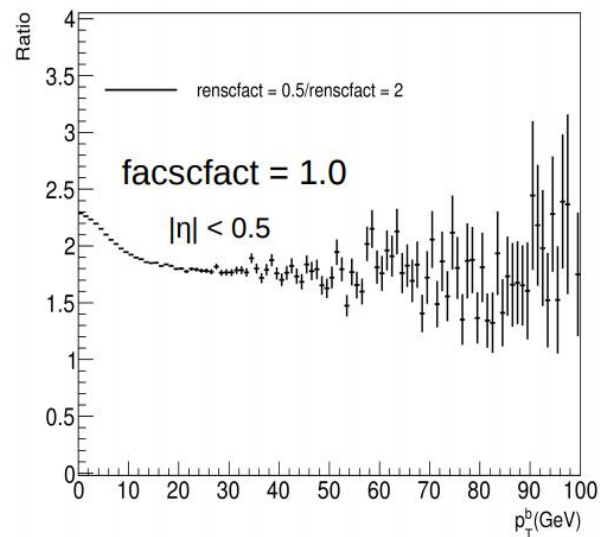
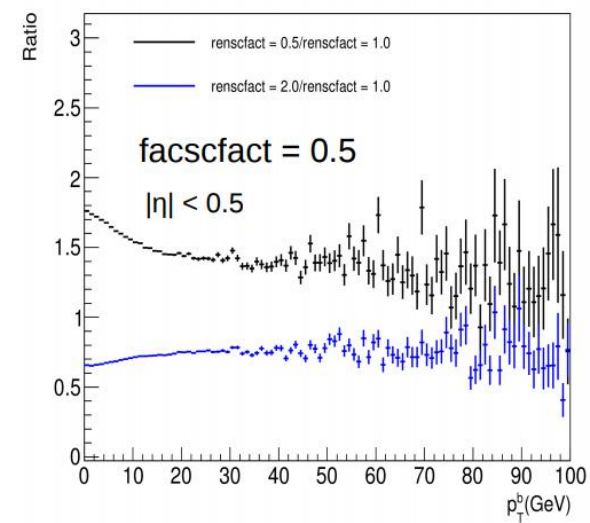
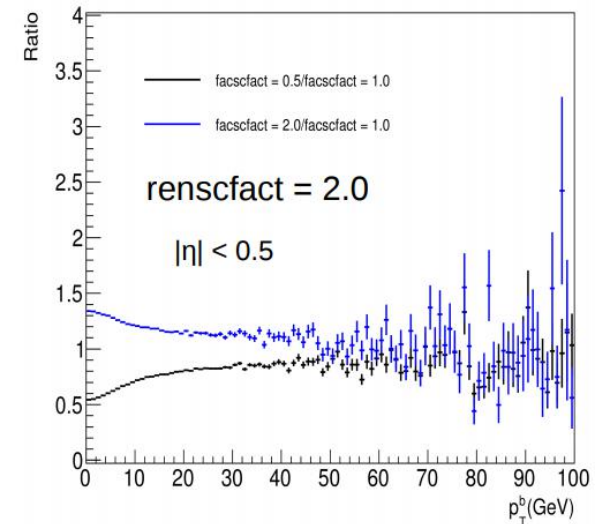
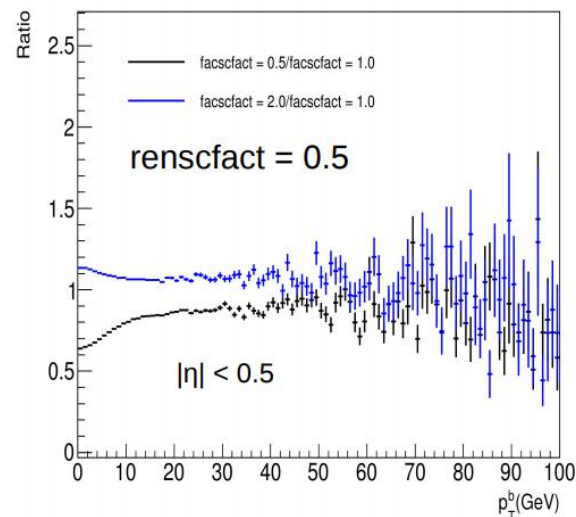
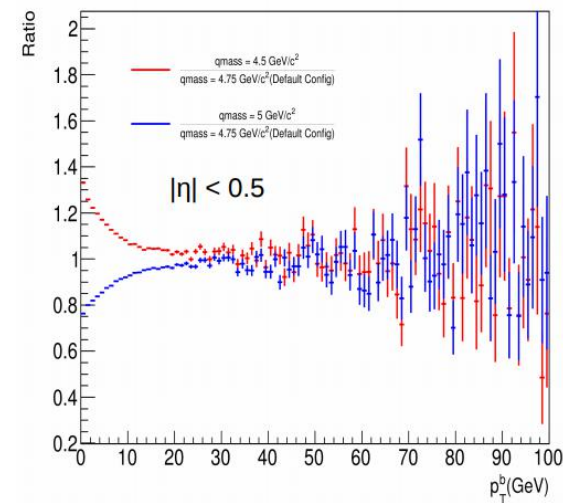
Hungarian Academy of Sciences
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count István Széchenyi

