

# Systematic exploration of multi-scale jet substructure in $p+p$ collisions at $\sqrt{s} = 200$ GeV by the STAR experiment

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Supported in part by

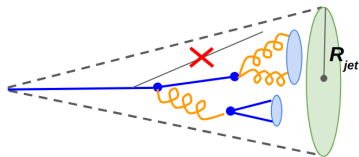
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# Jet substructure and SoftDrop

- Study of jet substructure can help understand partonic fragmentation and hadronization processes
- Our goal is to access parton showers through experimental observables
- Grooming technique called SoftDrop used to remove soft wide-angle radiation from the jet in order to mitigate non-perturbative effects
- Connects parton shower and angular tree



$$\frac{\min(p_{T,1}, p_{T,2})}{p_{T,1} + p_{T,2}} > z_{\text{cut}} \theta^\beta,$$

$$\text{where } \theta = \frac{\Delta R_{12}}{R_{\text{jet}}}$$

$p_{T,1}, p_{T,2}$  - transverse momenta of the subjects

$z_{\text{cut}}$  - threshold (0.1)

$\beta$  - angular exponent (0)

$\Delta R_{12}$  - distance of subjects

in the rapidity-azimuth plane

- Iterative SoftDrop used to study first, second, and third splits



# Substructure observables

## Momentum and angular observables

$z_g$	<b>shared momentum fraction</b>	$z_g \equiv \frac{\min(p_{T,1}, p_{T,2})}{p_{T,1} + p_{T,2}}$
$R_g$	<b>groomed radius</b>	first $\Delta R_{12}$ that satisfies SoftDrop condition
$k_T$	<b>splitting scale</b>	$k_T = z_g p_{T,\text{jet}} \sin R_g$

## Mass observables

$M$	<b>jet mass</b>	$M = \left  \sum_{i \in \text{jet}} p_i \right  = \sqrt{E^2 -  \vec{p} ^2}$
$M_g$	<b>groomed jet mass</b>	jet mass after grooming
$\mu$	<b>groomed mass fraction</b>	$\mu \equiv \frac{\max(m_{j,1}, m_{j,2})}{M_g}$



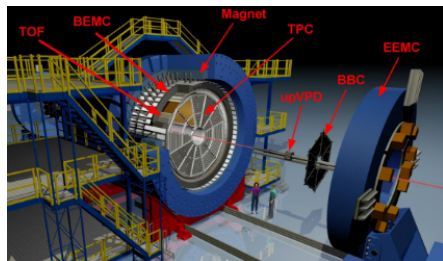
# STAR experiment

## TPC - Time Projection Chamber

- Detection of charged particles for jet reconstruction
- Transverse momenta of tracks:  
 $0.2 < p_T < 30 \text{ GeV}/c$

## BEMC - Barrel Electromagnetic Calorimeter

- Detection of neutral particles for jet reconstruction
- Granularity  $(\Delta\eta \times \Delta\phi) = (0.05 \times 0.05)$
- Tower requirements:  
 $0.2 < E_T < 30 \text{ GeV}$



### Dataset:

$p+p$ ,  $\sqrt{s} = 200 \text{ GeV}$ , 2012

### Algorithms:

anti- $k_T$ , Cambridge/Aachen

### Jets:

Full jets,  $20 < p_{T,\text{jet}} < 50 \text{ GeV}/c$

Full azimuthal angle,  $|\eta| < 1$





# Detector effects correction

- Measurement is affected by finite efficiency and resolution of the instrumentation
- Our goal is to deconvolve detector effects and obtain true distribution from measured one

