

A NEW LOOK INTO BACKGROUND SUBTRACTION FOR JET PHYSICS IN P+P AND A+A COLLISIONS

Alba Soto-Ontoso

[Yacine Mehtar-Tani, ASO, Marta Verweij arXiv:1904.12815]

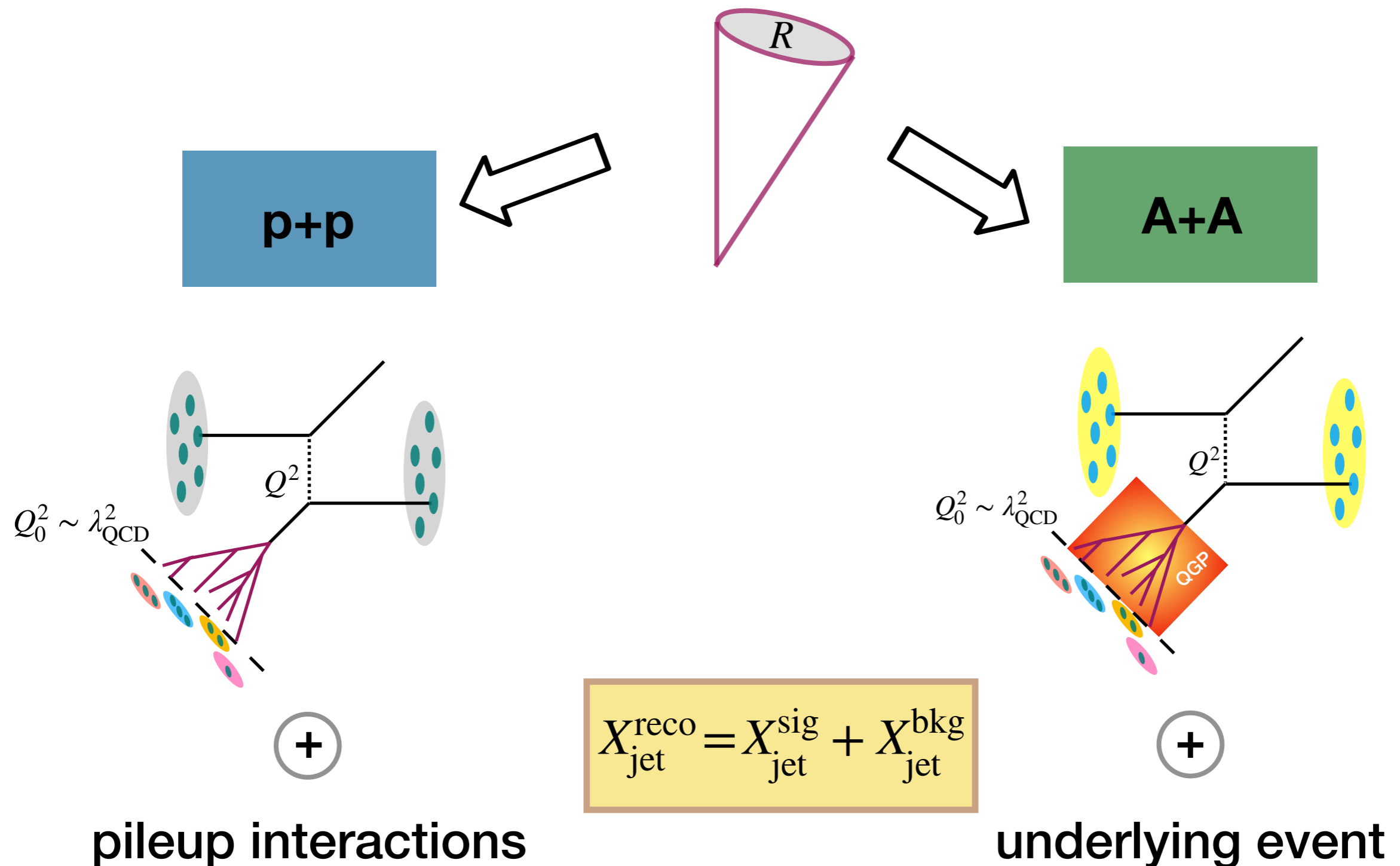
[Yacine Mehtar-Tani, ASO, Marta Verweij in preparation]

TH Heavy Ion Coffee

CERN, 18th September, 2019

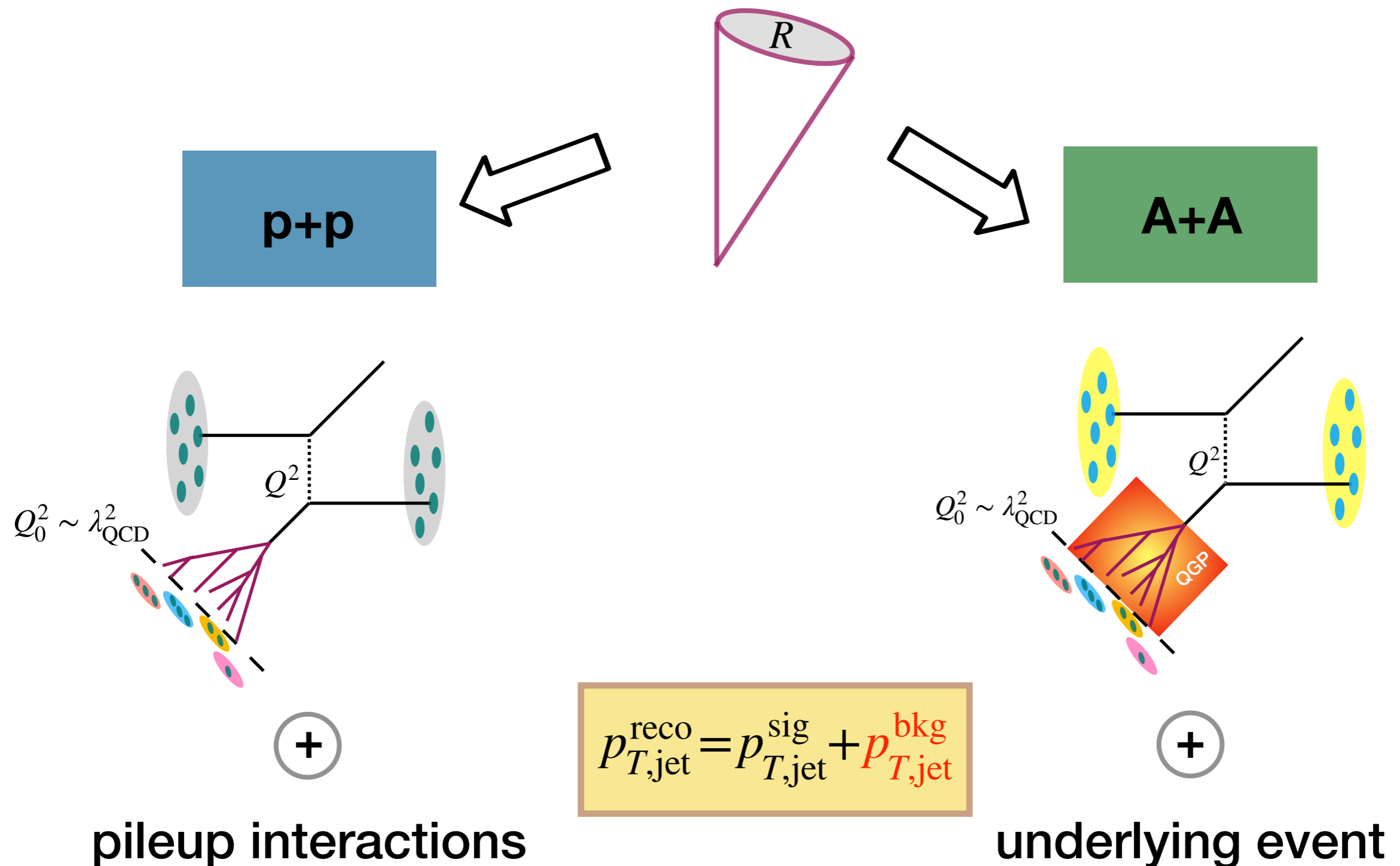
Jet reconstruction analyses

Cluster all particles in the event with k_T / anti- k_T / Cambridge-Aachen



Jet reconstruction analyses

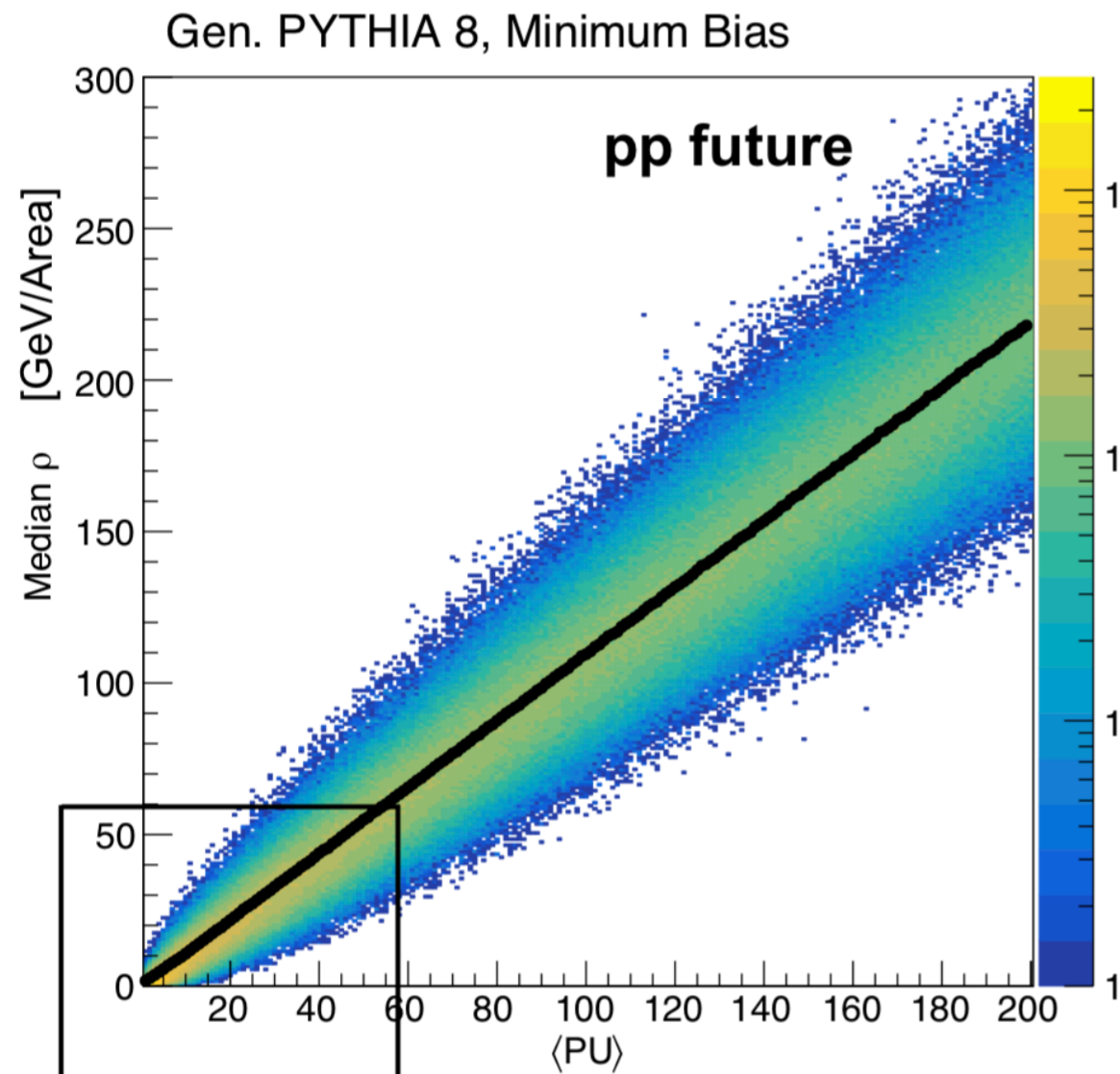
Cluster all particles in the event with k_T / anti- k_T / Cambridge-Aachen



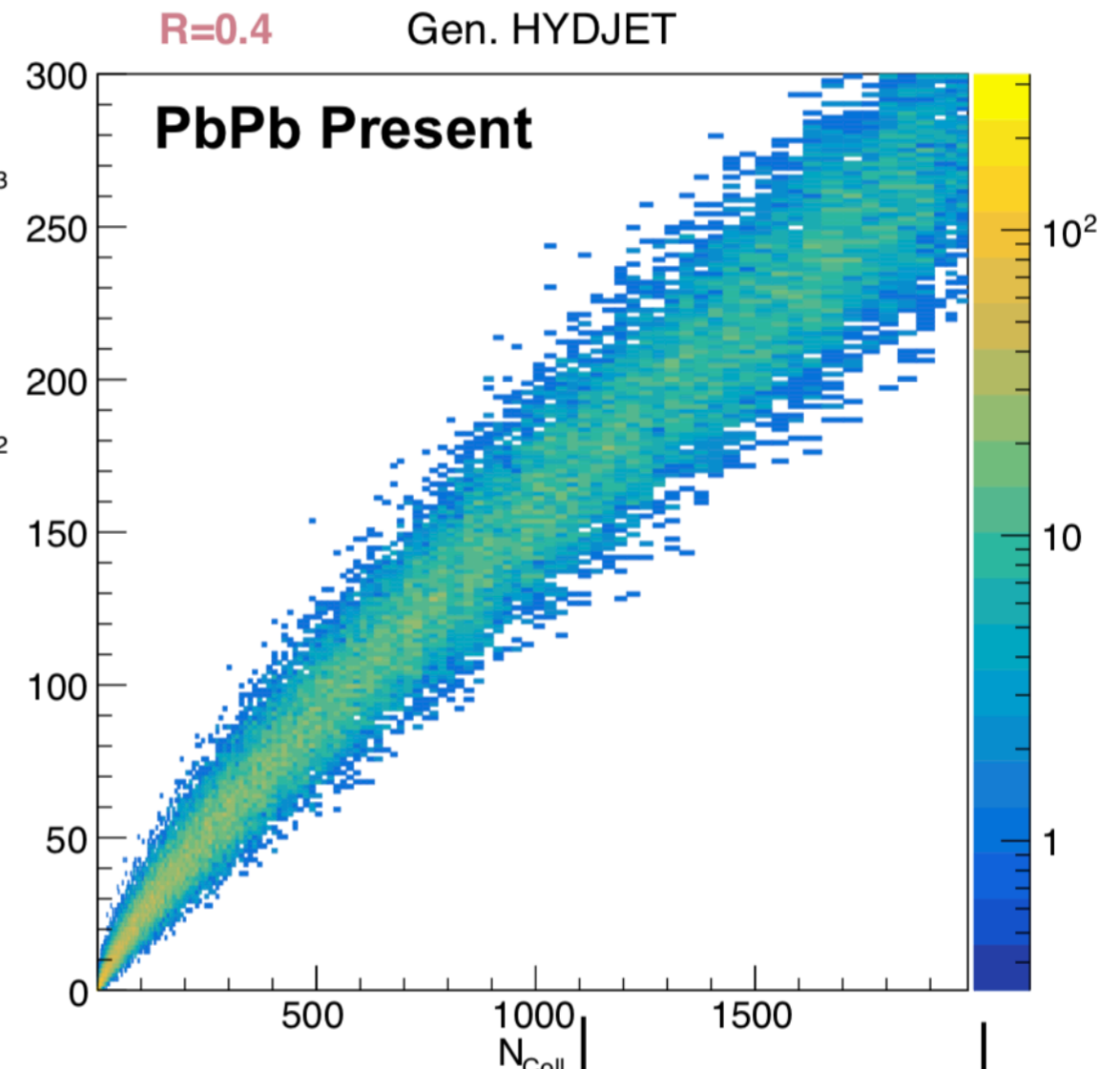
HL-LHC and HIC background contamination

Background momentum density per unit area

$$\rho = \frac{p_T^{\text{bkg}}}{A}$$



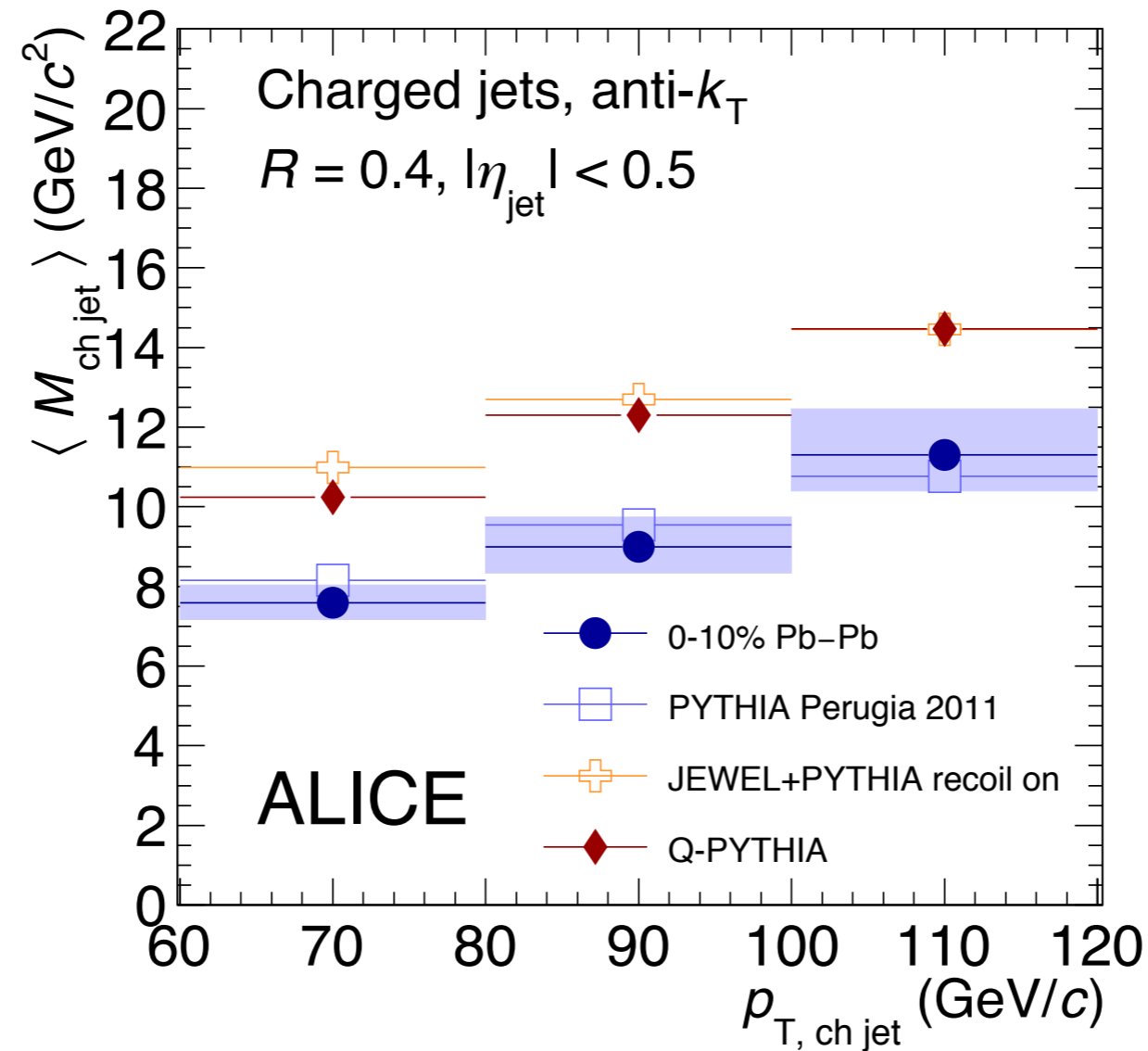
pp present



~0-10% PbPb at 5.02 TeV

[C. McGinn Talk at "The Definition of Jets in Large Background"]

