

# Latest jet cross-section measurements in proton–proton collisions by ATLAS

Jonathan Bossio



9 June 2020



- ▶ Jet cross-sections provide valuable information about the strong coupling constant,  $\alpha_s$ , and the structure of the proton
- ▶ Final states with only jets represent a background to many other processes at hadron colliders

The measurements presented here collectively probe:

- ▶ Precision QCD predictions and MC predictions
- ▶ Jet substructure (substructure observables, trimming and soft-drop)
- ▶ Jet quantities related to fragmentation
- ▶ Event shapes in multijet events

## Brief introduction on jets and grooming techniques

- ▶ Jets: collimated sprays of particles initiated by high-energy partons
- ▶ Grooming techniques (soft drop, trimming) remove soft and wide-angle radiation, making jets robust against pileup, final-state radiation and underlying event

**Trimming procedure:** Reconstruct very-small-R jets from constituents and discard those that have very low  $p_T$  fraction

### Soft-drop procedure:

- ▶ Jet constituents are reclustered using C/A algorithm
  - ▶ Which iteratively clusters the closest constituents in rapidity and azimuth
- ▶ Last step of clustering is undone, breaking the jet into two subjets ( $j_1$  &  $j_2$ )
- ▶ These subjets are then used to evaluate the soft-drop condition:

$$\frac{\min(p_{T,j1}, p_{T,j2})}{p_{T,j1} + p_{T,j2}} > z_{\text{cut}} \left( \frac{\Delta R_{12}}{R} \right)^\beta$$

- ▶  $z_{\text{cut}}$  and  $\beta$  regulates the sensitivity to soft and wide-angle radiation
- ▶ **Failed condition:** subjet w/ lower  $p_T$  is removed and the procedure is iterated with the remaining jet
- ▶ **Satisfied condition:** algorithm stops and resulting jet is the soft-drop jet

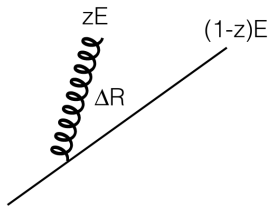
## Lund Plane measurement with charged particles

[\[arXiv:2004.03540\]](#)

[Submitted to PRL]

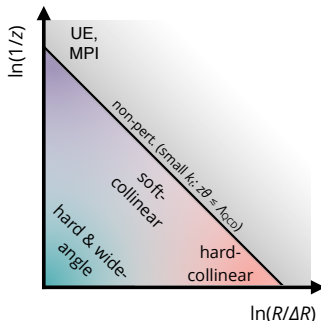
## Lund plane measurement with charged particles in Full Run 2 data

- ▶ In the soft gluon ('eikonal') picture of jet formation, a quark or gluon radiates a haze of relatively low energy and statistically independent gluons
- ▶ Emission pattern is  $\approx$  uniform in  $\ln(1/z) - \ln(1/\Delta R)$ 
  - ▶  $z$ : relative momentum fraction of the emitted gluon
  - ▶  $\Delta R$ : emission opening angle
- ▶ This space is called the **Lund plane**, where different physical effects factorize



- ▶ How to extract  $z$  &  $\Delta R$  proxies [ref]:
  - ▶ Recluster jet's constituents w/ C/A, reverse clustering history
  - ▶ At each step in the C/A declustering sequence, Lund plane is filled w/:
  - ▶  $z = p_T(j_2)/(p_T(j_1) + p_T(j_2))$  &  $\Delta R = \Delta R(j_1, j_2)$
  - ▶ Where  $j_1$  ( $j_2$ ) is the hardest (softer) of the proto-jet pair

### Regions of the Lund Plane



# Lund plane measurement with charged particles | Selections & Results

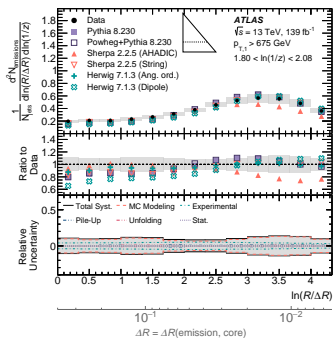
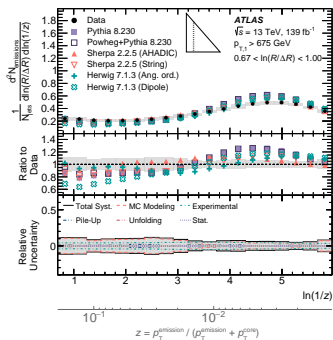
## Event selection and procedure:

- ▶ Leading anti- $k_t$   $R = 0.4$  jet  
 $p_{T>} > 650$  GeV,  $p_{T}^{\text{lead}} < 1.5 \times p_{T}^{\text{sublead}}$
- ▶ First two leading  $p_{T}$  jets w/  $|\eta| < 2.1$
- ▶ Tracks matched to jets ( $\Delta R = 0.4$ )  
are used to recluster C/A jets

## Uncertainties:

- ▶ Jets and tracks
- ▶ Unfolding, particle-detector matching  
and pile-up modelling
- ▶ **Fragmentation modelling**

Slices in the plane are done to quantitatively compare with MC



**No prediction describes the data in all regions, but Herwig 7.1.3 angle-ordered provides the best description across most of the plane**

# Measurement of Soft Drop jet Observables at 13 TeV

[Phys. Rev. D 101 (2020) 052007]

[arXiv:1912.09837]

# Soft Drop Jet Observables at 13 TeV (using 2016 ATLAS data)

## Jets:

- ▶ Soft drop applied to calorimeter- and track-based jets reconstructed w/ anti- $k_t$  algorithm and  $R = 0.8$
- ▶ Tracks are those that match the ungroomed jets using ghost-association

## Event selection:

- ▶ At least two jets with  $p_T^{\text{lead}} > 300$  GeV and within  $|\eta| < 1.5$
- ▶ Dijet topology enhanced by requesting  $p_T^{\text{lead}}/p_T^{\text{sublead}} < 1.5$

## Double-differentially cross sections measured for these observables:

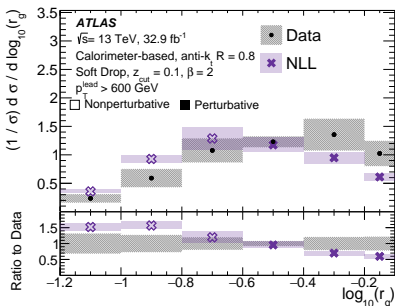
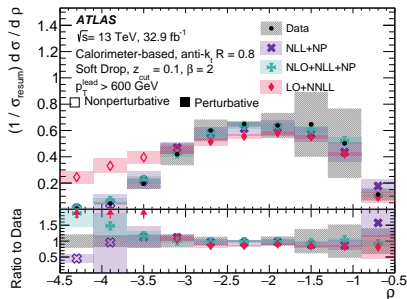
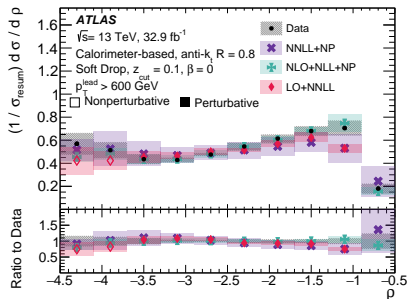
- ▶ Relative jet mass:  $\rho = m^2/p_T^2$  ( $m$  is groomed and  $p_T$  is ungroomed)
- ▶  $p_T$  balance  $z_g$  from the soft-drop condition  $\left( \frac{\min(p_{T,j1}, p_{T,j2})}{p_{T,j1} + p_{T,j2}} \right)$
- ▶  $r_g$ : Opening angle of the splitting ( $\Delta R_{12}$ )

## Systematics:

- ▶ Calorimeter-cell cluster uncertainties
- ▶ MC and pileup modelling and unfolding non-closure
- ▶ Tracking uncertainties are negligible



