



Recent probes of perturbative QCD calculations with jets at ATLAS and CMS

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Large momentum transfers seen at the LHC allow us to probe pQCD

- ▶ Jet final states can be sensitive to parton structure (PDFs), strong coupling, matrix elements

$$\sigma_{P_1, P_2 \rightarrow X} = \sum_{ij} \int dx_1 dx_2 f_{i, P_1}(x_i, \mu_f) f_{j, P_2}(x_j, \mu_f) \times \sigma_{ij \rightarrow X} \left(x_1 p_1, x_2 p_2, \alpha(\mu_r^2), \frac{Q^2}{\mu_f^2} \right)$$

Will discuss four recent ATLAS and CMS results, looking at jet final states sensitive to the above

- ▶ dijet and inclusive jet: probe NNLO calculation and scale choices
- ▶ triple differential dijet: constrain PDFs
- ▶ azimuthal correlations: compare MC generators
- ▶ azimuthal correlation ratio: $\alpha_s(Q)$ extraction at high Q

ATLAS: Dijet and inclusive jet

[arXiv:1711.02692](https://arxiv.org/abs/1711.02692)

measure inclusive jet and dijet double-differential cross sections:

$$\frac{d^2\sigma}{dp_T dy} = \frac{N_{\text{jets}}}{\mathcal{L} \Delta p_T \Delta y} \qquad \frac{d^2\sigma}{dm_{jj} dy^*} = \frac{N_{\text{dijet}}}{\mathcal{L} \Delta m_{jj} \Delta y^*}$$

- ▶ Use 3.2 fb^{-1} of $\sqrt{s} = 13 \text{ TeV}$ data
- ▶ Compare to NLO and state of the art NNLO calculations
- ▶ Also probe how the choice of scale effects on the inclusive jet calculations

Event selection and triggering

- ▶ Use a suite of single jet triggers to select inclusive jet events
- ▶ dijet selection: use trigger pairing with σ based on pairings

$$y^* = |y_1 - y_2|/2$$

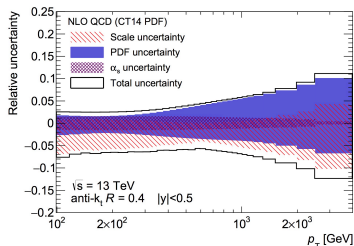
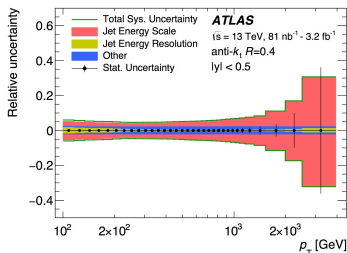
Data is corrected for detector effects using the iterative dynamically stabilised (IDS) unfolding method

NLO calculated using NLOJET++

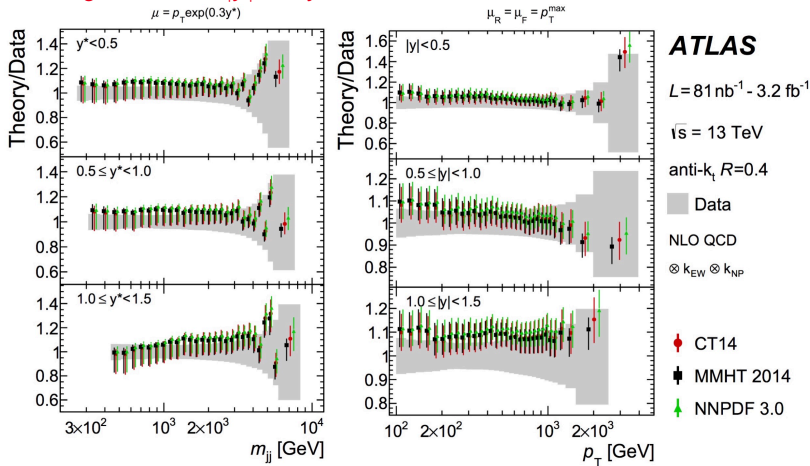
- ▶ Calculated using 6 different NLO PDF sets provided by LHCPDF6