POWHEG simulations

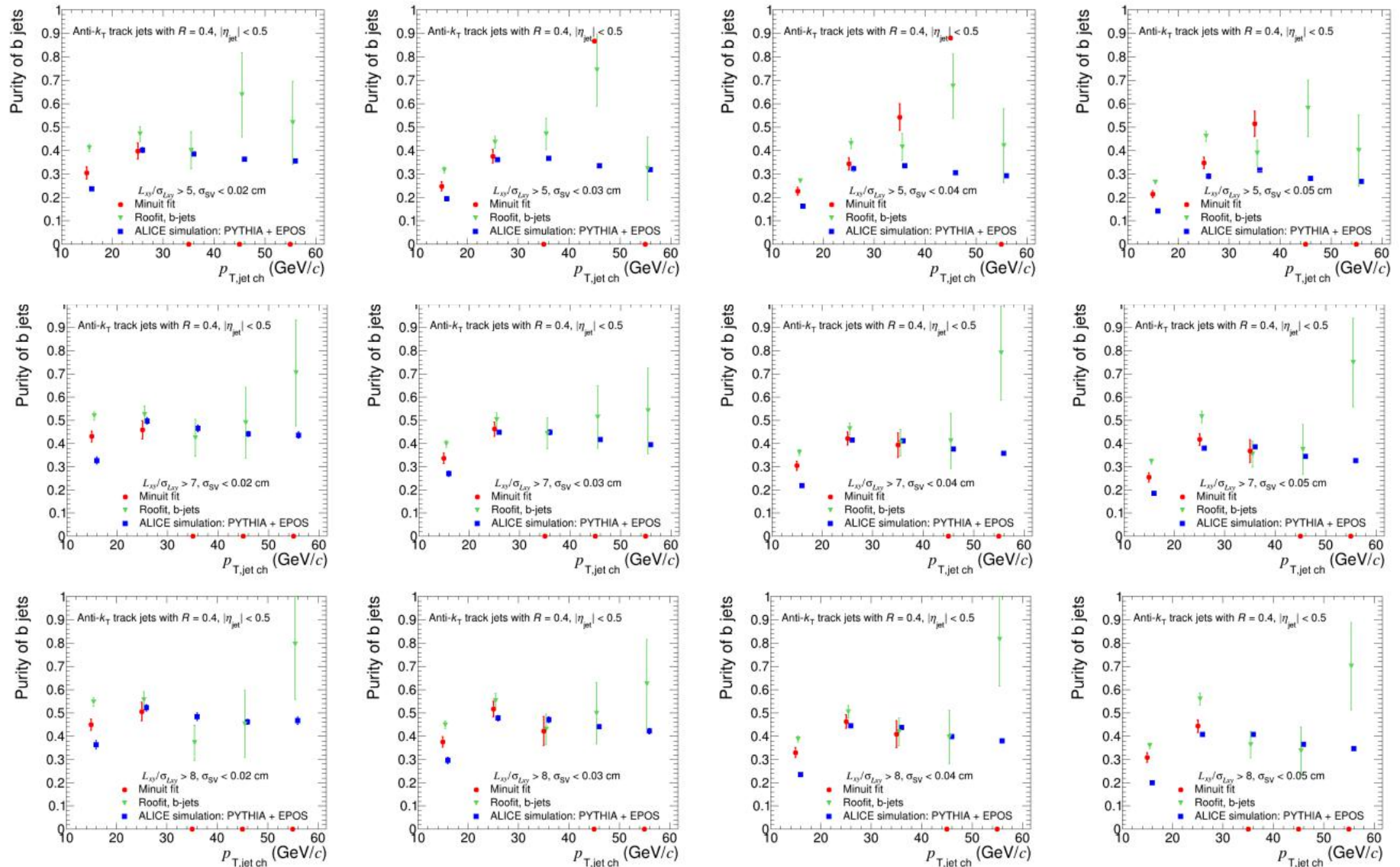


- Changes in "new" since the "old" one:
 - acceptance $|\eta| < 0.5$ instead of $|\eta| < 0.6$ scaled by 1/1.2
 - more suitable 1-GeV/c binning
 - Lorentz-boost applied
 - p-Pb nPDF applied
- No significant difference between "old" and "new"
- Marginal match to FONLL