## **(2+1)D unfolding** (D'Agostini. arXiv:1010.0632(2010))

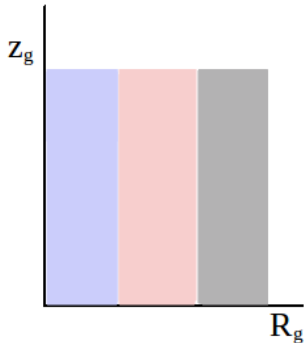
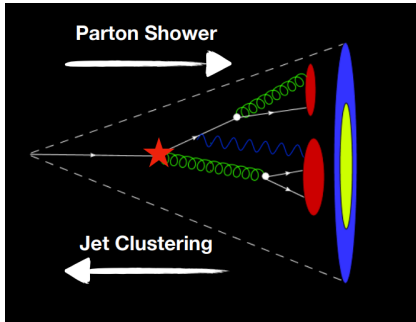
- 2D unfolding via Iterative Bayesian unfolding
- Correction on ensemble level for the 3<sup>rd</sup> dimension

## **MultiFold** (Andreassen et al. PRL 124, 182001 (2020))

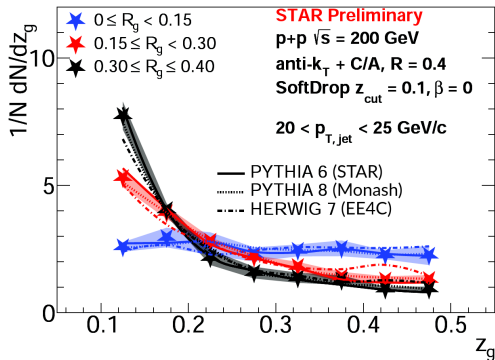
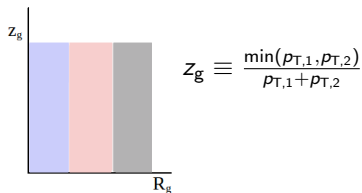
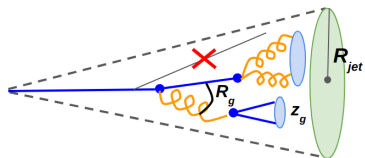
- Machine learning method
- New tool at RHIC
- All observables are simultaneously unfolded in an unbinned way



# Correlation between substructure observables at the first split



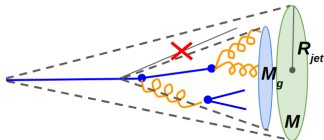
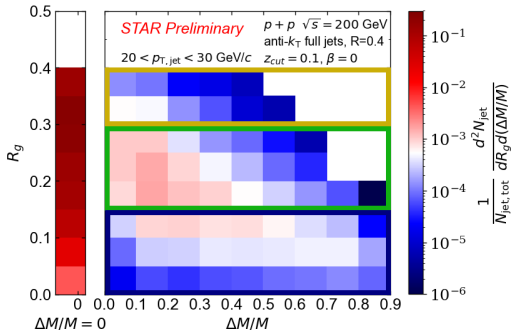
# $z_g$ vs. $R_g$ at the first split



- When we move from collinear hard splitting to softer wide angle splitting,  $z_g$  distribution becomes **steeper** and more **perturbative**
- MC models describe the trend of the data



# $R_g$ vs. $\Delta M/M$ at the first split



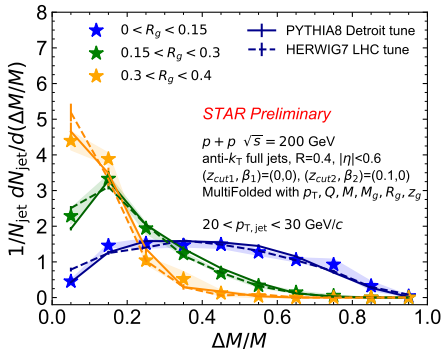
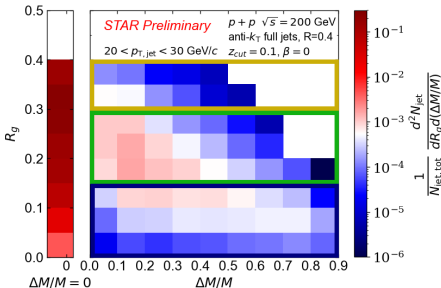
$$\Delta M = M - M_g \text{ [GeV]}$$

