

Jets and Jet Substructure



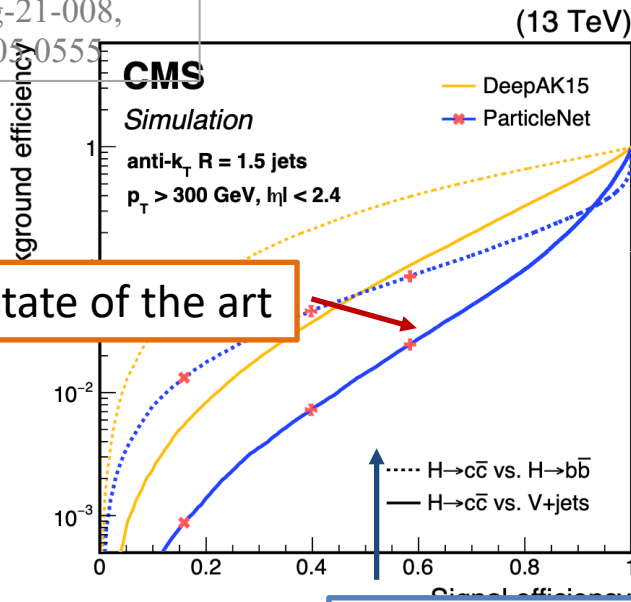
Matthew Schwartz
Harvard University
Aug 10, 2023

Recent progress in Jets

- Experimental progress
 - Improved calibration/modeling for run 3
 - Machine Learning methods expanding reach
 - Measuring many new observables
- Theory progress
 - Machine learning for tagging, measurement, searches
 - Resummed and fixed order calculations
 - Improvements in jet mass predictions
 - Energy-energy correlators
 - Top mass measurement improvements

Charm tagging

CMS-Hig-21-008,
arXiv:2205.05555



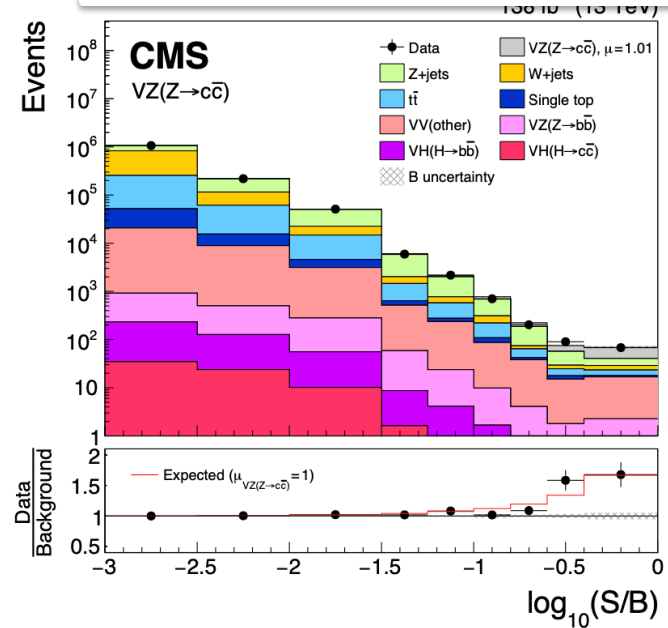
previous state of the art

“particle net”
ML algorithm

Particle net (arXiv:1902.08570)

- Graph neural network
- takes momenta of particles
- Includes tracks, particle id, etc

observation of $Z \rightarrow c\bar{c}$ at 5.7σ

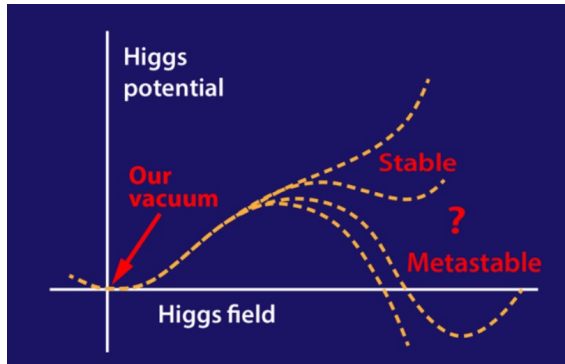


ratio to SM expectation

$$\mu_{VZ(Z \rightarrow c\bar{c})} = 1.01^{+0.23}_{-0.21}$$

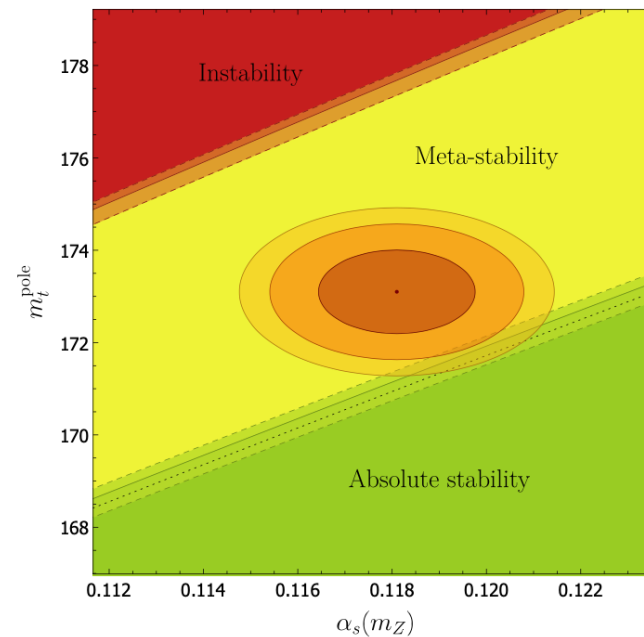
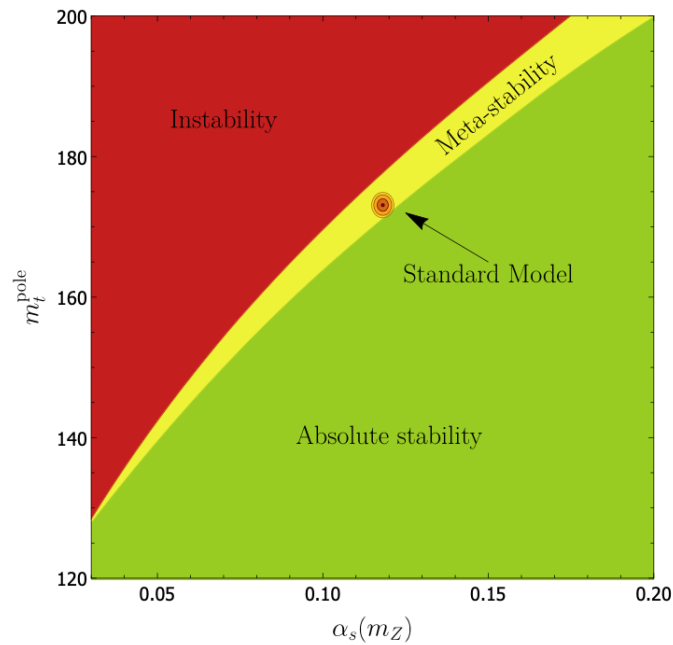
- Higgs to cc is now possible
 - Current limit is 14 x Standard Model
 - Statistics limited
 - Could be seen in run 3

Standard Model measurements are important!