POWHEG systematics



Data-driven fits to real data



Unfolding closure test - inclusive

Unfolding of inclusive spectrum of jets filtered PYTHIA

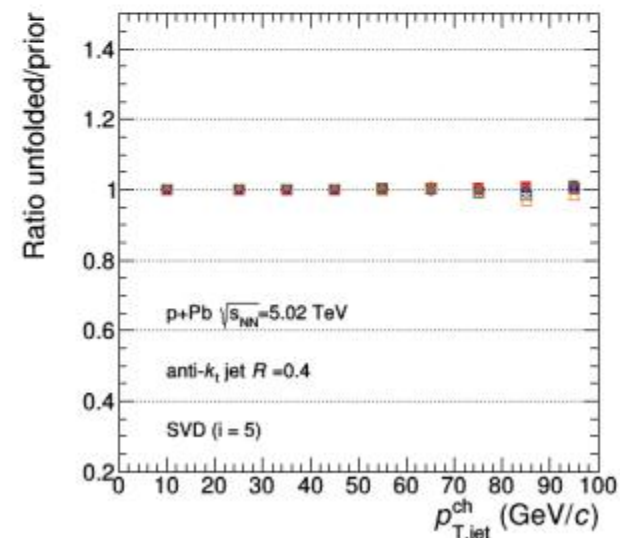
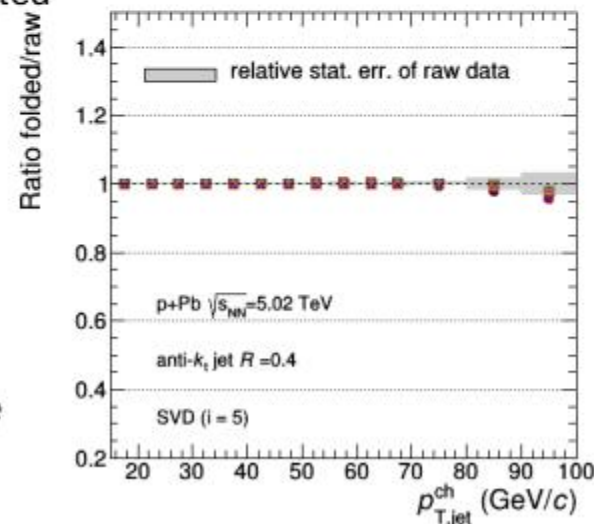
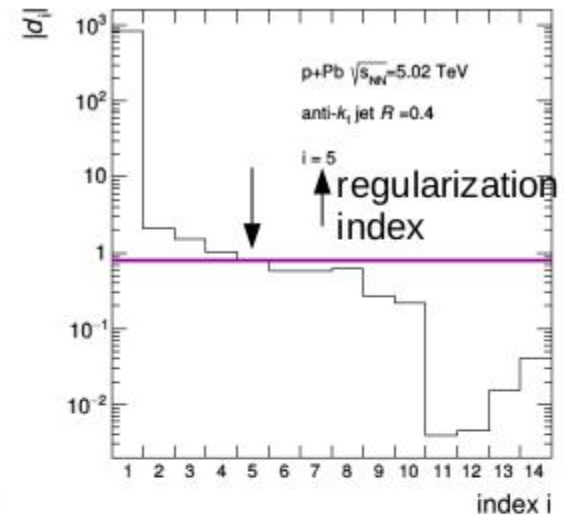
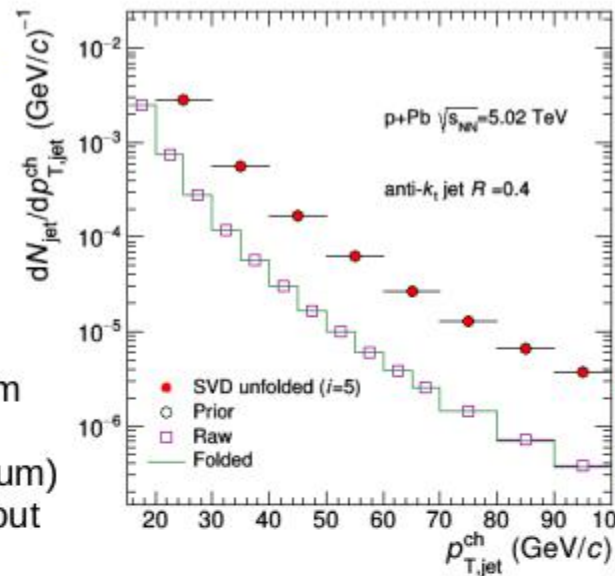
response matrix was filled with inclusive jet

normalization spectrum for response matrix (=here the true spectrum) are inclusive jets without the requirement that the jet was reconstructed

measured spectrum is projection of the response matrix on X axis

prior spectrum = true spectrum

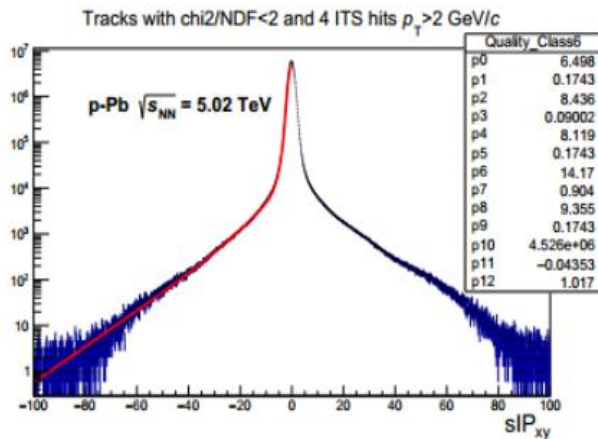
in this most ideal case everything works



● i=2 ■ i=3 ▲ i=4 ▼ i=5 ○ i=6 □ i=7 △ i=8

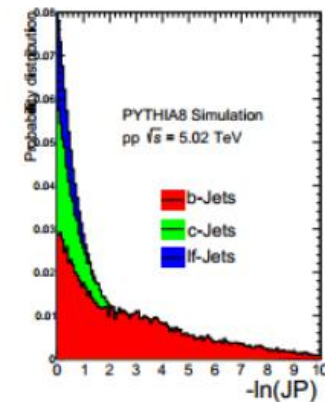
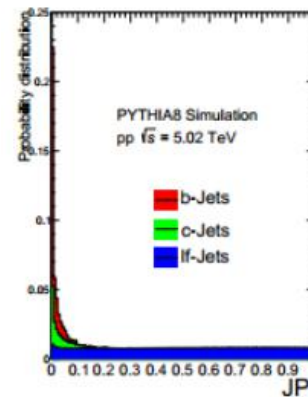
● i=2 ■ i=3 ▲ i=4 ▼ i=5 ○ i=6 □ i=7 △ i=8

Jet probability algorithm



- Calculate the track probability (P_{tr})

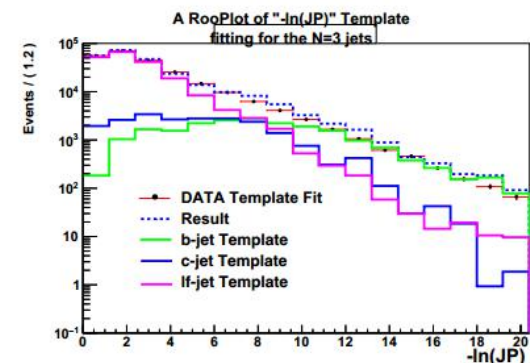
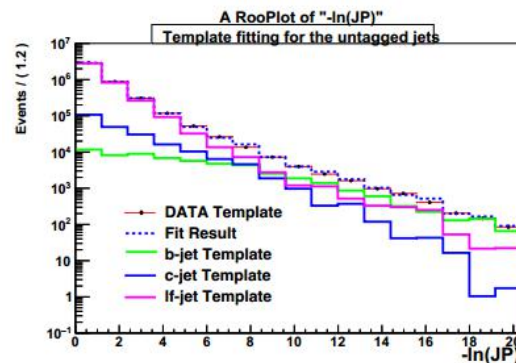
$$P_{tr}(S_{d_0}) = \frac{\int_{-\infty}^{-|S_{d_0}|} R(S) dS}{\int_{-\infty}^0 R(S) dS}$$