# Soft Drop observables | Comparisons with analytical predictions

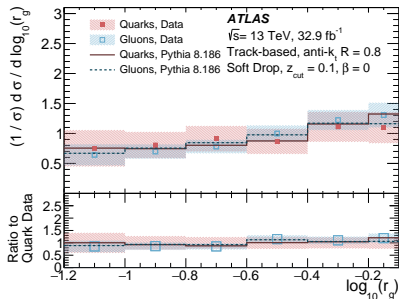
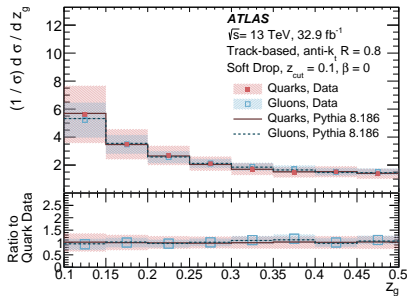
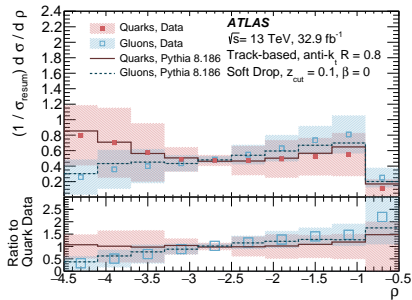


►  $\rho$ :

- Calculations are able to model the data in the resummation region ( $-3 \lesssim \rho \lesssim -1$ ) at the level of a 10% difference
- NLO+NLL models well data at the high values of  $\rho$ , while LO+NNLL and NNLL don't

- $r_g$ : The prediction is systematically higher than the data in regions where nonperturbative effects are large

# Soft Drop observables | Quark-gluon track-based distributions ( $\beta = 0$ )



- ▶ Gluon distribution tends towards higher values of  $\rho$  and larger splitting
- ▶ This becomes more apparent at larger values of  $\beta$
- ▶  $z_g$ : Very similar distributions for  $\beta = 0$ , while some differences begin to appear for  $\beta > 0$
- ▶ In general, data distributions are in agreement with MC predictions

**Properties of jet fragmentation using charged particles at**  
 **$\sqrt{s} = 13 \text{ TeV}$**

[Phys. Rev. D 100 (2019) 052011]

[arXiv:1906.09254]

## Jet fragmentation properties (using 2016 ATLAS data)

**Jets:** Calorimeter-based jets reconstructed with anti- $k_t$  and  $R = 0.4$

### Event selection:

- ▶ At least two jets with  $p_T^{\text{lead}} > 60$  GeV and within  $|\eta| < 2.1$
- ▶ Dijet topology enhanced by requesting  $p_T^{\text{lead}} / p_T^{\text{sublead}} < 1.5$

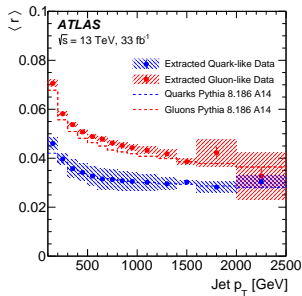
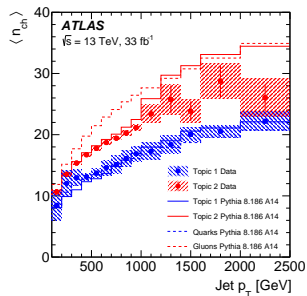
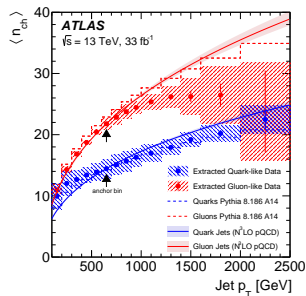
**Observables:** (two more were studied by are not discussed today)

- ▶ Charged-particle multiplicity ( $n_{\text{ch}}$ ): # of charged particles inside a jet
- ▶ Radial profile ( $\langle r \rangle$ ): Weighted average number of charged particles (weighted by radial distance w.r.t. jet axis)

### Systematics:

- ▶ Rate of fake and secondary tracks (result of interactions in detector material)
- ▶ Tracking efficiency
- ▶ MC modelling
- ▶ Unfolding uncertainty
- ▶ Jet energy scale and resolution uncertainties much smaller than the rest

# Jet fragmentation properties | Data vs MC and pQCD predictions



## ▶ $\langle n_{ch} \rangle$ :

- ▶ Predictions describe the quark-like data, but the gluon-like data have systematically fewer charged particles than the prediction
- ▶ Similar conclusions from the two jet flavour labelling approaches

## ▶ Radial profile:

- ▶ Predictions describe the quark-like data, but underestimate the gluon-like data

**Brand new results!**

**Measurement of hadronic event shapes in multijet final states at  
 $\sqrt{s} = 13 \text{ TeV}$**

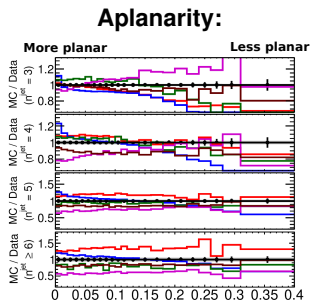
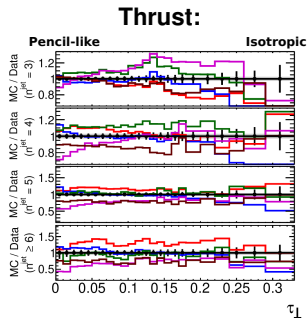
[ATLAS-CONF-2020-011]

# Measurement of hadronic event shapes in multijet final states

- ▶ Event shapes are a family of observables which are sensitive to event-wide energy flow:

- ▶ *Thrust*: How dijet-like is the event?
  - ▶ **Pythia (LO)** describes the data well in dijet-like topologies
  - ▶ **Sherpa** over-estimation becomes more pronounced as  $n^{\text{jet}}$  increases
  - ▶ Herwig7 (**dipole** & **ang. ord.** PS) show different trends
  - ▶ **MadGraph** underestimate the data

- ▶ *Aplanarity*: How planar is the event?
  - ▶ Sensitivity to PS algorithm selection in Herwig7 is apparent (**dipole** & **ang. ord.** PS)
  - ▶ **Sherpa** over-predicts events at high  $n^{\text{jet}}$
  - ▶ **Pythia** describes more-planar events best
  - ▶ **MadGraph** underestimate the data but agreement in shape



# Conclusions

- ▶ Many great jet cross section measurements from ATLAS
- ▶ Interesting results probing new techniques and different physics effects
- ▶ Most of the results are well modelled by predictions
- ▶ But some discrepancies are observed in some results
- ▶ Giving room to improve MC event simulations and pQCD predictions



## Back-up slides

### Following approach is used to extract the quark and gluon distributions

- ▶ Take the more forward ( $f$ ) and central ( $c$ ) of the two selected jets
- ▶ Using the fraction of quark jets  $f_q$  from MC we can extract the quark ( $h_i^q$ ) and gluon ( $h_i^g$ ) jet fragmentation properties separately by solving the following per bin  $i$  of an observable:

$$h_i^f = f_q^f h_i^q + (1 - f_q^f) h_i^g$$
$$h_i^c = f_q^c h_i^q + (1 - f_q^c) h_i^g$$

where the flavour of a jet in MC is defined as the type of the highest- $E$  parton

An alternative approach (**topic modelling**) is used by one of the analyses:

- ▶ This does not require the input of any fractions from MC
- ▶ We can extract distributions of 'topics'  $T_1$  and  $T_2$  (more in back-up)
- ▶ Used only for those observables which ensures that first topic is well aligned with quarks and second one is more gluon-like

In this, we can extract distributions of 'topics'  $T_1$  and  $T_2$

$$h_i^{T_1} = \frac{h_i^f - (\min_j \{h_j^f / h_j^c\}) \times h_i^c}{1 - \min_j h_j^f / h_j^c}$$
$$h_i^{T_2} = \frac{h_i^c - (\min_j \{h_j^c / h_j^f\}) \times h_i^f}{1 - \min_j h_j^c / h_j^f}$$

where:

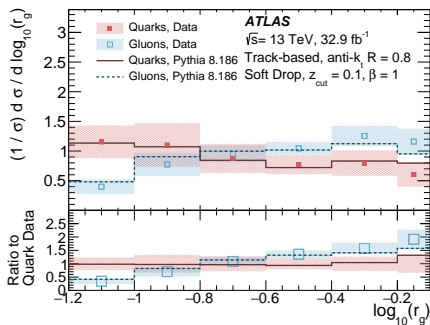
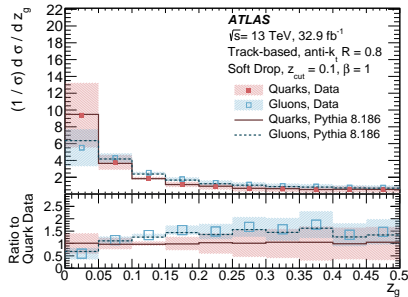
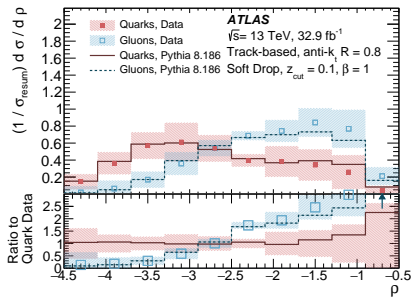
- ▶  $h_i$  is a bin of a histogram for an observable
- ▶  $f$  and  $c$  represent the forward and central regions
- ▶  $q$  and  $g$  represent quark or gluon

# Measurement of Soft Drop jet Observables at 13 TeV

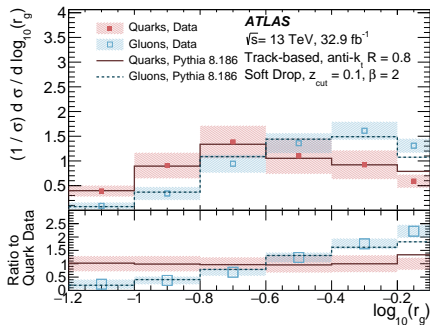
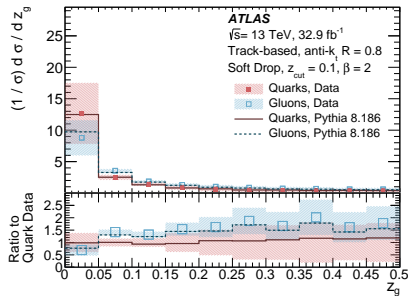
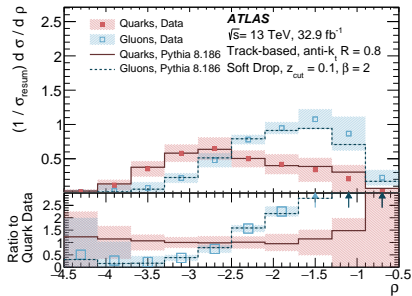
[[Phys. Rev. D 101 \(2020\) 052007](#)]

[[arXiv:1912.09837](#)]

# Soft Drop observables | Quark-gluon track-based distributions ( $\beta = 1$ )

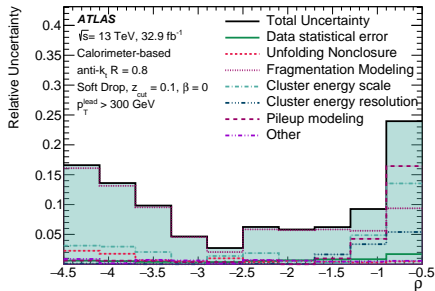


# Soft Drop observables | Quark-gluon track-based distributions ( $\beta = 2$ )

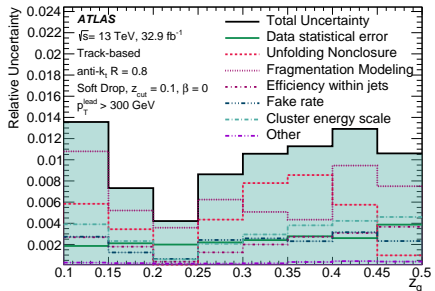
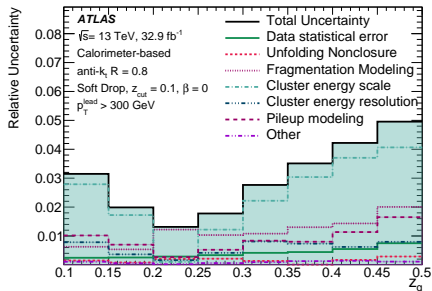
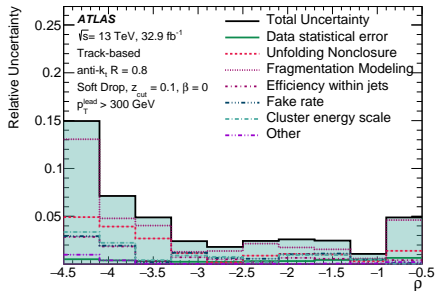


# Soft Drop observables | Systematics

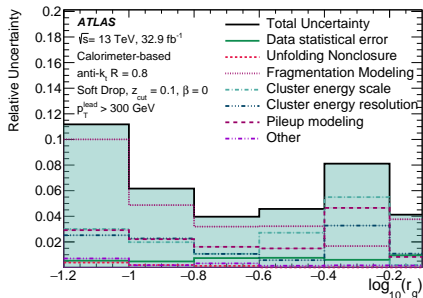
## Calorimeter-based



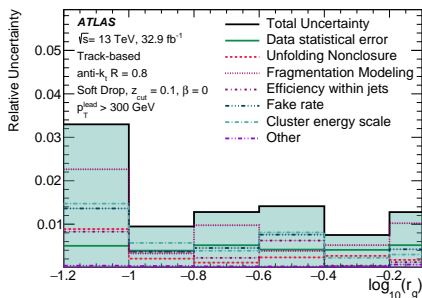
## Track-based



## Calorimeter-based



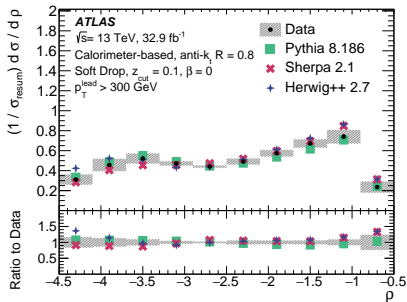
## Track-based



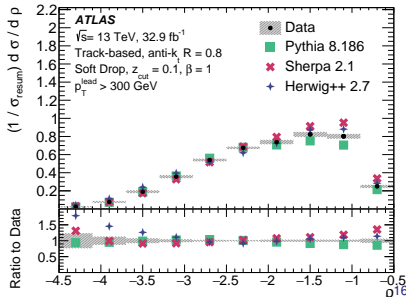
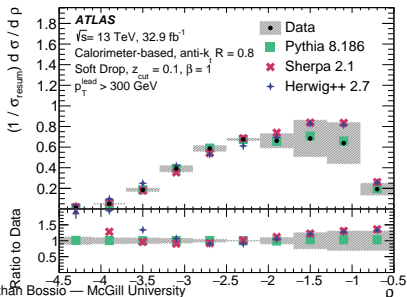
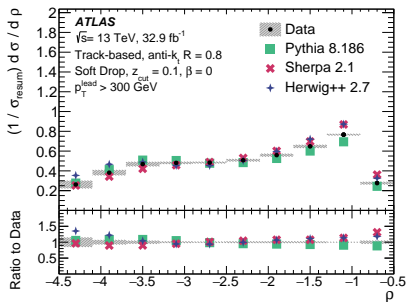