Need to improve measurements of top quark mass and α_s

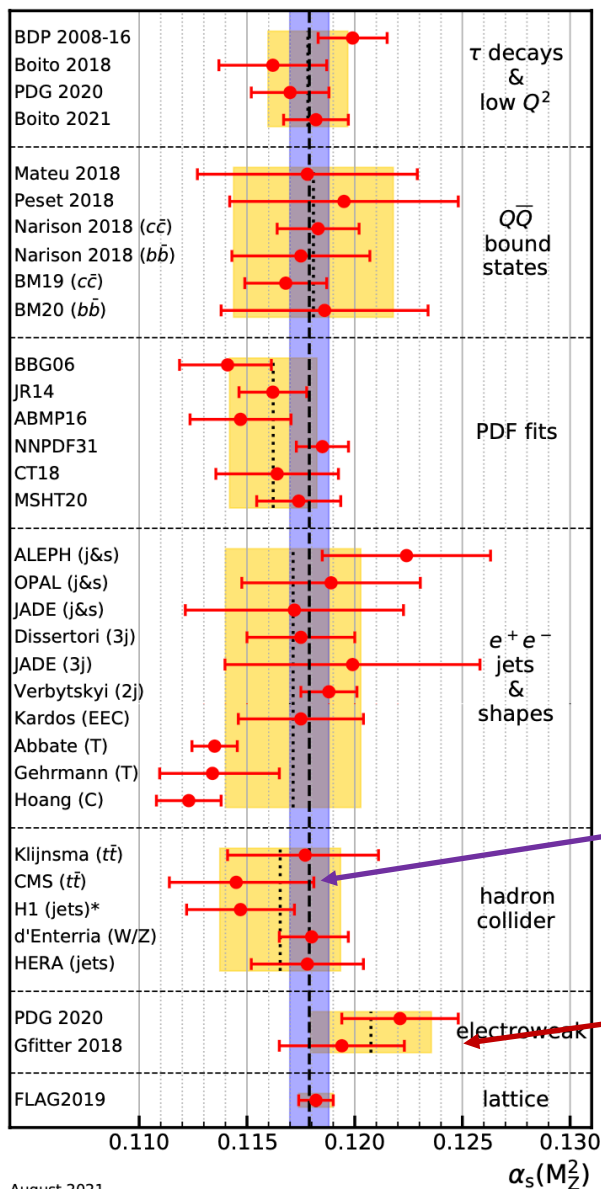
- Lifetime of the universe depends on their values
- Important measurements within LHC reach



Status of α_s , PDG 2021

$$\alpha_s(M_Z^2) = 0.1179 \pm 0.0009$$

0.8% uncertainty



Energy-energy correlators at e^+e^-

e^+e^- event shapes (thrust, C parameter)

Only LHC measurement included in 2021 PDG average

CMS $t\bar{t}$ total cross section

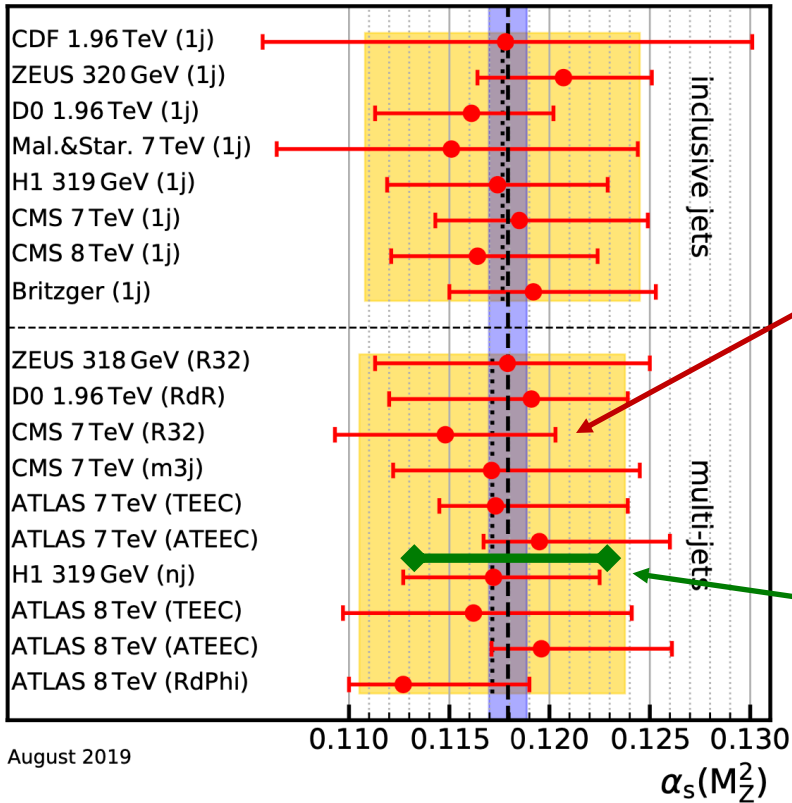
$$\alpha_s(M_Z^2) = 0.1145^{+0.0036}_{-0.0031}$$

Dominant measurements from Flavor Lattice Averaging Group

$$\alpha_s(M_Z^2) = 0.1182 \pm 0.0008$$

Hadron collider jet measurements

PDG 2021: inclusive and multi-jet measurements



CMS arXiv:1304.7498

- 3 to 2 jet cross section ratios (R32)
- 5 fb^{-1} @ 7 TeV CMS

$$\alpha_s(M_Z) = 0.1148 \pm 0.0014 (\text{exp.}) \pm 0.0018 (\text{PDF}) \pm 0.0050 (\text{theory})$$

$$= 0.1148 \pm 0.0055$$

5% uncertainty

ATLAS arXiv:2301.09351

- Transverse energy-energy correlators (TEECs)
- 139 fb^{-1} @ 13 TeV ATLAS

$$\alpha_s(m_Z) = 0.1175 \pm 0.0006 (\text{exp.})_{-0.0017}^{+0.0034} (\text{theo.})$$

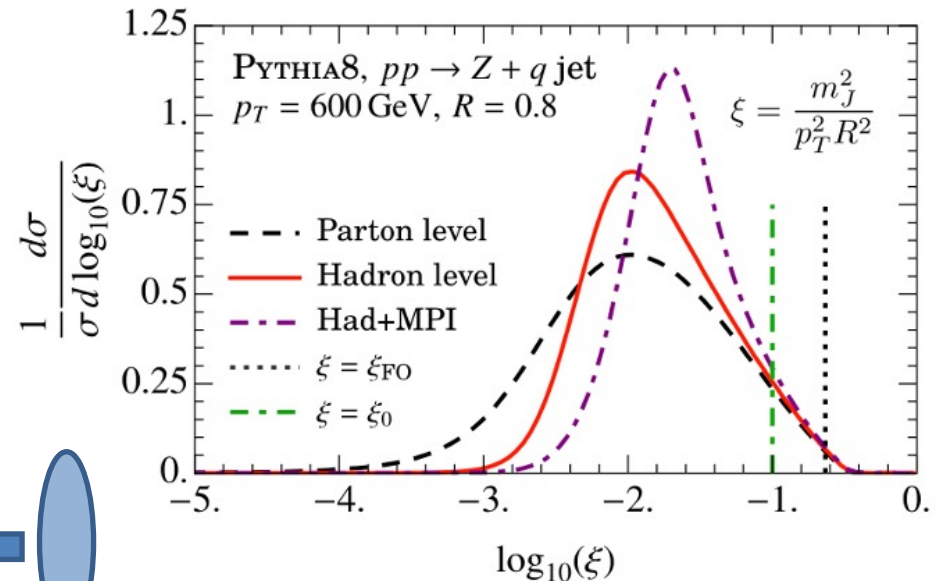
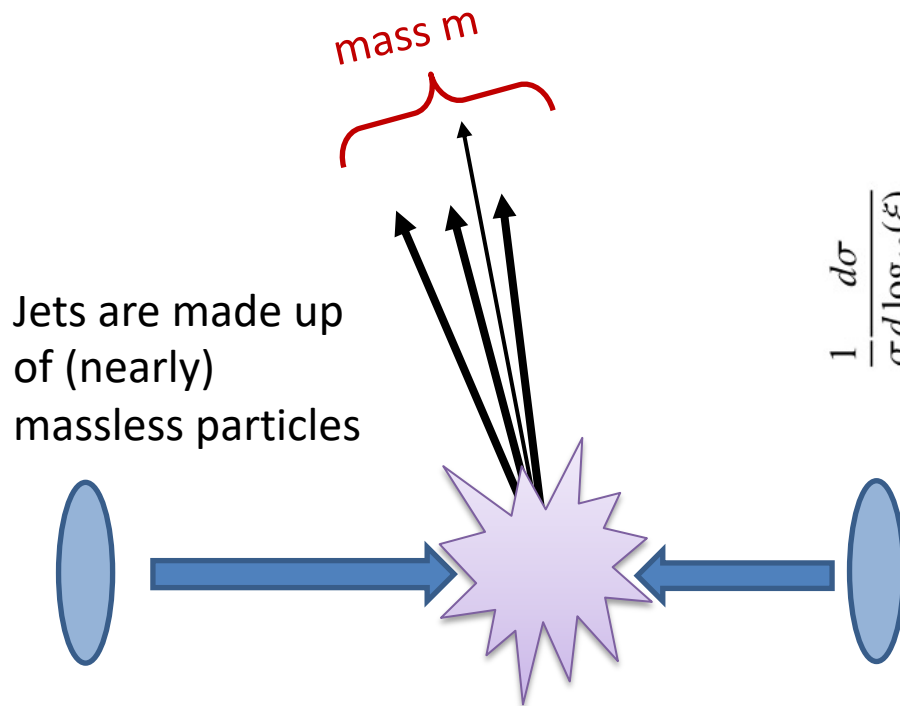
3% uncertainty

