

# A Novel Reclustering Algorithm for Jet Quenching

André Cordeiro

Based on: "Time reclustering for jet quenching studies"  
(L. Apolinário, A. Cordeiro, K. Zapp) Eur. Phys. J. C 81, 561 (2021)

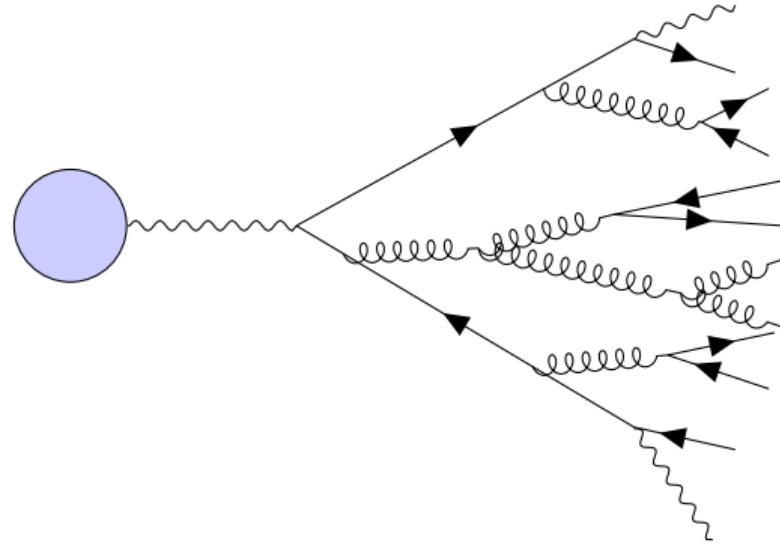
5 September, 2021



# QGP AND PARTON SHOWERS

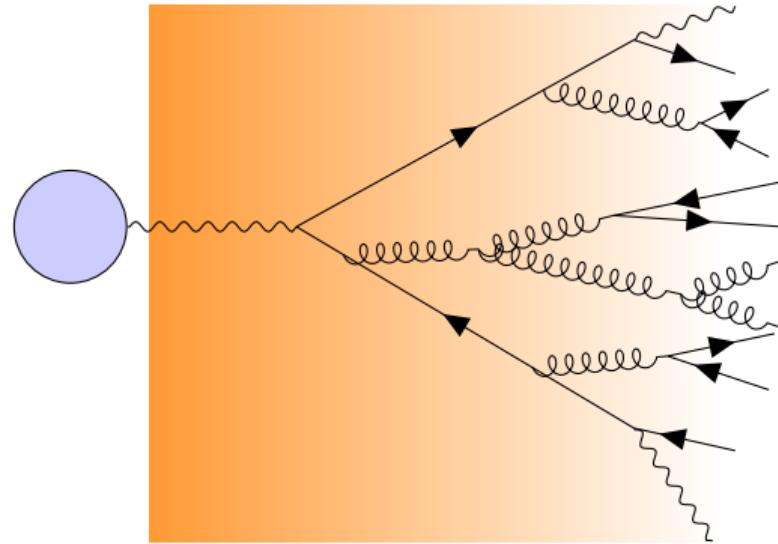
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- QCD shower as a series of emissions over a wide range of scales.
- One can construct jets — Multi-scale observables.



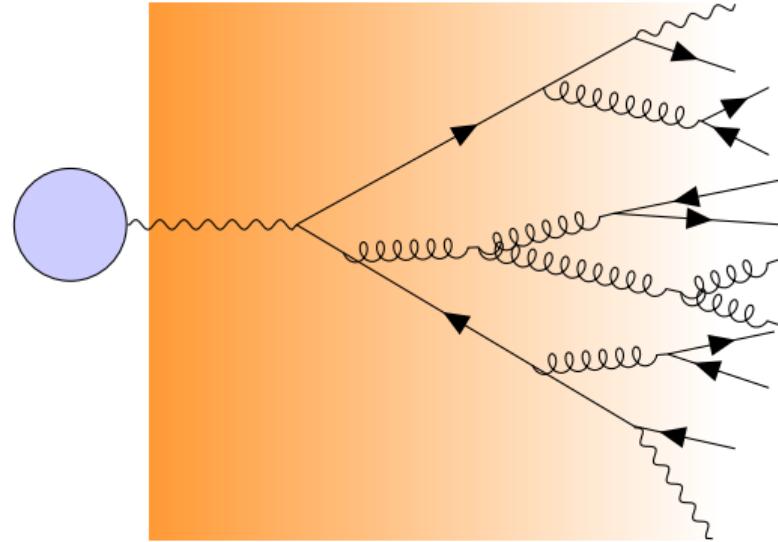
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**Probing jet substructure may allow for precision measurements of QGP structure.**

# JET RECONSTRUCTION AND SUBSTRUCTURE

# Jet Clustering

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$$d_{ij} = \min(p_{T,i}^{2p}, p_{T,j}^{2p}) \left( \frac{\Delta R_{ij}}{R} \right)^2$$