## Collinear Drop

- Probes the soft component of the jet
- Difference of an observable with two different SoftDrop settings of parameters  $(z_{\text{cut},1}, \beta_1)$  and  $(z_{\text{cut},2}, \beta_2)$
- Our case:  $(z_{\text{cut},1}, \beta_1) = (0, 0)$ ,  $(z_{\text{cut},2}, \beta_2) = (0.1, 0)$



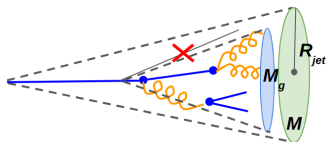
# $R_g$ vs. $\Delta M/M$ at the first split



- The  $\Delta M/M$  distribution is **anti-correlated** with  $R_g$ , which is consistent with angular ordering of the parton shower
- Large groomed jet radius  $\rightarrow$  little/no soft wide angle radiation (small  $\Delta M/M$ ) in the shower
- MC models describe the trend of the data

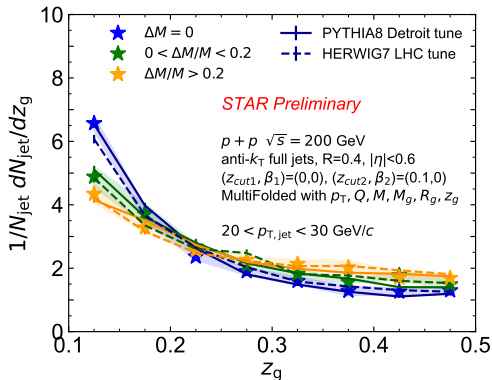


# $z_g$ vs. $\Delta M/M$ at the first split



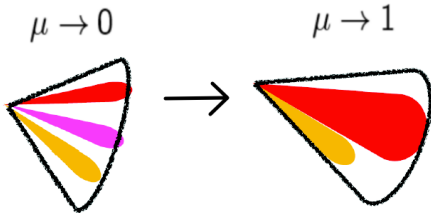
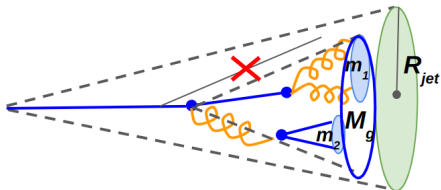
$$\Delta M = M - M_g \text{ [GeV]}$$

- The more mass that is groomed away relative to the ungroomed mass, the **flatter** and more **non-perturbative** the  $z_g$  distribution is
- The first splitting that passes SoftDrop can be non-perturbative  $\rightarrow$  application of the  $\Delta M = 0$  selection can filter out the jets with large non-perturbative contribution



# $\mu$ vs. $R_g$ at the first split

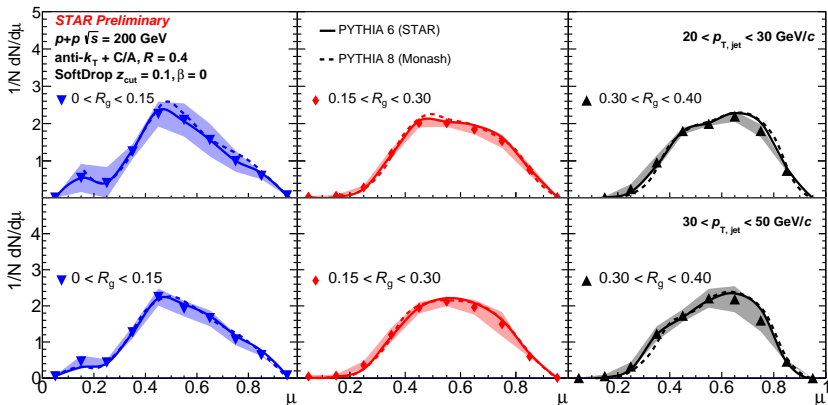
$$\mu \equiv \frac{\max(m_{j,1}, m_{j,2})}{M_g}$$