Can we get down to the 1% level with jets?

Jet mass measurements

Can we calculate the jet-mass distribution from first principles?

- Must avoid MC as much as possible to measure α_s

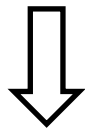
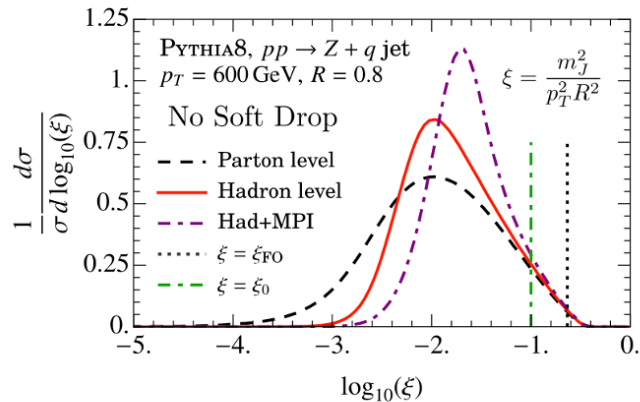


Very challenging theory calculation

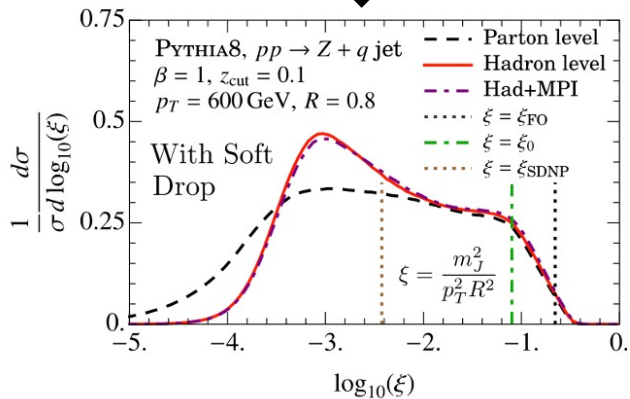
- Mass is sensitive to many things under poor theoretical control
 - Underlying event, pileup, hadronization corrections, etc.

Soft drop

- Removes soft (low energy) particles from a jet in a systematic way



apply soft drop



- Undo the clustering, starting from small angles

- Drop a particle if it is soft, meaning

$$\frac{p_{Ta}}{p_{Ta} + p_{Tb}} < z_{cut} \left(\frac{\theta_{ab}}{R} \right)^\beta$$

- If neither particle is soft at a given step, stop declustering and return the soft-drop jet

After soft drop

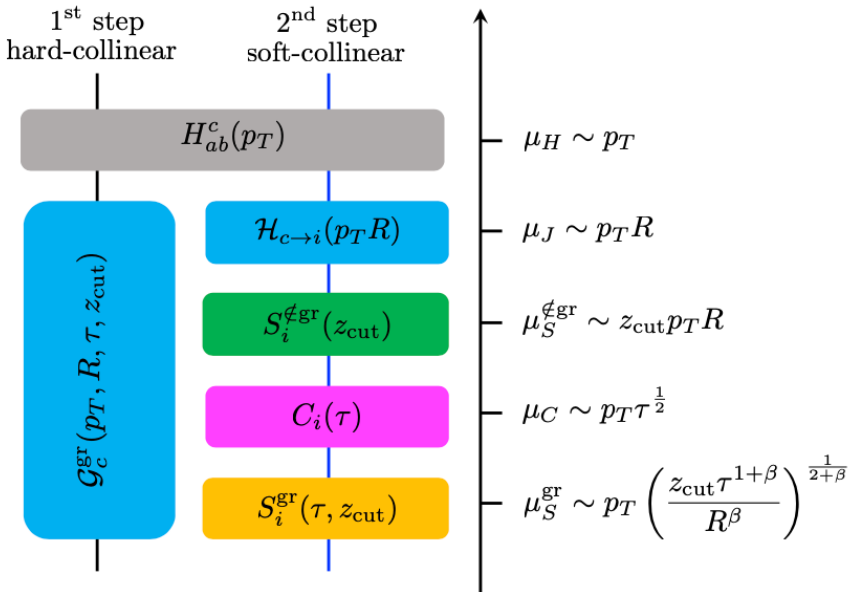
- Effect of MPI (underlying event) and pileup are tiny
- Region exists where hadronization corrections are small

Soft-drop jet mass offers potential for α_s measurement at the LHC

Factorization formula

$$\frac{d^3\sigma}{dp_T d\eta d\xi} = \sum_{abc} \int \frac{dx_a dx_b dz}{x_a x_b z} f_a(x_a, \mu) f_b(x_b, \mu) H_{ab}^c \left(x_a, x_b, \eta, \frac{p_T}{z}, \mu \right) \mathcal{G}_c(z, \xi, p_T, R, \mu)$$

PDFs \rightarrow $f_a(x_a, \mu) f_b(x_b, \mu)$
 Hard function \rightarrow H_{ab}^c
 jet function \rightarrow \mathcal{G}_c
 $\xi = \frac{m_J^2}{p_T^2 R^2}$
 soft-drop jet mass



Physics at many scales relevant

- Jet mass, jet energy (p_T)
- Collinear scale, soft scale, soft-collinear scale
- Soft drop cutoff scale ($z \tau$)
- ...