# Soft Drop observables | MC/data plots for $\rho$

## Calorimeter-based

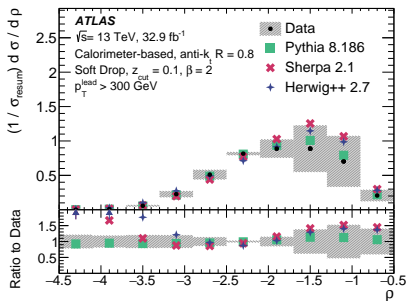


## Track-based

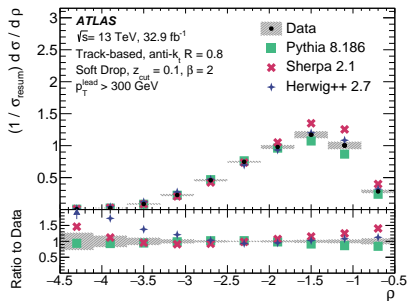


# Soft Drop observables | MC/data plots for $z_g$

## Calorimeter-based

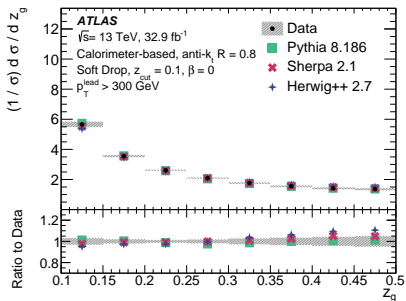


## Track-based

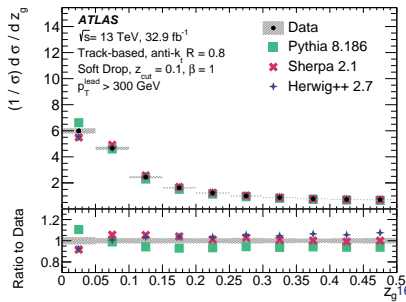
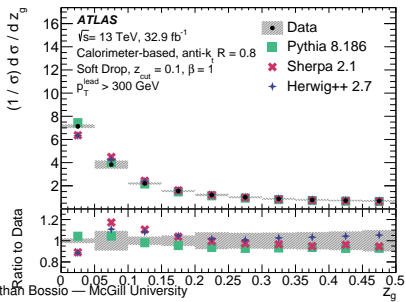
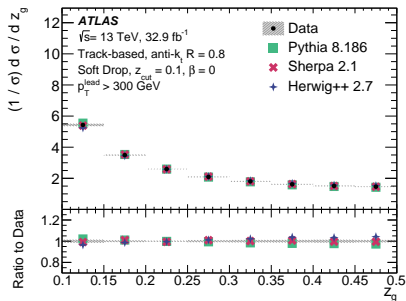


# Soft Drop observables | MC/data plots for $z_g$

## Calorimeter-based

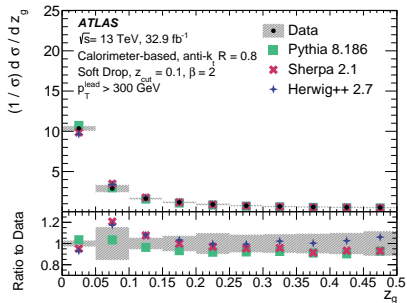


## Track-based

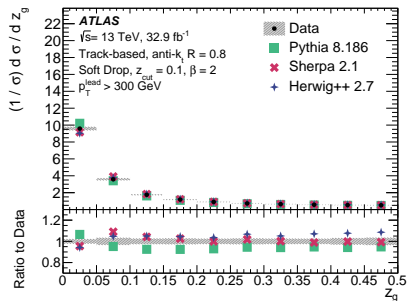


# Soft Drop observables | MC/data plots for $z_g$

## Calorimeter-based

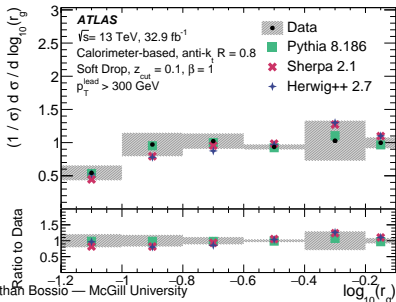
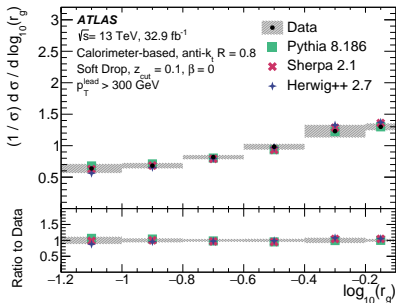


## Track-based

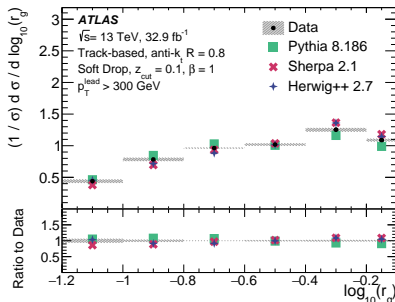
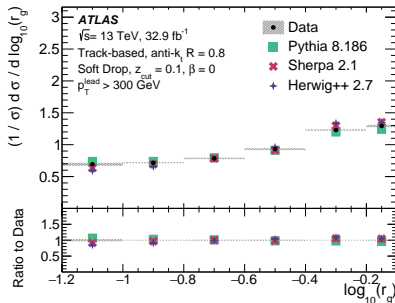


# Soft Drop observables | MC/data plots for $r_g$

## Calorimeter-based

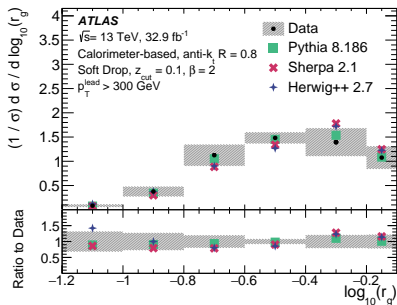


## Track-based

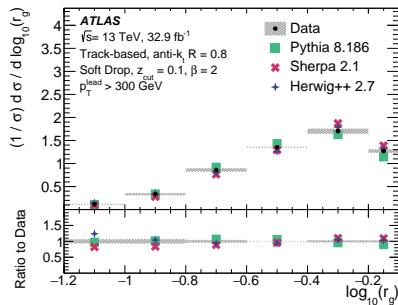


# Soft Drop observables | MC/data plots for $r_g$

## Calorimeter-based



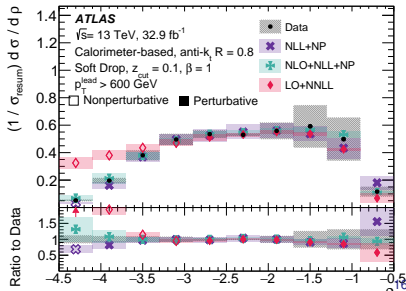
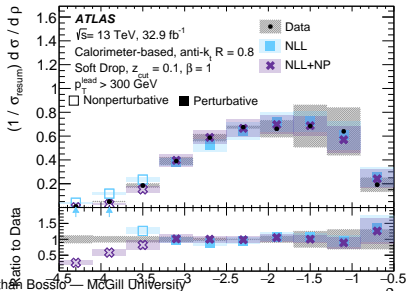
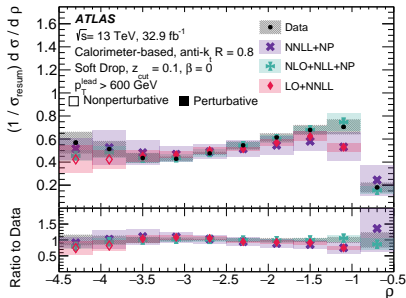
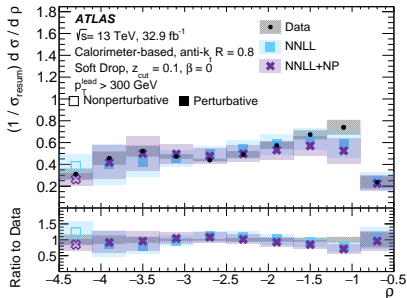
## Track-based