NNLO calculation provided by [J. Currie](#), [E. Glover](#) and [J. Pires](#)

- ▶ Non perturbative correction factors are derived bin by bin, comparing a LO MC with and without showering and hadronisation
- ▶ Electroweak corrections are taken from [S. Dittmaier](#), [A. Huss](#) and [C. Speckner](#)



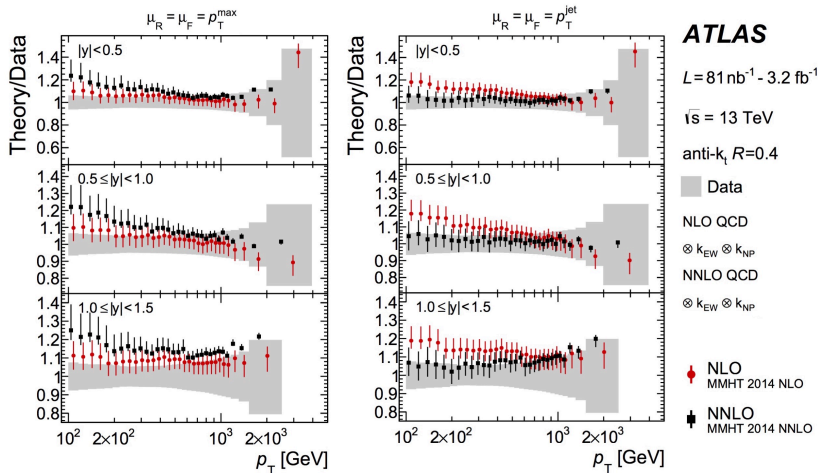
Showing a subset of $|y|$ and y^* bins and PDF sets



Fair agreement is seen in most of the phase space, with some tensions in the 1.5-2.5 y^* range for the dijet selection.

Tension between data and theory is observed in the inclusive measurement when considering the full phase-space

Showing NNLO/data comparison with different scale choices



Effects of two different scale choices considered

Either NLO or NNLO has better agreement based on choice

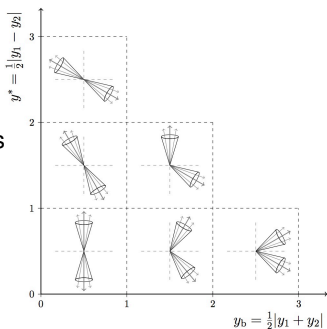
CMS: Dijet triple-differential

arXiv:1705.02628

Measure triple differential cross section using 19.7 fb^{-1} of 8 TeV data

$$\frac{d^3\sigma}{d\mathbf{p}_{T,\text{avg}} dy^* dy_b} = \frac{1}{\epsilon \mathcal{L}^{\text{eff}}} \frac{N}{\Delta p_{T,\text{avg}} \Delta y^* \Delta y_b}$$

- ▶ Comparisons made to NLO predictions
- ▶ Binning in y_b results in selections with different partonic subprocesses
- ▶ Data used to constrain PDFs and extract a value for α_s



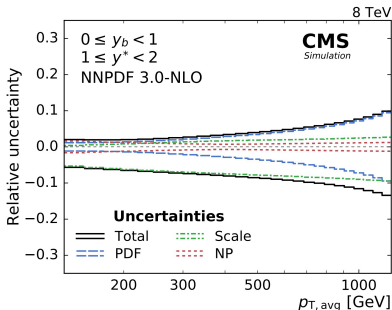
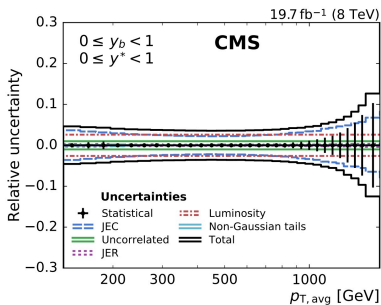
Use a suite of single jet triggers and select the leading two central ($|y| < 3.0$) jets from events with at least two jets