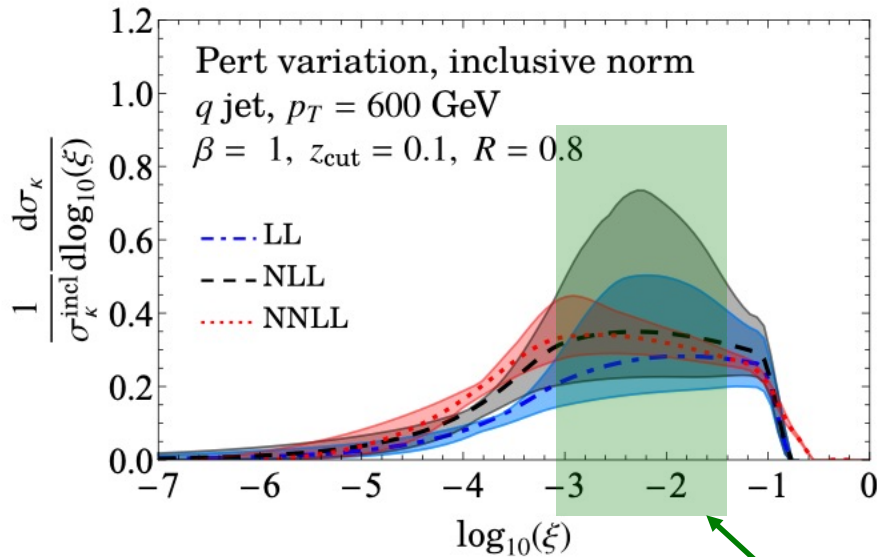
Although complicated, we can still understand it

- Frye, Larkoski, MDS, Yan 1603.09338 factorization
- Marzani et al. 1704.02210 power corrections
- Stewart, Hannesdottir, Pathak, MDS, Stewart 2210.04901
 - NNLL resummation with power corrections

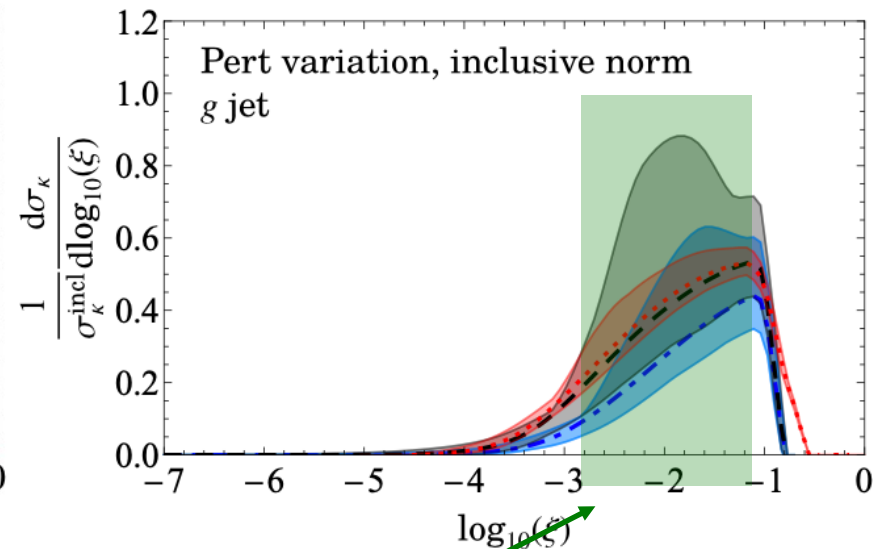
Perturbative Uncertainties

Stewart, Hannesdottir, Pathak, MDS, Stewart
arXiv:2210.04901

quarks



gluons



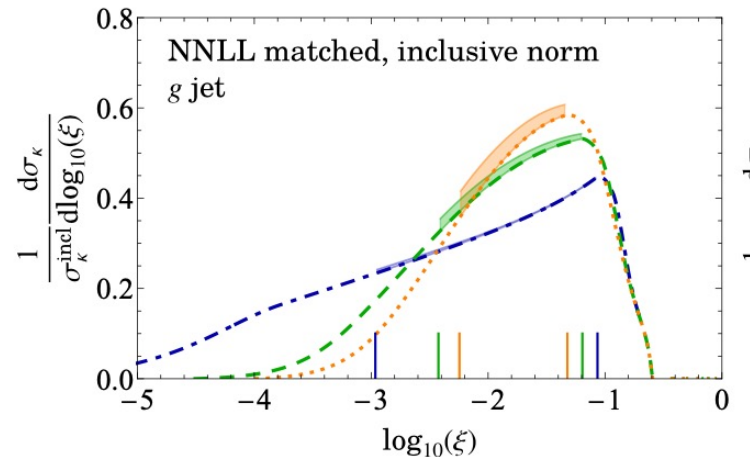
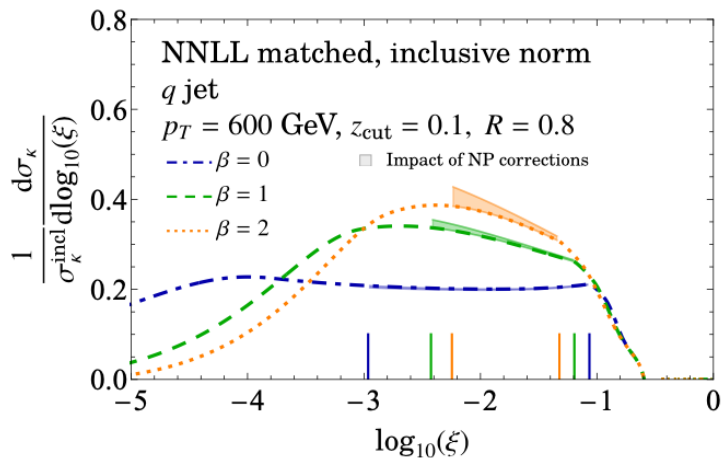
- Good perturbative control in fit region ($-3 < \log_{10} \xi < -1$)
- Good convergence from LL \rightarrow NLL \rightarrow NNLL

Non-perturbative Uncertainties

- Six non-perturbative shape-function parameters
- Central values fit to pythia

$$\Omega_{1,q}^{\oplus} = 0.55 \text{ GeV}, \quad \Upsilon_{1,0q}^{\ominus} = -0.73 \text{ GeV}, \quad \Upsilon_{1,1q}^{\ominus} = 0.90 \text{ GeV}, \quad \text{for quarks,}$$

$$\Omega_{1,g}^{\oplus} = 0.91 \text{ GeV}, \quad \Upsilon_{1,0g}^{\ominus} = -0.24 \text{ GeV}, \quad \Upsilon_{1,0g}^{\ominus} = 0.90 \text{ GeV}, \quad \text{for gluons.}$$

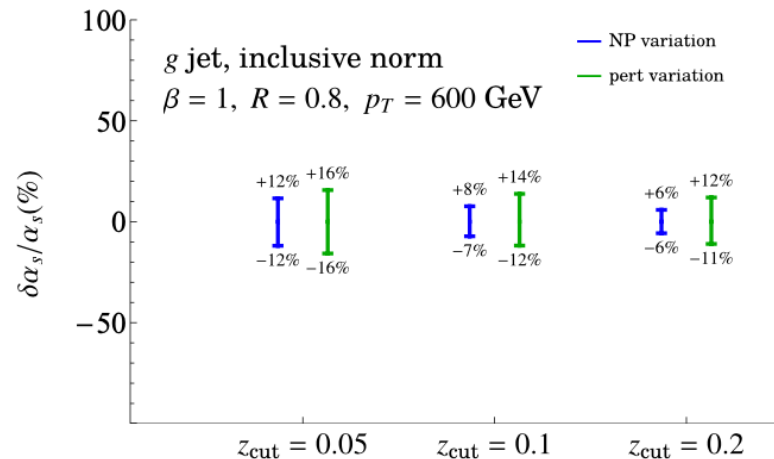
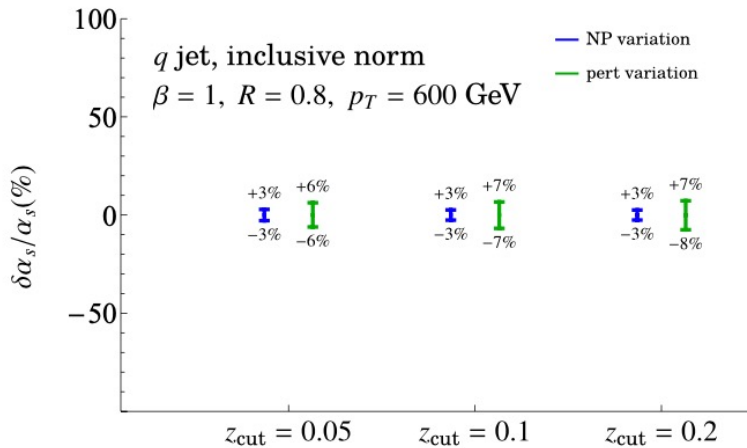