$p = 0$  : Cambridge – Aachen

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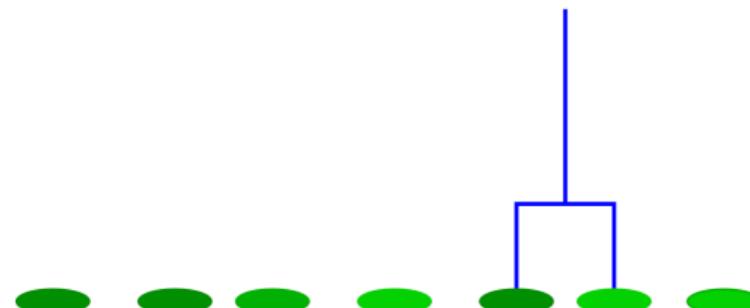
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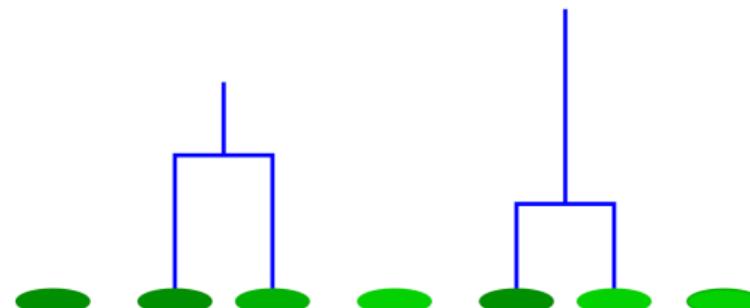
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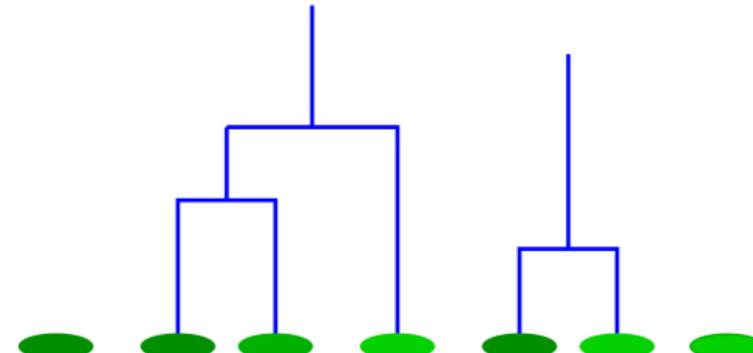
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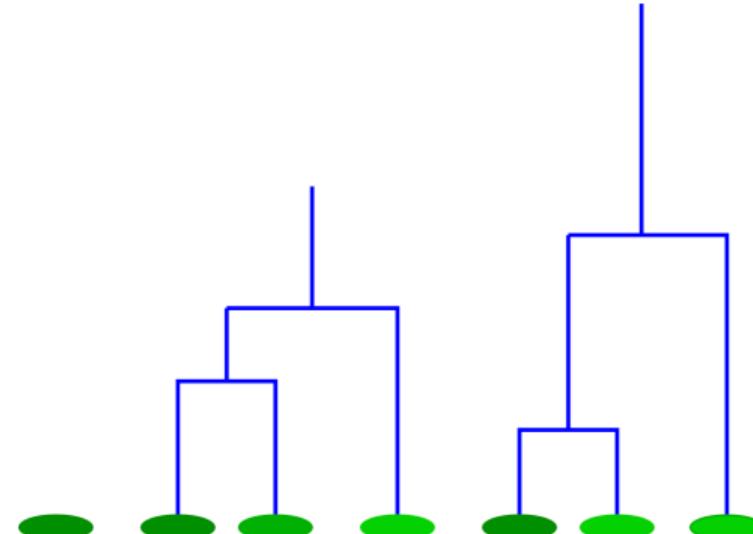
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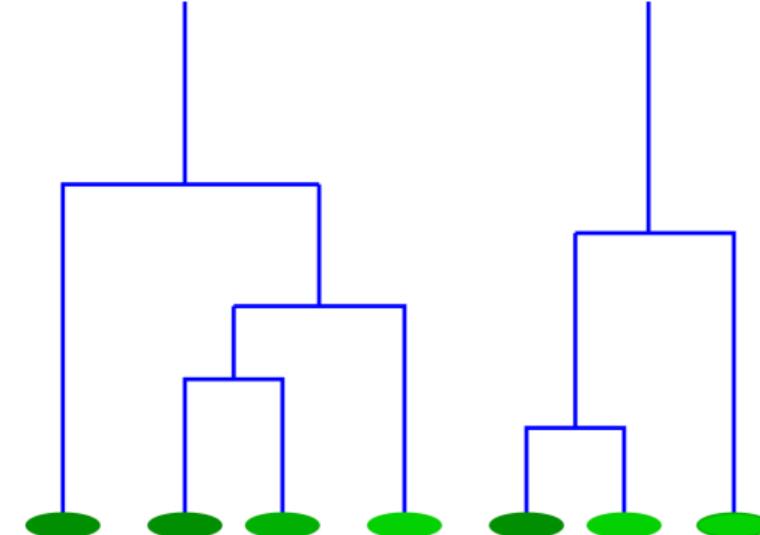
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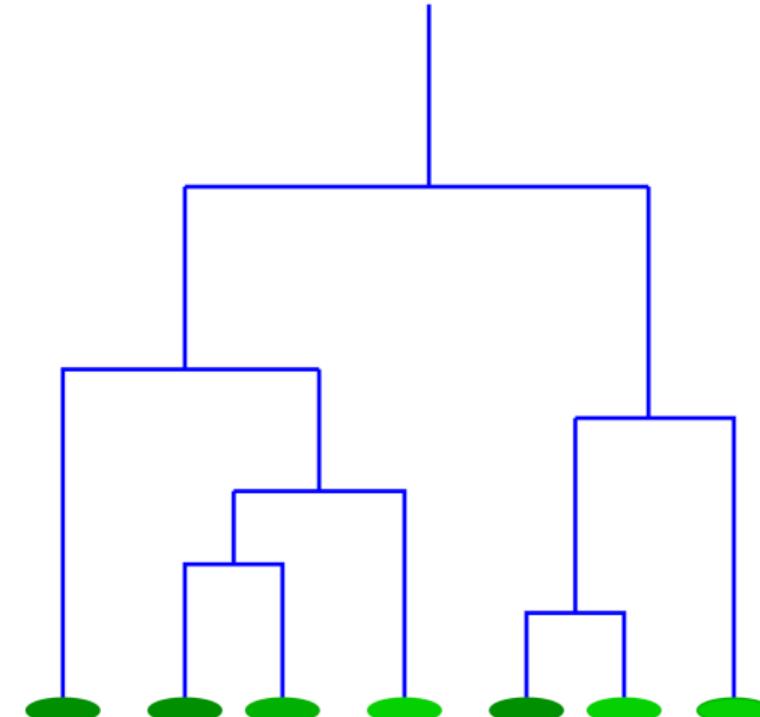
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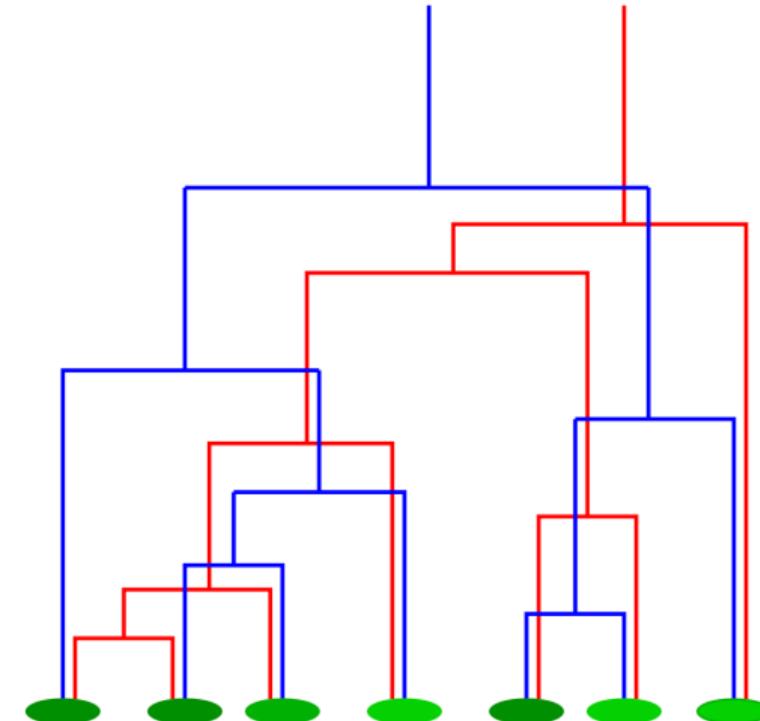
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# Estimating Time Scales

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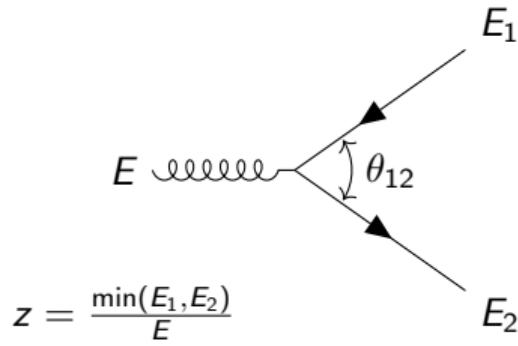
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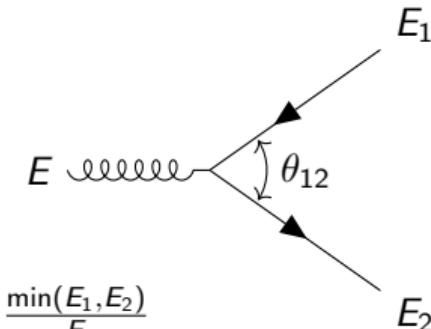


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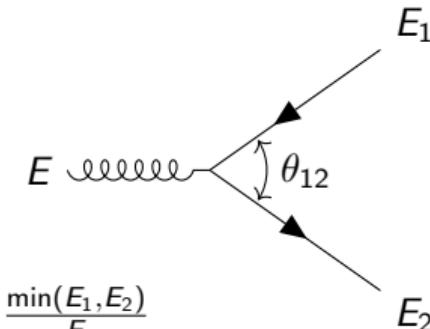
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**From now on called the  $\tau$  algorithm, can be used for jet reclustering.**

# MONTE CARLO ANALYSIS

# Monte Carlo Event Generators

## Initialization and Analysis

Sampled jet populations using vacuum references (PYTHIA8 + JEWEL Vac) and jet quenching (JEWEL w/ rec and JEWEL w/o rec).

Different orderings crucial in evaluating algorithm performance:  $k_T$ -ordering (PYTHIA8),  $Q^2$  ordering (JEWEL vacuum),  $Q^2$  ordering +  $\tau_{\text{form}}$  veto (JEWEL medium).