Background as a limiting factor



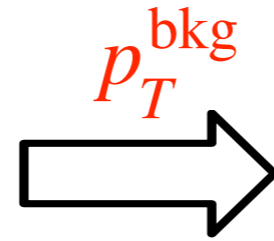
[ALICE Collab Phys. Lett. B776 (2018) 249-264]

Source	Pb-Pb			p-Pb		
	$p_{\text{T, ch jet}} < 80 </math> (GeV/c)$	80–100	100–120	60–80	80–100	100–120
Prior	1.0%	3.0%	5.0%	0	0	0
Background	3.0%	3.0%	5.0%	1.0%	0.5%	1.0%
Tracking efficiency	5.0%	5.0%	5.0%	3.0%	3.0%	3.0%
Unfolding (iterations, range)	1.0%	3.0%	4.0%	0.5%	1.0%	4.0%
Total	6.0%	8.0%	9.0%	3.5%	3.5%	4.5%

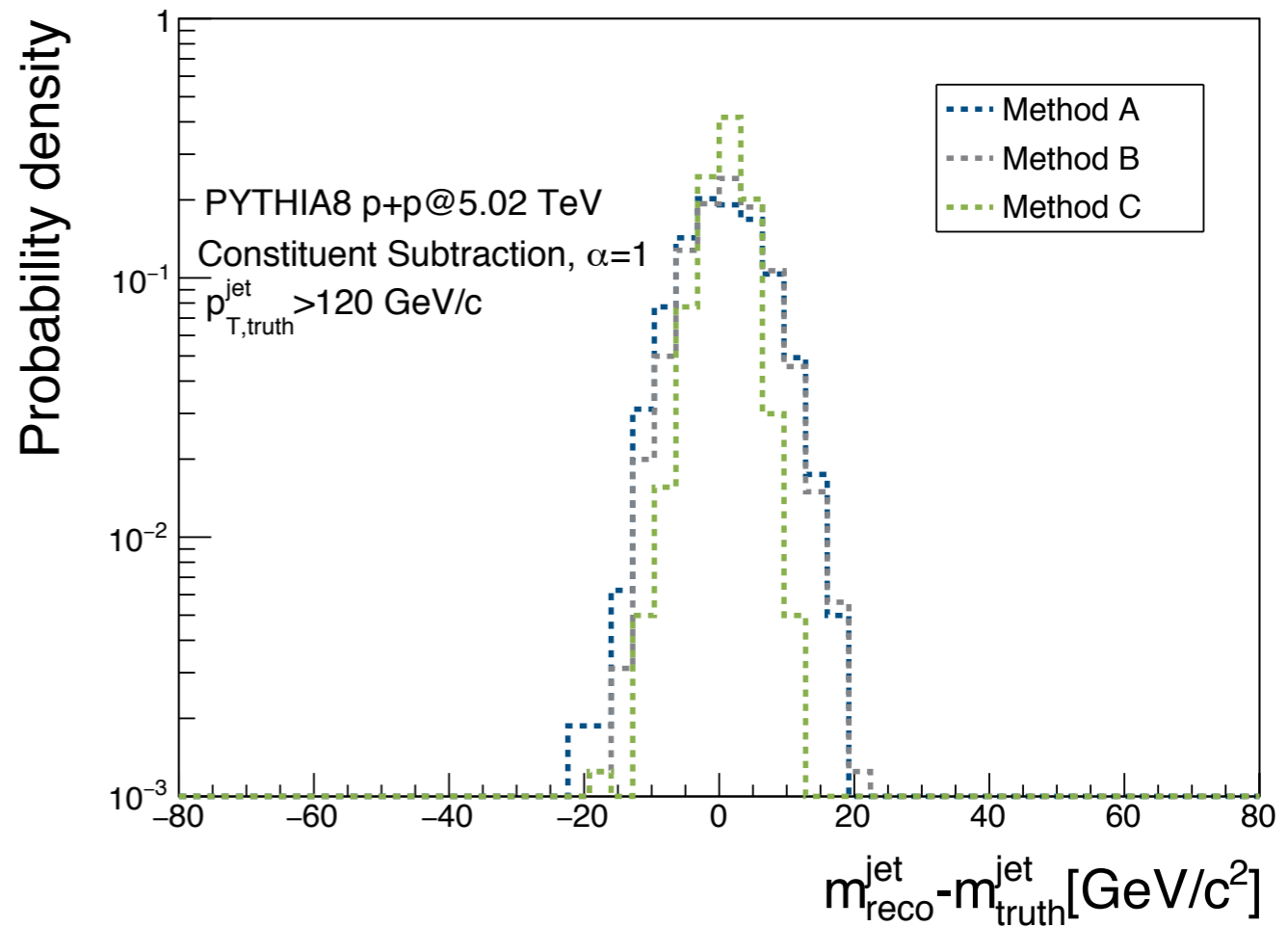
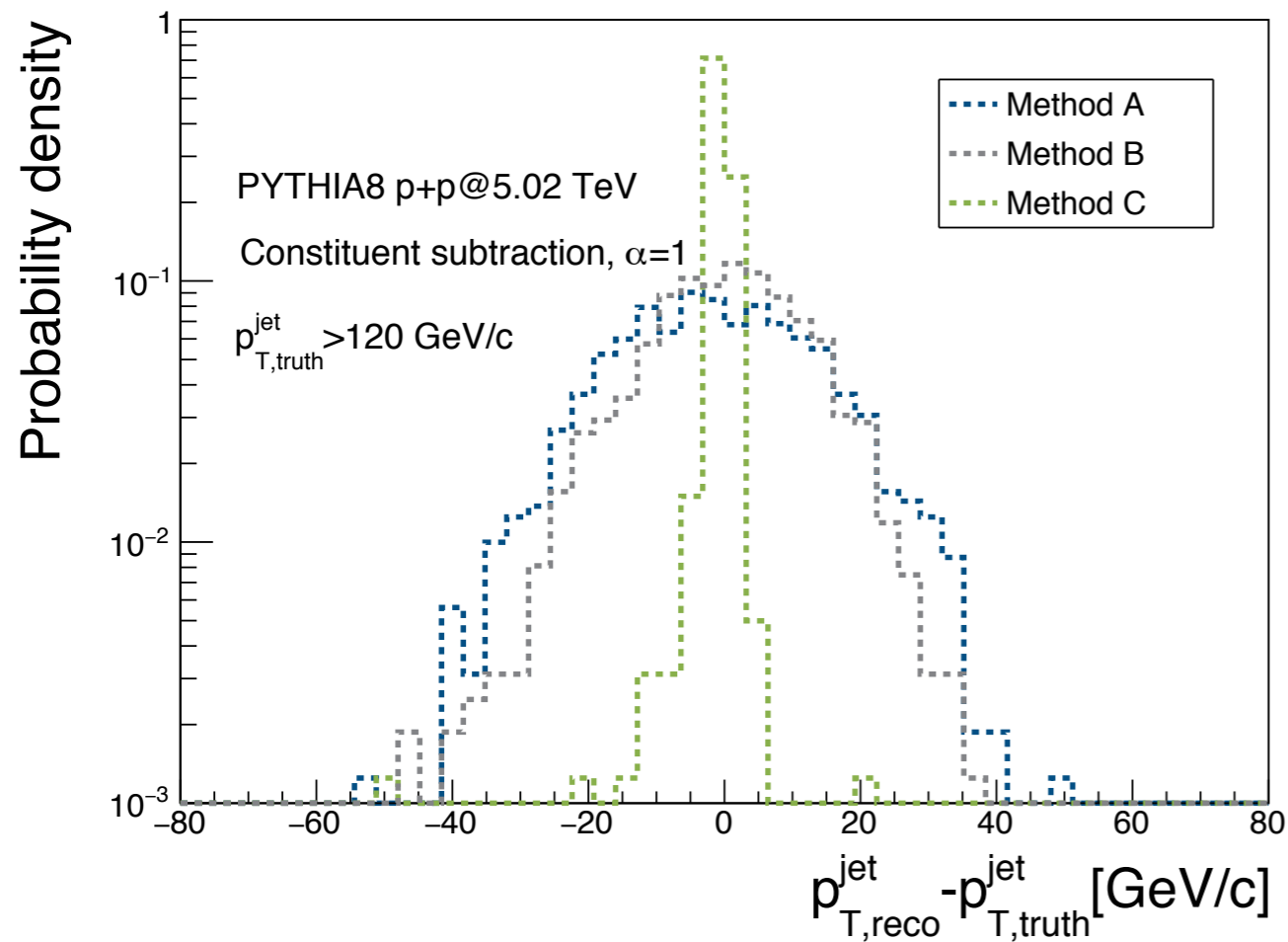
Table 1: Systematic uncertainty in mean jet mass from different sources in the 10% most central Pb–Pb collisions (left) and minimum-bias p–Pb collisions (right).

Background subtraction: two-step process

Background estimation



Subtraction method



Method A: Area median

Method B: SoftKiller

Method C: Truth

Simulation setup [\[https://github.com/JetQuenchingTools/JetToyHI\]](https://github.com/JetQuenchingTools/JetToyHI)

Background

- Randomly distributed in (η, ϕ)
- Thermal distribution with

$$\text{This talk } \begin{cases} |\eta| < 3, \\ \langle N_{\text{part}} \rangle = 7000, 12000 \\ \mu \equiv \langle p_{\text{T}} \rangle = 1.2, 0.7 \text{ GeV}/c \end{cases}$$

Signal

- Pythia8 dijet event

$$\text{This talk: } \begin{aligned} \langle \hat{p}_{\text{T}} \rangle &= 100 \text{ GeV} \\ \sqrt{s} &= 5.02 \text{ TeV} \end{aligned}$$

- JEWEL (back-reaction off)

$$\text{This talk: } \begin{aligned} \langle \hat{p}_{\text{T}} \rangle &= 150 \text{ GeV} \\ \sqrt{s} &= 5.02 \text{ TeV} \end{aligned}$$

PYTHIA8 + Thermal background == p+p collision @ HL-LHC

JEWEL w/ E-loss + Thermal background == Pb+Pb collision @LHC



BACKGROUND ESTIMATION

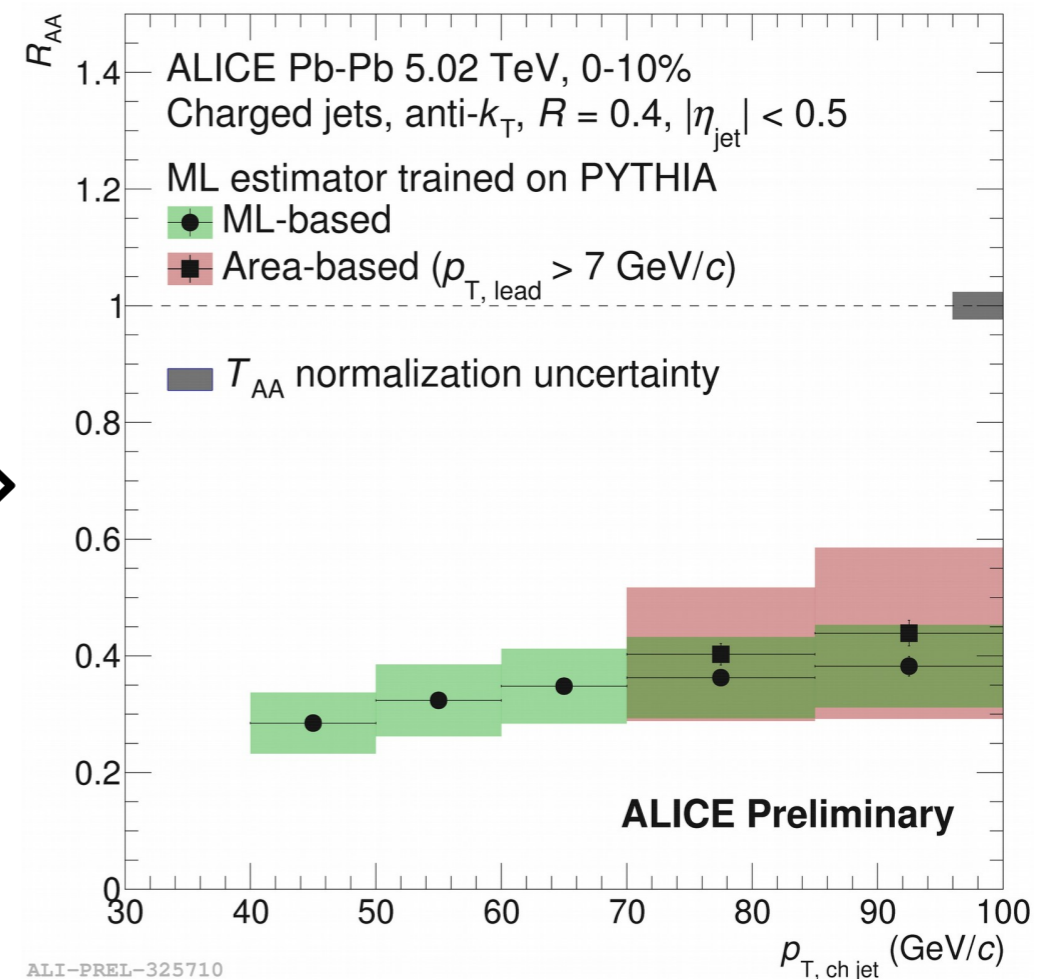
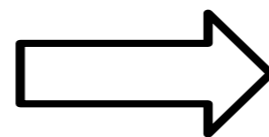
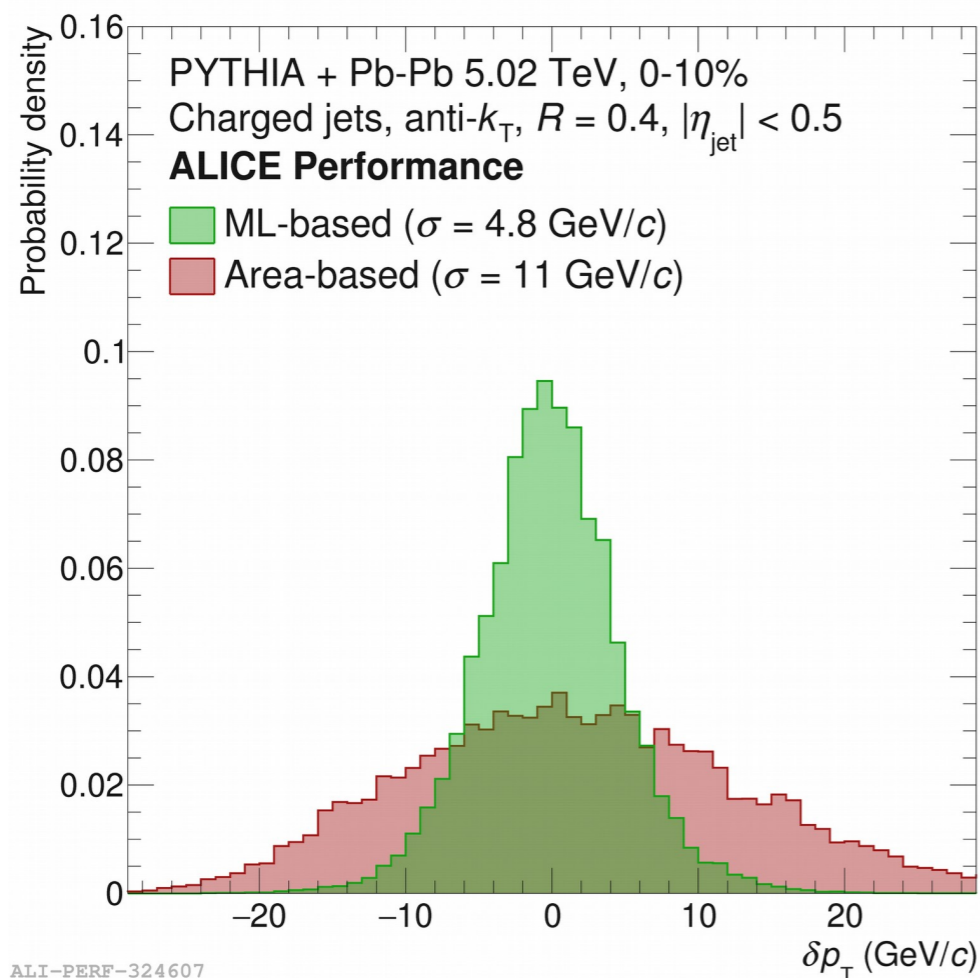
A decade of progress

- **Area-median:** an unbiased method [M. Cacciari, G.Salam Phys. Lett. B659 (2008) 119–126]

$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \text{med}(\rho) A_{\text{raw}}^{\text{jet}}$$

- **Supervised learning approach** [R. Haake, C.Loizides, PRC 99, 064904 (2019)]