Non-perturbative uncertainty relatively small

- As expected – that’s why we’re using soft-drop

α_s measurement prospects

Stewart, Hannesdottir, Pathak, MDS, Stewart
arXiv:2210.04901



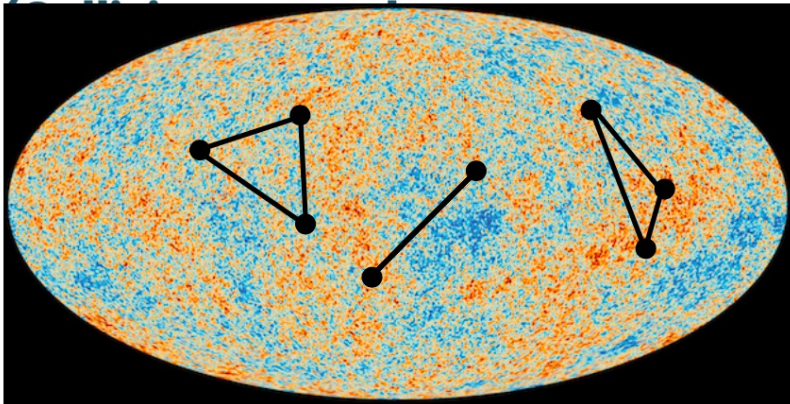
- Should be able to measure α_s at the < 10% level now
- Dominated by perturbative uncertainty
 - Fit range chosen to minimize non-perturbative effects
- Different parameter values and different energies can help reduce overall uncertainty

Possible 5% measurement in the future

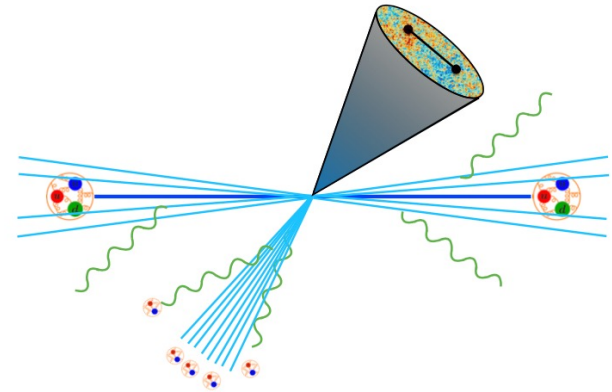
- Difficult to get to < 1% level competitive with world average
- Non-perturbative effects are irreducible below $\sim 3\%$

Energy-energy correlators

An alternative approach to studying jets



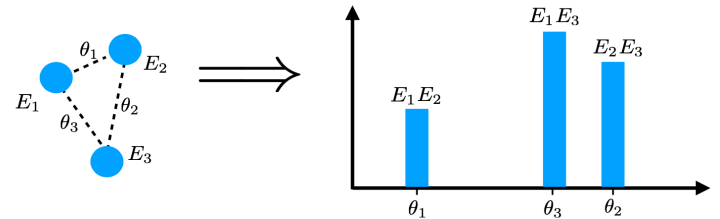
Correlation functions are standard tools in condensed matter and astronomy



Can also measure at colliders

Each event contributes multiple values of the observable

$$\frac{d\sigma}{d\theta} = \sum_{i,j} \int d\sigma \frac{E_i E_j}{Q^2} \delta(\theta - \theta_{ij}) \sim \langle \Psi | \mathcal{E}(\hat{n}_1) \mathcal{E}(\hat{n}_2) | \Psi \rangle$$



Energy-energy correlators

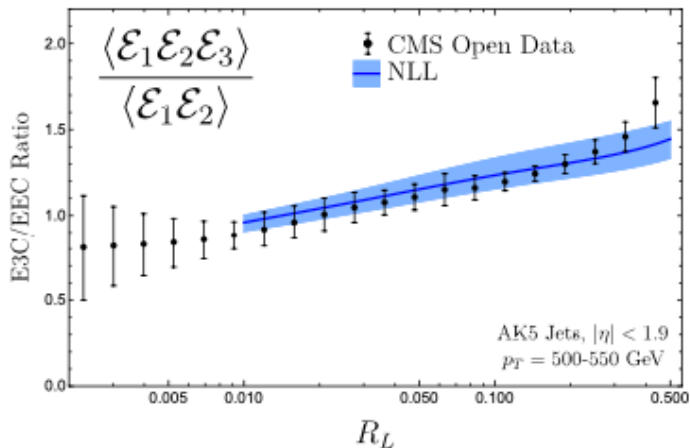
Lee, Mecaj and Moulton arXiv:2205.03414

$$\frac{d\Sigma}{dp_T d\eta d\{\zeta\}} = \sum_i \mathcal{H}_i(p_T/z, \eta, \mu) \quad (5)$$

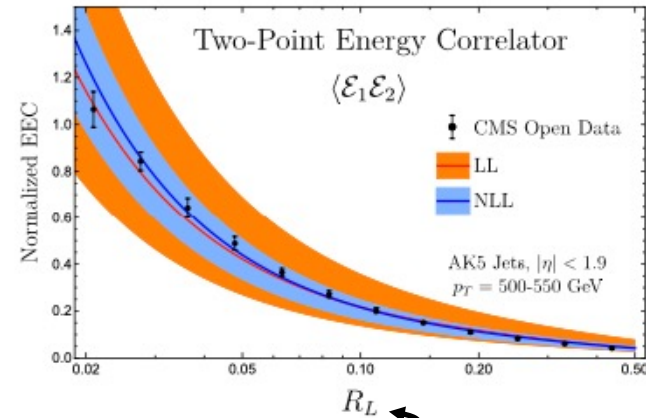
$$\otimes \int_0^1 dx x^N \mathcal{J}_{ij}(z, x, p_T R, \mu) J_j^{[N]}(\{\zeta\}, x, \mu).$$

EECs factorize and can be resummed

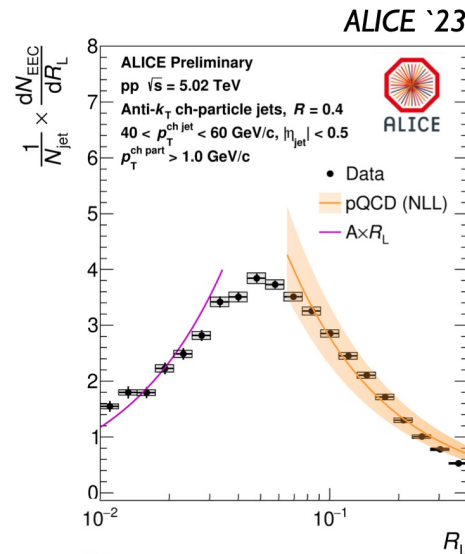
3 point function predicted too



- 2-point function
- Good agreement with theory and CMS open data



R_L ↗ $R_L =$ maximum angle between energy flow operators



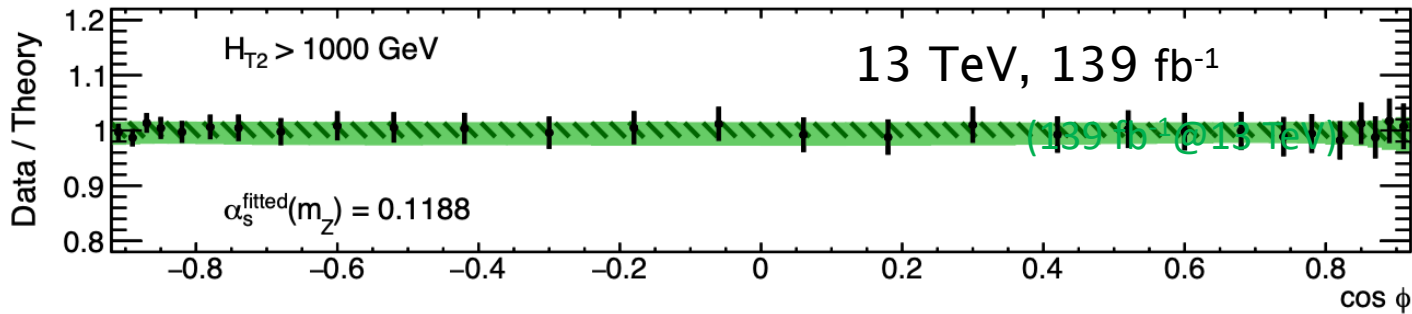
- Measured by ALICE
- good agreement with theory