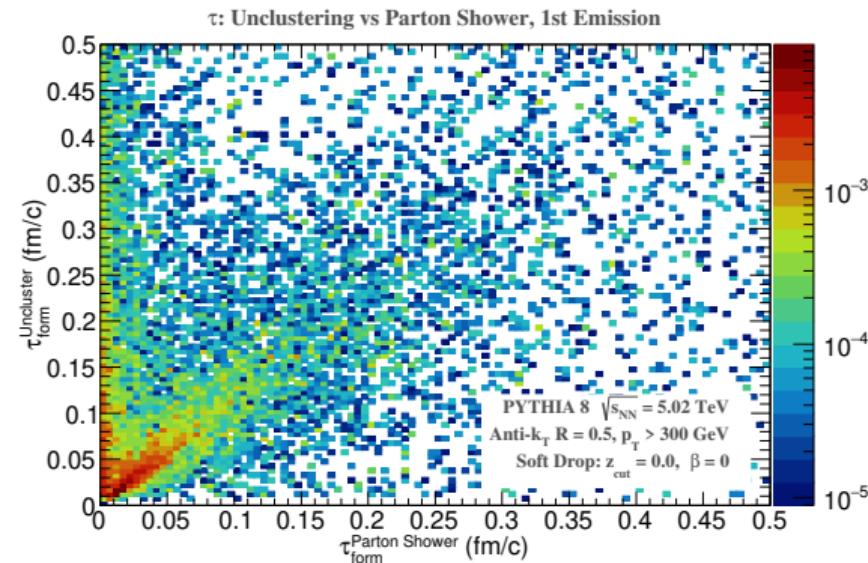
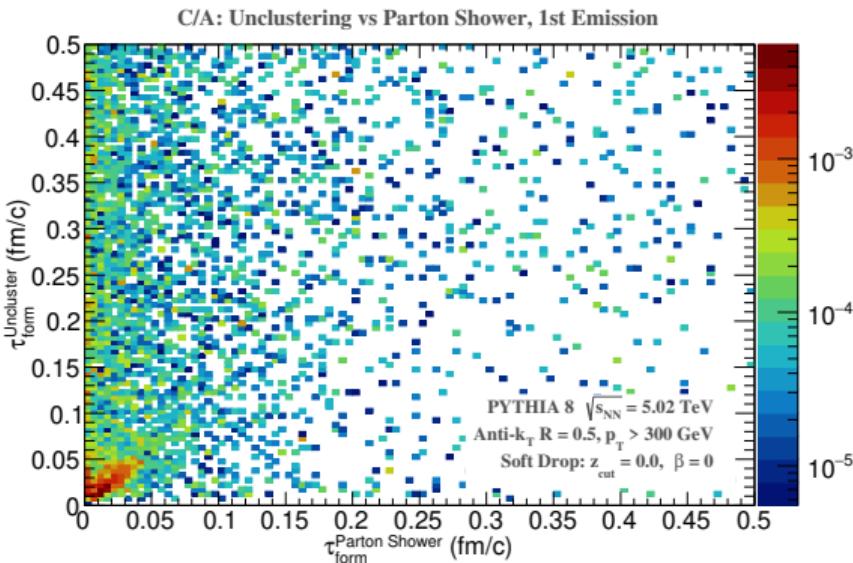
Hard scattering tuned for dijet events at  $\sqrt{s_{\text{NN}}} = 5.02 \text{ TeV}$ , trigger on leading jet.

Analysis of jets with  $p_T > 300 \text{ GeV}$  and  $|\eta| < 1.0$ .

Quenching effects achieved by a medium model + Bjorken expansion, with  $\tau_{\text{init}} = 0.4 \text{ fm/c}$  and  $T_{\text{init}} = 0.44 \text{ GeV}$ .

# Jet substructure & Parton Shower history

Correlation of 1<sup>st</sup> splitting time

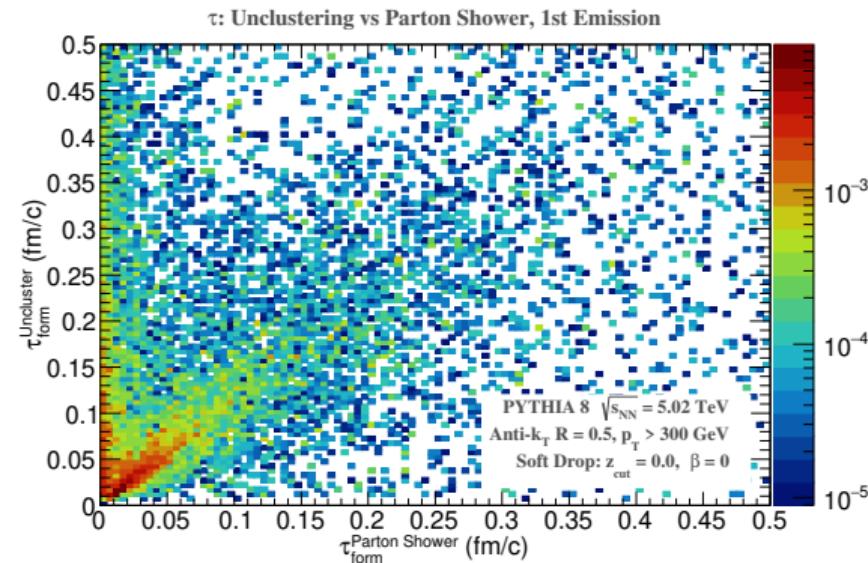
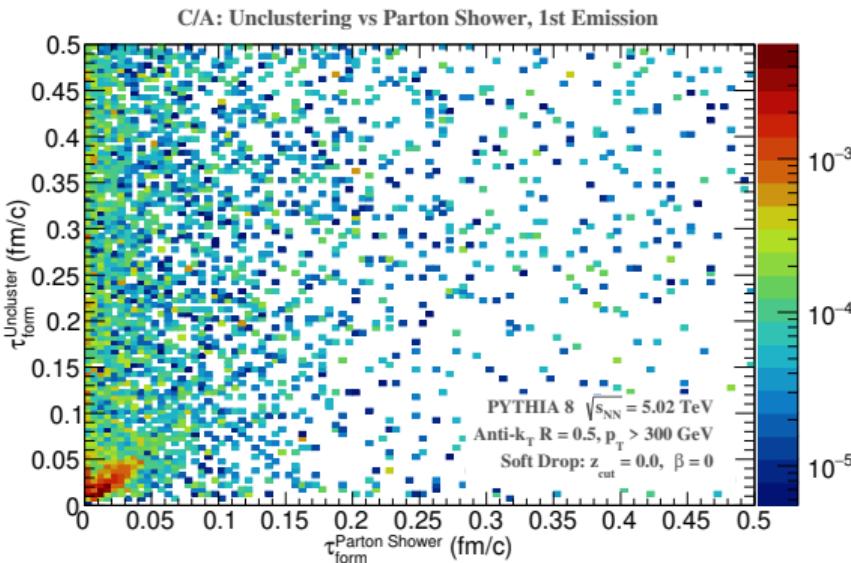


Look at correlations between proxies for  $\tau_{\text{form}}$ :

- $\tau_{\text{form}}^{\text{Uncluster}}$ : Jet Variable.
- $\tau_{\text{form}}^{\text{PartonShower}}$ : Monte Carlo Variable.

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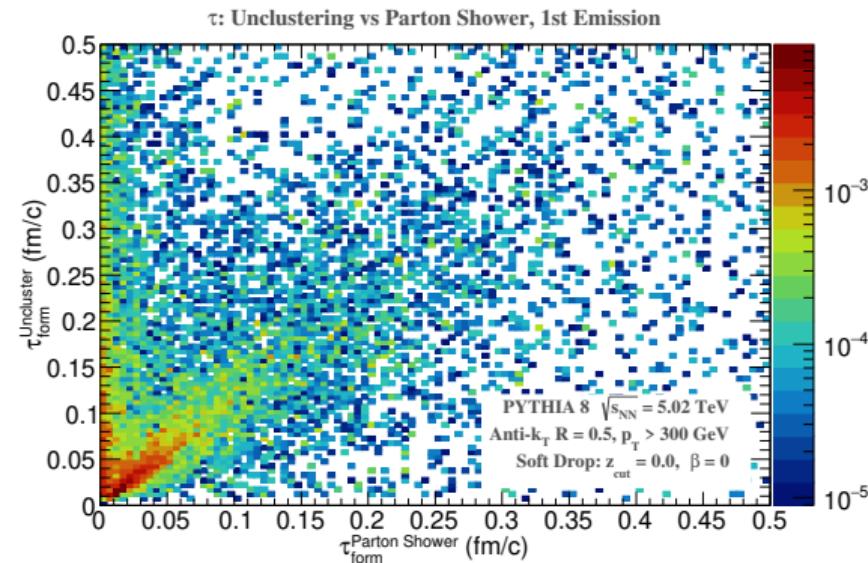
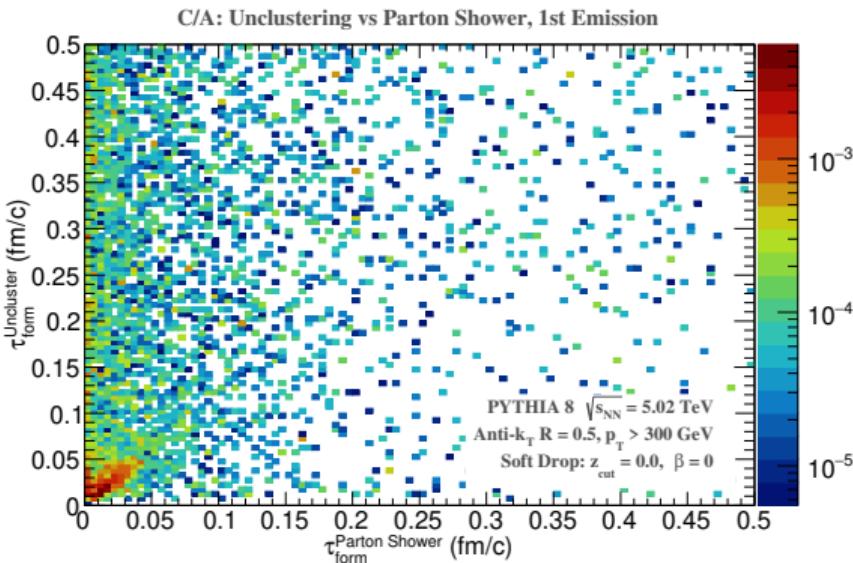
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- **Diagonal:** Correlation.
- **Vertical:** Emissions outside of jet cone.