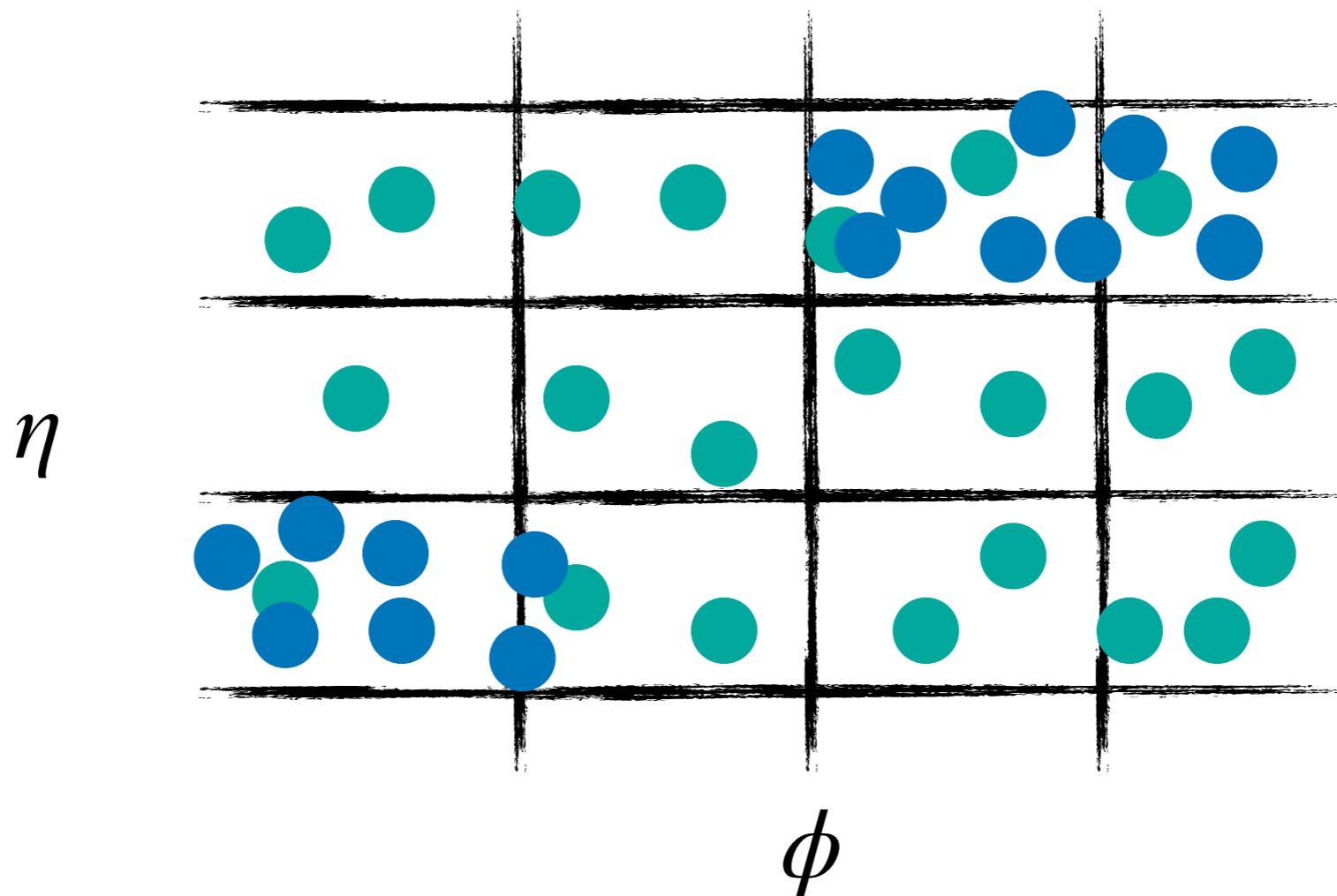
$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - p_{T,\text{ML}}^{\text{bkg}}$$



Area-median: an unbiased method

[M. Cacciari, G.Salam Phys. Lett. B659 (2008) 119–126]

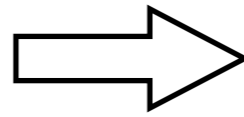
Cluster all particles (**sig+bkg**)
in the event with k_T algorithm



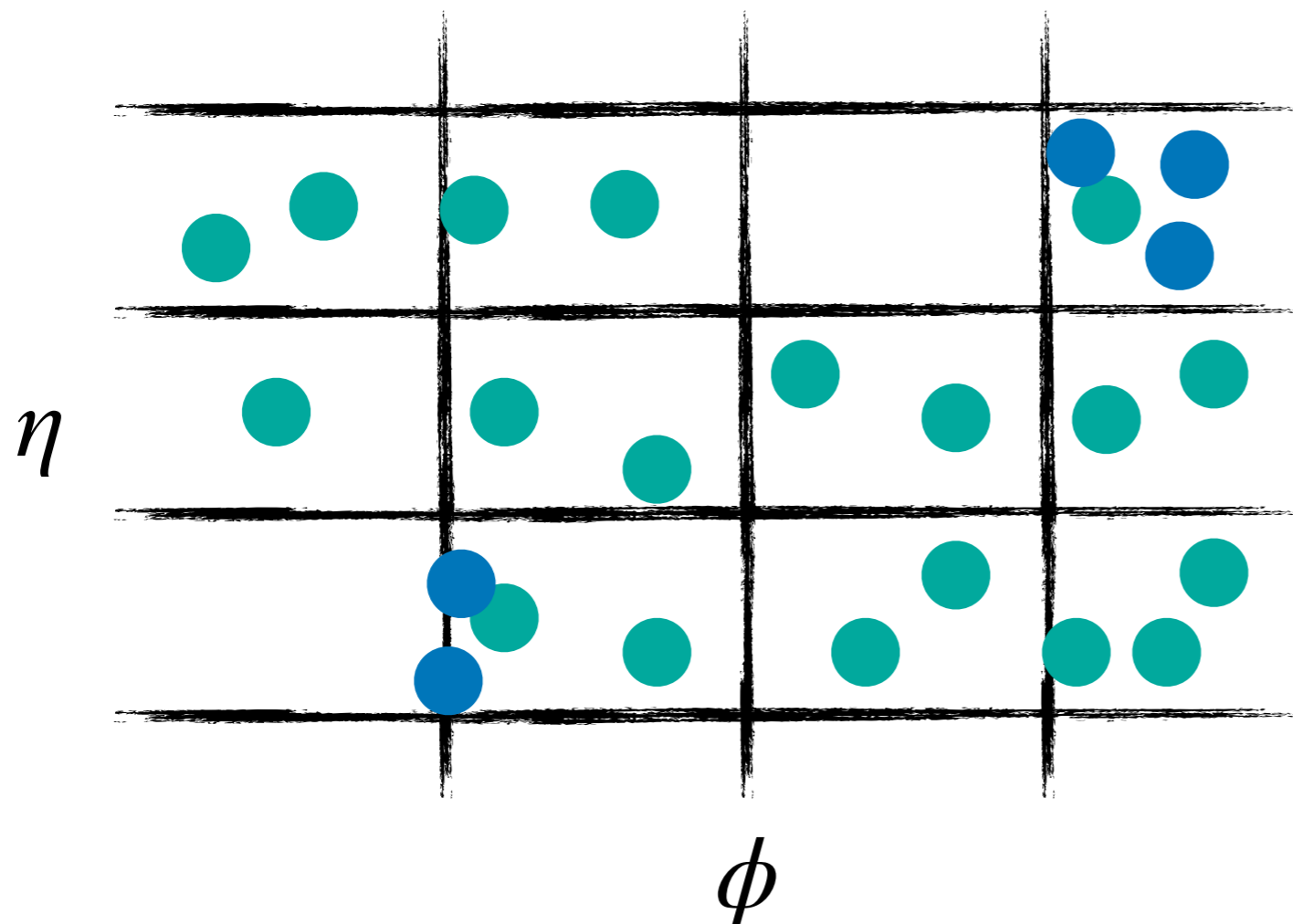
Area-median: an unbiased method

[M. Cacciari, G.Salam Phys. Lett. B659 (2008) 119–126]

Cluster all particles (sig+bkg) in the event with k_T algorithm



Remove the 2 hardest jets from the sample



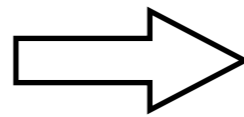
For every patch, i , compute

$$\rho_i = \frac{p_{T,i}}{A_i}$$

Area-median: an unbiased method

[M. Cacciari, G.Salam Phys. Lett. B659 (2008) 119–126]

Cluster all particles (**sig+bkg**) in the event with k_T algorithm

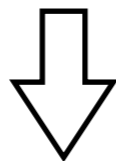
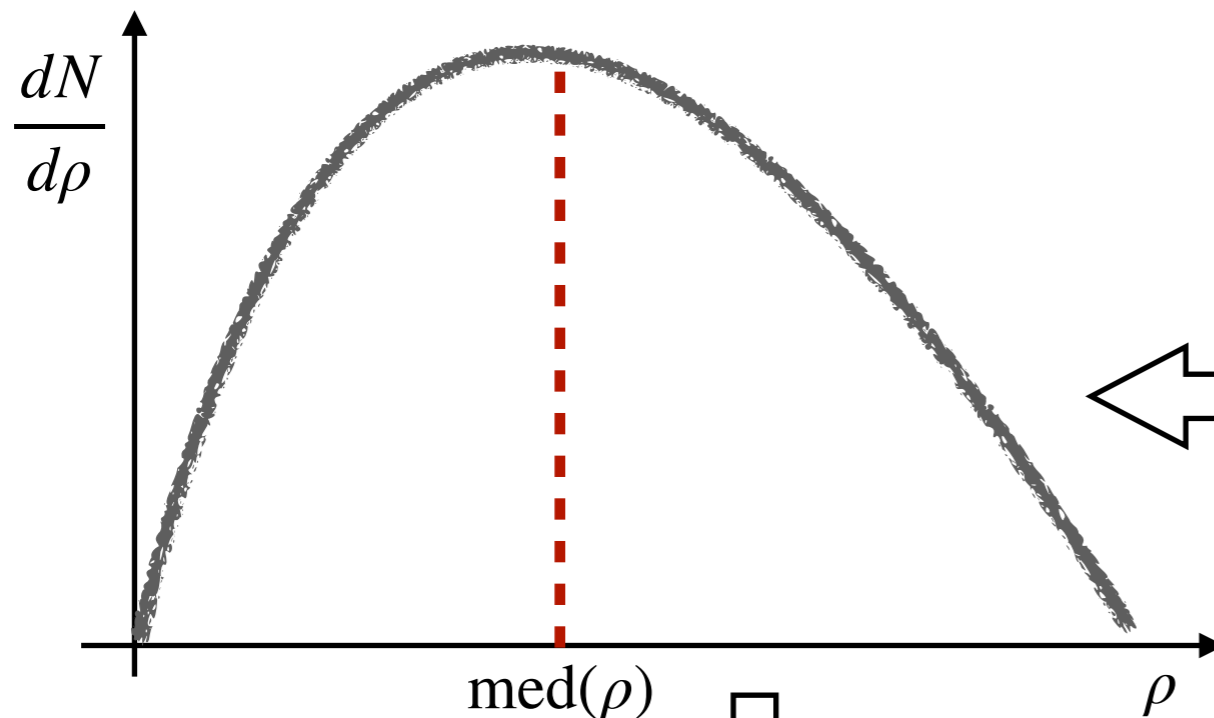
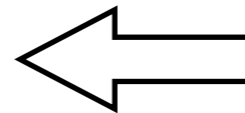


Remove the 2 hardest jets from the sample

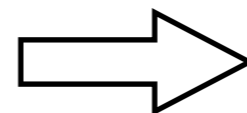


For every patch, i , compute

$$\rho_i = \frac{p_{T,i}}{A_i}$$



Cluster all particles (**sig+bkg**) in the event with **anti- k_T** algorithm

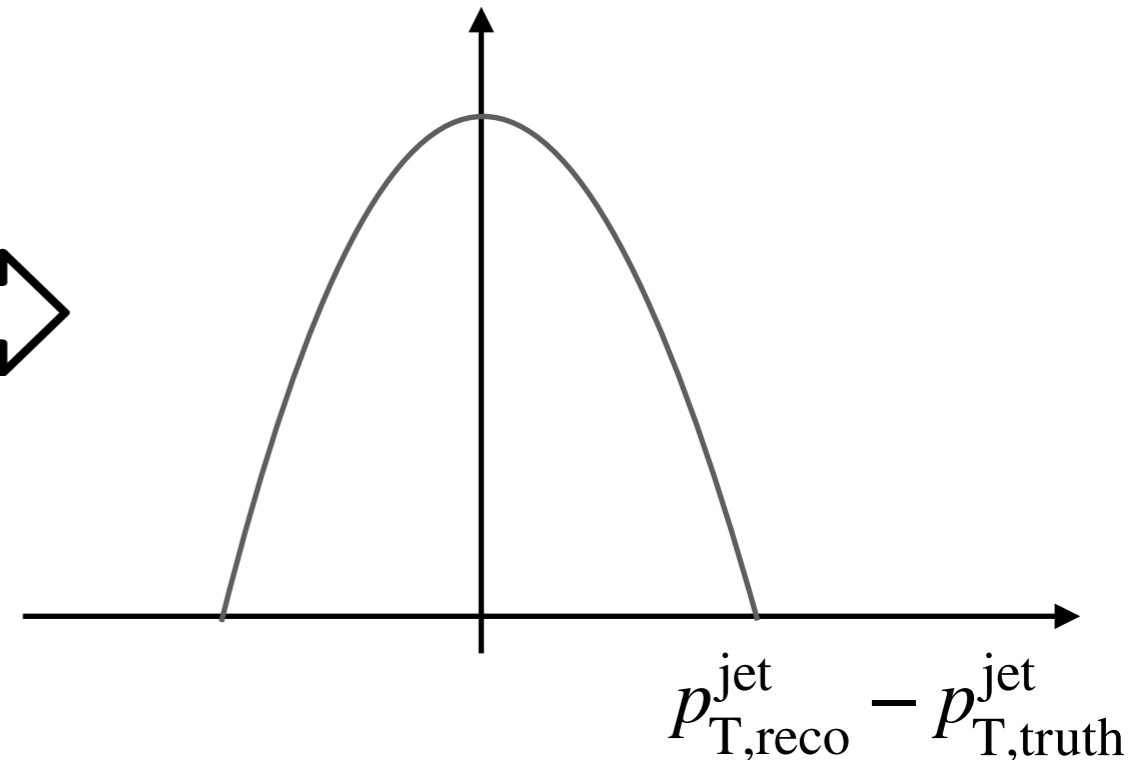
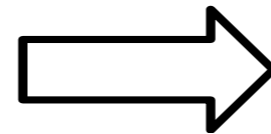


$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \text{med}(\rho) A_{\text{raw}}^{\text{jet}}$$

Area-median: an unbiased method

[M. Cacciari, G.Salam Phys. Lett. B659 (2008) 119–126]

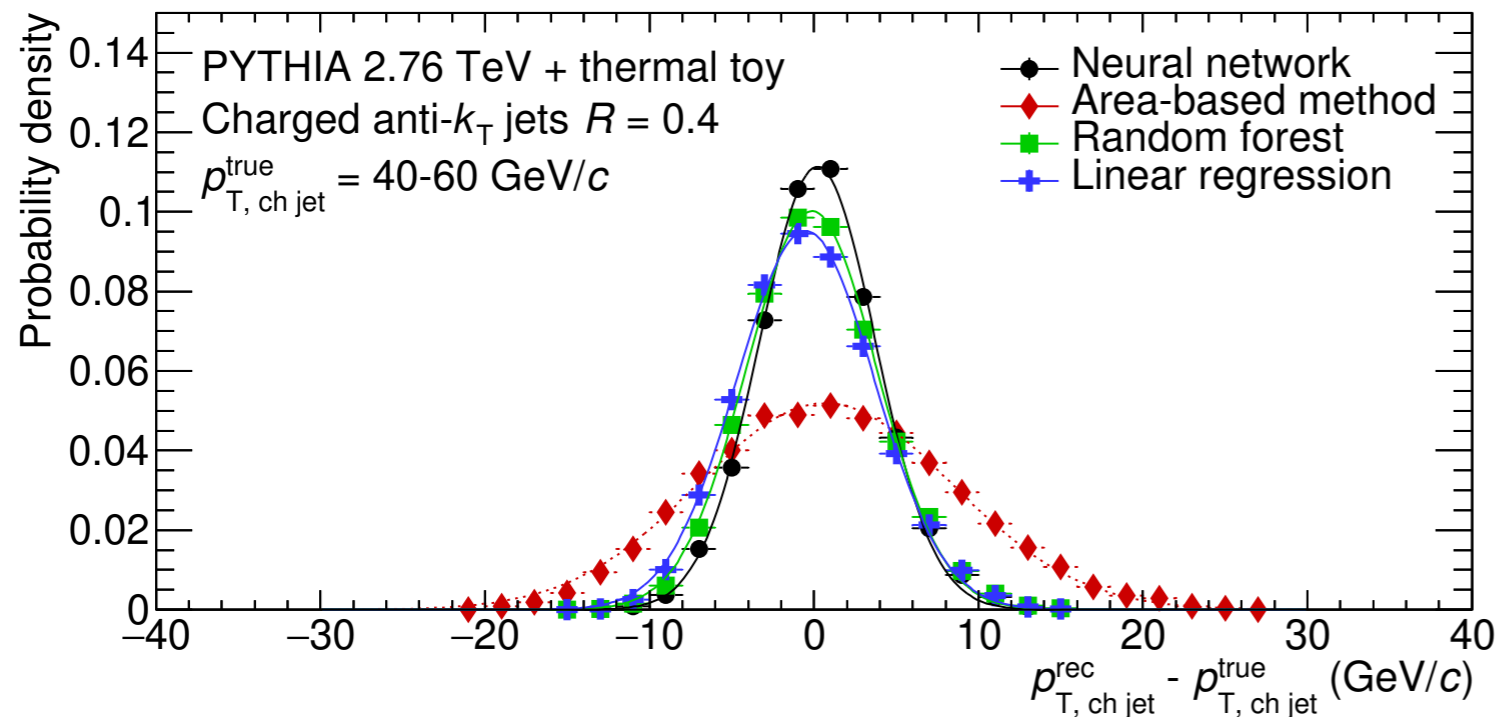
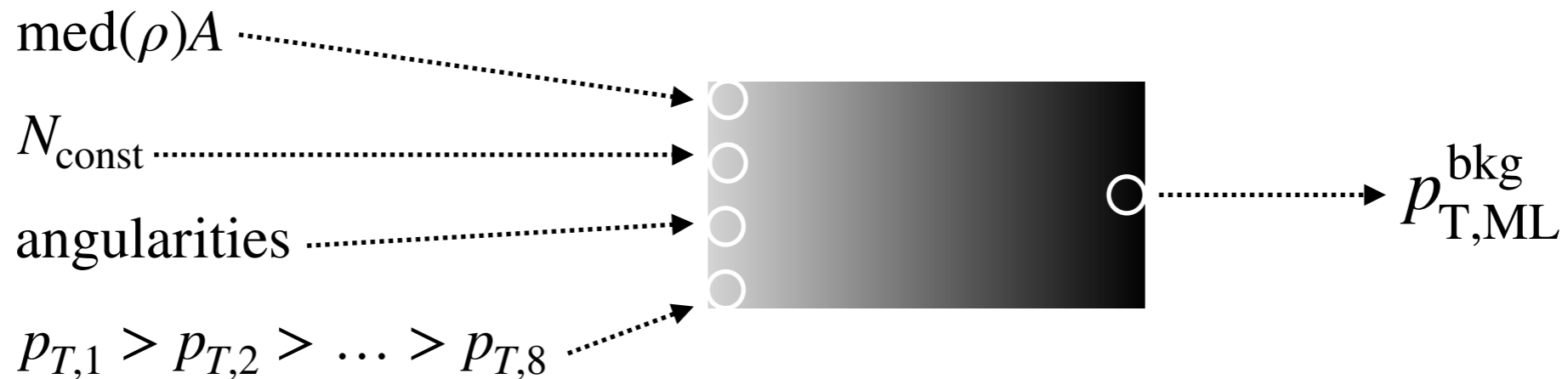
$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \text{med}(\rho) A_{\text{raw}}^{\text{jet}}$$