$\mu$  allows us to study mass sharing of the hard splitting



# $\mu$ vs. $R_g$ at the first split for two different $p_{T,jet}$ bins



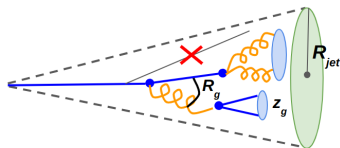
- Dependence on  $R_g$  much **weaker** than  $\Delta M/M$ , largely independent of  $p_{T,jet}$ , MC models agree with data
- $\mu$  shifts to smaller values at smaller angles, indicating a faster reduction of virtuality in the jet shower



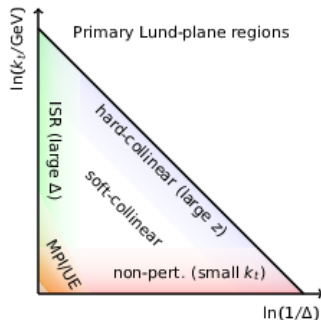


# $\log(k_T)$ vs. $R_g$ at the first split

Dreyer, Salam, Soyez, JHEP 12 (2018) 064



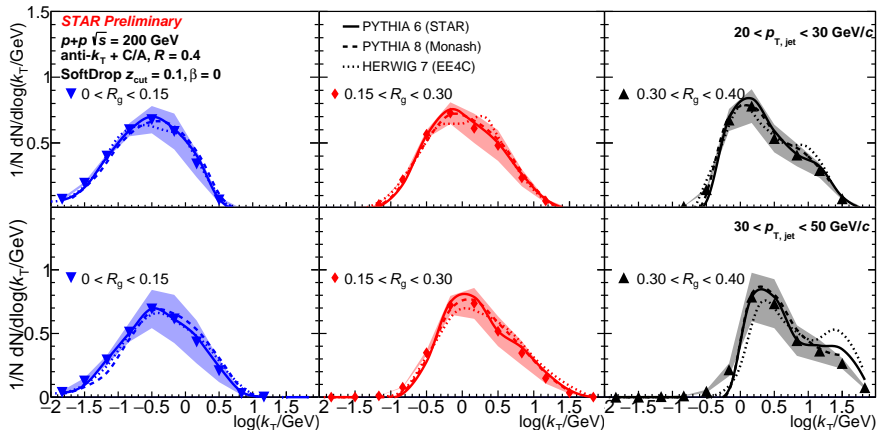
$$k_T = z_g p_{T,jet} \sin R_g$$



Cutting on  $R_g$  moves us to different  $k_T \rightarrow$  we are probing different parts of the Lund Plane



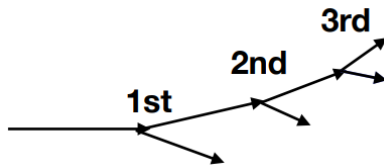
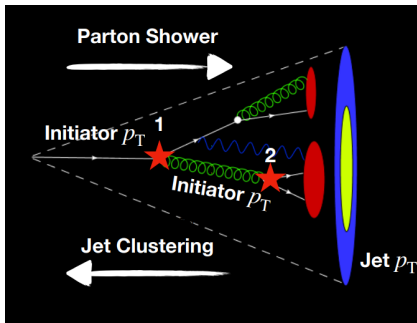
# $\log(k_T)$ vs. $R_g$ at the first split for two different $p_{T,jet}$ bins



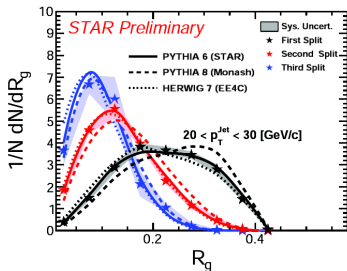
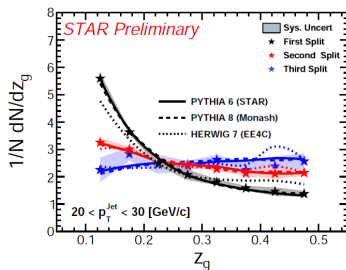
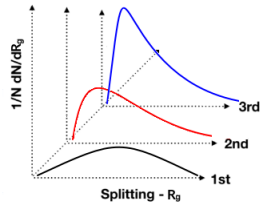
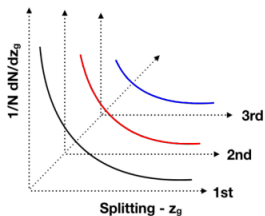
- $\log(k_T)$  has a **strong** dependence on  $R_g$  and **weak** dependence on  $p_{T,jet}$ , MC models describe the trend of the data
- 0 value corresponds to 1 GeV  $\rightarrow$  we move from **non-perturbative** to **perturbative** region