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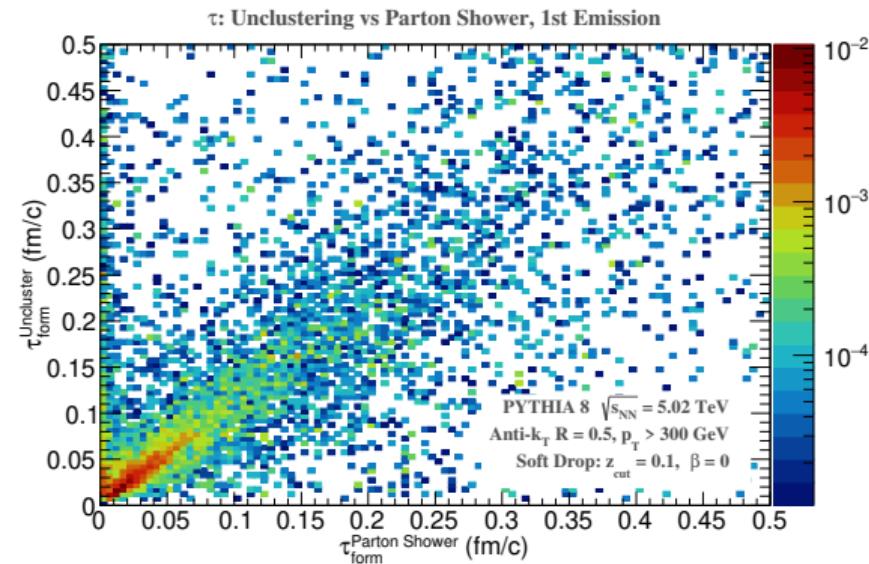
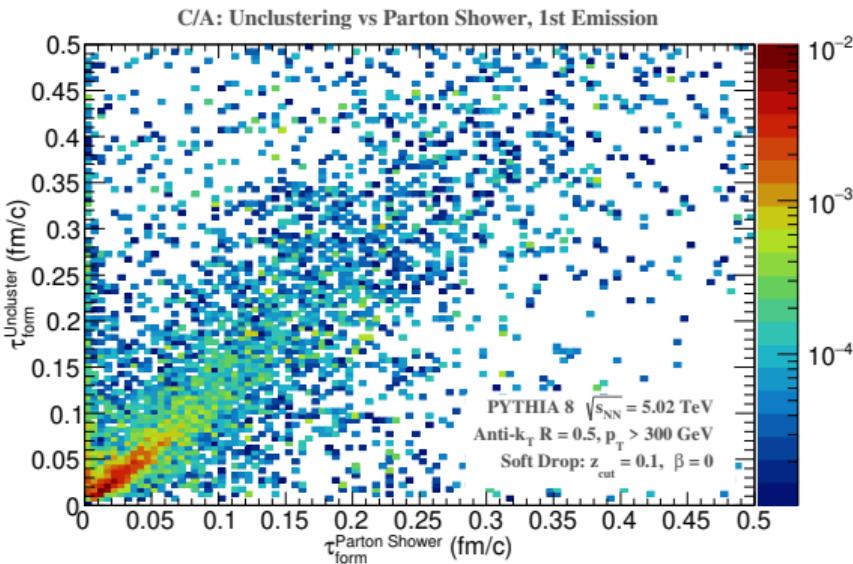
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**Apply SoftDrop:** 
$$z_g = \frac{\min(p_{T1}, p_{T2})}{p_{T1} + p_{T2}} > z_{\text{cut}}, \text{ with } z_{\text{cut}} = 0.1.$$

# Jet substructure & Parton Shower history

Correlation of 1<sup>st</sup> splitting time — Groomed Jets

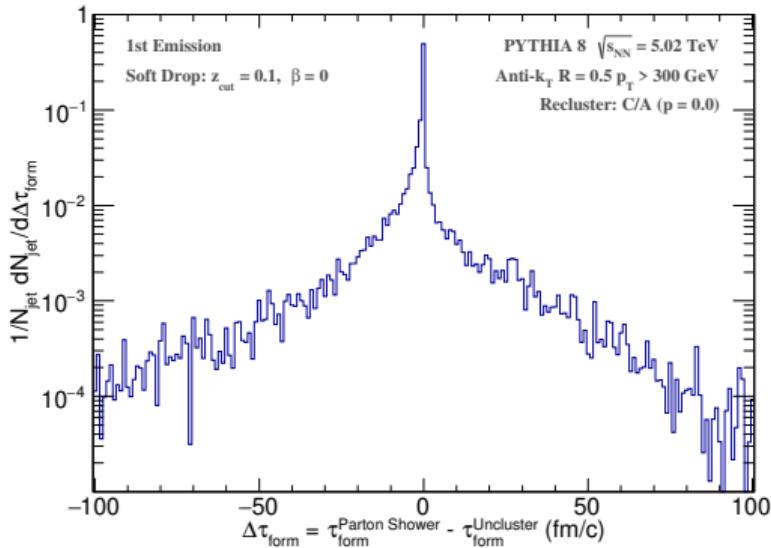


**Grooming improves considerably the obtained correlation.**

# Comparing the Generalised- $k_T$ Algorithms

## Vacuum Distributions

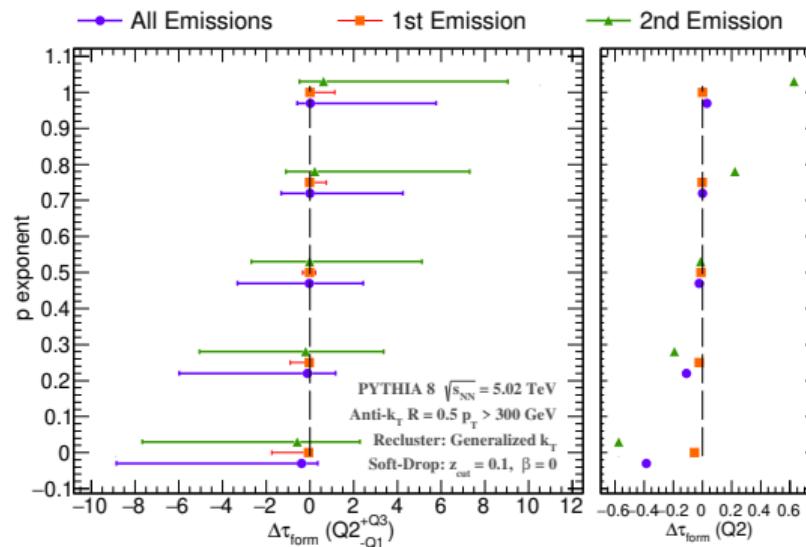
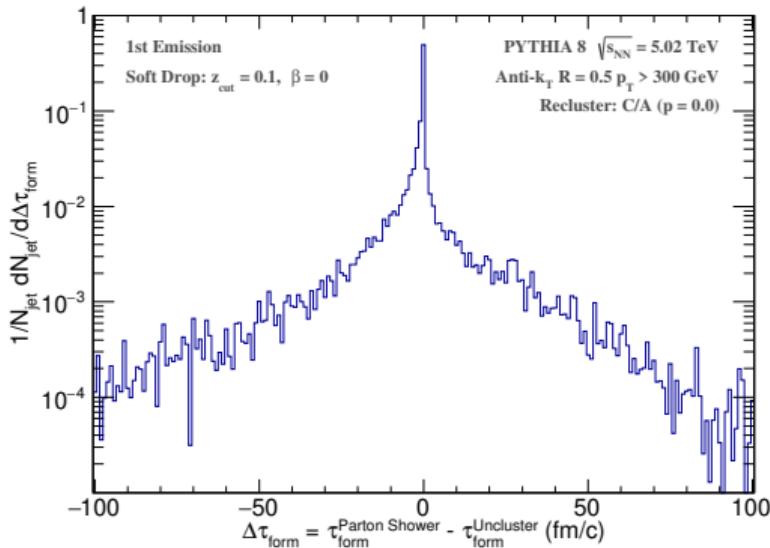
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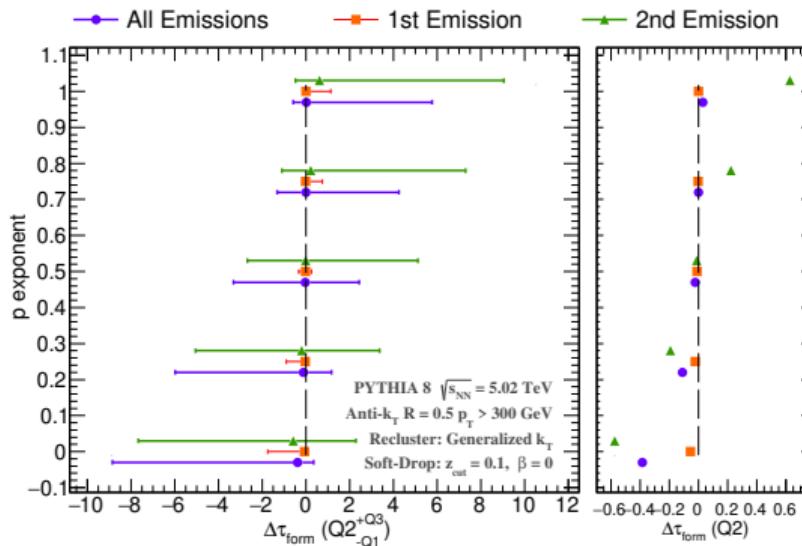
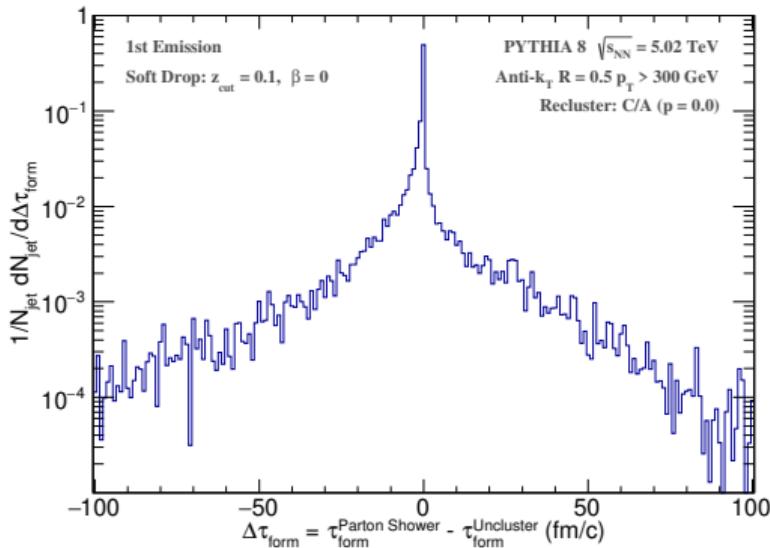