- Data-driven, unbiased, correct $\langle p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}} \rangle$
- No control over background fluctuations (needs unfolding)

Supervised learning approach

[R. Haake, C.Loizides, PRC 99, 064904 (2019)]

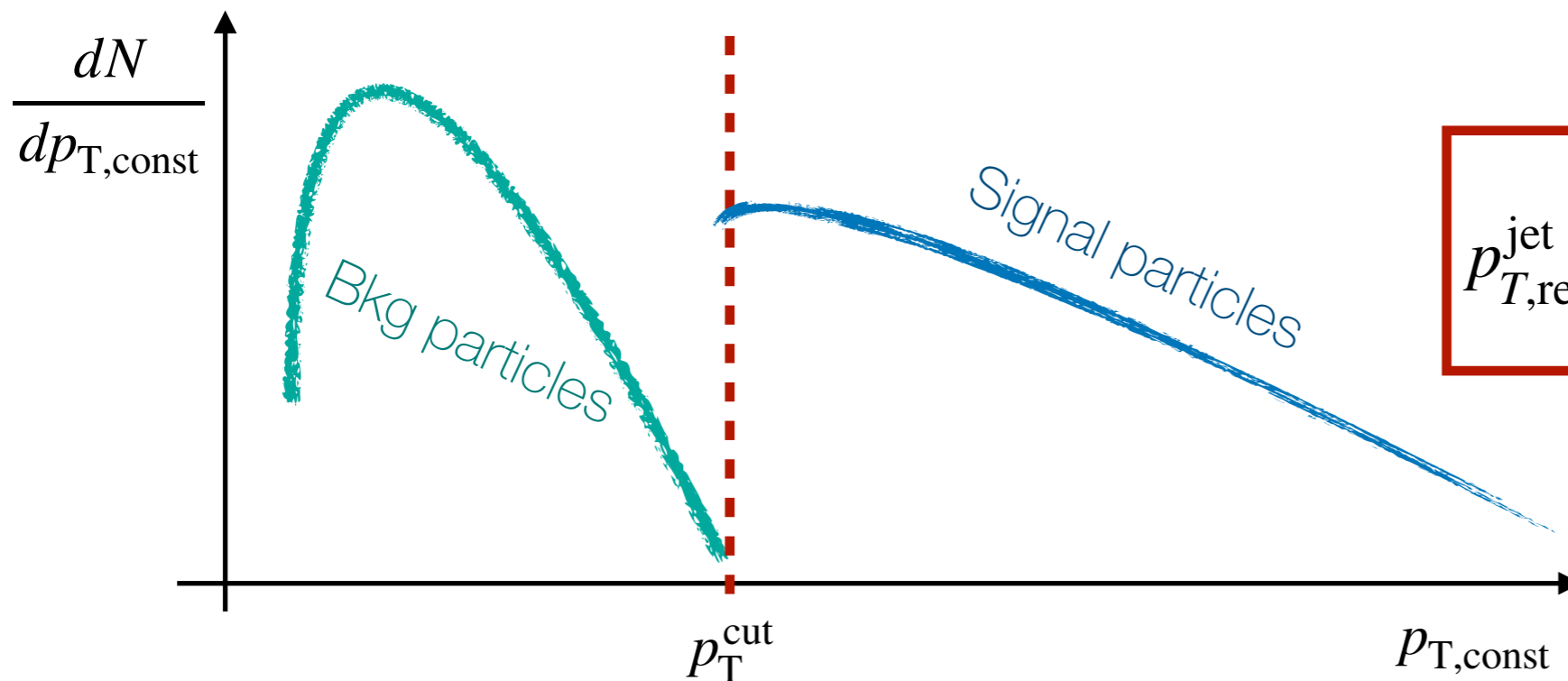


Estimator	Inclusive jets		Quark jets		Gluon jets	
Fit param. (GeV/c)	Mean	σ	Mean	σ	Mean	σ
Area-based method	0.7	7.8	0.8	7.3	0.9	7.7
Neural network	0.2	3.5	1.0	3.1	-0.5	3.6
Random Forest	-0.1	4.0	0.7	3.6	-0.8	3.9
Linear regression	-0.5	4.2	0.8	3.8	-1.4	4.1

TABLE I. Properties of the residual distributions for the considered estimators for $40 \leq p_{T, \text{ch jet}}^{\text{true}} < 60 \text{ GeV}/c$. Mean and σ represent mean and standard deviation of the Gaussian fits, respectively.

SoftKiller: introducing a soft p_T -cut

[M. Cacciari, G. Salam and G. Soyez Eur.Phys.J. C75 (2015) no.2, 59]



$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \sum_i^{p_{T,i} < p_T^{\text{cut}}} p_{T,i}$$

If there is no signal below the p_T -cut, exact handle on background fluctuations

$$\langle p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}} \rangle = 0$$

$$\sigma(p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}}) = 0$$

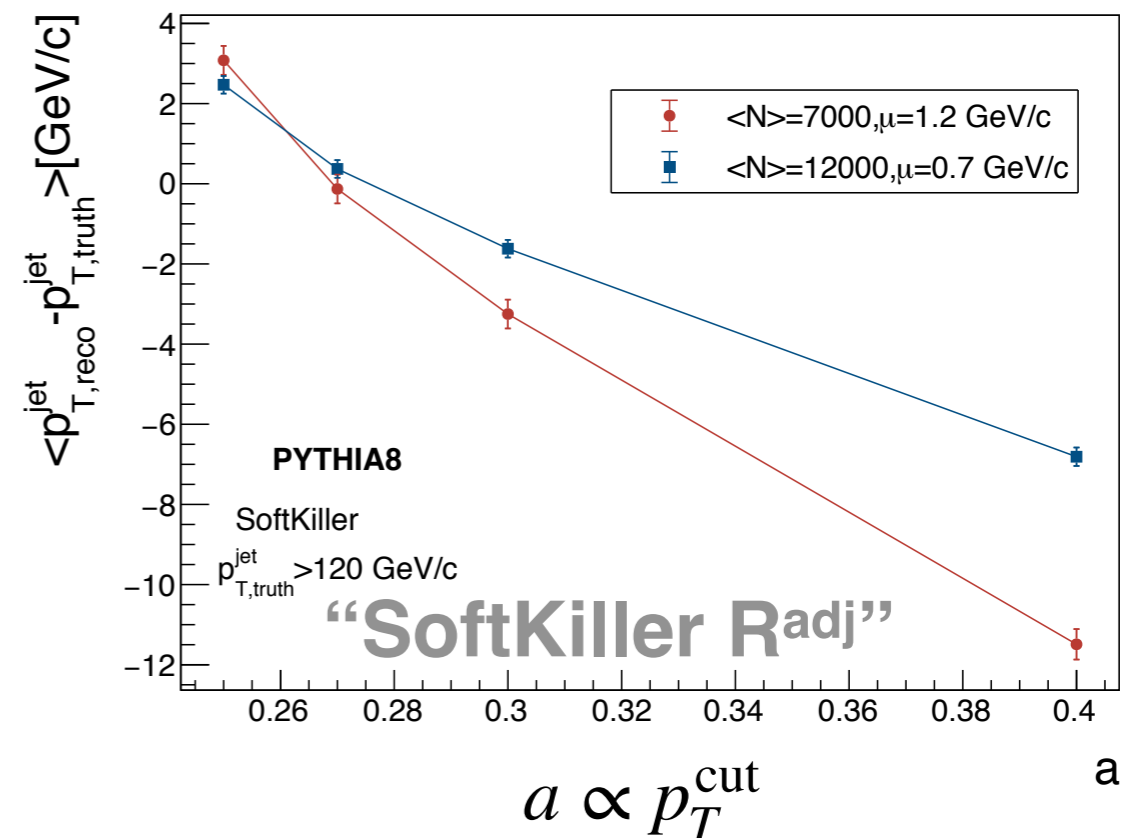
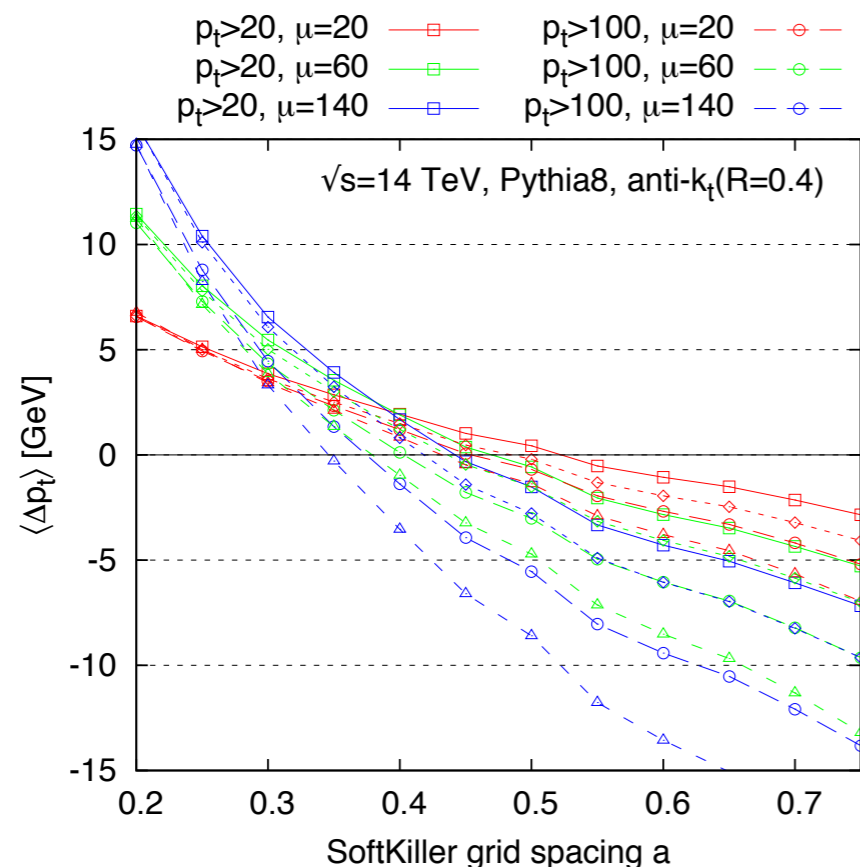
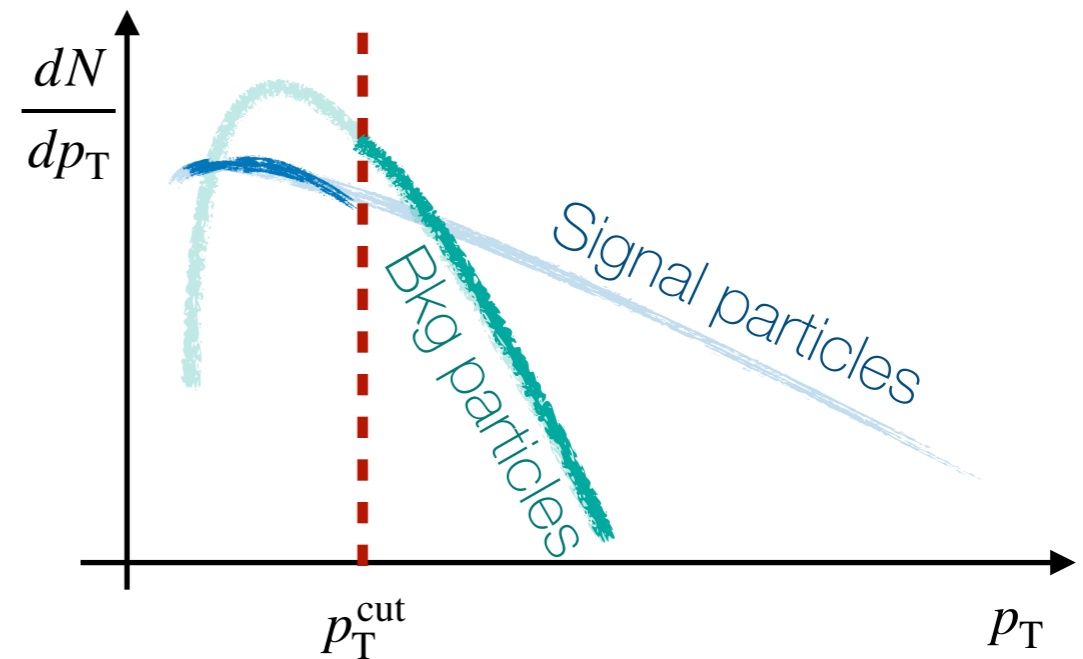
SoftKiller: introducing a soft p_T -cut

[M. Cacciari, G. Salam and G. Soyez Eur.Phys.J. C75 (2015) no.2, 59]

- Choose p_T -cut such that $p_{T,<}^{\text{sig}}$ and $p_{T,>}^{\text{bkg}}$ balance each other

$$\langle p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}} \rangle \sim 0$$

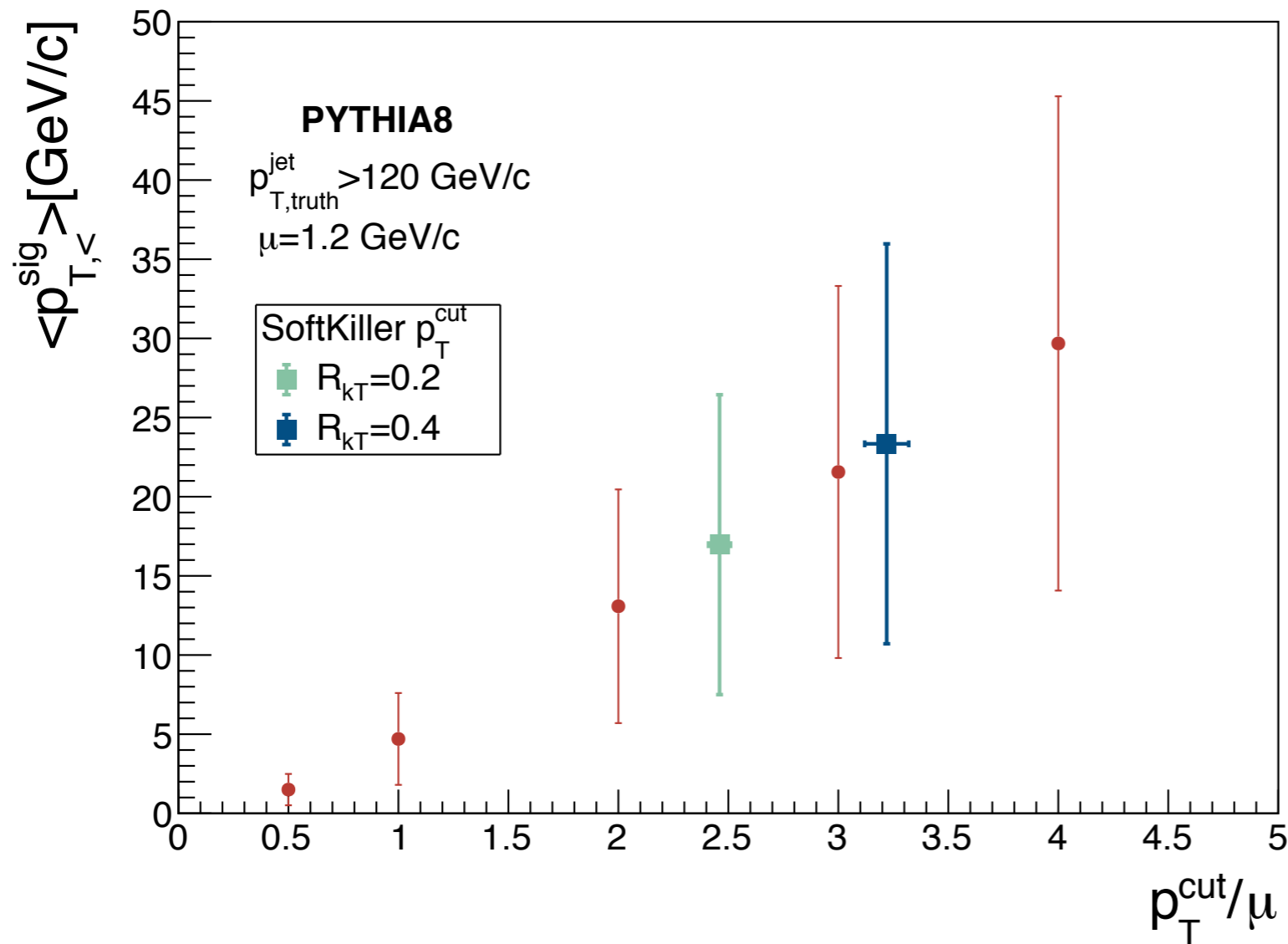
- But...this actually requires a large degree of fine tuning