# Evolution of the splitting observables as we travel along the jet shower



# $z_g$ and $R_g$ distributions at 1<sup>st</sup>, 2<sup>nd</sup>, and 3<sup>rd</sup> splits



- Going from 1<sup>st</sup>  $\rightarrow$  3<sup>rd</sup> split
  - $z_g$  distribution becomes **flatter**
  - $R_g$  distribution becomes **narrower**
- Collinear emissions are enhanced when going from 1<sup>st</sup> to 3<sup>rd</sup> split



# Summary

## Correlation at the first split

- New methods for the unfolding were applied (MultiFold, (2+1)D unfolding)
- $z_g$ ,  $\Delta M/M$ ,  $\log(k_T)$  have a **weak** dependence on  $p_{T,\text{jet}}$  and a **strong** dependence on  $R_g$
- Study of different Lund Plane regions allows us to observe the correlations between jet substructure observables

## Splits along the shower

- Observed significantly harder/symmetric splitting at the third/narrow split compared to the first and second splits

Selecting on the split number along the jet clustering tree results in similar change in  $z_g$  distributions as selecting on  $R_g$  or  $\Delta M/M$  at the first split

Jet substructure measurements at RHIC energies allow to disentangle perturbative (early, wide splits) and mostly non-perturbative dynamics (late, narrow splits) within jet showers, and test validity of MC models



Thank you for your attention!



Back up



# Jet clustering algorithms

- Jets are defined using algorithms

## Anti- $k_T$ algorithm

- $d_{ij} = \frac{\min(1/p_{Ti}^2, 1/p_{Tj}^2)\Delta R_{ij}^2}{R}$ ,  $d_{iB} = 1/p_{Tj}^2$
- Clustering starts from the particles with the highest transverse momentum

## Cambridge/Aachen (C/A) algorithm

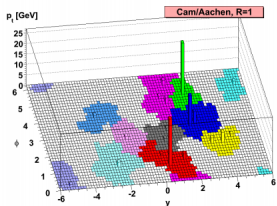
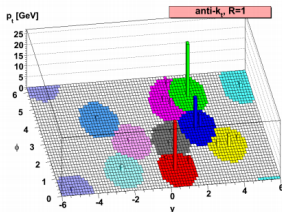
- $d_{ij} = \Delta R_{ij}^2/R^2$ ,  $d_{iB} = 1$
- Particles are clustered exclusively based on angular separation, ideal to be used to resolve jet sub-structure

$d_{iB}$  - distance of the particle  $i$  from the beam

$p_T$  - transverse momentum

$\Delta R_{ij}$  - distance between the particle  $i$  and  $j$

$R$  - jet resolution parameter



Cacciari, Salam, Soyez,  
JHEP 0804:063 (2008)





- Grooming technique used to remove soft wide-angle radiation from the jet
- Connects parton shower and angular tree
  - 1 Jets are first found using the anti- $k_T$  algorithm
  - 2 Recluster jet constituents using the C/A algorithm
  - 3 Jet  $j$  is broken into two sub-jets  $j_1$  and  $j_2$  by undoing the last stage of C/A clustering
  - 4 Jet  $j$  is final SoftDrop jet, if sub-jets pass the condition on the right, otherwise the process is repeated



Larkoski, Marzani, Thaler, Tripathy, Xue,  
Phys. Rev. Lett. 119, 132003 (2017)