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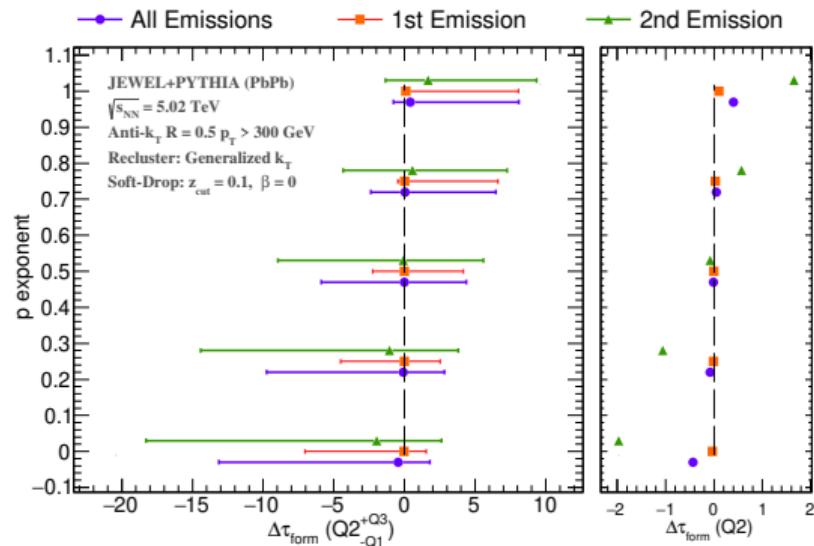
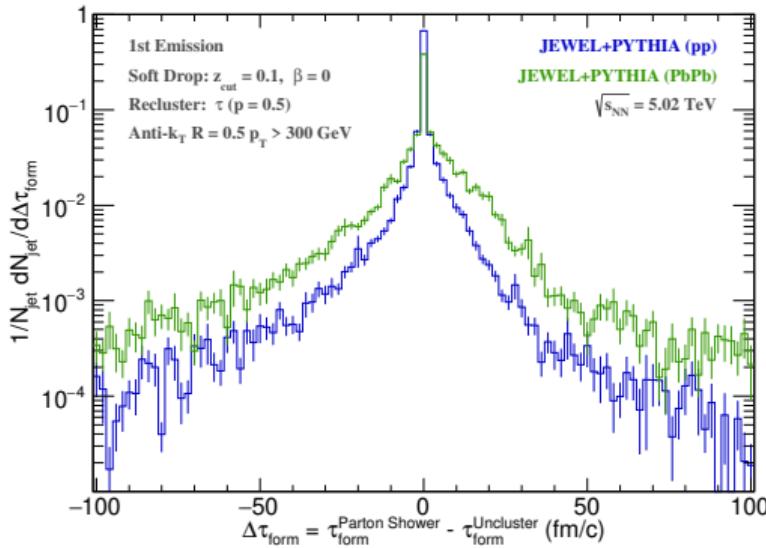
- Represent distribution by *Median  $\pm$  Quartiles*

The  $\tau$  algorithm returns the most centred distributions for PYTHIA8.

# Comparing the Generalised- $k_T$ Algorithms

## Medium Distributions

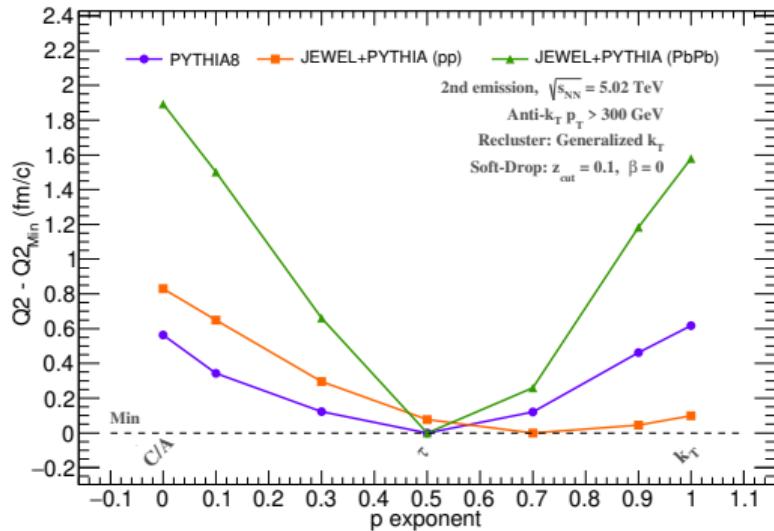
- For medium-modified jets the distributions widen, but the trend remains.



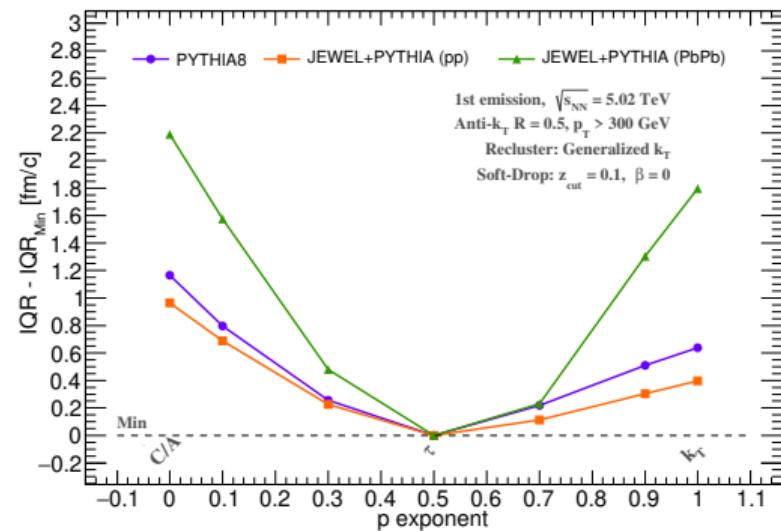
# Comparing Monte-Carlo Event Generators