# Soft Drop observables | Comparisons to analytical predictions

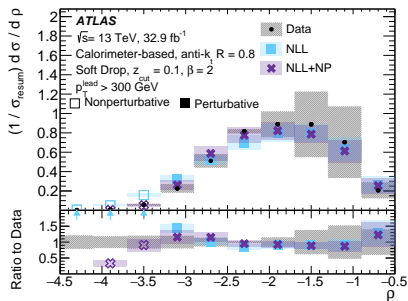
Low  $p_T$

High  $p_T$

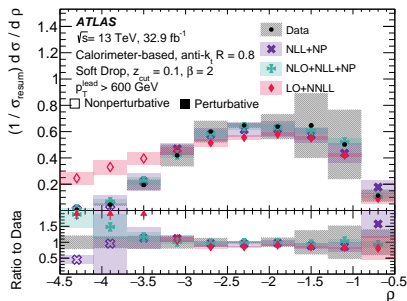


# Soft Drop observables | Comparisons to analytical predictions

Low  $p_T$



High  $p_T$

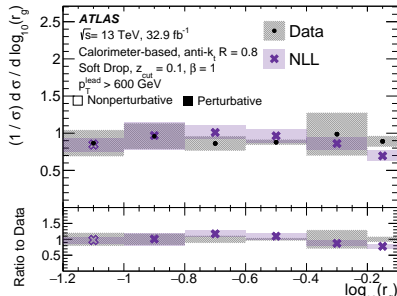
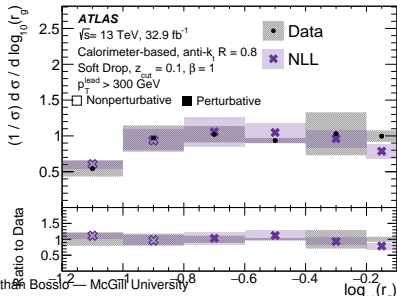
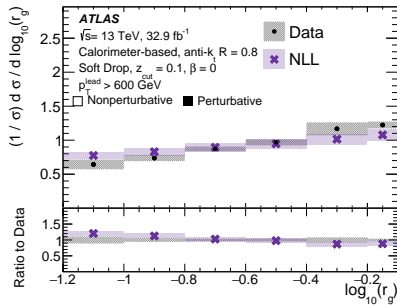
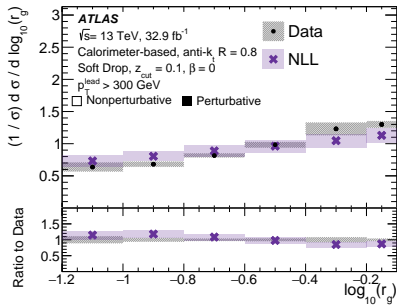




# Soft Drop observables | Comparisons to analytical predictions

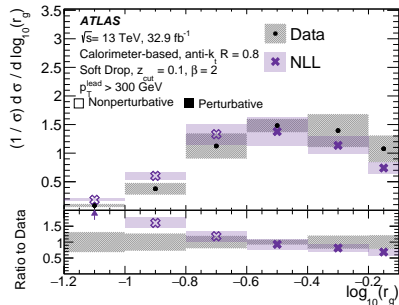
Low  $p_T$

High  $p_T$

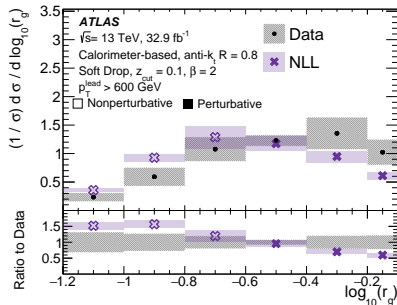


# Soft Drop observables | Comparisons to analytical predictions

Low  $p_T$

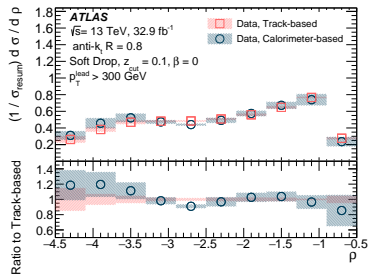


High  $p_T$

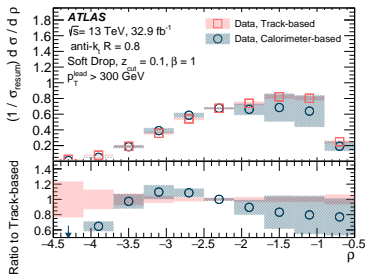


# Soft Drop observables | Calorimeter- vs track-based $\rho$

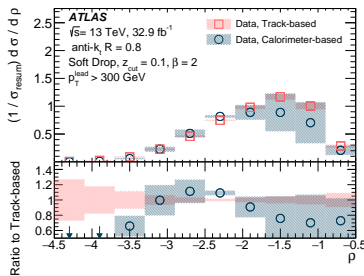
$\beta = 0$



$\beta = 1$

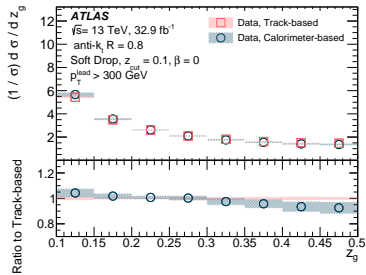


$\beta = 2$

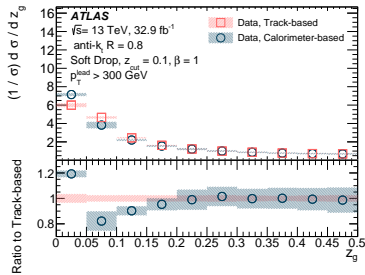


# Soft Drop observables | Calorimeter- vs track-based $z_g$

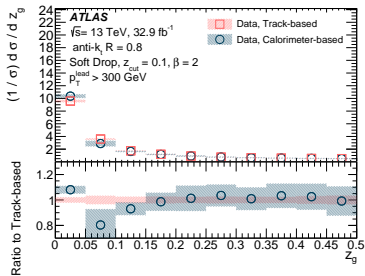
$\beta = 0$



$\beta = 1$

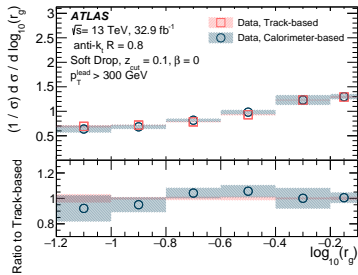


$\beta = 2$

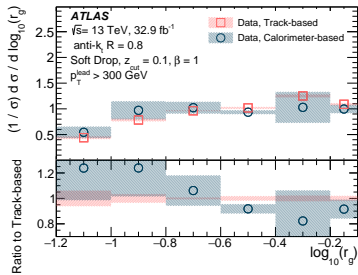


# Soft Drop observables | Calorimeter- vs track-based $r_g$

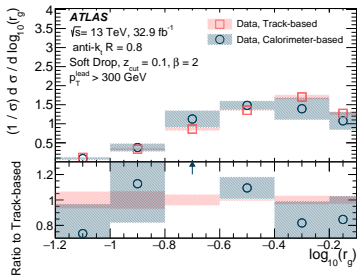
$\beta = 0$



$\beta = 1$



$\beta = 2$



**Properties of jet fragmentation using charged particles at**  
 **$\sqrt{s} = 13 \text{ TeV}$**

[Phys. Rev. D 100 (2019) 052011]

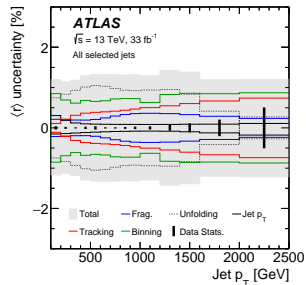
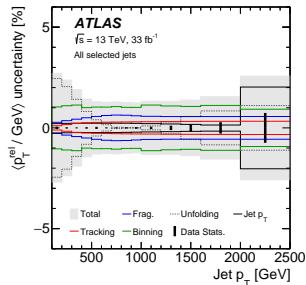
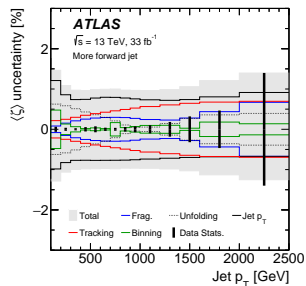
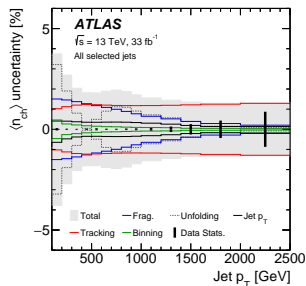
[arXiv:1906.09254]

Summed fragmentation function: The distribution of the fragmentation fraction is studied inside jets summed over charged-hadron types

Where the fragmentation function describes the probability of finding a hadron  $h$  with energy fraction  $z$  of the parton  $p$  that has energy  $E$ .