- **Shared momentum fraction**  $z_g$

$$z_g = \frac{\min(p_{T,1}, p_{T,2})}{p_{T,1} + p_{T,2}} > z_{\text{cut}} \theta^\beta,$$

$$\text{where } \theta = \frac{\Delta R_{12}}{R}$$

- **Groomed radius**  $R_g$  - first  $\Delta R_{12}$  that satisfies SoftDrop condition

$p_{T,1}, p_{T,2}$  - transverse momenta of the subjets

$z_{\text{cut}}$  - threshold (0.1)

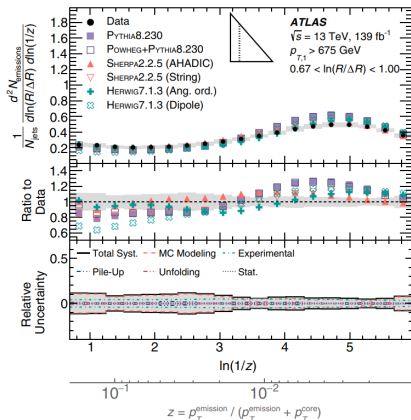
$\beta$  - angular exponent (0)

$\Delta R_{12}$  - distance of subjets in the rapidity-azimuth plane



# Lund Plane measurement

- Previous ATLAS measurement uses Lund jet plane
- Significant differences in varying hadronization models at high  $p_{T,\text{jet}}$  at the LHC  $\rightarrow$  we want to study this at lower  $p_{T,\text{jet}}$ , where non-perturbative effects are expected to be larger
- While Lund jet plane integrates over all splits, we focus on the first split



ATLAS, Phys. Rev. Lett. **124**, 222002 (2020)



# Data analysis

- $p + p$  collisions at  $\sqrt{s} = 200$  GeV, 2012
- $\sim 11$  million events analyzed

## Event and track selection

- Transverse momenta of tracks:  $0.2 < p_T < 30$  GeV/c
- Tower requirements:  $0.2 < E_T < 30$  GeV

## Jet reconstruction

- Jets reconstructed with anti- $k_T$  algorithm, reclustered with the C/A algorithm
- Transverse momenta of jets:  $15 < p_{T,\text{jet}} < 40$  GeV/c
- Resolution parameters:  $R = 0.4, R = 0.6$
- SoftDrop parameters:  $z_{\text{cut}} = 0.1, \beta = 0$

$$\frac{\min(p_{T,1}, p_{T,2})}{p_{T,1} + p_{T,2}} > z_{\text{cut}} \left( \frac{\Delta R_{12}}{R} \right)^\beta$$



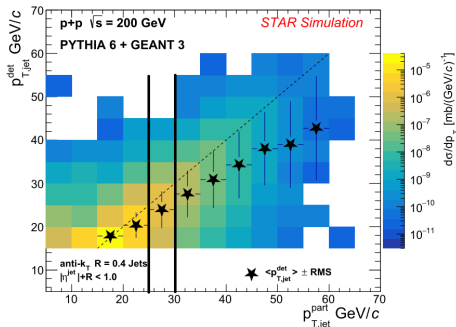
# 2D Bayesian Unfolding

- 2D iterative Bayesian method implemented in the RooUnfold
- Procedure has following steps:
  - ① The jets at the detector and particle level are reconstructed separately
  - ② Jets are matched based on  $\Delta R < 0.6$
  - ③ Jets without match - missed jet (particle level) and fake jets (detector level)
  - ④ Response between detector level and particle level for observables is constructed
- We use RooUnfold response which contains Matches and Fakes
  - Unfolding is done separately for  $p_T^{det}$  intervals 15-20, 20-25, 25-30, 30-40 GeV/c
- Then unfolded spectra are weighted with values from our projection and put together
- Together with trigger missed and unmatched weighted spectra we get our fully unfolded spectrum