Median and IQR dependence on the algorithm

## Distribution Median for the 2<sup>nd</sup> emission



## Distribution IQR for the 1<sup>st</sup> emission

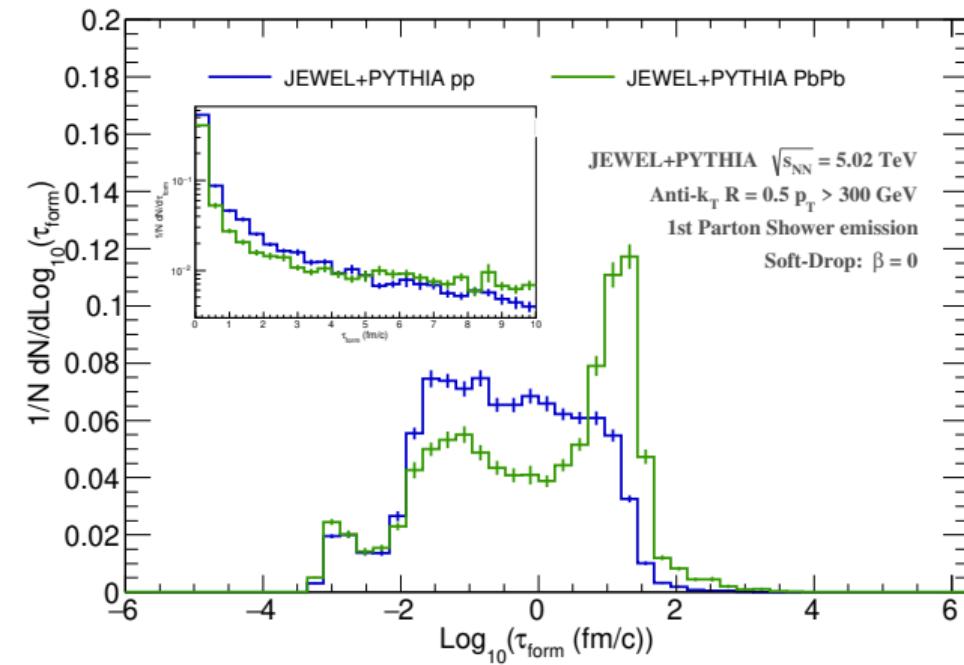


- Generally, the  $\tau$  algorithm minimizes both the median and the width of the  $\Delta\tau_{\text{form}}$  distribution.

# APPLICATIONS TO JET QUENCHING STUDIES

# Jet Quenching by the QGP

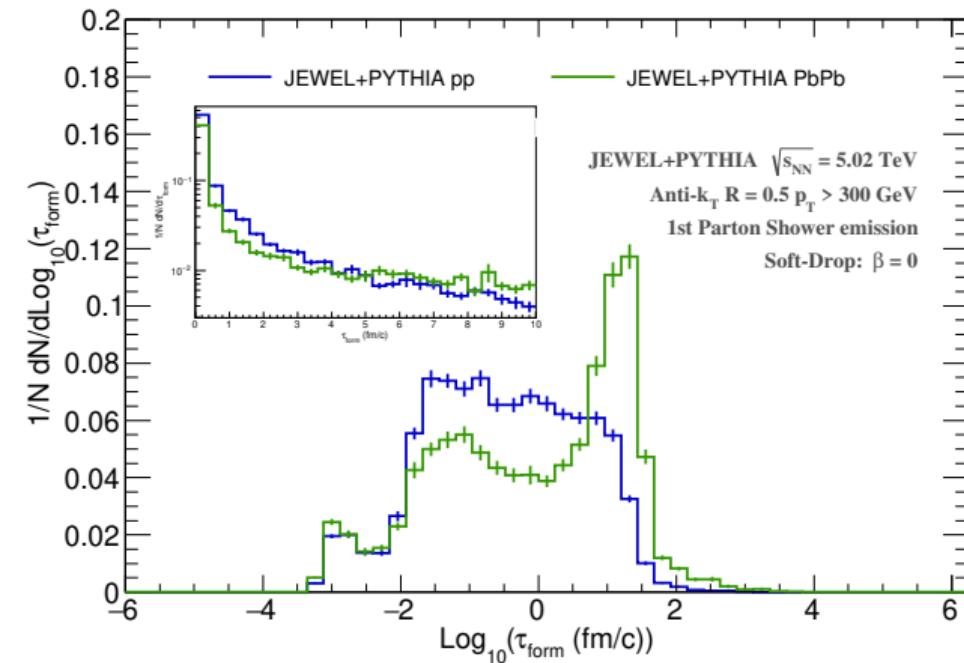
## Jet Quenching Classifier



- Longer  $\tau_{\text{form}} \longleftrightarrow$  Harder fragmentation.
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- Jets with harder fragmentation more likely to stay in the  $p_T > 300 \text{ GeV}$  region.
- Can select different jets using the  $\tau_{\text{form}}$  variable.

# Jet Quenching by the QGP

## Nuclear Modification Factor

To study shower interactions with the QGP, we introduce the nuclear modification factor,

$$R_{AA}(p_T) = \frac{N_{\text{evt}}^{pp}}{N_{\text{evt}}^{AA}} \frac{dN_{\text{jet}}^{AA}/dp_T}{dN_{\text{jet}}^{pp}/dp_T},$$

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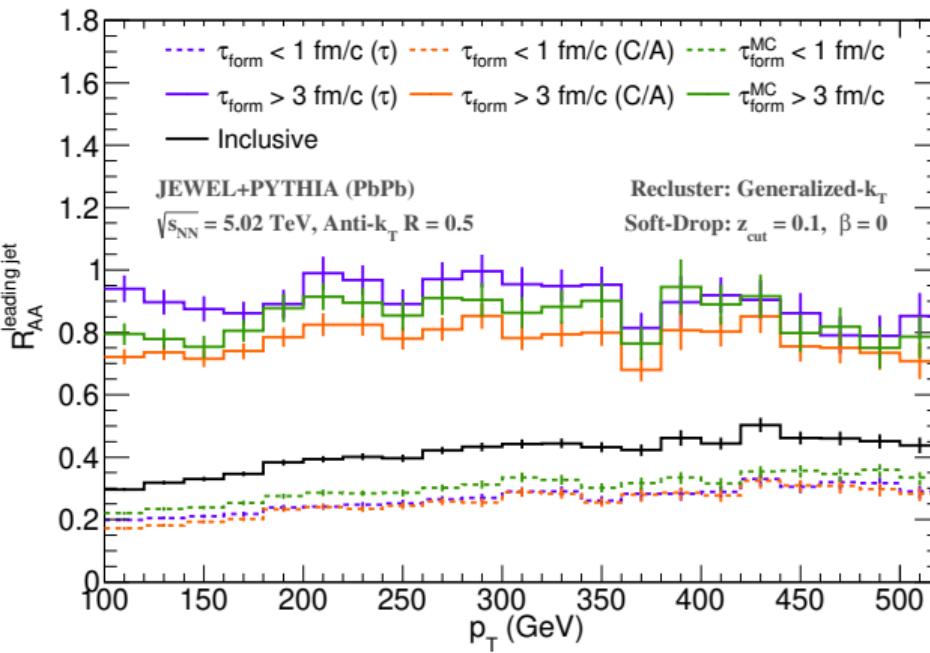
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Compute  $R_{AA}$  for different jet populations:

- “Early jets” —  $\tau_{\text{form}} < 1 \text{ fm/c}$
- “Late jets” —  $\tau_{\text{form}} > 3 \text{ fm/c}$

# Jet Quenching by the QGP

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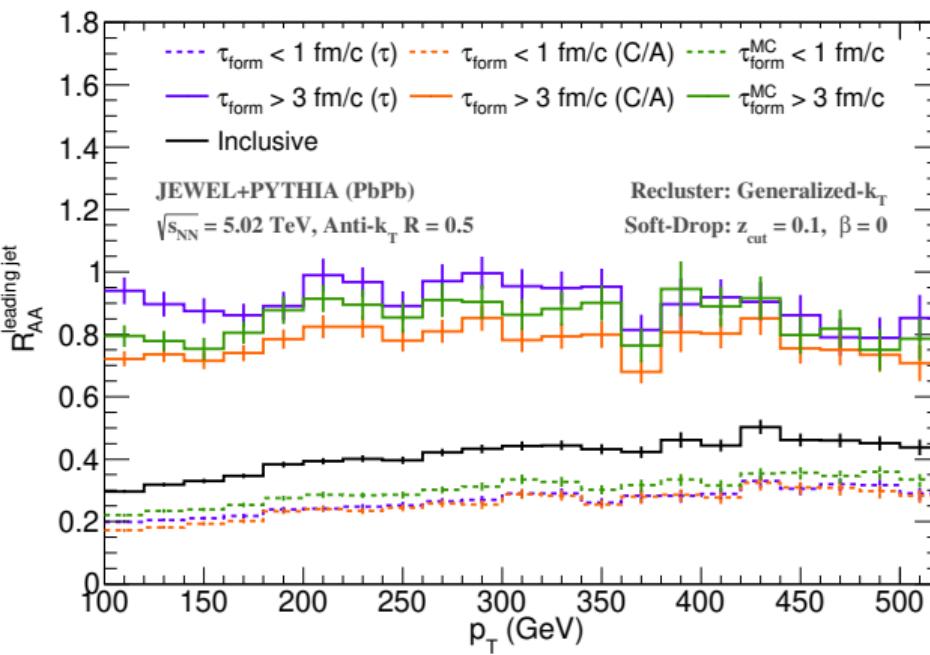