- Calculate the jet probability:

$$JP = \prod \times \sum_{k=0}^{N_{trk}-1} \frac{(-\log \prod)^k}{k!} \quad \text{where}$$

$$\prod = \prod_{i=1}^{N_{trk}} P_{tr}$$



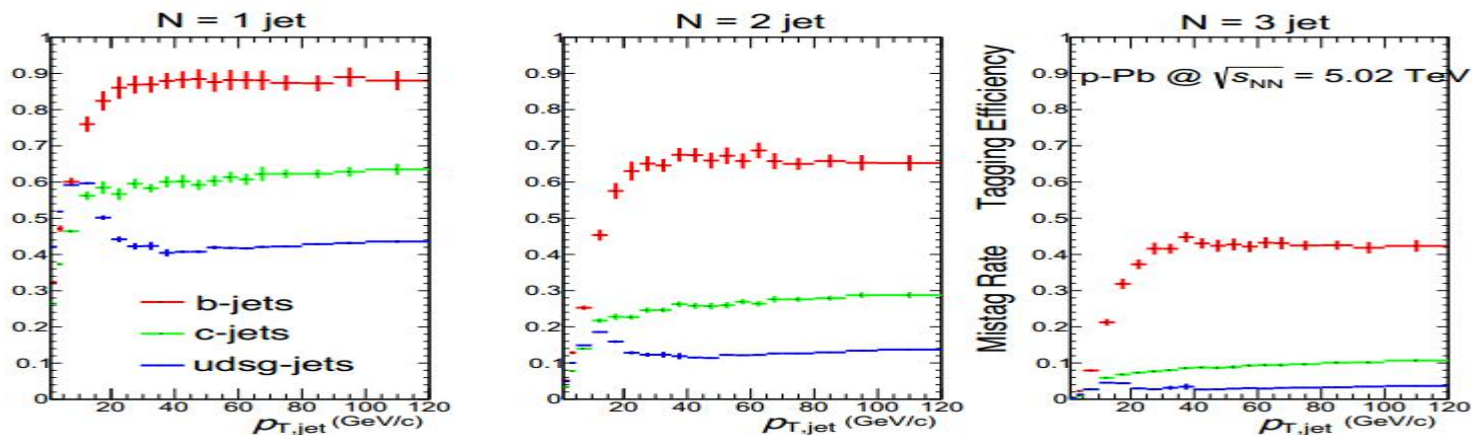
- Construct the $-\ln(JP)$ templates for the b,c, lf-jet from MC
- Apply the track counting tagger.
- Fit to data for tagged and untagged $-\ln(JP)$ distributions.

- Determine efficiency by: $\epsilon_b = \frac{f_b^{\text{tag}} \cdot N_{\text{data}}^{\text{tag}}}{f_b^{\text{untag}} \cdot N_{\text{data}}^{\text{untag}}} C_b$

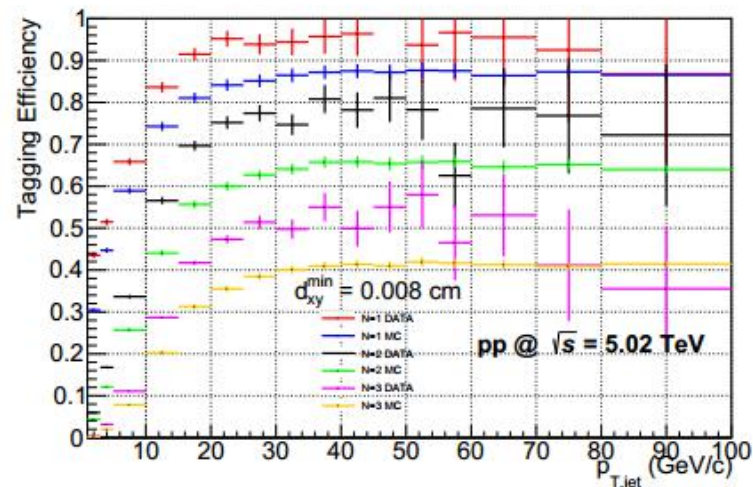
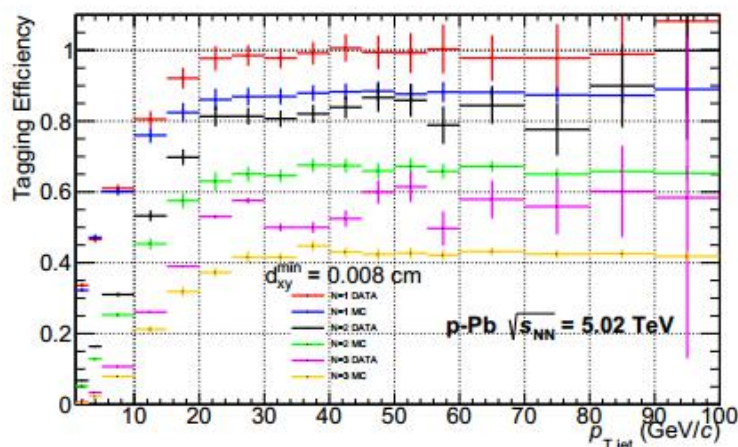
Efficiencies

MC efficiency

$$\epsilon_b(p_{T, \text{ch. jet}}^{\text{det}}) = \frac{N_b^{\text{tagged}}(p_{T, \text{ch. jet}}^{\text{det}})}{N_b^{\text{Total}}(p_{T, \text{ch. jet}}^{\text{det}})}$$