EECs for α_s

- Transverse energy-energy correlators (EECs)
 - Measures EEC using p_T instead of energy
 - Fit to NNLO theory

$$\frac{1}{N} \sum_{A=1}^N \sum_{ij} \frac{E_{Ti}^A E_{Tj}^A}{\left(\sum_k E_{Tk}^A\right)^2} \delta(\cos \phi - \cos \varphi_{ij}),$$

ATLAS arXiv:2301.09351

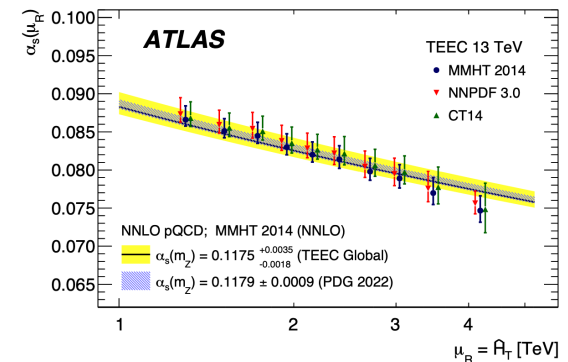


$$\alpha_s(m_Z) = 0.1175 \pm 0.0006 (\text{exp.})_{-0.0017}^{+0.0034} (\text{theo.})$$

3% uncertainty

$$0.1175 \pm 0.0001 (\text{stat.}) \pm 0.0006 (\text{sys.})_{-0.0011}^{+0.0032} (\mu) \pm 0.0011 (\text{PDF}) \pm 0.0002 (\text{NP}) \pm 0.0005 (\text{mod.})$$

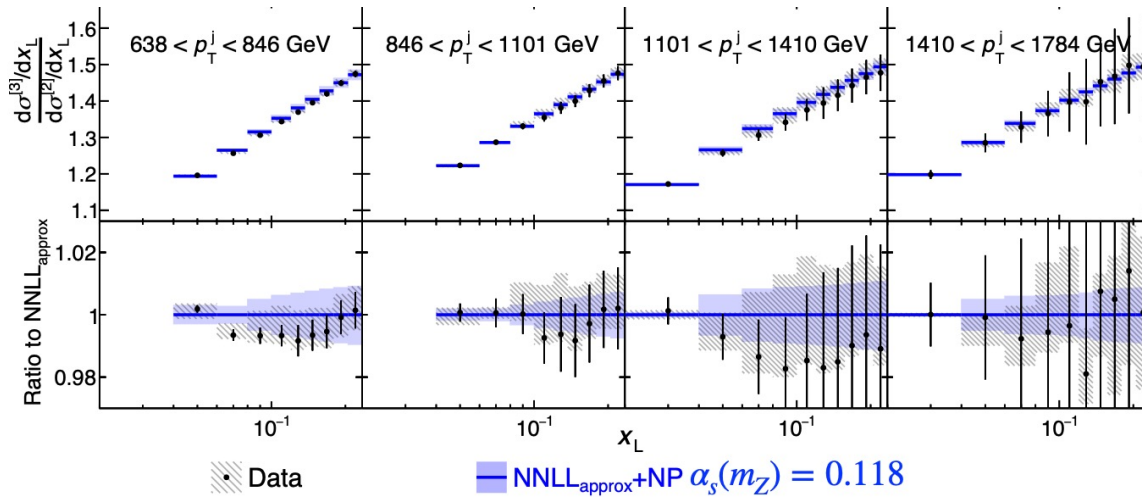
- Does not include any resummation
- Monte carlo used to include non-perturbative effects
- Scale uncertainties dominate



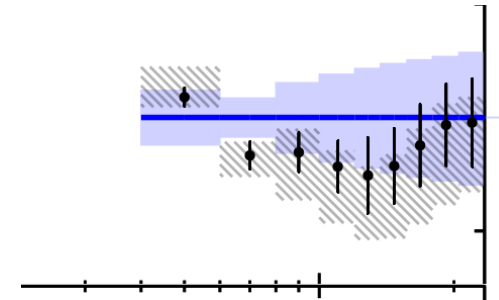
EECs for α_s

Chenfeng Lu CMS, July 31, 2023
(talk at Boost)

- E3C/E2 in high- p_T jets
 - compared to NNLL theory (Chen et al. arXiv:2307.07510)



$468 < p_T^j < 638 \text{ GeV}$



$$\alpha_s(m_Z) = 0.1229^{+0.0014(stat.)+0.0030(theo.)+0.0023(exp.)}_{-0.0012(stat.)-0.0033(theo.)-0.0036(exp.)}$$

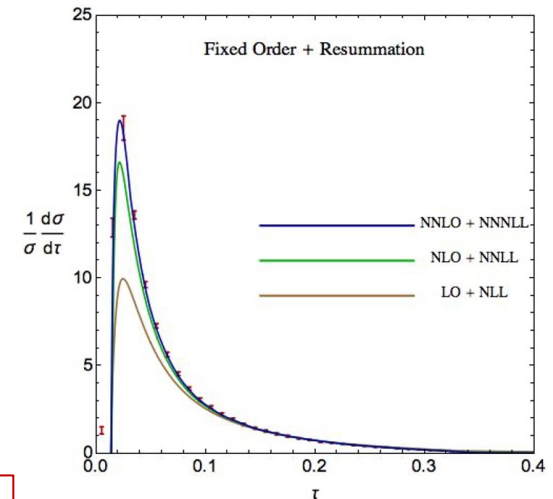
4% uncertainty

- Hadronization taken from average of pythia + herwig (3%)
 - Should use theory model
- Paper not published, hard to assess

e^+e^- event shapes

Thrust, C parameter

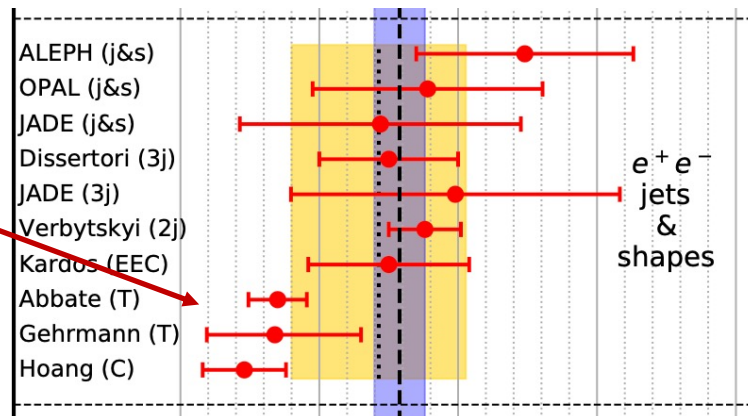
- NNLO fixed order Gerhmann et al, 0711.4711
- NNNLL resummation Becher, MDS 0803.0342.
- Power corrections Abbate et al 1006.0308
- Renormalon subtractions
- heavy quark corrections
- Electroweak corrections