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← →  
Untested region, correlation might deteriorate

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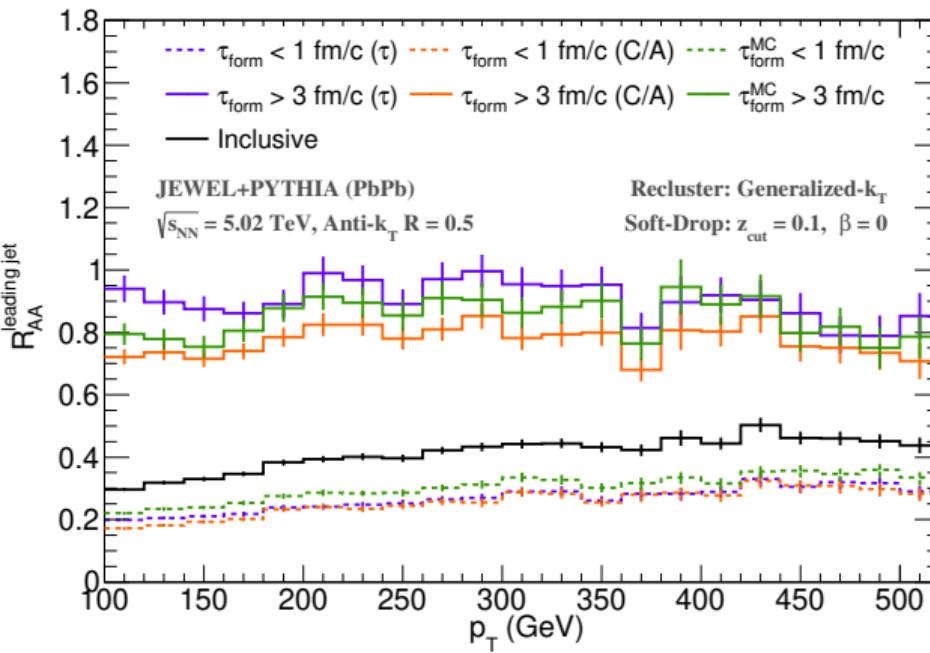
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**$\tau$  algorithm can be used as a jet quenching classifier.**

# Summary

- With an appropriate choice of parameters, the generalized- $k_T$  ( $p = 0.5 \Rightarrow \tau$ ) algorithms can be used to produce a jet history (mostly) ordered in time.

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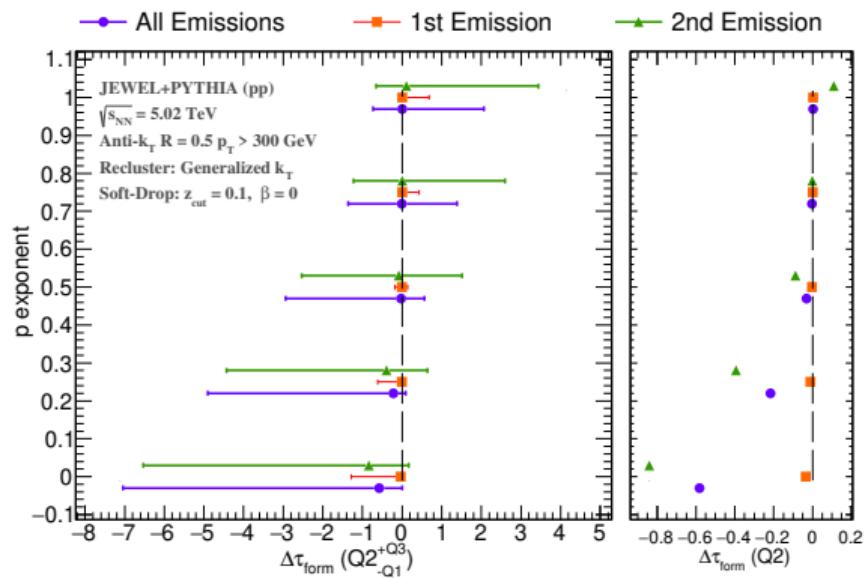
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- With the use of grooming techniques,  $\tau$  clustered jets show an improvement over C/A.
- A more accurate estimation of  $\tau_{\text{form}}$  provides an inclusive jet-quenching classifier, allowing for precision QGP studies.

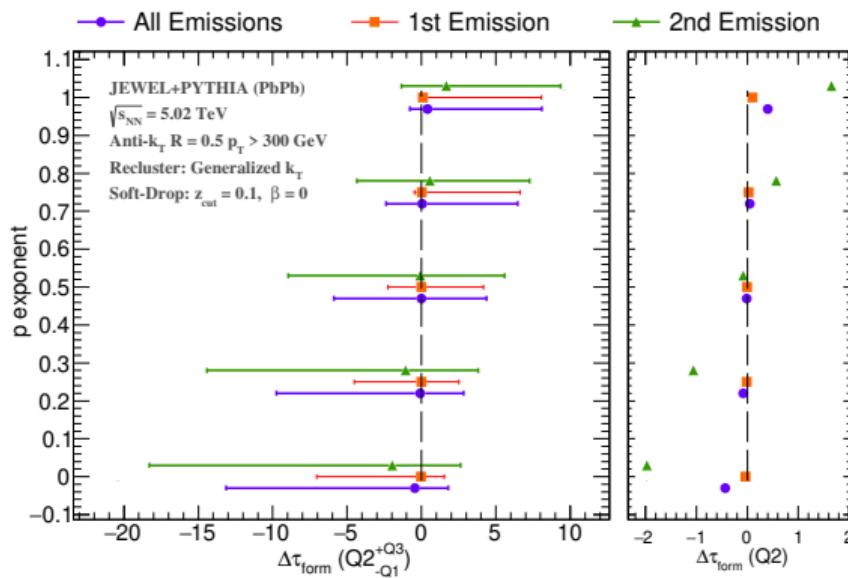
# BACKUPS

# Comparing Monte-Carlo Event Generators

Median and IQR dependence on the algorithm



JEWEL+PYTHIA (pp)

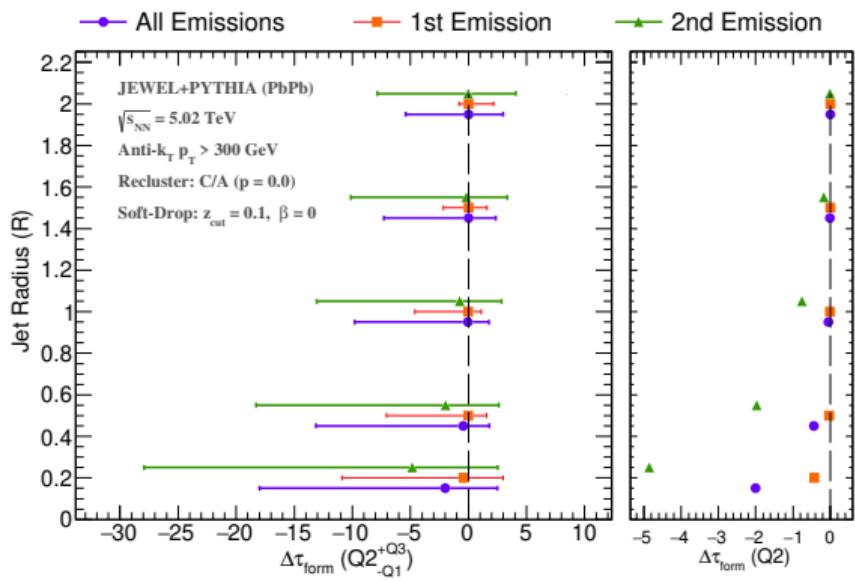


JEWEL+PYTHIA (PbPb)

**Reclustering algorithms other than C/A produce a more centered  $\Delta\tau_{\text{form}}$  distribution, especially for the second emissions, allowing for more accurate substructure.**