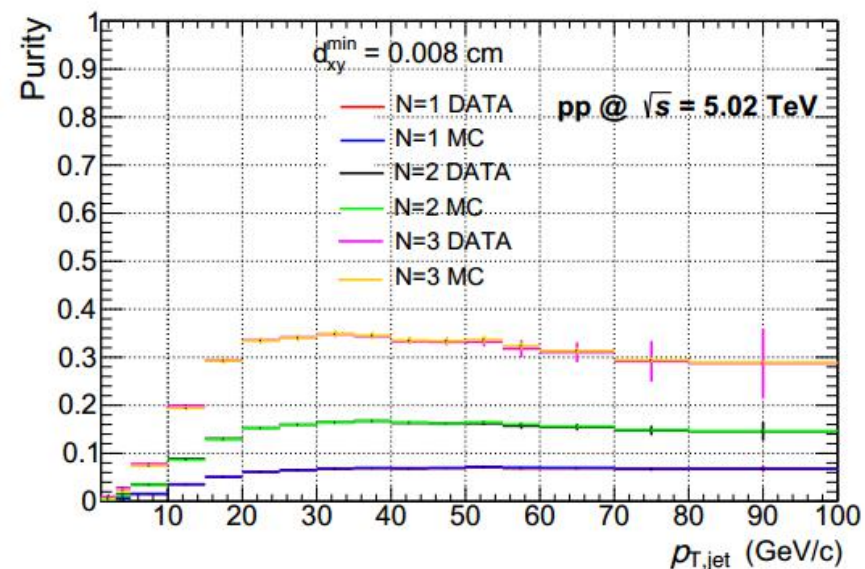
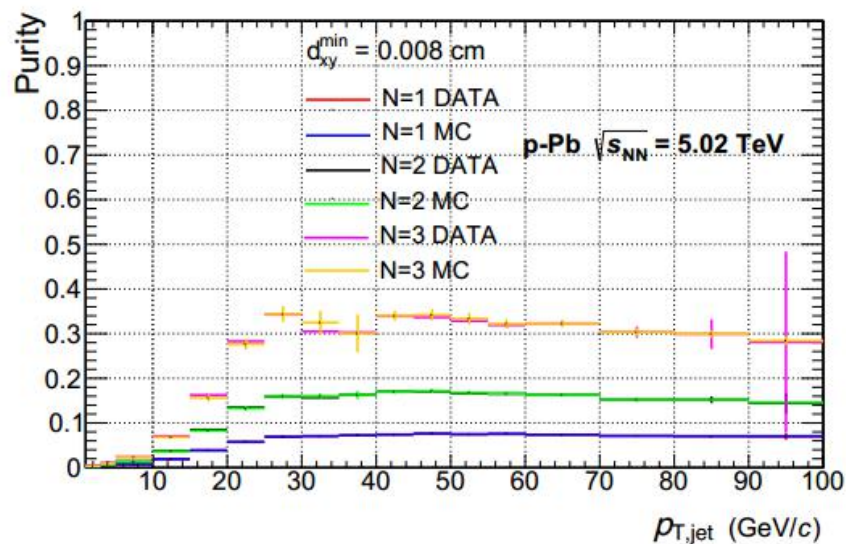


Data-driven efficiency underestimated by MC!



b-jet purity

- b-jet purity: $P = \frac{N_{b\text{-jet}}^{\text{tagged}}}{N_{\text{total}}^{\text{tagged}}}$.
- $N_{b\text{-jet}}^{\text{tagged}}$ can be estimated using data-driven methods:
 - by fitting templates for the JP.



- b-jet tagging purity is consistent between data and MC.

Underlying event

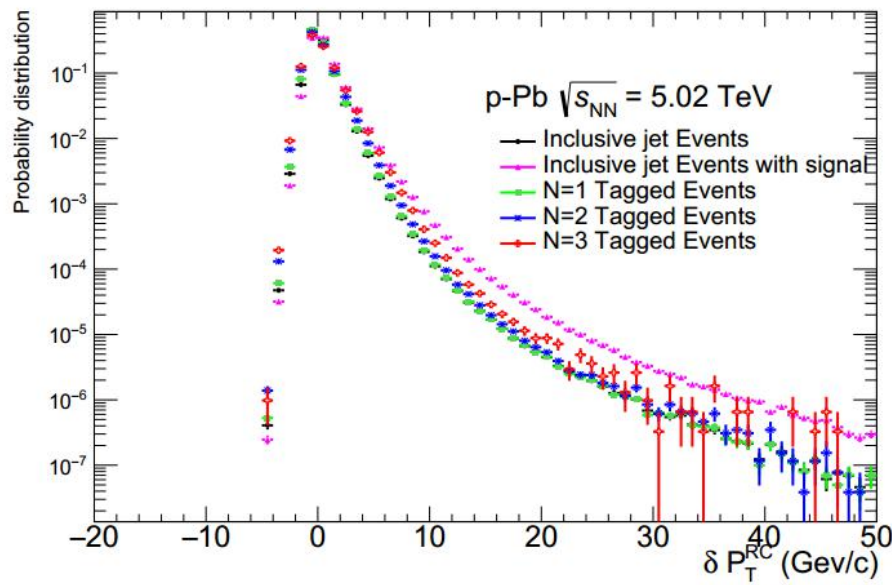
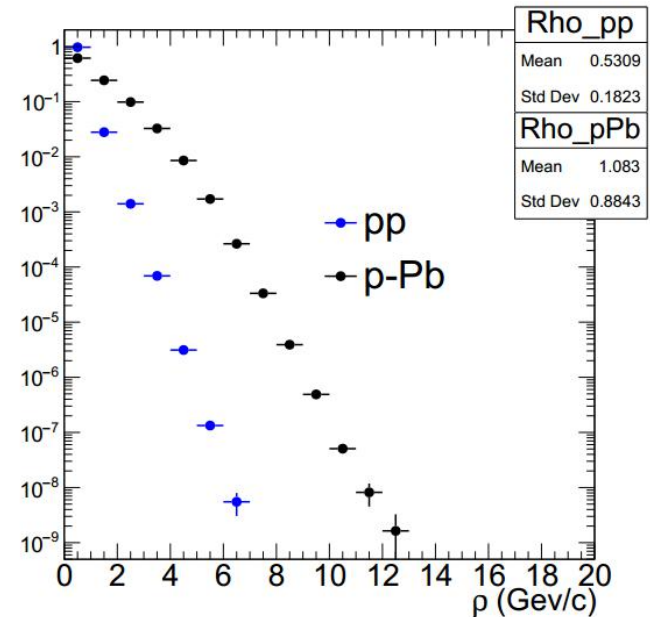
- UE density:

$$\rho = \text{median}\left(\frac{p_{T,i}}{A_i}\right) \cdot C, \text{ where}$$

$$C = \frac{\text{CoveredArea}}{\text{TotalArea}}.$$

- Correct the jet p_T :

$$p_{T,j}^{\text{Sub}} = p_{T,j} - \rho A_j.$$



- UE fluctuation for unfolding
- Random cone method

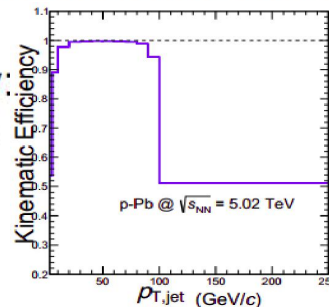
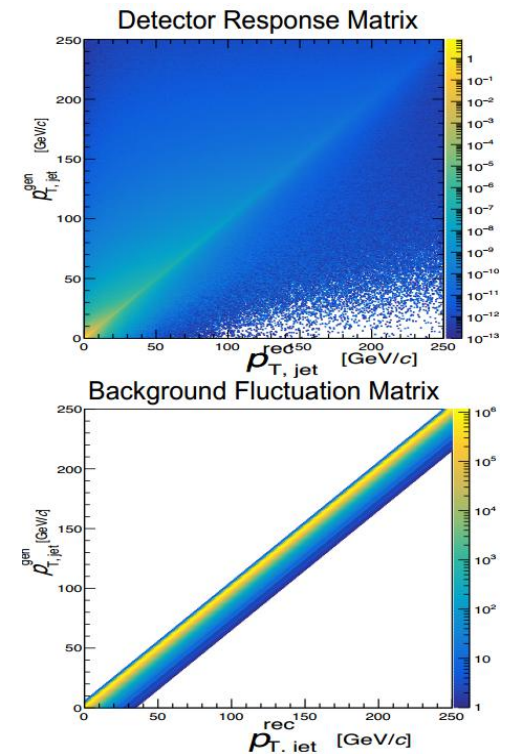
$$\delta p_T = p_T^{\text{RC}} - \rho \pi R^2$$

- If overlap with signal jet, throw again

Unfolding

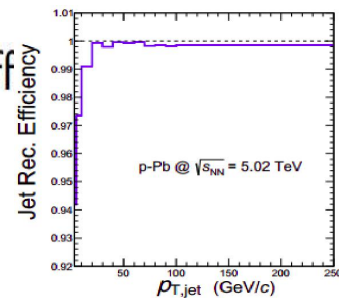
- Correction for detector effects.
⇒ Detector response (DR) matrix is needed.
- The DR is built by matching jets at the detector level to jet in the generated level:

$$\Delta R_{jet1, jet2} = \sqrt{(\eta_{jet1} - \eta_{jet2})^2 + (\phi_{jet1} + \phi_{jet2})^2} < 0.25$$
- Correction for UE fluctuations (for p–Pb collisions).
⇒ background fluctuation (F) matrix need,
- The F matrix built from the δp_T distribution.
- The SVD unfolding was used (A. Hoecker et al).
- Prior: PYTHIA b-jet spectrum (jet-jet MC).
 - Combine both matrices for p–Pb : $R = F \times DR$.
 - Closure test shows that the measured spectrum is correctly unfolded