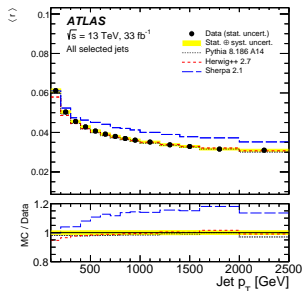
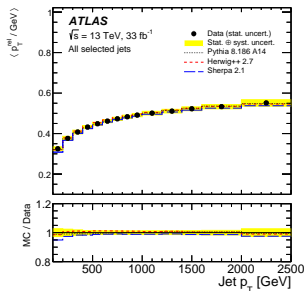
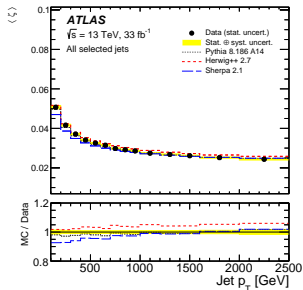
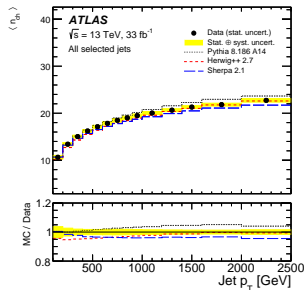
Transverse momentum:  $p_{\text{T}}^{\text{rel}} = p_{\text{T}}^{\text{charged particle}} \sin\Delta\phi$ , where  $\Delta\phi$  is the angle b/w the charged particle and the jet axis in the transverse plane

# Jet fragmentation properties | Uncertainties

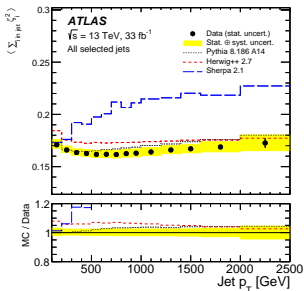
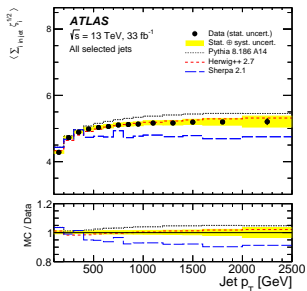
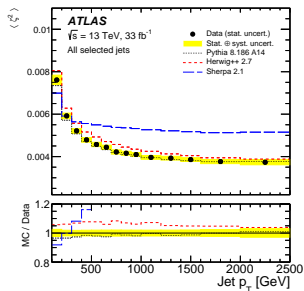
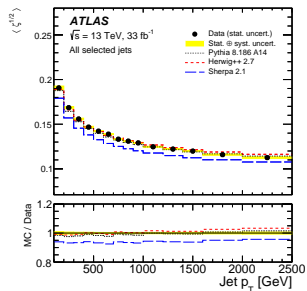




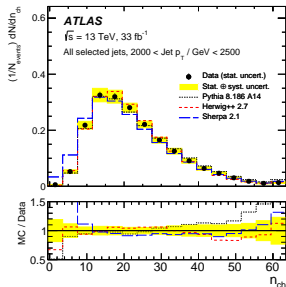
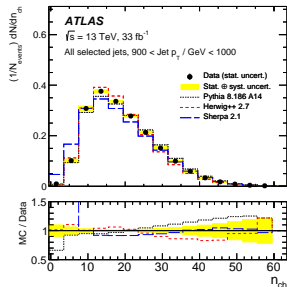
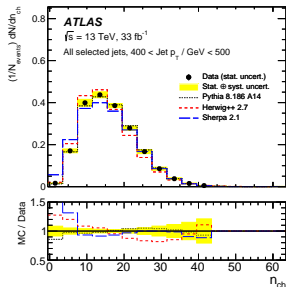
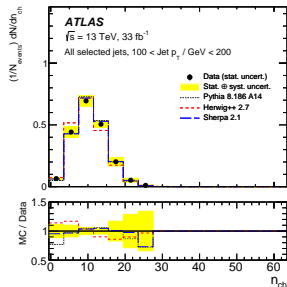
# Jet fragmentation properties | Unfolded data vs MC



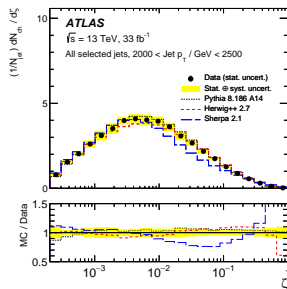
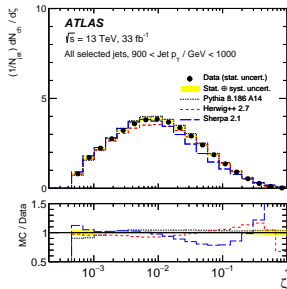
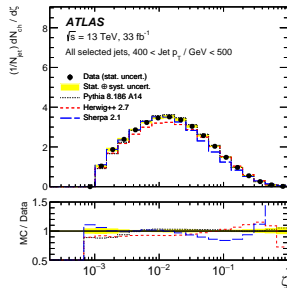
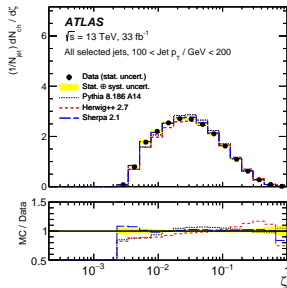
# Jet fragmentation properties | Unfolded data vs MC



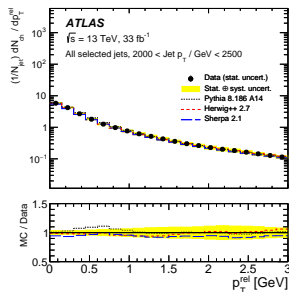
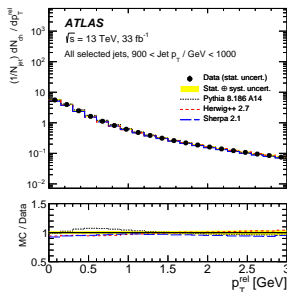
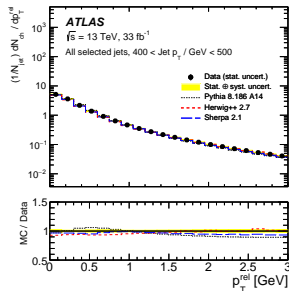
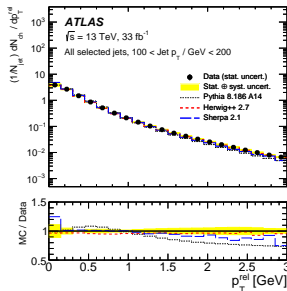
# Jet fragmentation properties | Distribution of the observables