Impact of signal contamination

Any background estimator that relies on a soft pT-cut is sensitive to QCD radiation below this threshold

[Mehtar-Tani, ASO, Verweij arXiv:1904.12815]

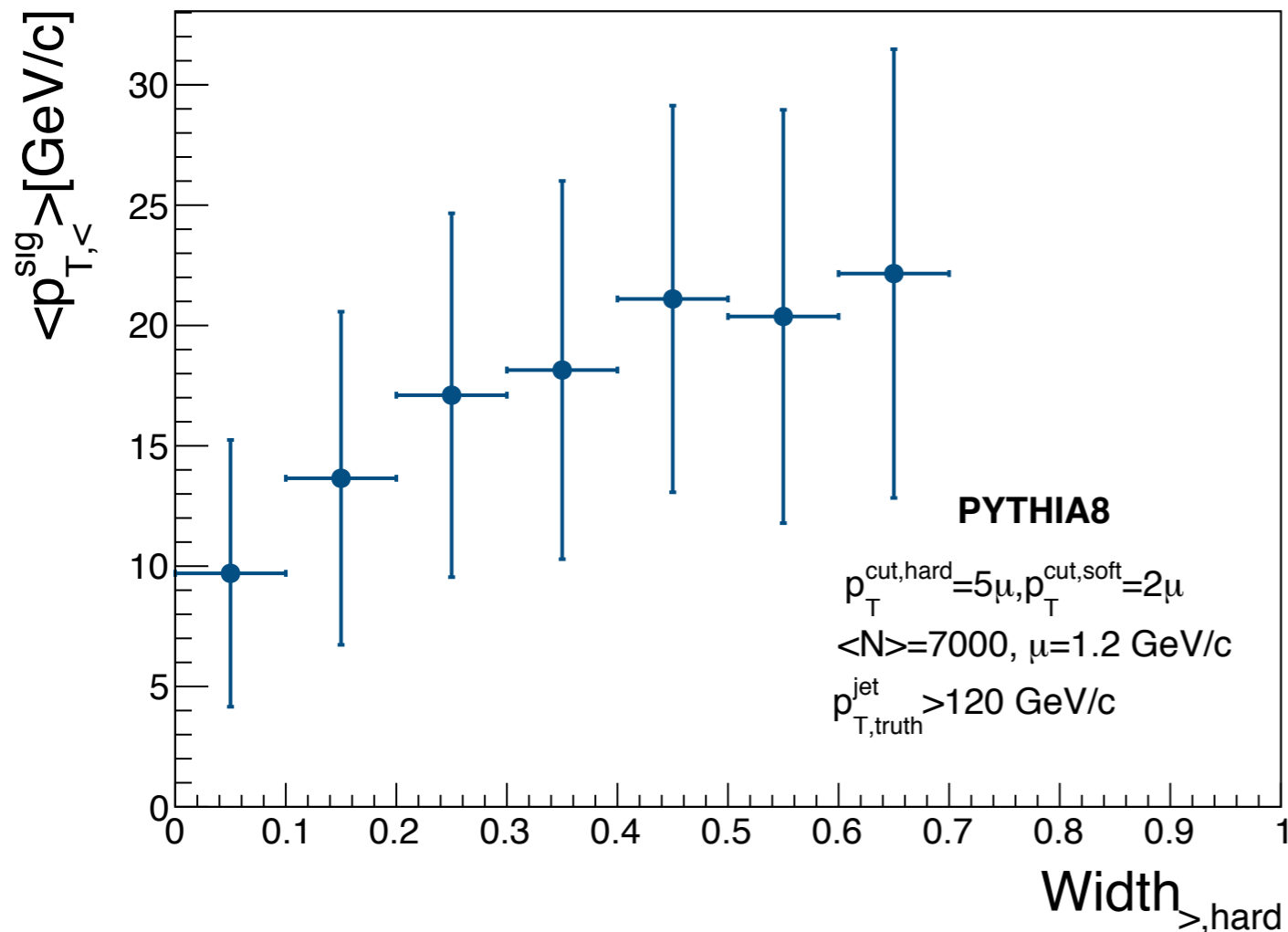
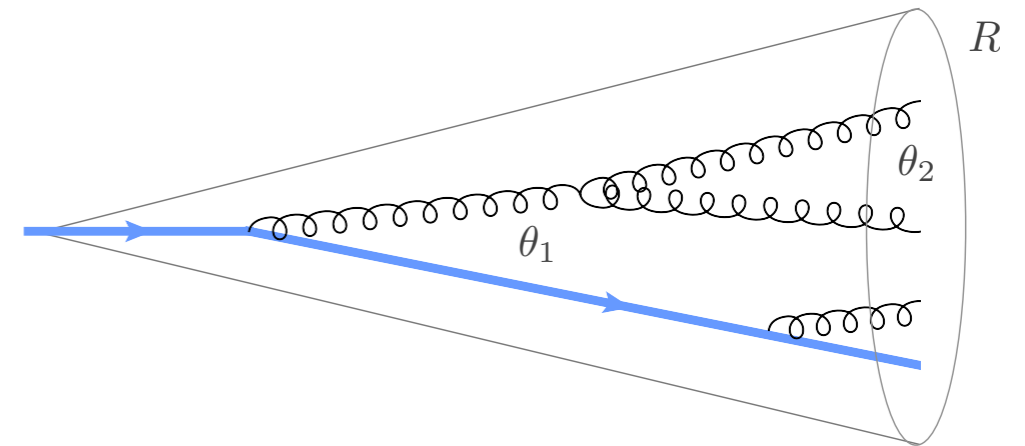
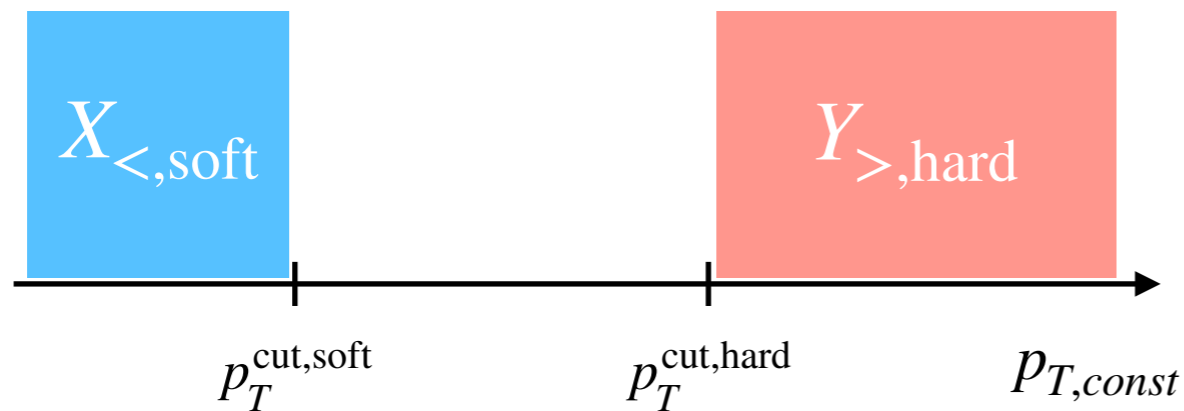


$$p_{T,<sup>sig>} = \sum_{i \in \text{sig}}^{p_{T,i} < p_T^{cut}} p_{T,i}$$

$$\langle p_{T, \text{reco}}^{\text{jet}} - p_{T, \text{truth}}^{\text{jet}} \rangle \propto - \langle p_{T,<sup>sig>} \rangle + \dots$$

$$\sigma(p_{T, \text{reco}}^{\text{jet}} - p_{T, \text{truth}}^{\text{jet}}) \propto \sigma(p_{T,<sup>sig>})$$

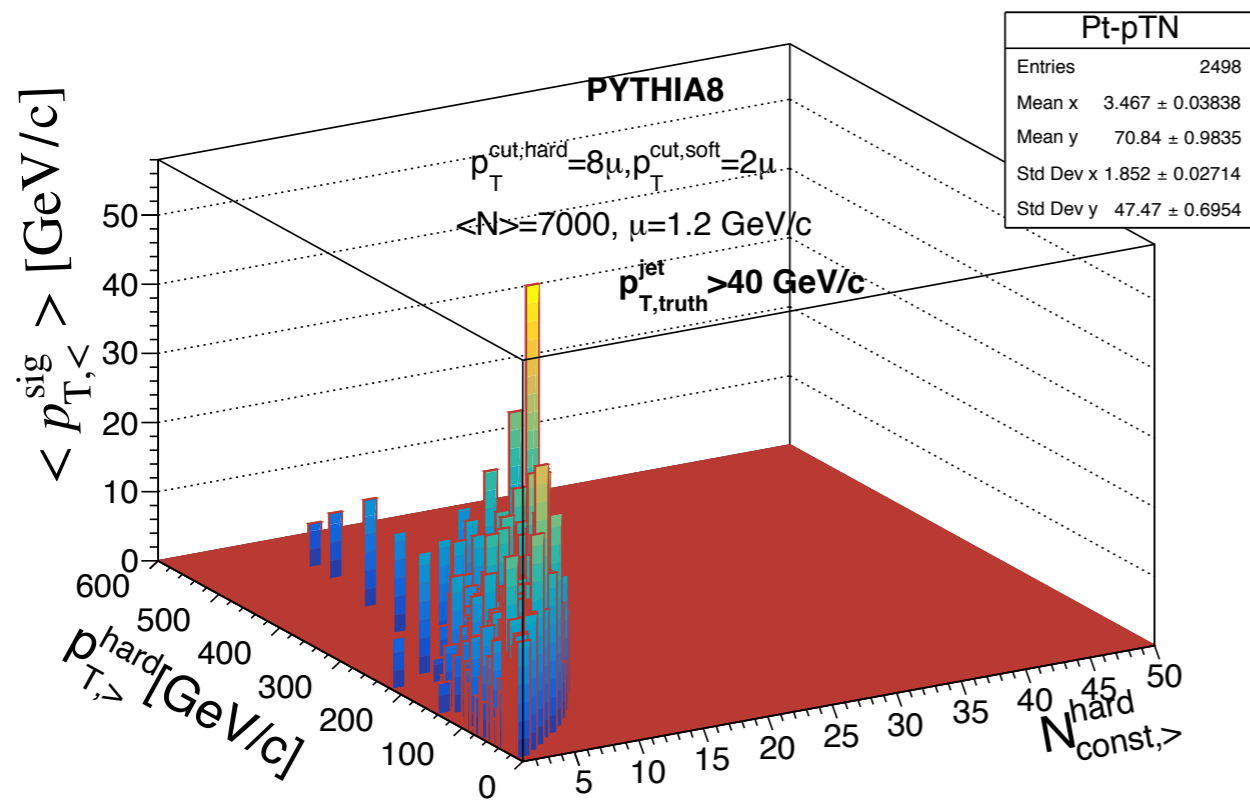
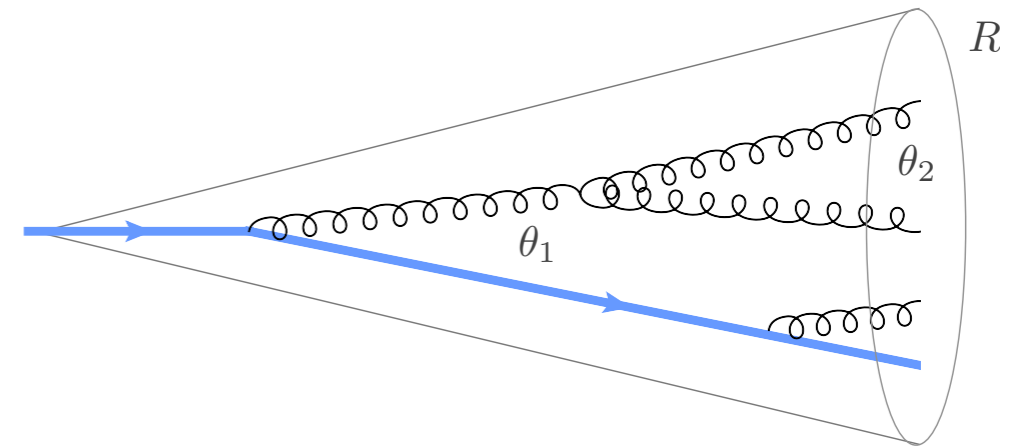
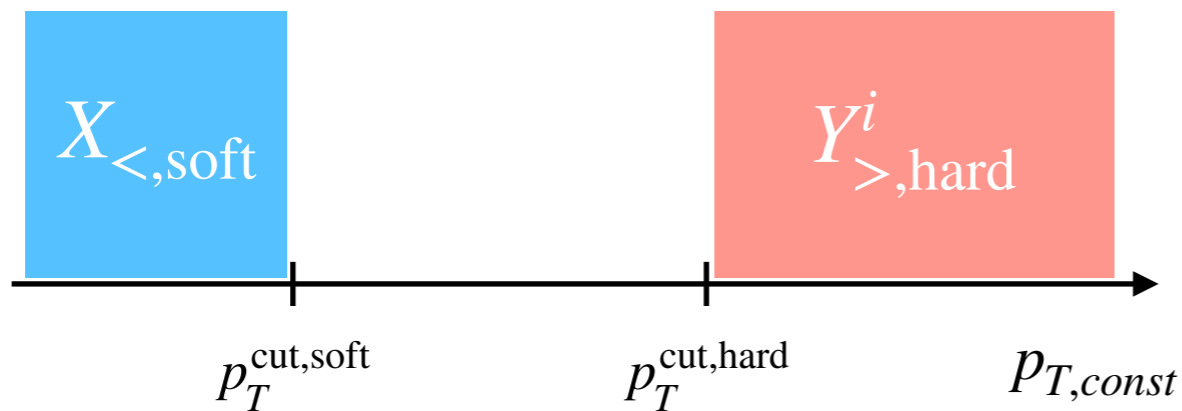
Hard-soft correlations within a QCD jet



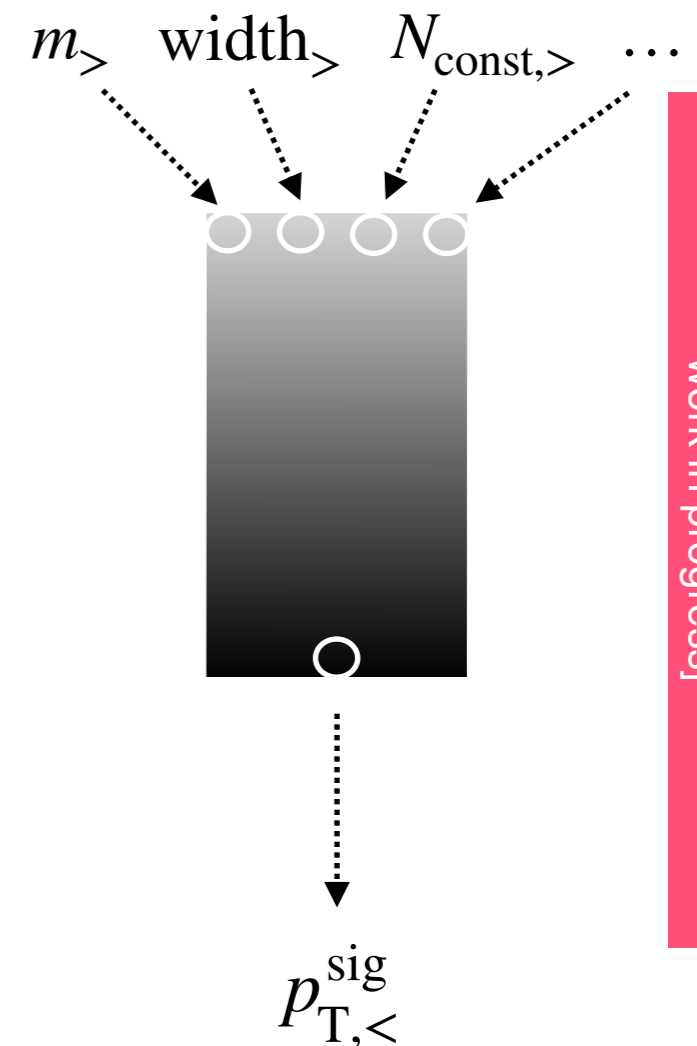
$\langle p_{T,<}^{\text{sig}} \rangle$ and $\sigma(p_{T,<}^{\text{sig}})$ can be mitigated by knowing/measuring jet properties in the hard sector i.e. a bkg free environment

$$\text{width}_{>,hard} = \sum_i^{p_{T,i} > p_T^{\text{cut,hard}}} \frac{p_{T,i}}{p_{T,>}} \frac{\theta_i}{R}$$