Distributions corrected to particle level using iterative D'Agostini algorithm

- ▶ Response matrix uses pseudo-events weighted to NLO prediction, smeared using the jet p_T
- ▶ Jet energy correction is the largest experimental uncertainty

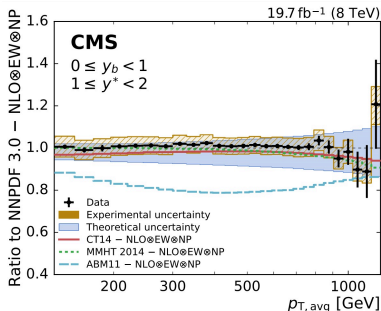
NLO predictions calculated using NLOJET++

- ▶ non perturbative corrections are applied by comparing LO MC with and without hadronisation and MPI
- ▶ Electroweak correction also applied, from [arXiv: 1210.0438](https://arxiv.org/abs/1210.0438)



good agreement between data and NLO,
 expect in regions of high $p_{T,avg}$ and y_b which are sensitive to high- x PDF values

ABM11 PDFs underestimate the data for $y_b < 2.0$

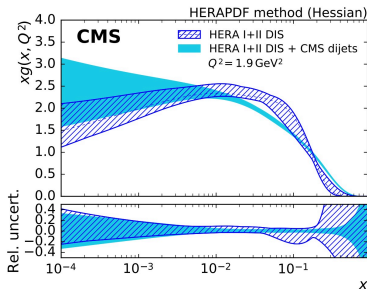


Constraints on PDFs are obtained by a fit including the results and HERA DIS data.

- ▶ Fits are performed using the XFitter framework, at NLO
- ▶ comparisons are made on the quality of the fit

Data set(s)	n_{dof}	χ^2	χ^2/n_{dof}	χ^2	χ^2/n_{dof}
HERA data	1040	1211.00	1.16	—	—
HERA & CMS data	1162	—	—	1372.52	1.18

- Uncertainty in the gluon pdf significantly reduced in the high x region, some reduction also seen in valence and sea quarks
- Also noticeable change in the shape



By repeating the fit while leaving $\alpha_s(M_Z)$ as a free parameter, one obtains

$$\alpha_s(M_Z) = 0.1199 \pm 0.0015(\text{exp})_{-0.0020}^{+0.0032}(\text{mod})$$

which is in agreement with previous CMS and ATLAS measurements and the world average.

CMS: Azimuthal correlations

arXiv:1712.05471

Consider leading two jets in 2,3,4 inclusive jet events and measure

$$\frac{1}{\sigma} \frac{d\sigma}{d\Delta\phi_{12}} \quad \text{and} \quad \frac{1}{\sigma} \frac{d\sigma}{d\Delta\phi_{2j}^{\min}}$$

- ▶ $\Delta\phi_{2j}^{\min}$ is sensitive to lower p_T jets and adds additional information
- ▶ Compare to various LO and NLO predictions

Used 35.9 fb^{-1} of 13 TeV data

- ▶ Use a selection of five single jet triggers to select events with at least one jet with $p_T > 200 \text{ GeV}$
- ▶ Study different MC generators at different orders
- ▶ Evaluate performance of parton showers and matching

Use a matrix inversion algorithm to correct for detector effects

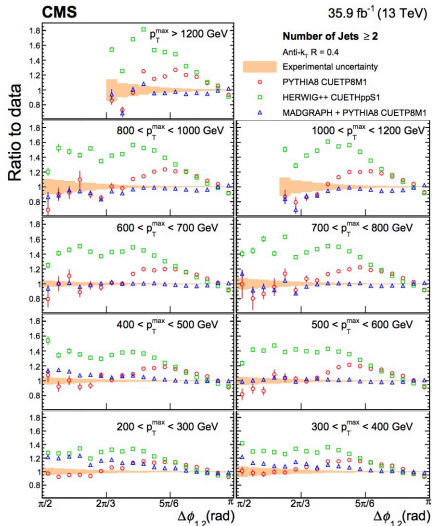
JES, JER and unfolding systematics are the largest experimental uncersts