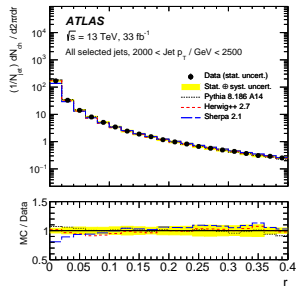
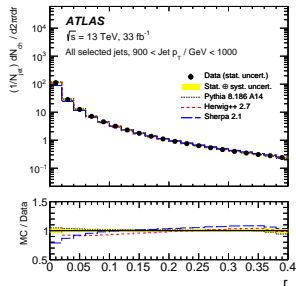
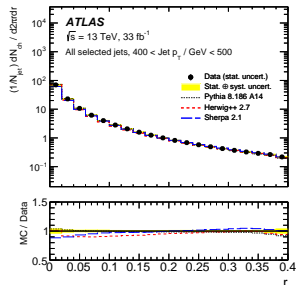
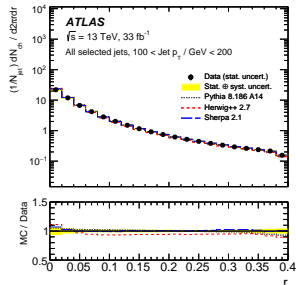
# Jet fragmentation properties | Distribution of the observables



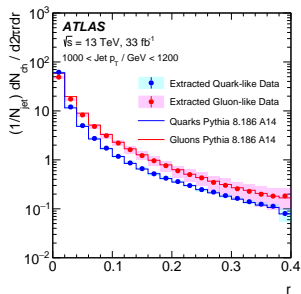
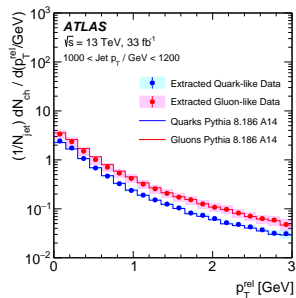
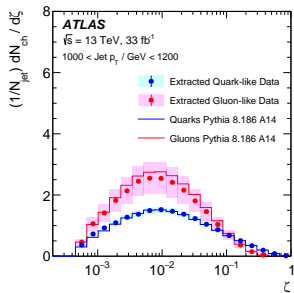
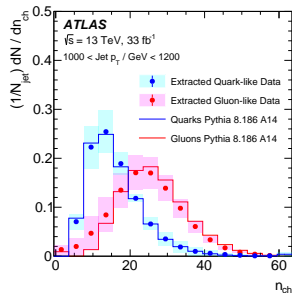
# Jet fragmentation properties | Unfolded data vs MC



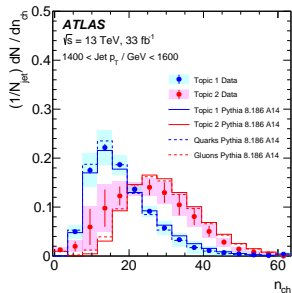
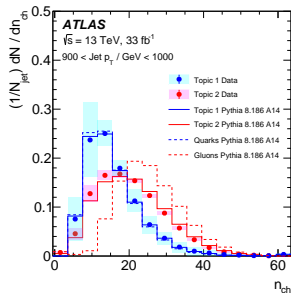
# Jet fragmentation properties | Unfolded data vs MC



# Jet fragmentation properties | Topic distributions



# Jet fragmentation properties | Uncertainties



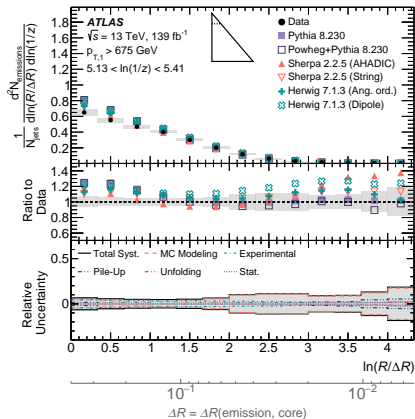
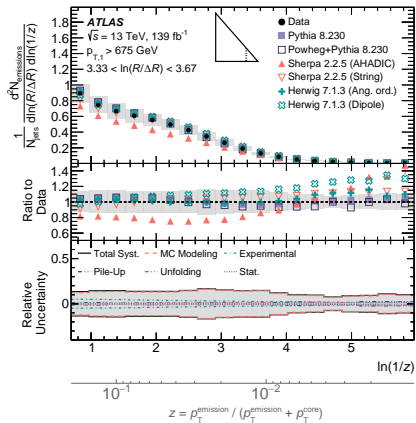


## Lund Plane measurement with charged particles

[\[arXiv:2004.03540\]](#)

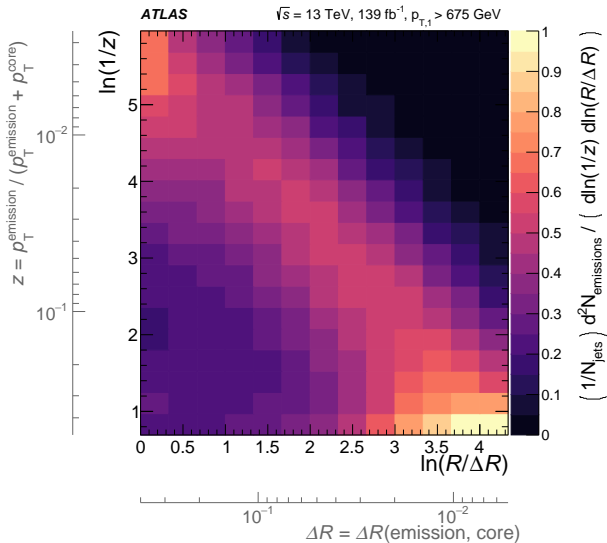
[Submitted to PRL]

# Lund plane measurement with charged particles



# Lund plane measurement with charged particles

## Unfolded Lund jet plane in data

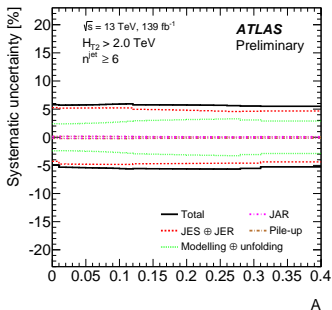
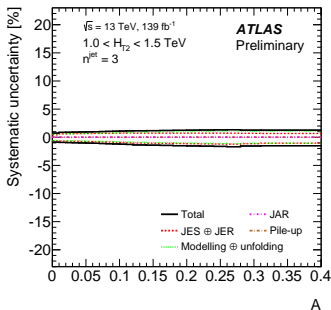
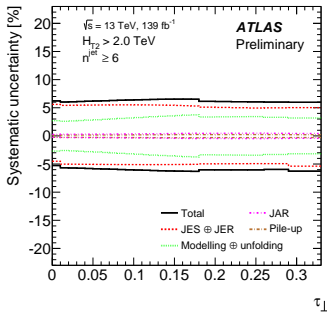
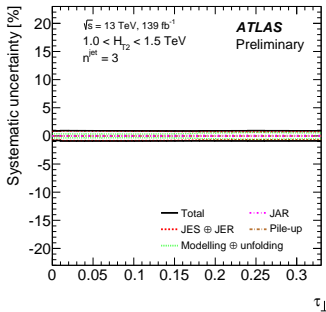


**Brand new results!**

**Measurement of hadronic event shapes in multijet final states at  
 $\sqrt{s} = 13 \text{ TeV}$**

[ATLAS-CONF-2020-011]

# Measurement of hadronic event shapes in multi-jet final states



# Measurement of the jet substructure observables at 13 TeV

[JHEP 08 (2019) 033]

[arXiv:1903.02942]

# Measurement of jet substructure observables at $\sqrt{s} = 13$ TeV

## Jet substructure observables in $t\bar{t}$ , W boson and dijet events

- ▶ anti- $k_t$   $R = 1.0$  jets groomed using trimming ( $R_{\text{sub}} = 0.2$ ,  $f_{\text{cut}} = 5\%$ ) and soft-drop ( $\beta = 0$ ,  $\zeta_{\text{cut}} = 0.1$ )
- ▶ Unfolded data distributions are compared to various MC event generators
- ▶ Cluster-level uncertainties on the overall shape and scale of the observables

$$e_n^{(\beta)} \equiv \frac{E_{\text{CF}n}(\beta)}{E_{\text{CF}1}(\beta)^n} \quad ; \quad E_{\text{CF}1}(\beta) \equiv \sum_{i \in J} p_{T_i}$$