Hard-soft correlations within a QCD jet

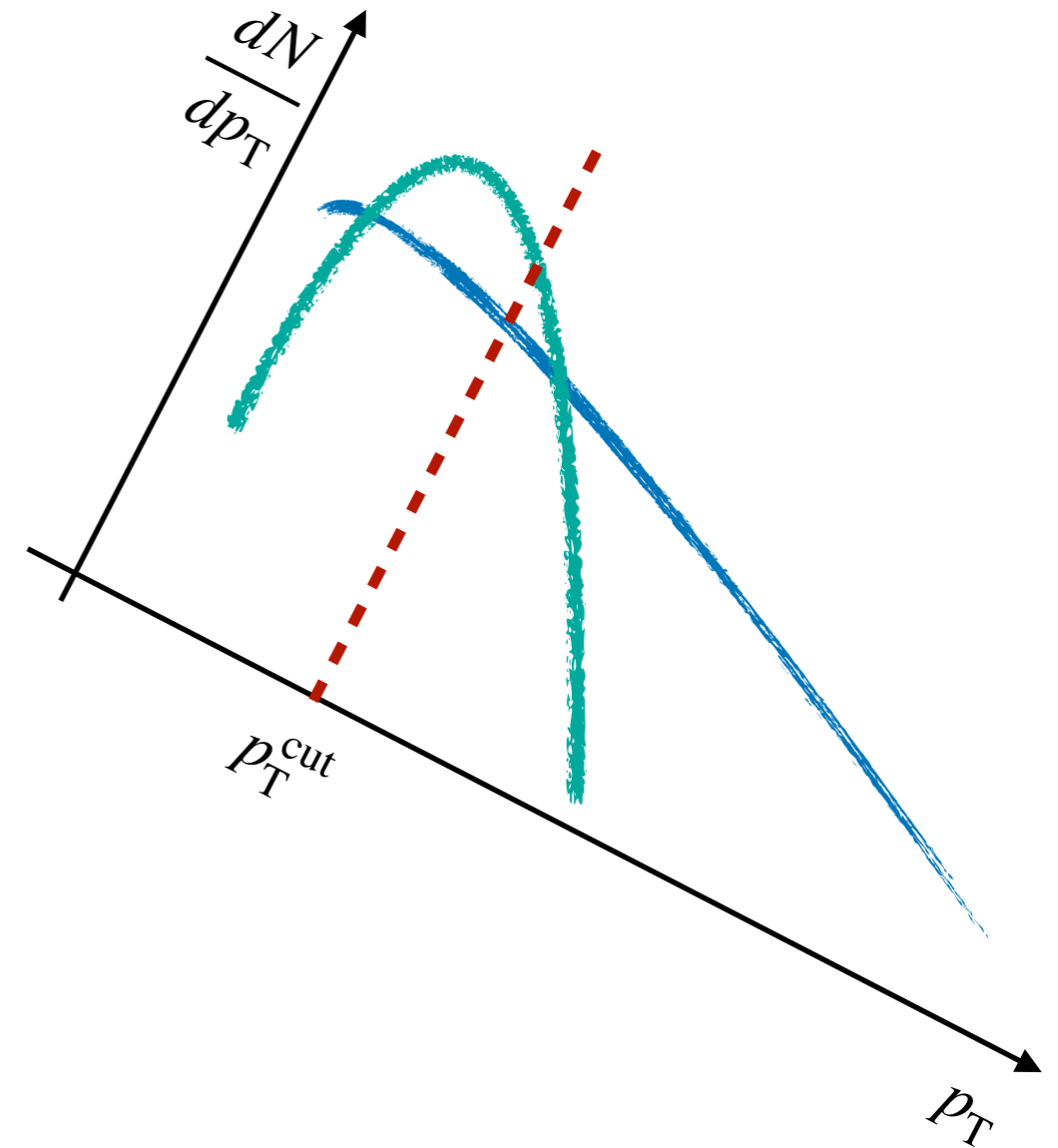
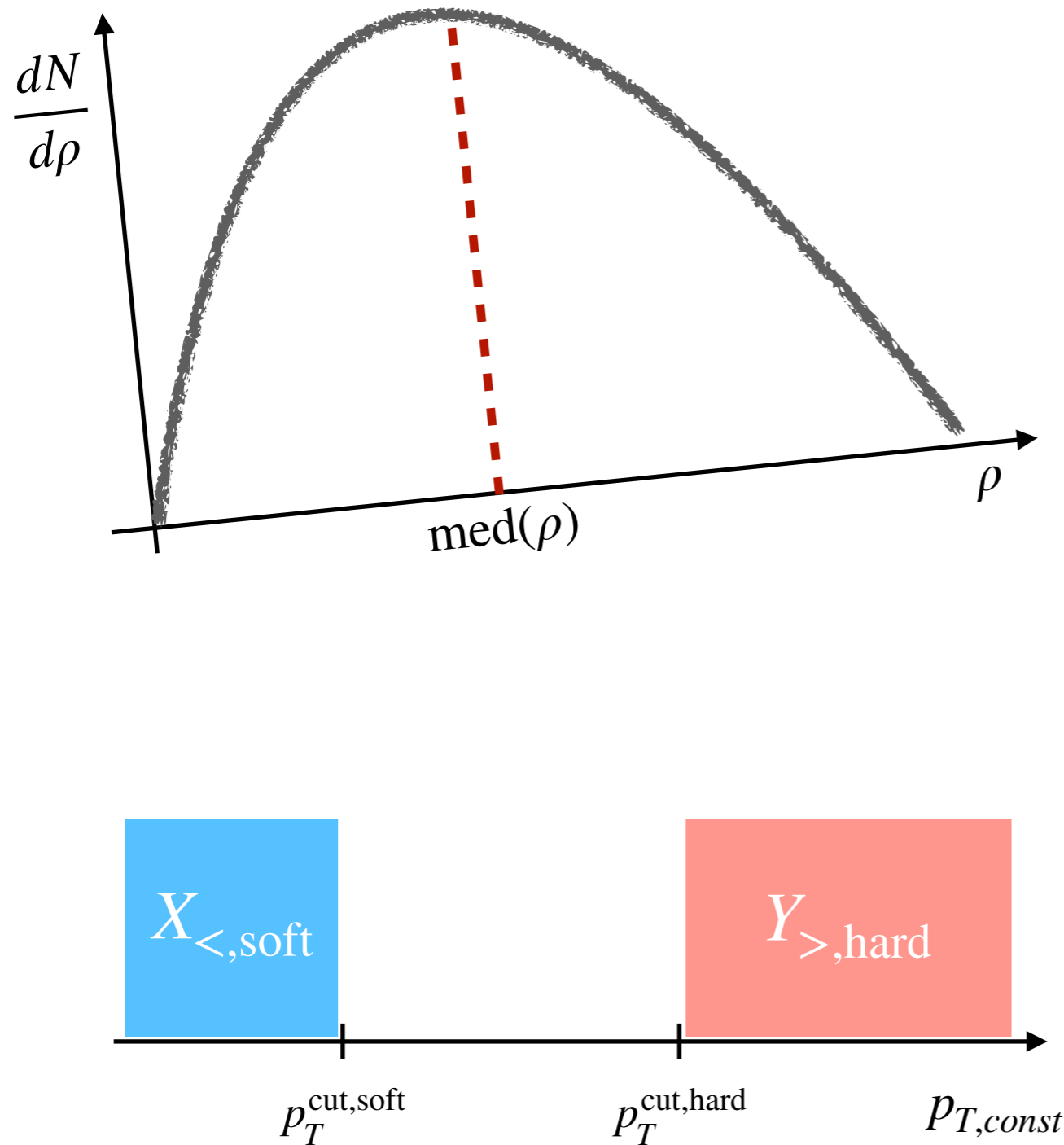


or



The ρ -correction approach

[Mehtar-Tani, ASO, Verweij arXiv:1904.12815]

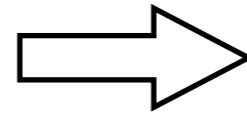


The ρ -correction approach

[Mehtar-Tani, ASO, Verweij arXiv:1904.12815]

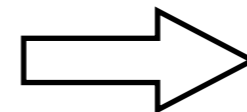
1

Cluster all particles (sig+bkg) in the event with k_T algorithm, $R=0.4$



Remove the 2 hardest patches from the sample

For each patch, $\rho_{>} = \frac{1}{A} \sum_i^{p_{T,i} > p_T^{\text{cut,soft}}} p_{T,i}$



med($\rho_{>}$)

2

Re-cluster the whole event (sig+bkg) in the event with anti- k_T algorithm, $R=0.4$

For each jet,

$$p_{T,<} = \sum_i^{p_{T,i} < p_T^{\text{cut,soft}}} p_{T,i}$$

3

For a given $p_T^{\text{cut,soft}}$ measure $\langle p_{T,<}^{\text{sig}} \rangle$ in low pileup proton-proton collisions

The ρ -correction approach

[Mehtar-Tani, ASO, Verweij arXiv:1904.12815]

$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \text{med}(\rho_{>}) A_{\text{raw}}^{\text{jet}} - p_{T,<} + \langle p_{T,<}^{\text{sig}} \rangle$$

- Neglecting $\langle p_{T,<}^{\text{sig}} \rangle$ -term, $\begin{cases} p_T^{\text{cut,soft}} = 0 : \text{Area-median} \\ p_T^{\text{cut,soft}} = p_T^{\text{cut,SK}} : \text{SoftKiller } R_{KT} \end{cases}$
- Role of $\langle p_{T,<}^{\text{sig}} \rangle$: correct for the signal contamination below $p_T^{\text{cut,soft}}$
 - If measured inclusively, ρ -correction
 - If measured as a function of $\text{width}_{>,\text{hard}}$, ρ -correction HS_{corr}

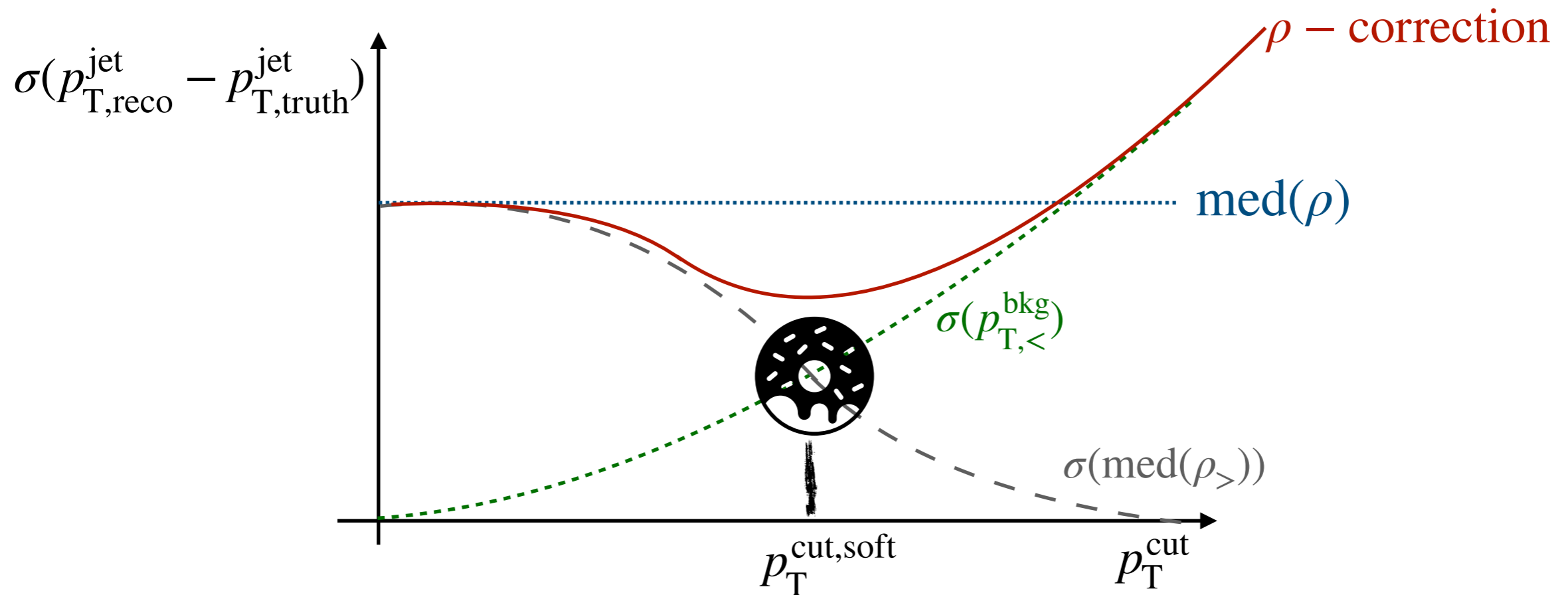
Value of $p_T^{\text{cut,hard}}$? Large enough to be in the background-free region

$$p_T^{\text{cut,hard}} = 5\mu \quad (\mu : \text{average momentum of bkg constituents})$$

The ρ -correction approach

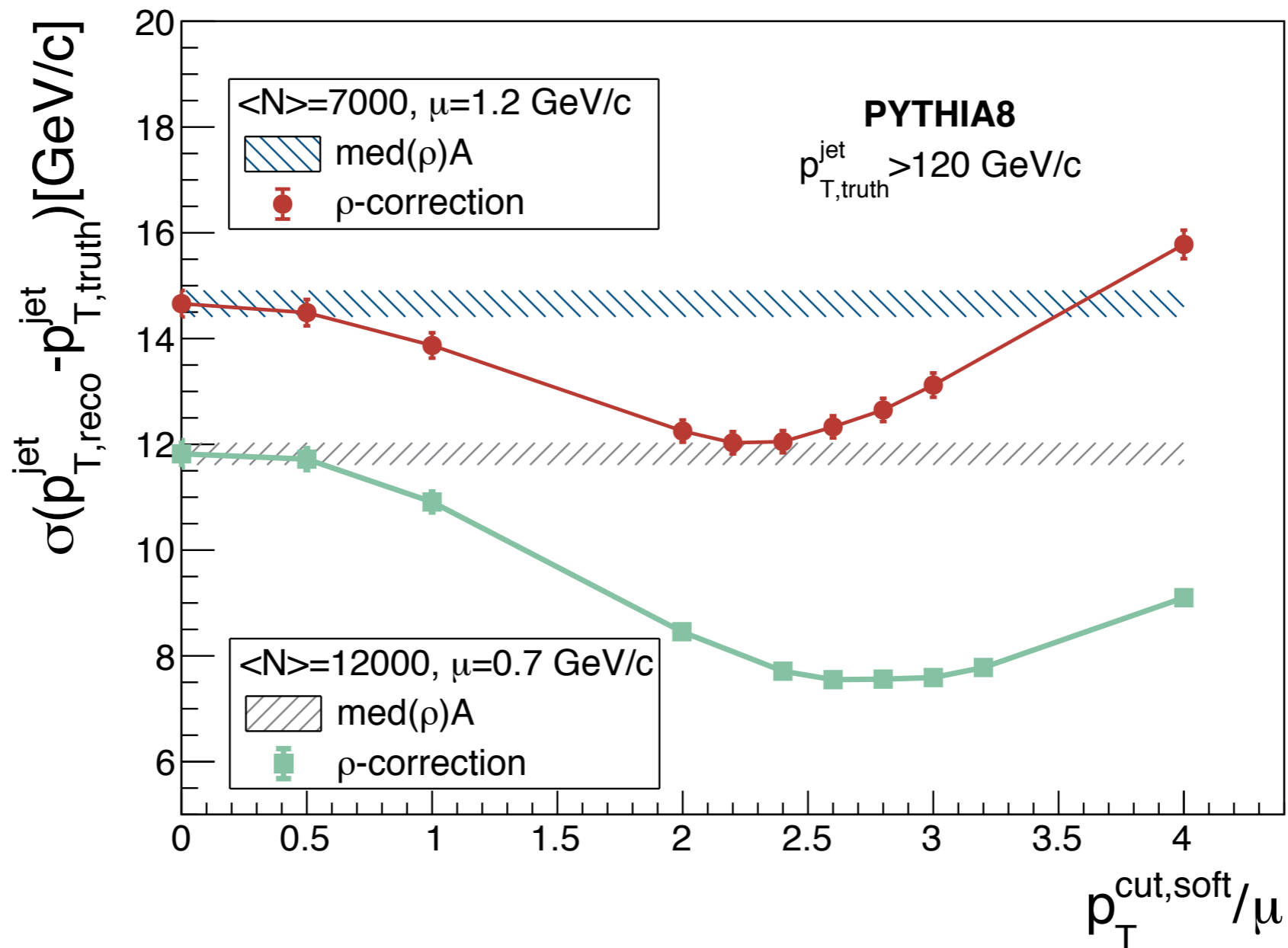
[Mehtar-Tani, ASO, Verweij arXiv:1904.12815]

- Prescription to choose $p_T^{\text{cut,soft}}$: minimize $\sigma(p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}})$



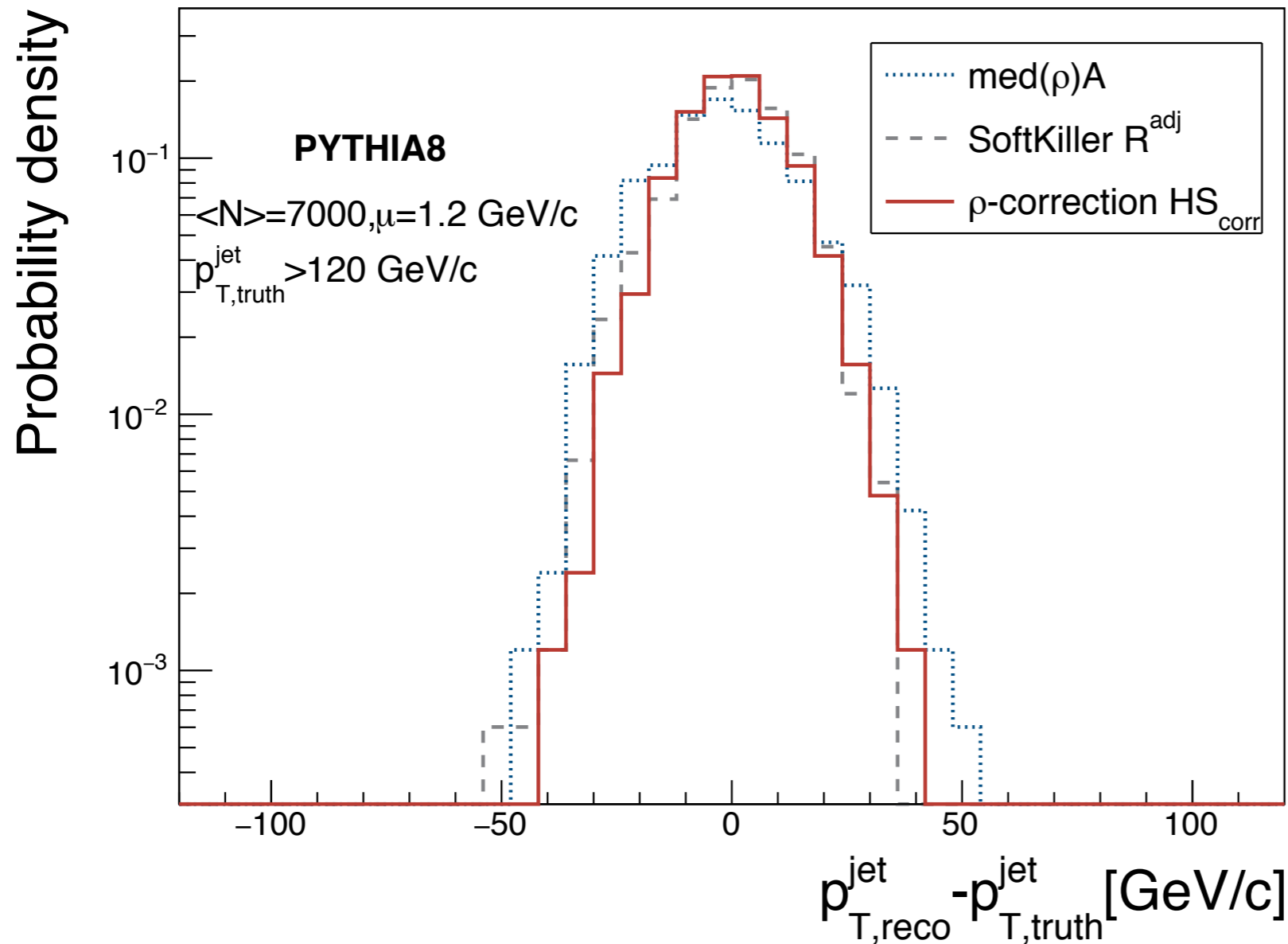
$$p_{T,\text{reco}}^{\text{jet}} = p_{T,\text{raw}}^{\text{jet}} - \text{med}(\rho_{>})A_{\text{raw}}^{\text{jet}} - p_{T,<} + \langle p_{T,<}^{\text{sig}} \rangle$$