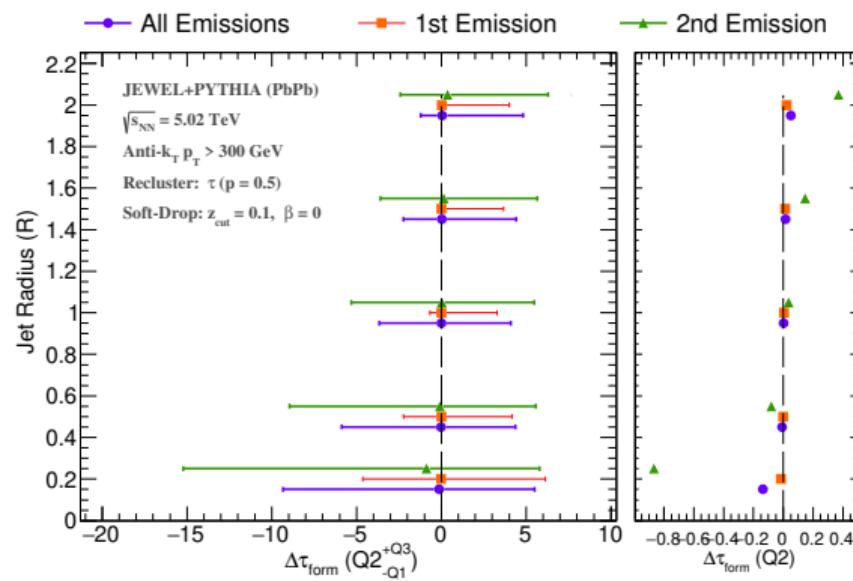
# Impact of the Grooming Parameters

Correlation as a function of the Jet Radius



C/A reclustering

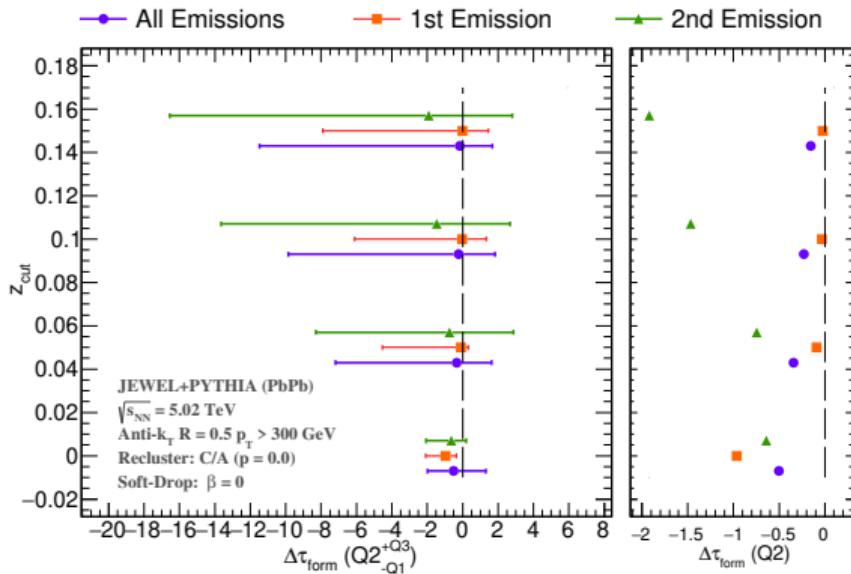
A jet radius of  $R = 1$  seems to be a good compromise for both algorithms.



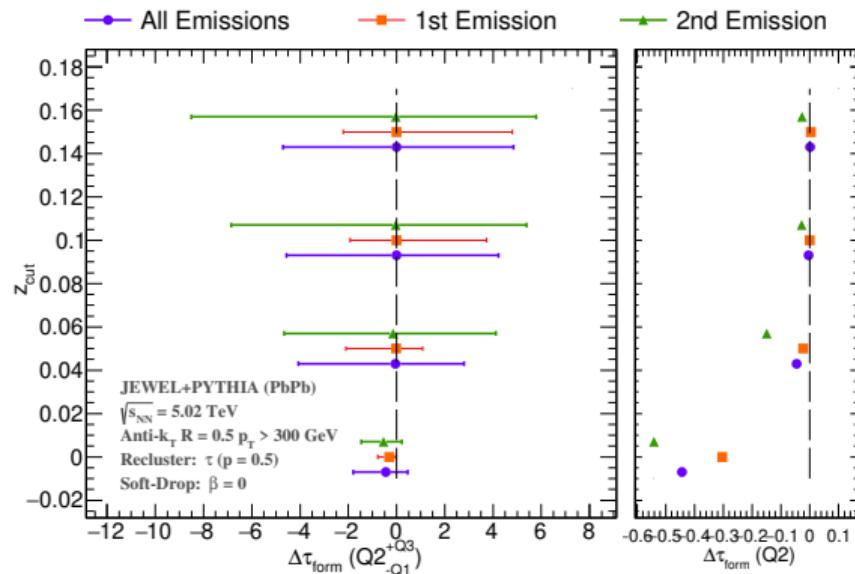
$\tau$  reclustering

# Impact of the Grooming Parameters

Correlation as a function of  $z_{\text{cut}}$



C/A reclustering

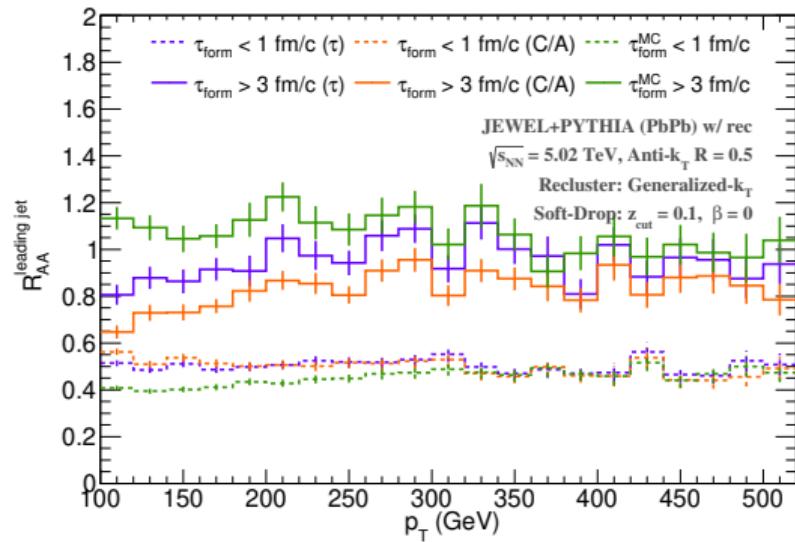
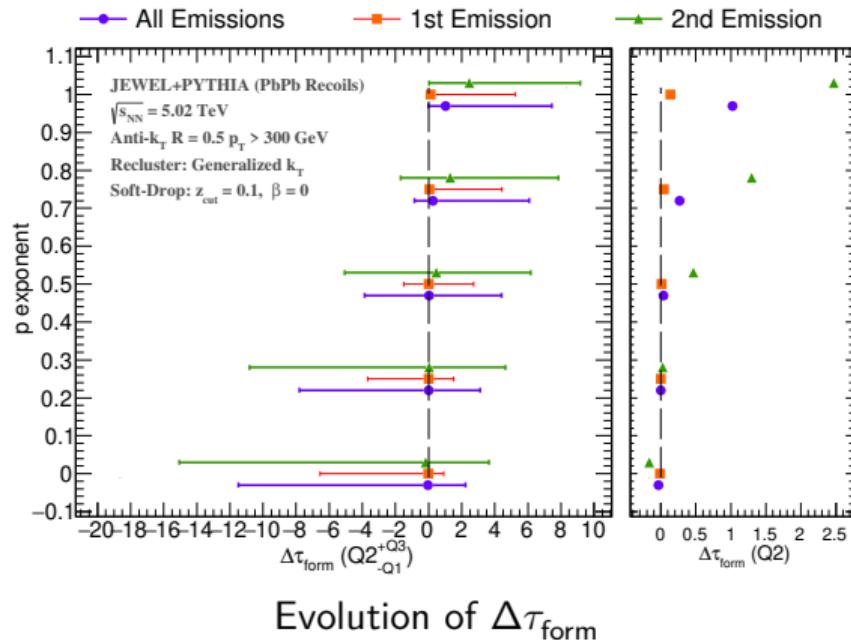


$\tau$  reclustering

Generally, increasing  $z_{\text{cut}}$  shifts the distribution to  $\Delta\tau_{\text{form}} \sim 0$ , while also broadening it.

# JEWEL PbPb with Medium Recoils

Evolution of  $\Delta\tau_{\text{form}}$  and jet quenching



Nuclear modification

When medium recoils are turned on, the nuclear modification effects are even more noticeable.