There were a number of event generators used for comparison

Matrix element generator	Simulated diagrams	PDF set	Tune
PYTHIA 8.219 [9]	2→2 (LO)	NNPDF2.3LO [14, 15]	CUETP8M1 [13]
HERWIG++ 2.7.1 [10]	2→2 (LO)	CTEQ6L1 [16]	CUETHppS1 [13]
MADGRAPH5_aMC@NLO 2.3.3 [17, 18] + PYTHIA 8.219 [9]	2→2, 2→3, 2→4 (LO)	NNPDF2.3LO [14, 15]	CUETP8M1 [13]
POWHEG V2.Sep2016 [20-22] + PYTHIA 8.219 [9]	2→2 (NLO), 2→3 (LO)	NNPDF3.0NLO [28]	CUETP8M1 [13]
POWHEG V2.Sep2016 [20-22] + PYTHIA 8.219 [9]	2→3 (NLO), 2→4 (LO)	NNPDF3.0NLO [28]	CUETP8M1 [13]
POWHEG V2.Sep2016 [20-22] + HERWIG++ 2.7.1 [10]	2→2 (NLO), 2→3 (LO)	NNPDF3.0NLO [28]	CUETHppS1 [13]
HERWIG 7.0.4 [23]	2→2 (NLO), 2→3 (LO)	MMHT2014 [29]	H7-UE-MMHT [23]

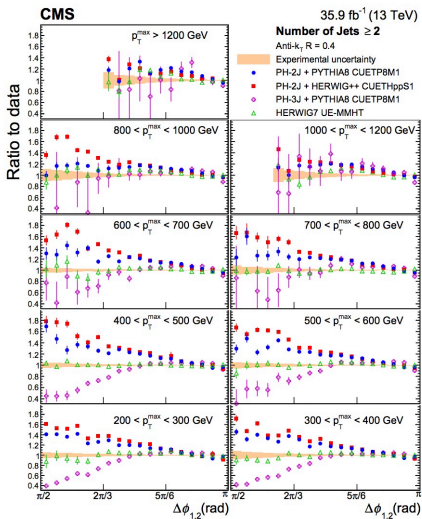
Largest theoretical uncertainty comes from parton showering

- ▶ Evaluated using Pythia8 by scaling renormalisation scale for ISR and FSR independently up and down
- ▶ For $\Delta\phi_{12}$ range from $< 5\%$ at π upto 40-60% at $(\approx \pi/2)$ for greater p_T^{\max} in the 2-jet case. Don't exceed 20% for the 3,4-jet case



LO results

- ▶ MadGraph+Pythia8 provides the best description of the data
- ▶ Pythia8 performs better than Herwig++



NLO results

- ▶ HERWIG7 provides the best description of the data
- ▶ PowHeg 2j and 3j have large deviations from the measured data
- ▶ Herwig++ and Pythia8 use different α_s values for (I)FSR and have a different upper scale for PS emissions
- ▶ HERWIG 7 uses MC@NLO method of combining PS with particle level, which here seems to perform better than the POWHEG method

ATLAS: Azimuthal decorrelations

arXiv:1805.04691

Measure the following ratio:

$$R_{\Delta\phi}(H_T, y^*, \Delta\phi_{\max}) = \frac{\frac{d^2\sigma_{\text{dijet}}(\Delta\phi_{\text{dijet}} < \Delta\phi_{\max})}{dH_T dy^*}}{\frac{d^2\sigma_{\text{dijet}}(\text{inclusive})}{dH_T dy^*}}$$

- ▶ Ratio has smaller dependence on PDFs in α_s extraction and running studies

Additional cuts applied on y_{\max}^* , y_{\max}^b and p_{T1}/H_T

- ▶ ensure that jets are within $|y| < 2.5$ and are thus well measured
- ▶ Reduces contributions from events with 4 or more jets, less sensitive to higher orders in α_s