Thrust: $\alpha_s(m_Z) = 0.1135 \pm 0.0011$ 1.0% uncertainty

C parameter: $\alpha_s(M_Z^2) = 0.1123 \pm 0.0015$ 1.3% uncertainty

Values are low compared to world average

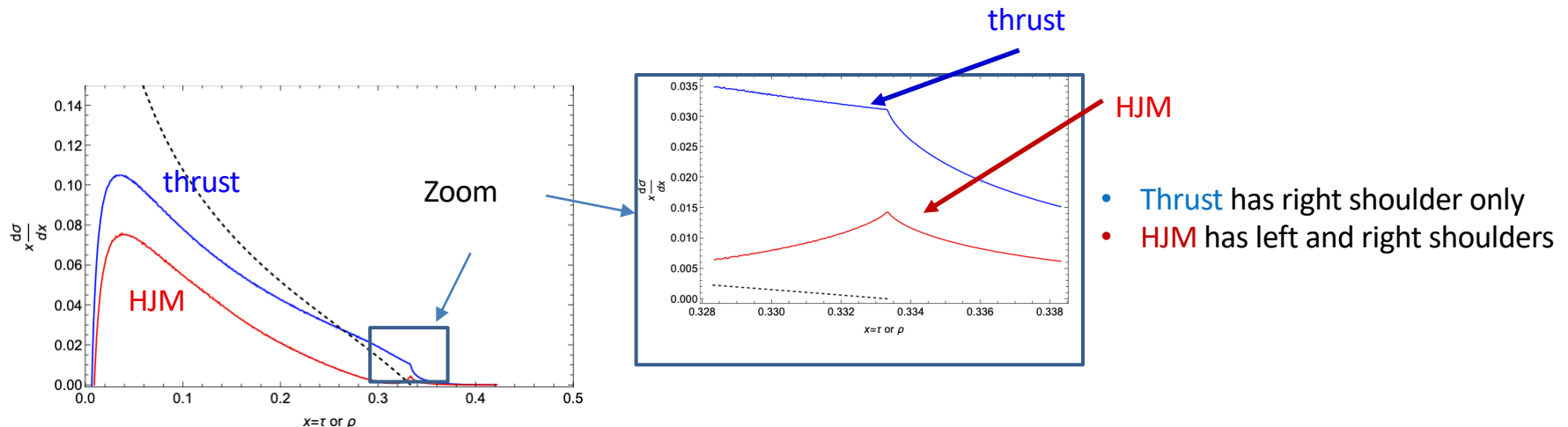


Heavy jet mass

- NNLO fixed order Gerhmann et al, 0711.4711
- NNNLL resummation Chien, MDS 1005.1644
- Salam and Wicke 0102343
 - "Fits for α_s from Heavy Jet Mass come out 10% smaller than for thrust"
- Consistency with other event shapes needed to **validate** methodology

Heavy jet mass is qualitatively different from other event shapes

- Differs from thrust and C parameter in the 3-jet region
- Has a left "**Sudakov Shoulder**" Catani and Webber 9710333



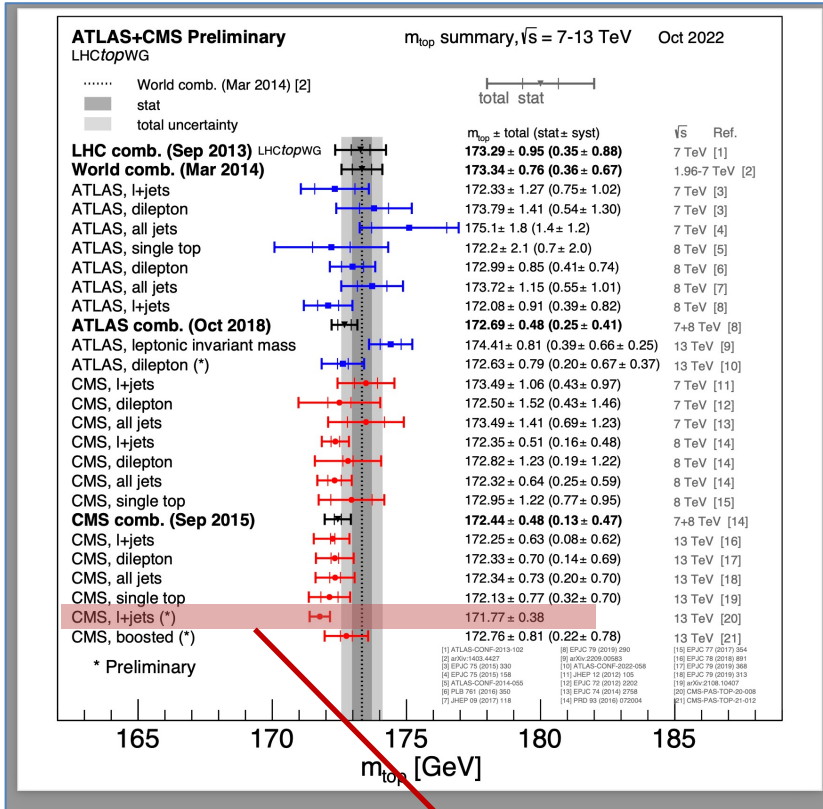
Power corrections in 3-jet region Nason & Zanderighi 2301.03607

- Sudakov Shoulder resummation MDS et al 2205.05702, 2306.08033
- Consistency with thrust and C parameter at 1% level would be convincing
- Stay tuned...

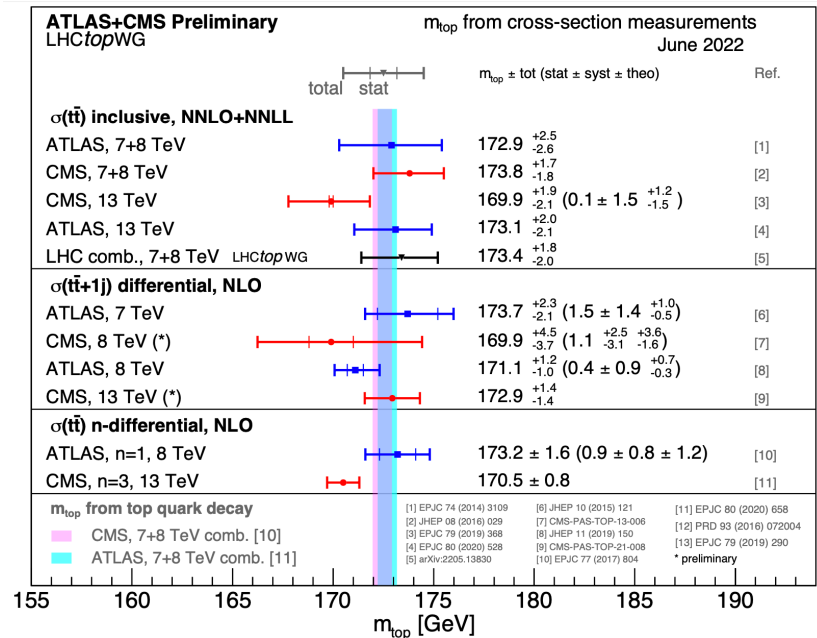
Top quark mass

<https://twiki.cern.ch/twiki/bin/view/LHCPhysics/LHCtopWGSummaryPlots>

Direct measurements



Indirect measurements (cross sections)