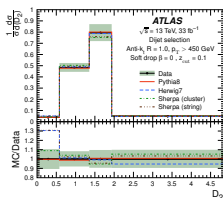
$$E_{\text{CF}2}(\beta) \equiv \sum_{i < j \in J} p_{T_i} p_{T_j} (\Delta R_{ij})^\beta$$

$$E_{\text{CF}3}(\beta) \equiv \sum_{i < j < k \in J} p_{T_i} p_{T_j} p_{T_k} (\Delta R_{ij} \Delta R_{ik} \Delta R_{jk})^\beta$$

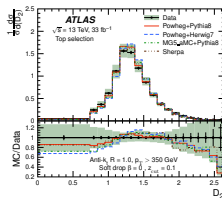
Observable:  $D_2^{(\beta)} \equiv \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^3}$

In general, reasonable agreement within uncertainties, with some discrepancies

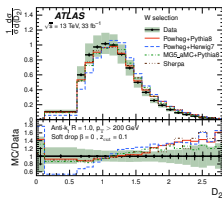
### Dijet selection



### $t\bar{t}$ selection



### W selection



# Measurement of jet substructure observables at $\sqrt{s} = 13$ TeV

Jet substructure observables in  $t\bar{t}$ , W boson and dijet events

- ▶ anti- $k_t$   $R = 1.0$  jets groomed using trimming ( $R_{\text{sub}} = 0.2$ ,  $f_{\text{cut}} = 5\%$ ) and soft-drop ( $\beta = 0$ ,  $\zeta_{\text{cut}} = 0.1$ )
- ▶ Unfolded data distributions are compared to various MC event generators
- ▶ Cluster-level uncertainties on the overall shape and scale of the observables

Observable:  $D_2^{(\beta)} \equiv \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^3}$

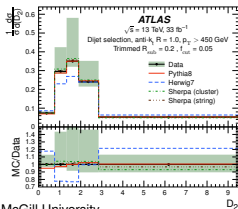
$$e_n^{(\beta)} \equiv \frac{E_{\text{CF}n}(\beta)}{E_{\text{CF}1}(\beta)^n} \quad ; \quad E_{\text{CF}1}(\beta) \equiv \sum_{i \in J} p_{T_i}$$

$$E_{\text{CF}2}(\beta) \equiv \sum_{i < j \in J} p_{T_i} p_{T_j} (\Delta R_{ij})^\beta$$

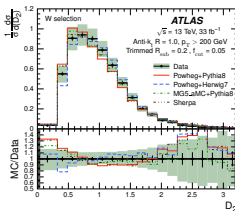
$$E_{\text{CF}3}(\beta) \equiv \sum_{i < j < k \in J} p_{T_i} p_{T_j} p_{T_k} (\Delta R_{ij} \Delta R_{ik} \Delta R_{jk})^\beta$$

In general, reasonable agreement within uncertainties, with some discrepancies

Dijet selection



W selection





**Properties of  $g \rightarrow b\bar{b}$  at small opening angles at  $\sqrt{s} = 13$  TeV**

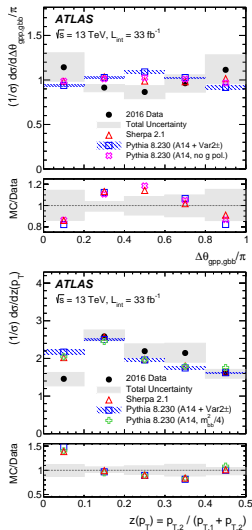
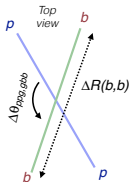
[Phys. Rev. D 99 (2019) 052004]

[arXiv:1812.09283]

# Properties of $g \rightarrow b\bar{b}$ at small opening angles at $\sqrt{s} = 13$ TeV

Main background in analyses involving boosted Higgs decaying into  $b$ -quarks

- ▶  $R = 1$  anti- $k_t$  trimmed jets w/  $p_T > 450$  GeV and  $|\eta| < 2$
- ▶ At least two  $R = 0.2$  anti- $k_t$  jets from tracks ghost-matched to  $R = 1.0$  jets
- ▶ Leading track-jet  $b$ -tagged by MV2c10 algorithm (60% efficiency)
- ▶ The contribution from  $R = 1.0$  jets that don't have 2  $b$ -tagged jets is subtracted from data using template fits
- ▶ Systematics: Jet energy scale, unfolding and theoretical modeling uncertainties dominate
- ▶ Significant differences observed b/w data and MC predictions



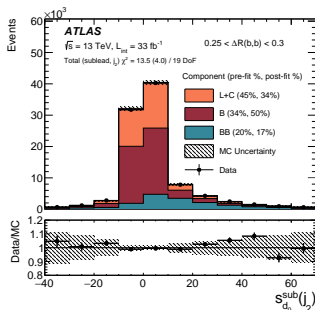
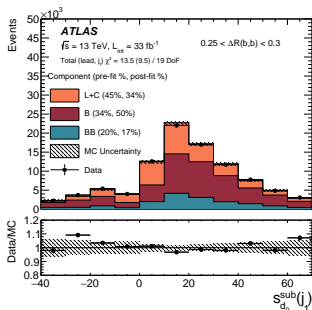
## Properties of $g \rightarrow b\bar{b}$ at small opening angles at $\sqrt{s} = 13$ TeV

Summary of systematic uncertainty sizes for each observable for the normalized differential cross sections

	$\Delta R(b, b)$	$\Delta\theta_{\text{ppg, gbb}}$	$z(p_T)$	$\log(m_{bb}/p_T)$
Calorimeter jet energy	2–3%	2–3%	2–6%	2–4%
Flavor tagging	<1%	<1%	<1%	<1%
Tracking	1–2%	1–2%	2–4%	1–2%
Background fit	1%	1%	1–2%	2%
Unfolding method	2–3%	2%	2–4%	2–5%
Theoretical modeling	3–10%	2–13%	3–10%	4–11%
Statistical	1%	1%	2%	1%
Total	3–10%	3–10%	3–14%	4–12%

## Properties of $g \rightarrow b\bar{b}$ at small opening angles at $\sqrt{s} = 13$ TeV

- ▶ The contribution from large- $R$  jets that do not have two associated track-jets containing B-hadrons is subtracted from data, before correcting for detector effects
- ▶ Correction factors are determined from data template fits to the signed impact parameter distribution ( $s_{d_0}$ ) and applied for each bin of the four observables
- ▶ In each bin, the distribution of  $s_{d_0}$  is fitted to data using templates from simulation while letting the fraction of each flavor component float in the fit.
- ▶ For a given track,  $s_{d_0} = s_j |d_0| / \sigma(d_0)$ , where  $d_0$  is the transverse impact parameter relative to the beam-line,  $\sigma(d_0)$  is the uncertainty in  $d_0$  from the track fit, and the variable  $s_j$  is the sign of  $d_0$  with respect to the jet axis:  $s_j = +1$  if  $\sin(\phi_{\text{jet}} - \phi_{\text{track}}) \cdot d_0 > 0$  and  $s_j = -1$  otherwise.



# Properties of $g \rightarrow b\bar{b}$ at small opening angles at $\sqrt{s} = 13$ TeV

Main background source in analyses involving boosted Higgs decaying into  $b$ -quarks

- ▶  $R = 0.2$  anti- $k_t$  jets from tracks are ghost-matched to  $R = 1.0$  anti- $k_t$  trimmed jets
- ▶ The contribution from  $R = 1.0$  jets that don't have 2 track-jets containing B-hadrons is subtracted from data using template fits
- ▶ Unfolding to the particle level
- ▶ **Significant differences observed b/w data and MC predictions**