Use a set of single and multi jet triggers in each H_T bin

$$y_{\max}^* < 2.0, \quad y_{\max}^b < 0.5, \quad p_{T1}/H_T > 1/3$$

pQCD calculations at fixed order in α_s with NP corrections

- ▶ Calculations carried out using NLOJET++
- ▶ predictions for $R_{\Delta\phi}$ are calculated at NLO, expect for $\Delta\phi_{\max} = 2\pi/3$ (4 jet quantity)
- ▶ evolution of α_s computed at a NLL approximation

A set of various PDF sets were used

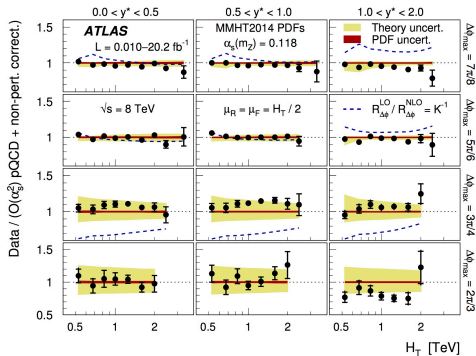
- ▶ Sets obtained for a series of discrete α_s values with $\Delta\alpha_s = 0.001$
- ▶ continuous dependance obtained by interpolation
- ▶ MMHT2014 used as nominal: largest range of values (0.108 - 0.128)

uncertainties

- ▶ Uncertainties on pQCD by varying μ_f and μ_r
- ▶ MMHT2014 PDF uncertainties used, an envelope of the results obtained with other sets is also used
- ▶ NP corrections obtained from [M. Wobisch, et al.](#)

All predictions (including LO) are consistent with the data

A subset of the datapoints are used for the α_s extraction

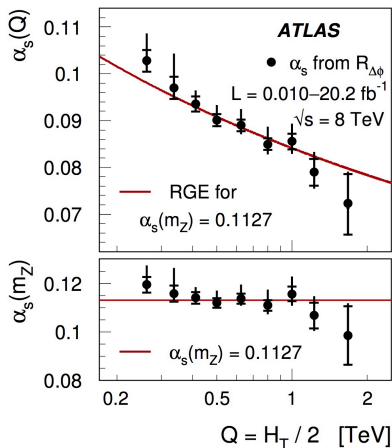


- ▶ points where calculation is most reliable (scale dependence)
- ▶ data points are combined if their phase space is orthogonal
- ▶ largest cancellation of PDF uncertainty
- ▶ smallest stat uncertainty

The datapoints from the region $0 < y^* < 0.5$ and $0.5 < y^* < 1$ for $\Delta\phi_{\max} = 7\pi/8$ are used
 Scale uncertainties are the largest systematic

α_s extracted as a function of $Q = H_T/2$

- ▶ the results are extracted from a Minuit χ^2 fit
- ▶ Nine $\alpha_s(Q)$ values are extracted $262 < Q \leq 1675$ GeV
- ▶ Separate χ^2 fits are made for scale variations and also for CT14, NNPDF2.3, ABMP16, HERAPDF 2.0
- ▶ Biggest difference of +0.0029 observed with HERAPDF 2.0
- ▶ A series of systematic studied also investigated the effect of other analyses choices (suggest result is rather independent of the analysis choice)



Final value: $\alpha_s(m_Z) = 0.1127^{+0.0063}_{-0.0027}$, consistent with global value

Large scope for doing precision measurements at the LHC

- ▶ Tensions seen in some region of phase space
- ▶ New complimentary measurements of α_s
- ▶ Help reduce PDF uncertainty in certain regions of phase space

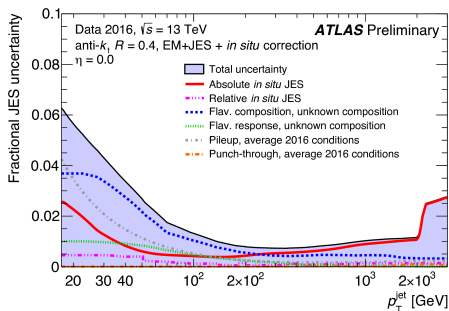
BACKUP

Multi-step Jet energy calibration

1. **Pile up correction:** Based on jet area, μ and N_{PV}
2. **Jet energy Scale:** Energy corrected for mean detector response in η , p_T
3. **Global sequential:** Based on topology and associated tracks
4. **In situ calibration:** measurements used to correct remaining data/MC difference