PYTHIA8 + Thermal background



Plateau-like region around the optimal $p_T^{cut, soft}$.
No need to fine-tune

PYTHIA8 + Thermal background



ρ -correction HS_{corr}

- 23% improvement w.r.t area-median
- 6% improvement w.r.t SoftKiller R^{adj}

[GeV/c]	Mean	Standard deviation
med(ρ)A	-1.51 ± 0.36	14.66 ± 0.25
SoftKiller R^{adj}	0.29 ± 0.29	12.10 ± 0.21
ρ -correction	0.37 ± 0.31	12.03 ± 0.21
ρ -correction HS_{corr}	0.49 ± 0.28	11.39 ± 0.21

JEWEL w/ E-loss + Thermal background

Hard to know $p_{T,\text{truth}}^{\text{jet}}$ due to energy loss + cannot switch off the background

⇒ $p_T^{\text{cut,soft}}$ minimizes $\sigma(p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}})$ in JEWEL w/o E-loss

⇒ Apply $\langle p_{T,<}^{\text{sig}} \rangle$ as measured in low pileup p+p

[GeV/c]	Mean	Standard deviation	Δ_{truth}
med(ρ)A	-0.61 ± 0.17	14.73 ± 0.12	10.1
SoftKiller R^{adj}	2.90 ± 0.12	10.91 ± 0.09	7.55
ρ -correction	2.96 ± 0.12	10.92 ± 0.09	7.6
ρ -correction HS _{corr}	2.04 ± 0.12	10.34 ± 0.08	6.9

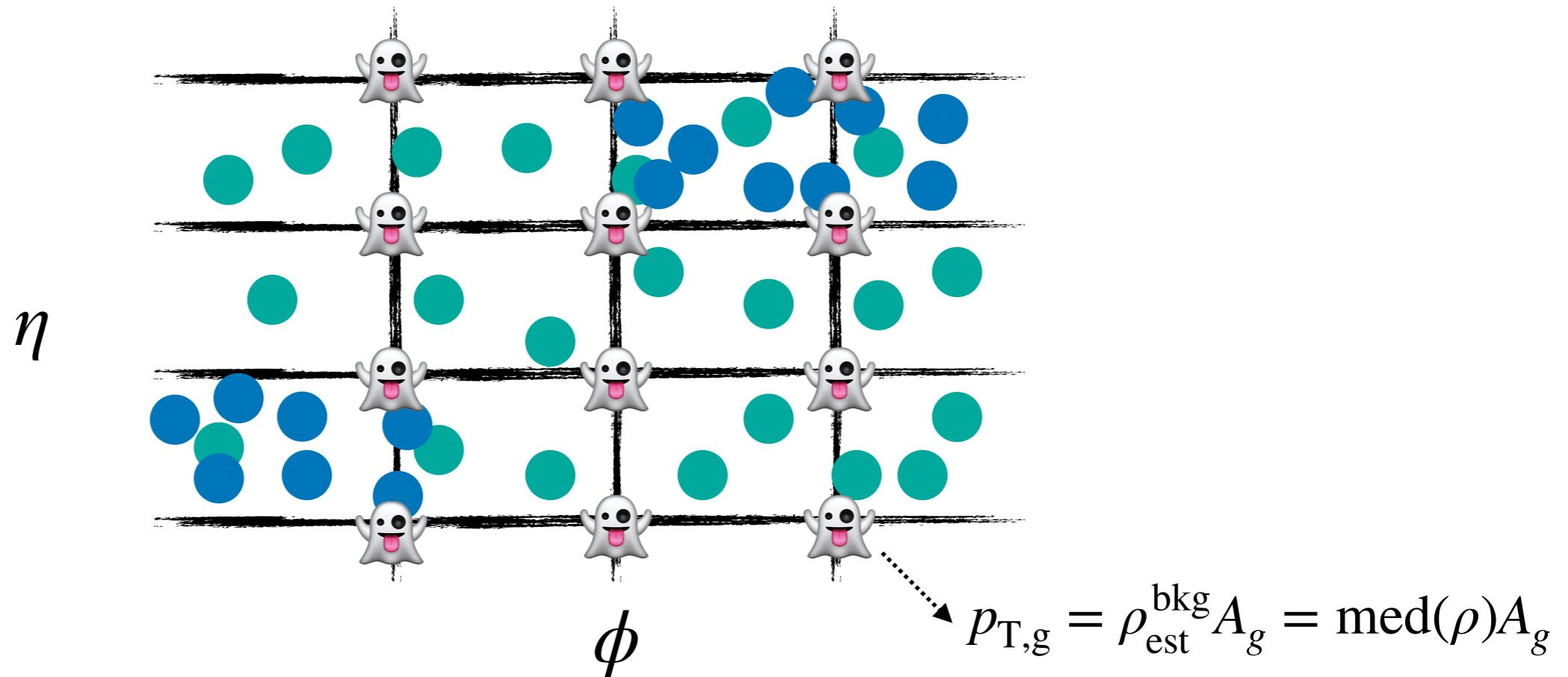
$[-\Delta_{\text{truth}}, \Delta_{\text{truth}}]$: interval in $p_{T,\text{reco}}^{\text{jet}} - p_{T,\text{truth}}^{\text{jet}}$ centered around 0 and containing 1/2 jets

Better resolution but not straightforward to control the bias due to in-medium modifications

SUBTRACTION METHOD

Constituent subtraction

[P. Berta, M. Spousta, D. Miller and R. Leitner JHEP 1406 (2014) 092]



1) For each jet, compute all the ghost-to-particle distances

$$\Delta R_{i,k} = p_{T,i}^\alpha \sqrt{(y_i - y_k^g)^2 + (\phi_i - \phi_k^g)^2} \quad \text{and sort them from smallest to largest.}$$

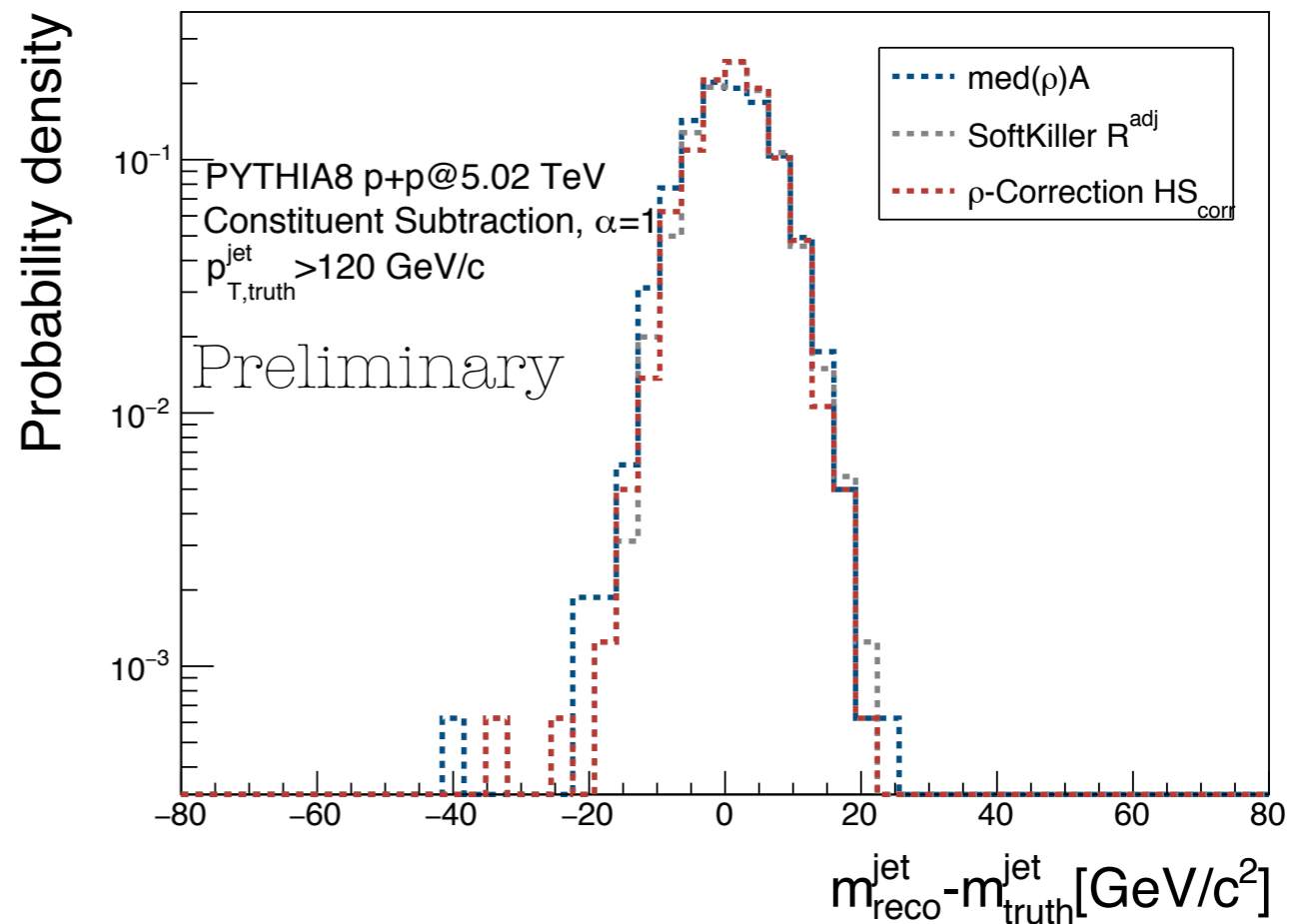
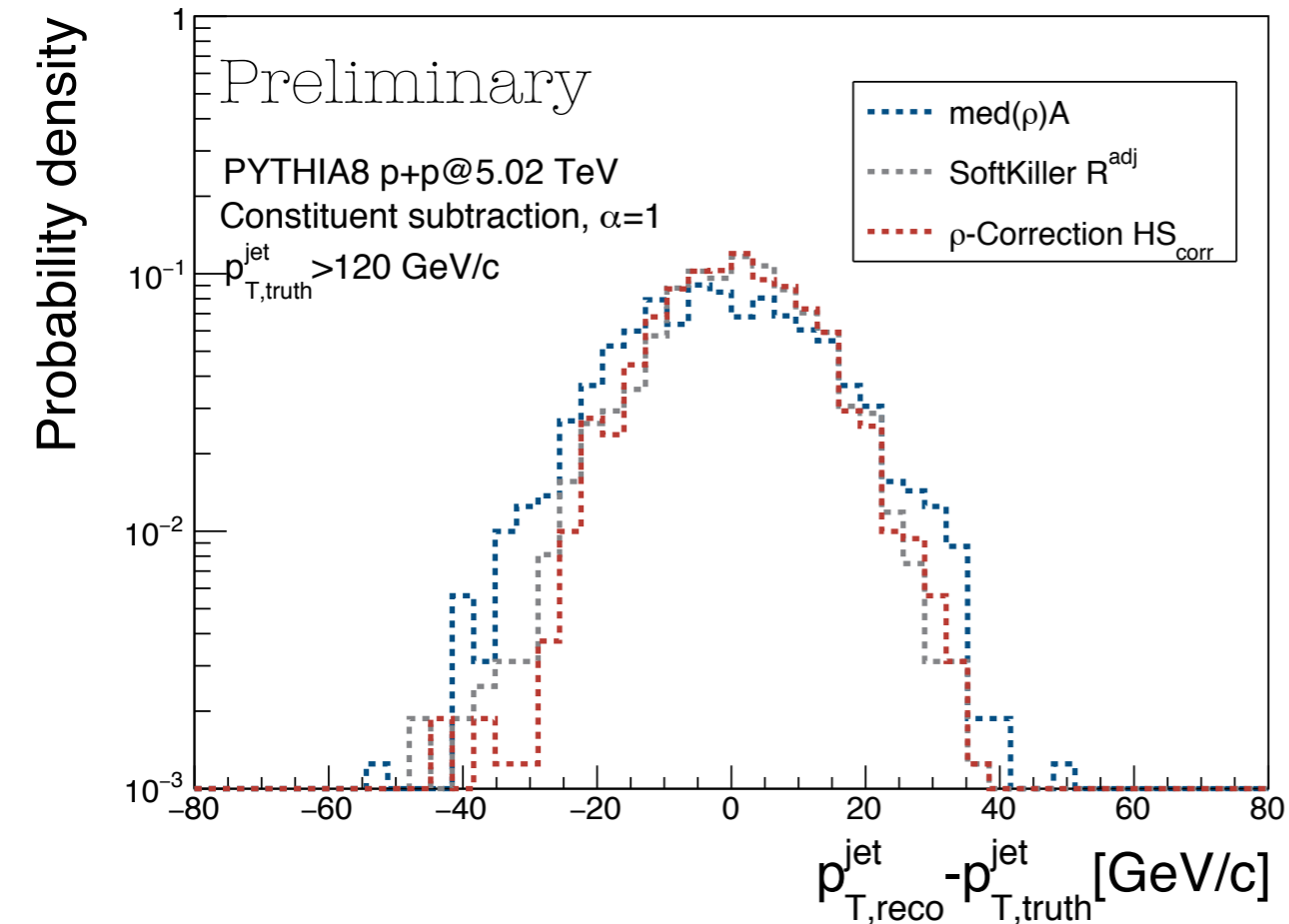
- If $p_{T,i} > p_{T,k}^g$: $p_{T,i}^- = p_{T,k}^g$ && $p_{T,k}^g = 0$

2) For each pair,

- If $p_{T,i} < p_{T,k}^g$: $p_{T,i} = 0$ && $p_{T,k}^g - = p_{T,i}$

3) Remove every particle/ghost with $p_T = 0$

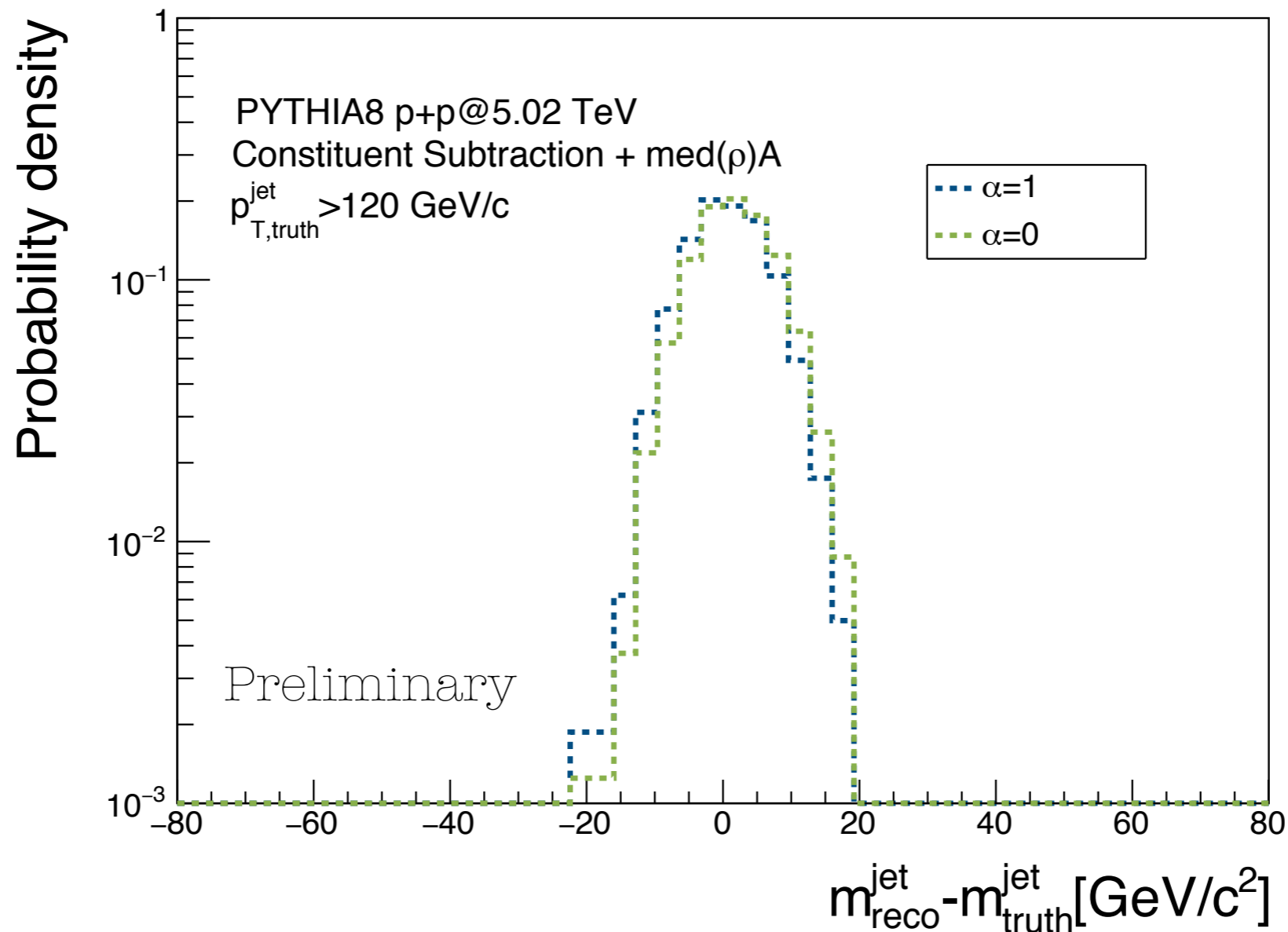
Constituent subtraction beyond area-median



Improved mass and pT resolutions just by having
a more precise estimate for p_T^{bkg}

Constituent subtraction: the role of α

$$\Delta R_{i,k} = p_{T,i}^{\alpha} \sqrt{(y_i - y_k^g)^2 + (\phi_i - \phi_k^g)^2}$$