# Correction in 2+1D for $z_g$ , $R_g$ , and $p_{T,jet}$

- Results are in 3D  $\rightarrow z_g$  vs.  $R_g$  is unfolded in 2D and correction for  $p_{T,jet}$  in 1D is needed
  - For each particle-level  $p_{T,jet}$  bin, we do projection of this bin into detector-level  $p_{T,jet}$ , and get the weights from detector-level  $p_{T,jet}$  bins



STAR, Phys. Lett. B 811 (2020) 135846

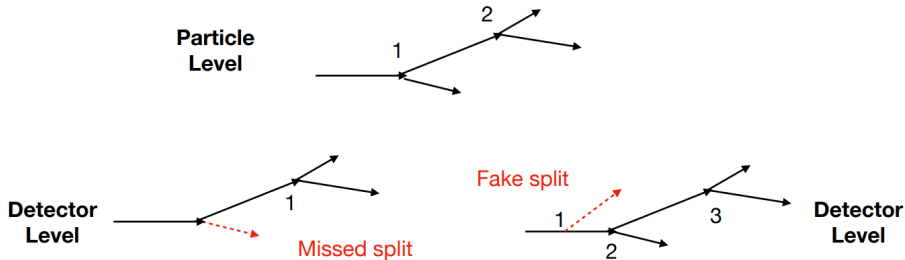
- We unfold  $z_g$  vs.  $R_g$  via iterative Bayesian unfolding in 2D using RooUnfold and unfolded spectra for each detector-level  $p_{T,jet}$  bin are weighted and summed
- Additional corrections for trigger and jet finding efficiencies are applied

Details on systematic uncertainties available in back up

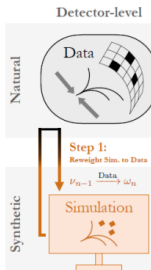


# Correction in 2+1D for $p_{T,jet/initiator}$ , $z_g$ , $R_g$

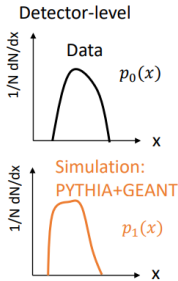
- Splits can be affected by detector efficiency and resolution
- Observables at a given split are smeared
- Splitting hierarchy is modified going from particle level to detector level



# MultiFold



Where does the machine learning part come in?



E.g., Iteration 1, step 1:

Weights:  $w(x) = p_0(x)/p_1(x)$  Ok for 1D

$$\approx f(x)/(1 - f(x))$$

[\(Andreassen and Nachman PRD 101, 091901 \(2020\)\)](#)

where  $f(x)$  is a neural network and trained with the binary cross-entropy loss function

to distinguish jets coming from data vs from simulation

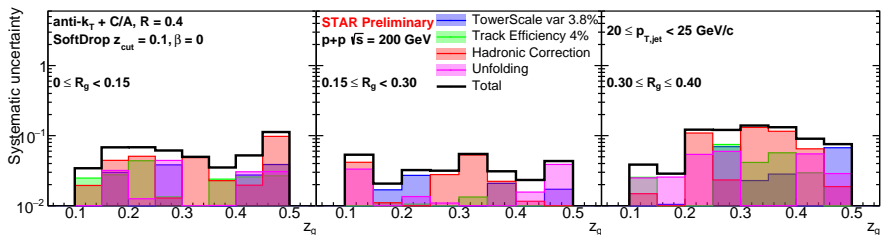
Unfolding  $\rightarrow$  Reweighting histograms  
 $\rightarrow$  Classification  $\rightarrow$  Neural network

See backup slides for details of the neural networks.



# Systematic uncertainties

- Systematic uncertainties estimated by varying the detector response
  - Hadronic correction - fraction of track momentum subtracted is varied
  - Tower scale variation - tower gain is varied by 3.8%
  - Tracking efficiency - efficiency is varied by 4%
  - Unfolding - iterative parameter is varied from 4 to 6
- Systematics due to prior shape variation will be included in the final publication



$$0 \leq R_g < 0.15$$

$$0.15 \leq R_g < 0.30$$

$$0.30 \leq R_g \leq 0.40$$