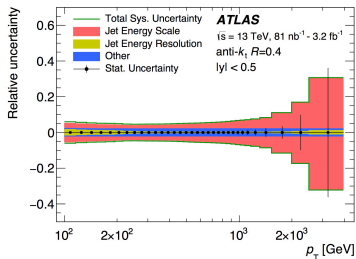
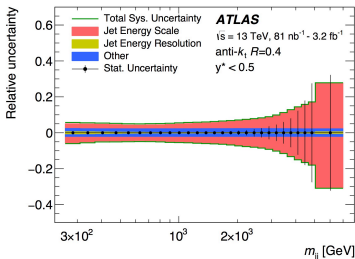
Energy scale uncertainties

The systematic uncertainties of the above steps are combined as independent components



Data is corrected for detector effects using the iterative dynamically stabilised (IDS) unfolding method

- ▶ statistical uncertainties are propagated through the unfolding using an ensemble of pseudo-experiments (bootstrap method)
- ▶ the various JES uncertainties are propagated through the unfolding using ± 1 sigma variations and pseudo data (bootstrap method) to evaluate statistical significance

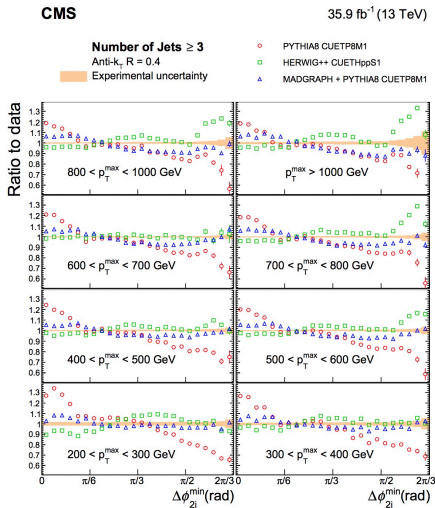


P values from the comparison between data and the NLO predictions for inclusive jet selection

Rapidity ranges	P_{obs}				
	CT14	MMHT 2014	NNPDF 3.0	HERAPDF 2.0	ABMP16
$p_{\text{T}}^{\text{max}}$					
$ y < 0.5$	67%	65%	62%	31%	50%
$0.5 \leq y < 1.0$	5.8%	6.3%	6.0%	3.0%	2.0%
$1.0 \leq y < 1.5$	65%	61%	67%	50%	55%
$1.5 \leq y < 2.0$	0.7%	0.8%	0.8%	0.1%	0.4%
$2.0 \leq y < 2.5$	2.3%	2.3%	2.8%	0.7%	1.5%
$2.5 \leq y < 3.0$	62%	71%	69%	25%	55%
$p_{\text{T}}^{\text{jet}}$					
$ y < 0.5$	69%	67%	66%	30%	46%
$0.5 \leq y < 1.0$	7.4%	8.9%	8.6%	3.4%	2.0%
$1.0 \leq y < 1.5$	69%	62%	68%	45%	54%
$1.5 \leq y < 2.0$	1.3%	1.6%	1.4%	0.1%	0.5%
$2.0 \leq y < 2.5$	8.7%	6.6%	7.4%	1.0%	3.6%
$2.5 \leq y < 3.0$	65%	72%	72%	28%	59%
χ^2/dof all $ y $ bins					
$p_{\text{T}}^{\text{max}}$	419/177	431/177	404/177	432/177	475/177
$p_{\text{T}}^{\text{jet}}$	399/177	405/177	384/177	428/177	455/177