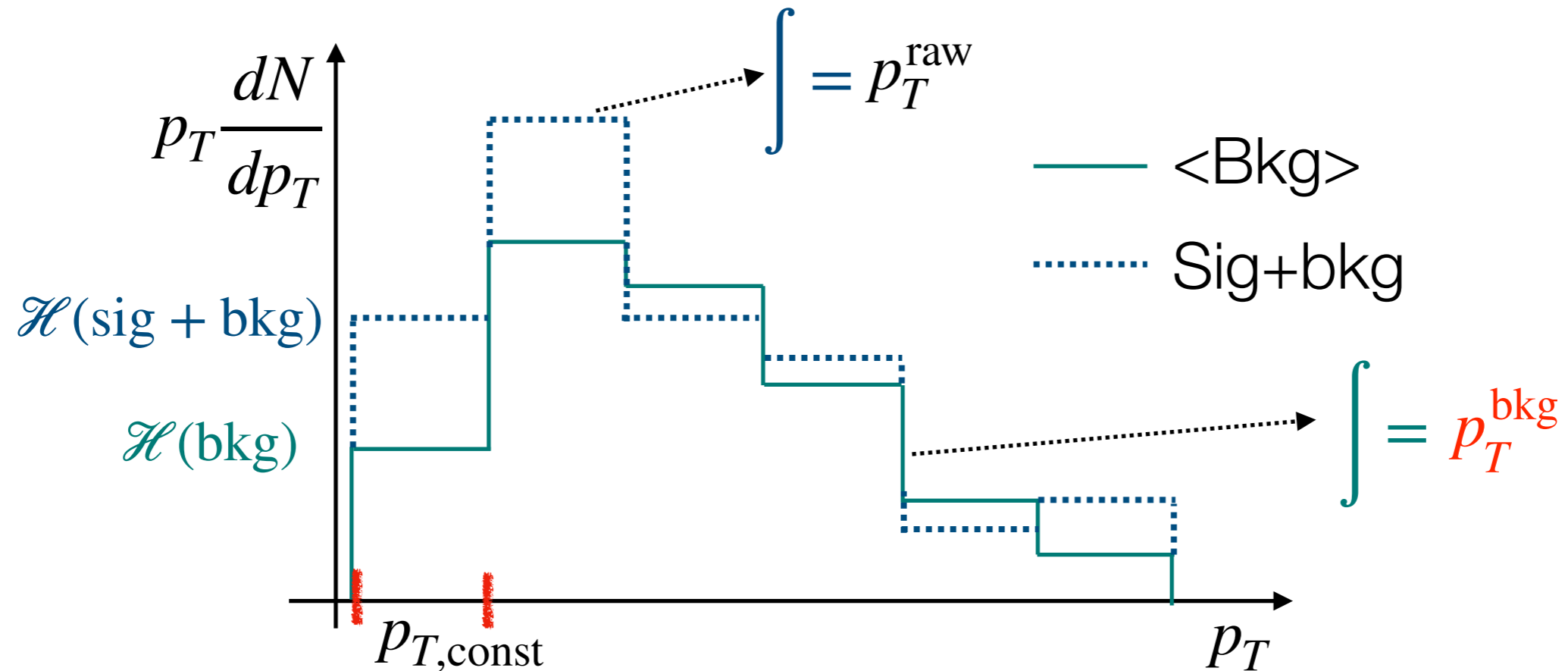


$\Delta m_{\alpha=0 \rightarrow \alpha=1} \sim 1 \text{ GeV}/c^2$ as due to the lack of pT information

pTSampler Subtractor

[Mehtar-Tani, ASO, Verweij, In preparation]

Exploit that background and signal particles have distinct angular and momentum distributions

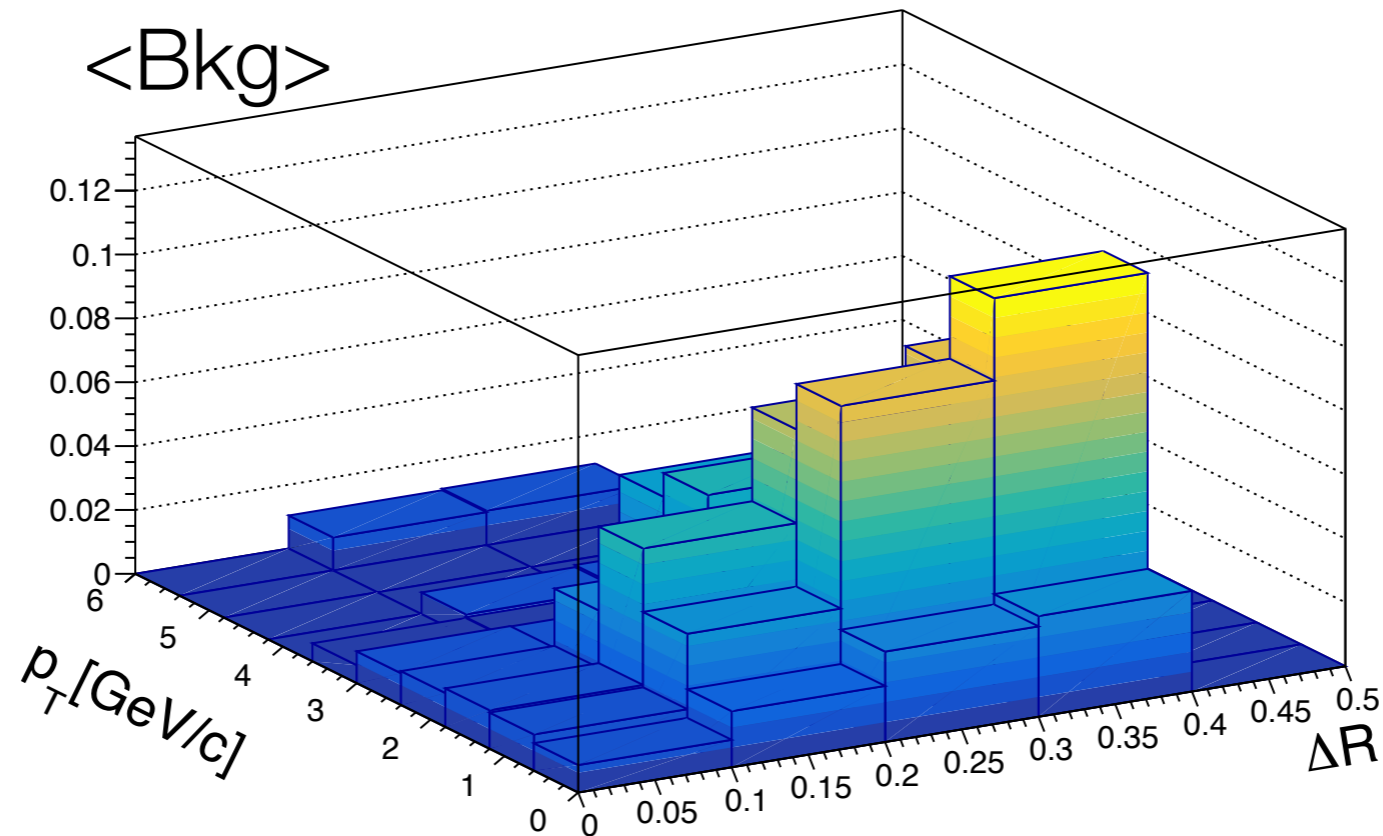


$$\mathcal{P}(p_i \in \text{bkg}) = \frac{\mathcal{H}_{\text{part}}(\text{bkg})}{\mathcal{R}} \quad \text{with} \quad \mathcal{R} \in (0, \mathcal{H}(\text{sig} + \text{bkg}))$$

pTSampler Subtractor

[Mehtar-Tani, ASO, Verweij, In preparation]

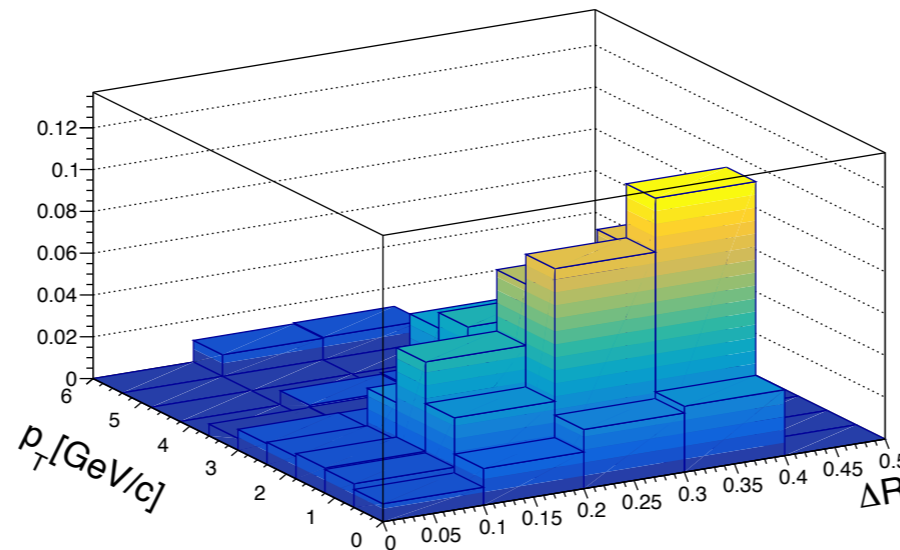
Exploit that background and signal particles have distinct angular and momentum distributions



$$\mathcal{P}(p_i \in \text{bkg}) = \frac{\mathcal{H}_{\text{part}}(\text{bkg})}{\mathcal{R}} \quad \text{with} \quad \mathcal{R} \in (0, \mathcal{H}(\text{sig} + \text{bkg}))$$

1 Cluster all particles (sig+bkg) in the event with anti- k_T $R=0.4$

2 For each jet, generate 5 cones with $\eta = -\eta_{\text{jet}}, R = 0.4, \phi \in (0, 2\pi)$

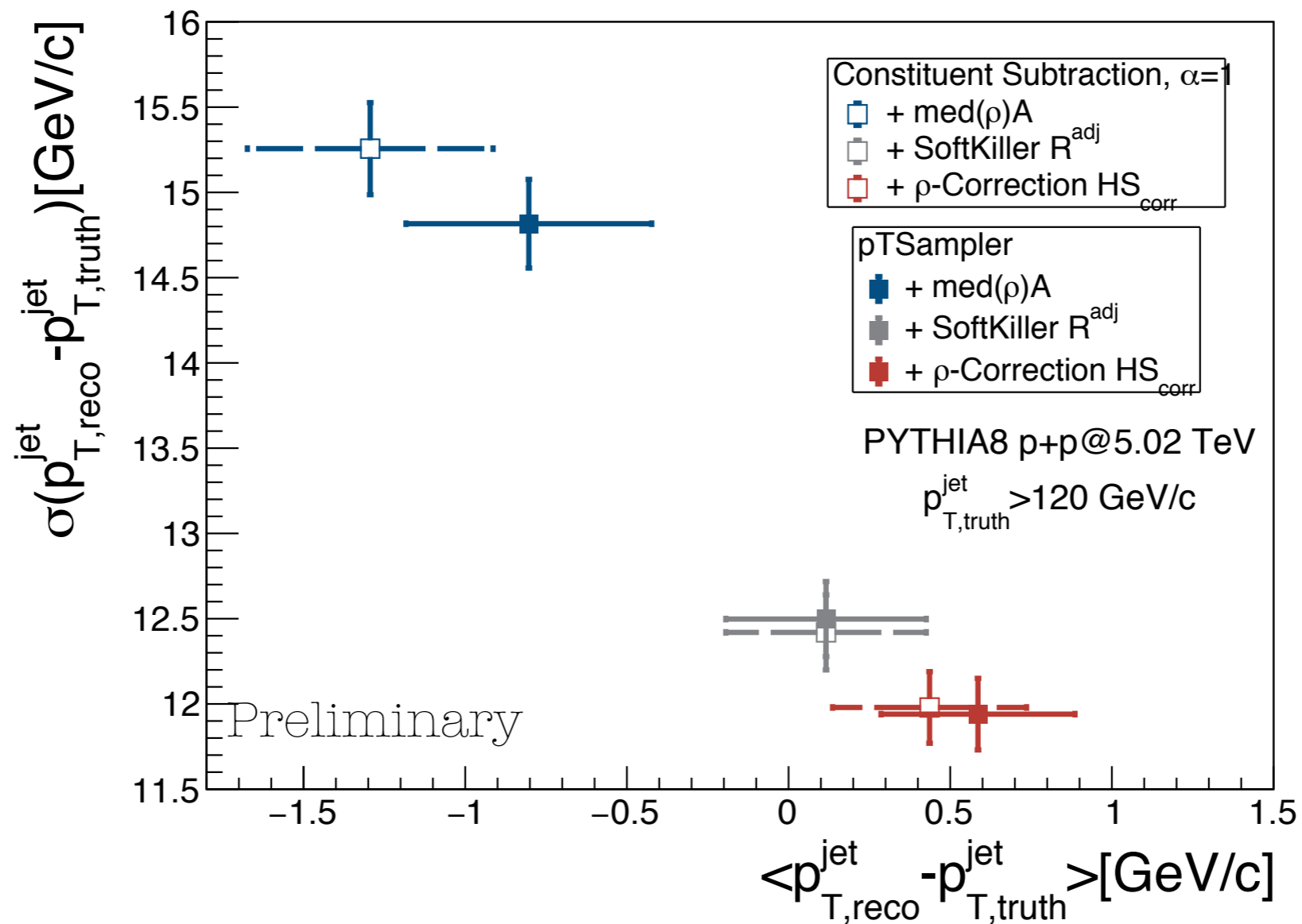


3 For each particle in the jet obtain its $\mathcal{P}(p_i \in \text{bkg})$

4 Sort the list of particles from largest to smallest probability

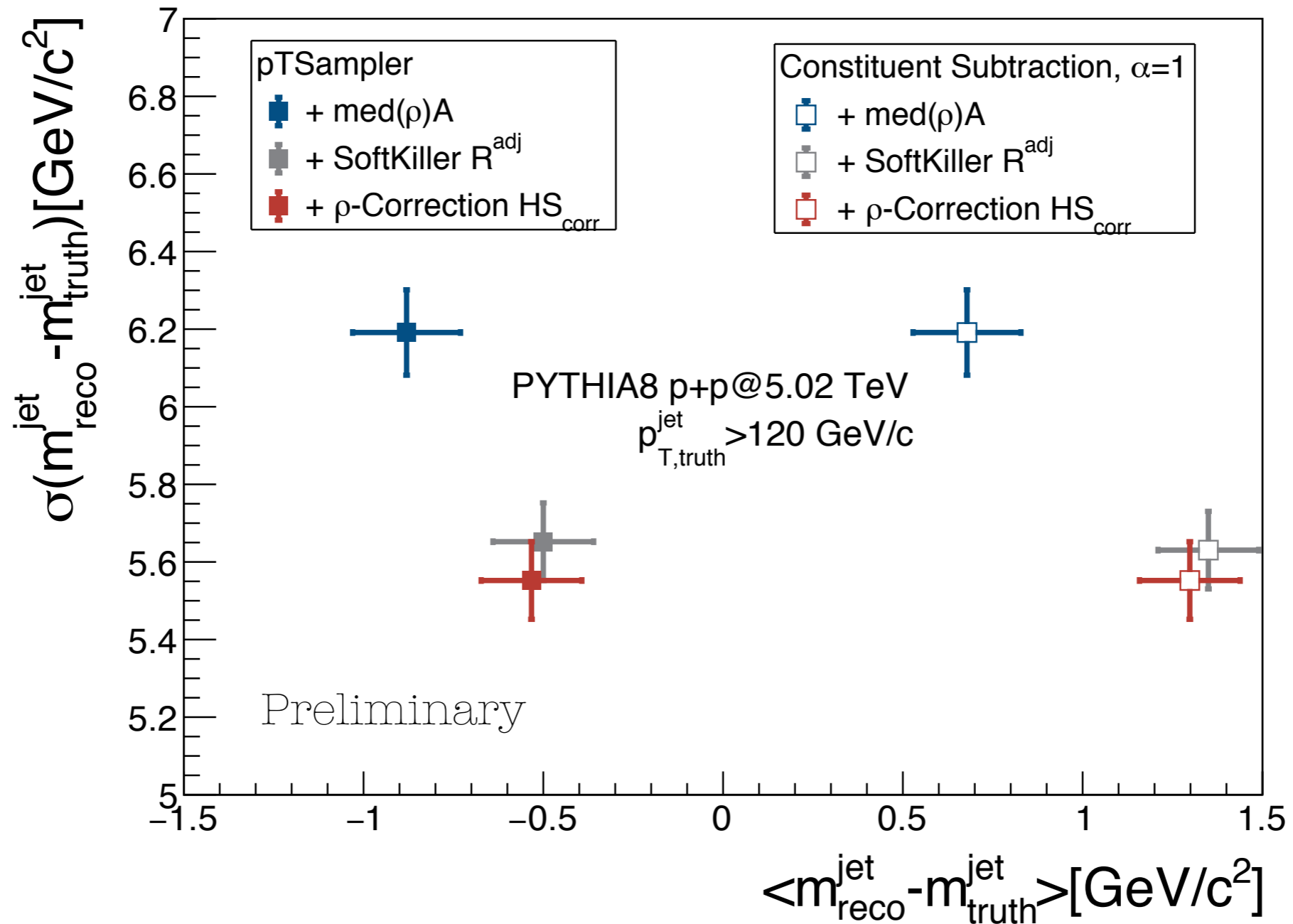
5 Add particles to the background list until $p_T^{\text{bkg}} > p_{T,\text{est}}^{\text{bkg}}$

pT reconstruction



CS/pTSampler + rhoCorrection results into a 20% improvement on the pT resolution

Mass reconstruction



pTSampler leads to an improvement of $\sim 1 \text{ GeV}$ on the jet mass response wrt to CS when using rhoCorrection

Wrap-up

- Jets are instrumental both for BSM searches and to bolster our knowledge about in-medium modifications and QGP
- Overwhelming and fluctuation background is a showstopper for precision physics
- Very sophisticated background subtraction methods on the market
- Our goal: obtain similar performance in a simpler fashion thanks to QCD

ρ -correction+pTSampler

Keep tuned and find more on: <https://github.com/aontososo/JetToyHI>

Soon in FastJet/contrib