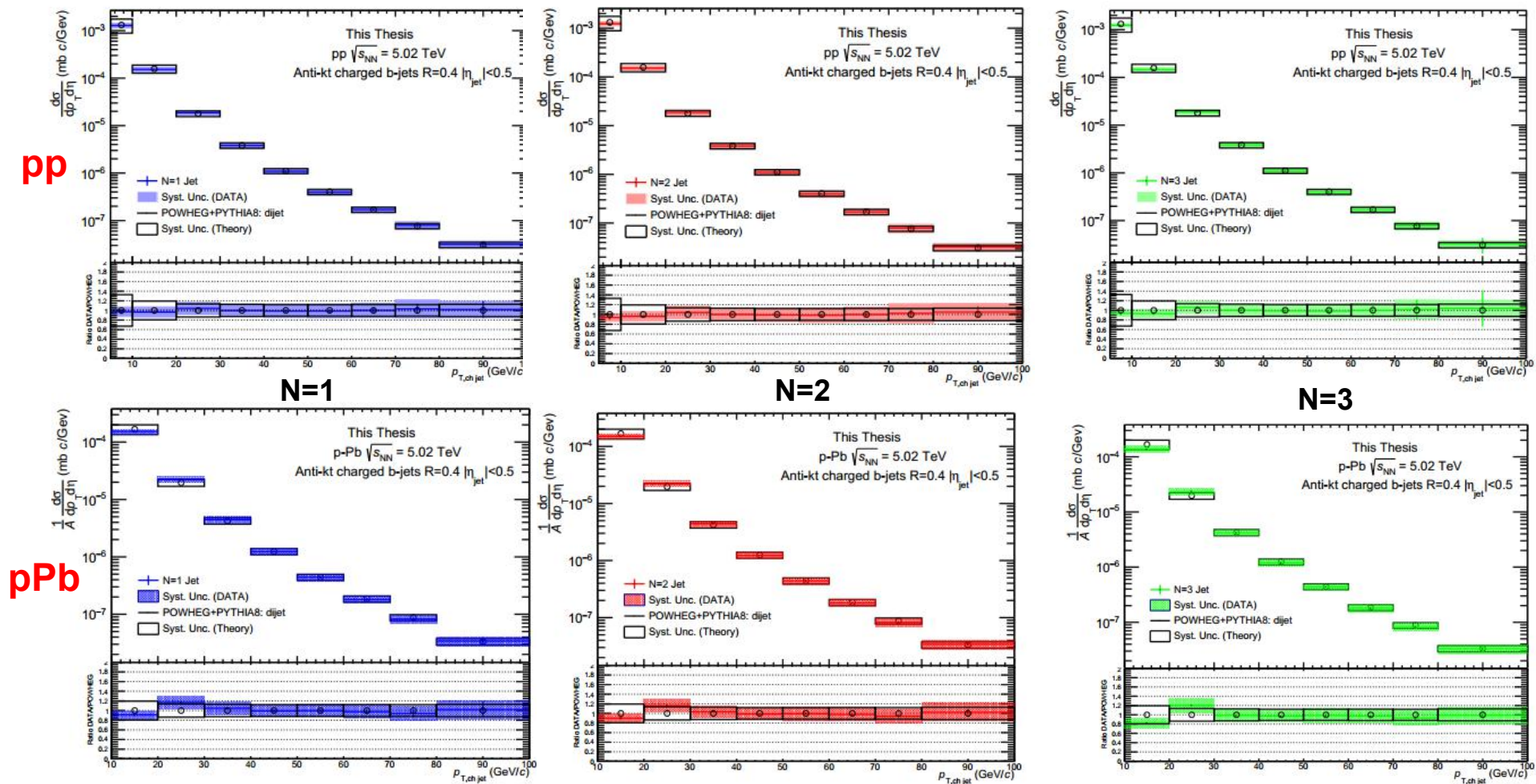
- b-jet reconstruction eff

$$\epsilon_{jet}^{rec} = \frac{N_{jet}^{matched}}{N_{jet}^{total}}$$



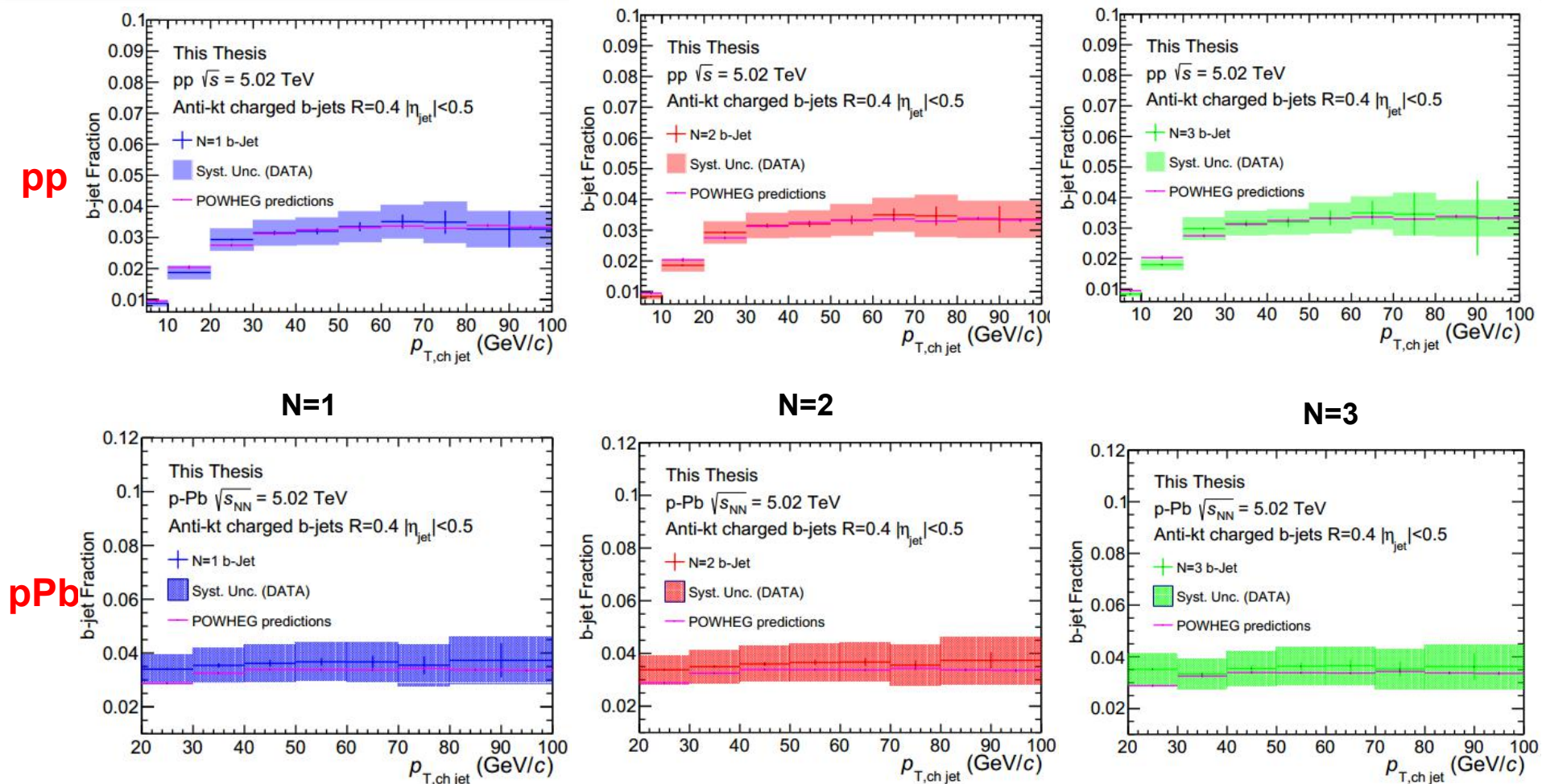
- Correct for kinematic efficiency: fraction of remaining jets after rebinning.