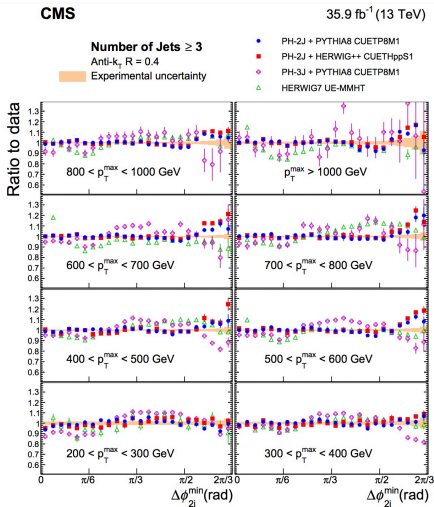
P values from the comparison between data and the NLO predictions for the dijet selection

y^* ranges	P_{obs}				
	CT14	MMHT 2014	NNPDF 3.0	HERAPDF 2.0	ABMP16
$y^* < 0.5$	79%	59%	50%	71%	71%
$0.5 \leq y^* < 1.0$	27%	23%	19%	32%	31%
$1.0 \leq y^* < 1.5$	66%	55%	48%	66%	69%
$1.5 \leq y^* < 2.0$	26%	26%	28%	9.9%	25%
$2.0 \leq y^* < 2.5$	43%	35%	31%	4.2%	21%
$2.5 \leq y^* < 3.0$	45%	46%	40%	25%	38%
all y^* bins	8.1%	5.5%	9.8%	0.1%	4.4%



LO results

- ▶ Pythia8 has a larger deviation from the data than Herwig++
- ▶ MADGRAPH has reasonable agreement, but gets worse in the 4-jet case



NLO results

- ▶ PowHeg 2j with either Pythia8 or Herwig++ provides the best agreement with data
- ▶ PowHeg 3j results are stat limited at high p_T^{\max} but have a worse agreement

Consider anti- k_t $R=0.6$ jets within detector acceptance ($|\eta| < 4.9$)

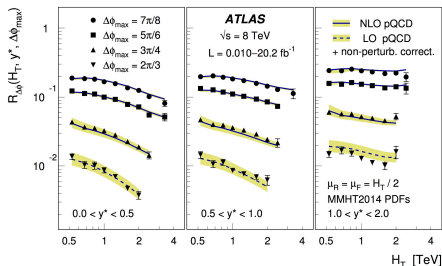
- ▶ multijet jet energy calibration is applied (pile up, area, JES, residual)
- ▶ Jet quality cuts applied to remove remaining pile up jets

The $R_{\Delta\phi}$ distributions are unfolded bin-by-bin to correct for detector effects

- ▶ Bin width is set be larger than $\Delta\phi$ resolution
- ▶ Cross checked using iterative unfolding procedure
- ▶ Corrections are small, uncertainties typically below 1%

62 sources of systematic uncertainty considered

- ▶ Mainly from the jet energy calibration
- ▶ Also includes angular and energy resolution
- ▶ typically between 1% and 1.5%



Use a matrix inversion algorithm to correct for detector effects

- ▶ Reponse matrix created by the convolution of the generator level observables with the $\Delta\phi$ resolution
- ▶ Cross checked using samples with full detector sim
- ▶ Bin width set to be between 5 to 10 times $\Delta\phi$ resolution

Consider three main sources of systematic uncertainty

JES: 3% (at $\pi/2$) to 0.1% (at π) $\Delta\phi_{12}$ uncertainty and a 0.1% to 2% $\Delta\phi_{2j}^{\min}$

JER: 1% (at $\pi/2$) to 0.1% (at π) $\Delta\phi_{12}$ uncertainty and $< 0.5\%$ $\Delta\phi_{2j}^{\min}$ uncertainty

unfolding Tested by changing choice of generator and varying the $\Delta\phi$ resolution.
Total uncertainty 0.2%

ATLAS: Softdrop mass

arXiv:1711.08341

New grooming techniques allow for more precise calculations

$$\text{measure } \frac{1}{\sigma} \frac{d\sigma}{d \log_{10} \rho^2}, \text{ where } \rho = m^{\text{softdrop}} / p_{\text{T}}^{\text{ungroomed}}$$