Production cross-section (pp and pPb)



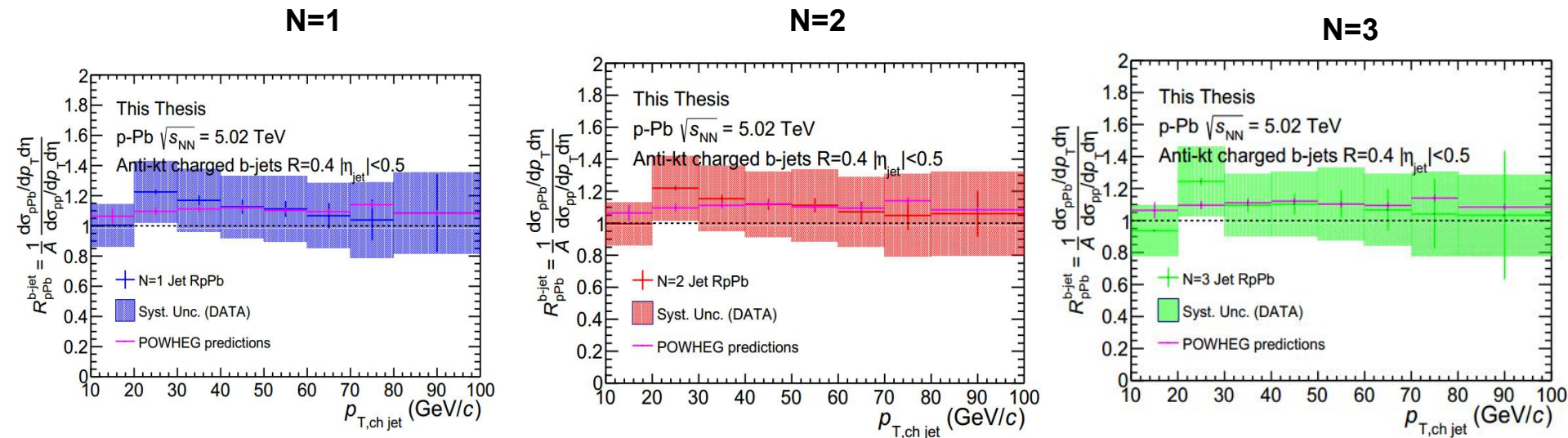
- Production is consistent within N=1,2,3
- Consistent with POWHEG w/ scale variation

b-jet fraction (pp and pPb)



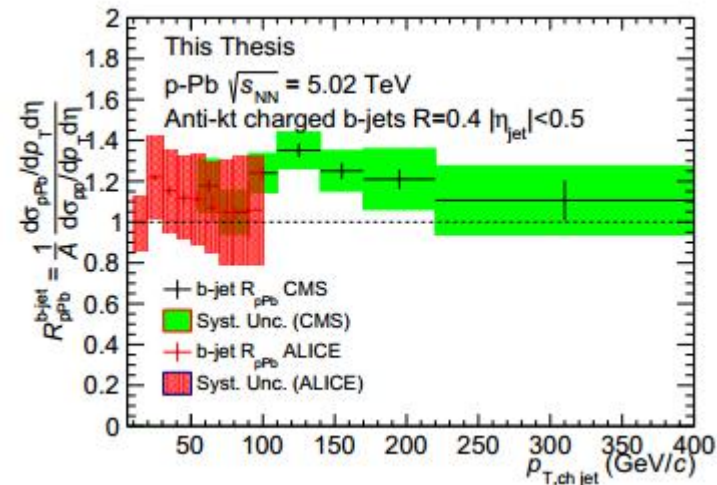
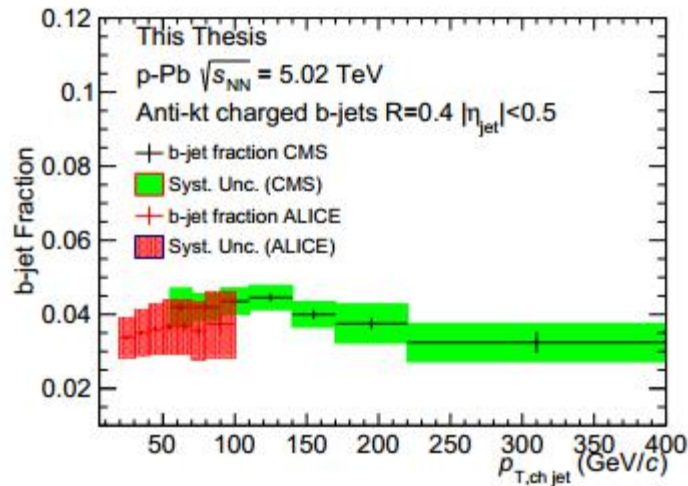
- Production is consistent within $N=1,2,3$ and POWHEG

R_{pPb} of b-jets



- The R_{pPb} of b-jets for $N=1,2,3$ is...
 - consistent with unity
 - and with theory predictions within uncertainties
- The interaction of the b-jet with the cold nuclear matter has no effect on the b-jet within uncertainties.

Comparison to CMS and ToDo



ToDo (Hadi)

- Change the jet probability distribution, and use another discriminator for the templates used in the tagging efficiency determination.
- Use another distribution to fit the purity.
- Cancel the correlated uncertainties on the b-jet fraction and the R_{pPb}