- ▶ Jet substructure techniques have been widely used for tagging
- ▶ General procedures now exist for understanding IR and collinear safe observables at LL accuracy
- ▶ softdrop: removes NGLs, allow predictions beyond LL
- ▶ softdrop mass has been calculated at both NLO with NLL and LO NNLL accuracy

Measure for three different softdrop parameters ($\beta = 0, 1, 2$)

The ungroomed jet p_{T} is used since, in some cases ($\beta = 0$), its collinear unsafe

softdrop grooming: start with a Cambridge/Aachen (angular) jet

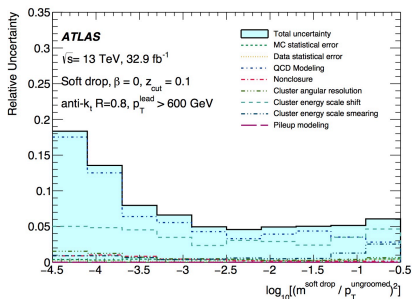
- ▶ un-do the clustering, at each step check the following (soft drop) condition for the protojets, j_i and j_j

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

- ▶ if the condition passes passes, terminate the algorithm, else discard discard the branch with the lowest p_T and iterate

Use a single jet trigger and select leading two Anti- k_t jets with $|\eta| < 1.5$

- ▶ Use iterative bayesian method to correct for detector effects
- ▶ experimental uncertainties, apply variations to calo-cell clusters
- ▶ QCD fragmentation, compare Pythia, Sherpa and Herwig++

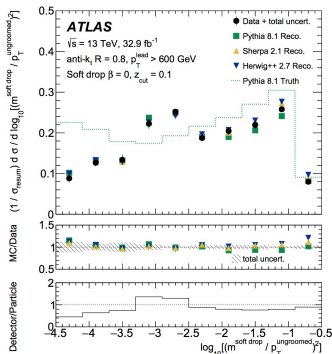


Results compared to

- ▶ Predictions from Pythia, Sherpa, Herwig++ generators
- ▶ NLO+NNLL predictions taken from S. Marzani, L. Schunk and G. Soyez [arXiv:1704.02210](https://arxiv.org/abs/1704.02210)
- ▶ LO+NNLL prediction from C. Frye et al. [arXiv:1704.02210](https://arxiv.org/abs/1704.02210), [arXiv:0808.1269](https://arxiv.org/abs/0808.1269)

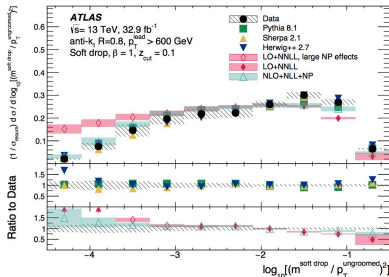
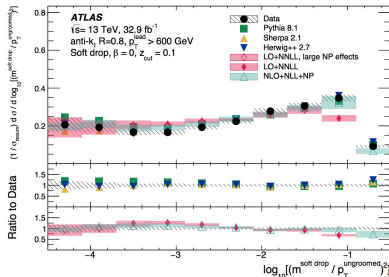
expect accuracies to differ in different regions of $\log_{10}(\rho^2)$

- ▶ resummation dominates:
 $-3.7 < \log_{10}(\rho^2) < -1.7$
- ▶ soft and collinear emissions
 $\log_{10}(\rho^2) < -3.7$
NP effects are larger
- ▶ Fixed order region:
 $\log_{10}(\rho^2) > -1.7$
Wide angle emissions



MC generators and LO+NNLL prediction should be most accurate in the resummation region

NLO+LL should be more accurate at $\log_{10}(\rho^2) > -1.7$



As β increases, soft drop removes less radiation so NP corrections become more important

- ▶ good agreement for all predictions and MC within the resummation region
- ▶ at large β values, larger difference between MC and LO+NNLL
- ▶ at low $\log_{10}(\rho^2)$ the LO+NNLL starts to over predict the data