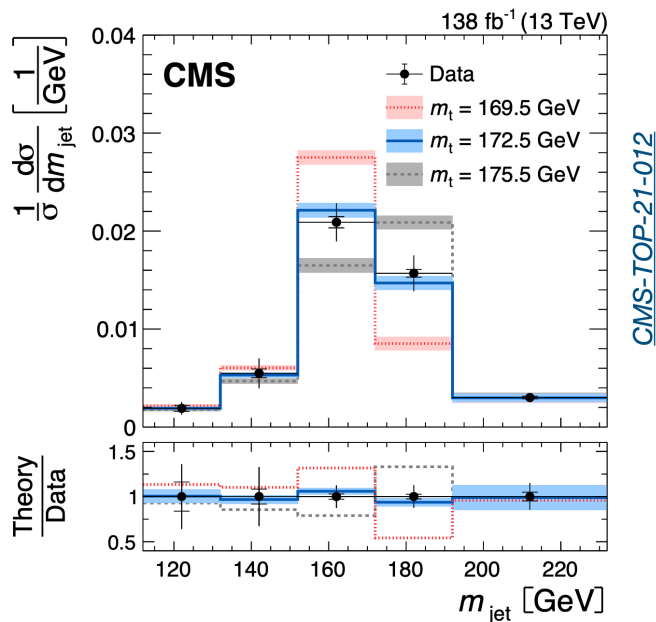
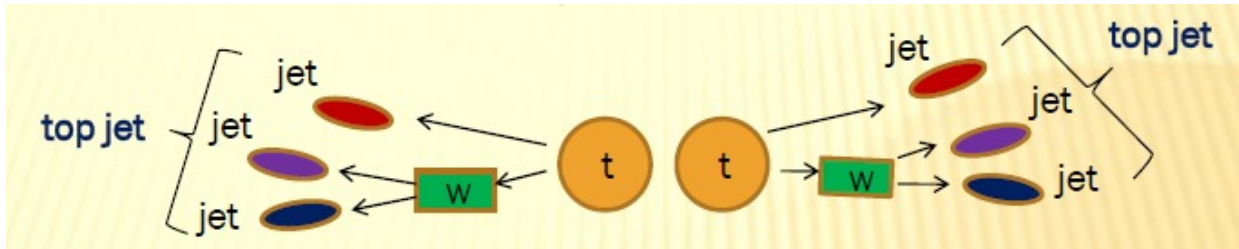
Best measurement in lepton+jets channel

$$m_t = 171.77 \pm 0.38 \text{ GeV}$$

Top quark mass with fat jets

Fully hadronic channel is very challenging

- Huge multijet backgrounds make tt event identification impossible
- In boosted regime, tops become collimated and easier to see



CMS 2023 arXiv:2211.01456

top quark mass measurement in boosted tt events

$$m_t = 173.06 \pm 0.24 \text{ (stat)} \pm 0.61 \text{ (exp)} \pm 0.47 \text{ (model)} \pm 0.23 \text{ (theo)} \text{ GeV}$$

$$= 173.06 \pm 0.84 \text{ GeV.}$$

jet energy scale, etc.

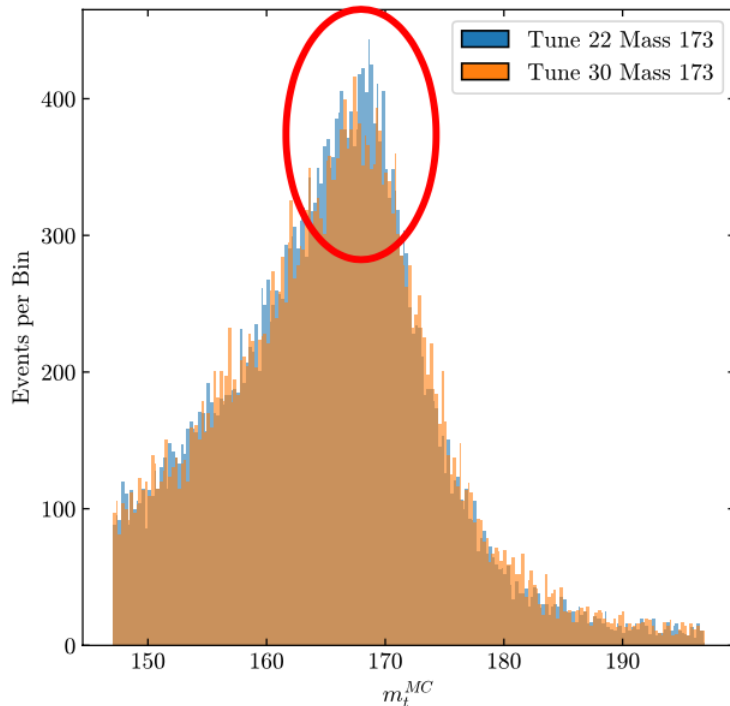
Monte Carlo modeling

perturbative theory

What is the top quark mass?

- Top quark is unstable, has color and charge
 - No well-defined pole-mass in quark propagator
 - **MS-bar top mass is well-defined** but hard to relate to data
 - useful for **indirect measurements** like **cross section**
 - Most experiments measure the **“Monte Carlo” mass**

Different MC tunes with same top mass give different distributions



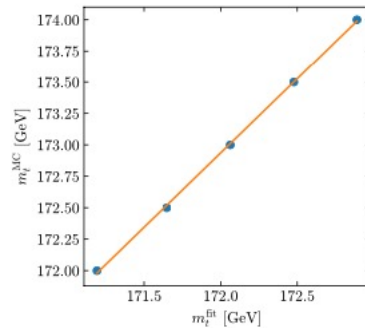
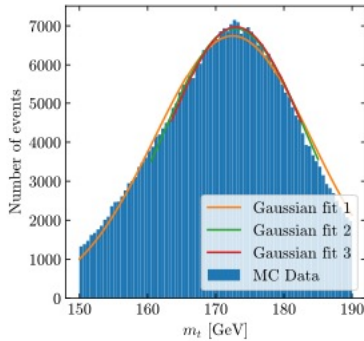
- differences largely soft physics
- tuning uncertainty reduced with jet grooming (trimming, soft-drop)

	without W calibration		with W -calibration	
No grooming	530 MeV		200 MeV	(-62%)
Trimming	530 MeV	(0.0%)	170 MeV	(-68%)
Soft drop	390 MeV	(-26%)	140 MeV	(-74%)
e^+e^-	110 MeV	(-79%)	50 MeV	(-90%)

Andreassen, MDS arXiv:1705.07135

Event ensembles

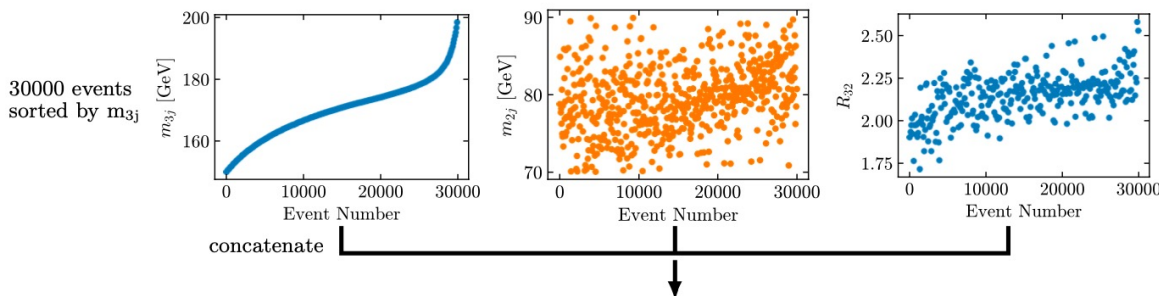
Flescher, Fraser, Hutchison, Osdiek, MDS arXiv:2011.04666



Fitting to peak/histogram shapes is inefficient

- Peak throws out useful information in tails
- Often need awkward parameterization of shape
- Why not just use all the information?

- For each event, measure m_{3j} (top mass), m_{2j} (W mass) and m_{3j}/m_{2j}
- Combine into one large array, sorted by m_{3j}

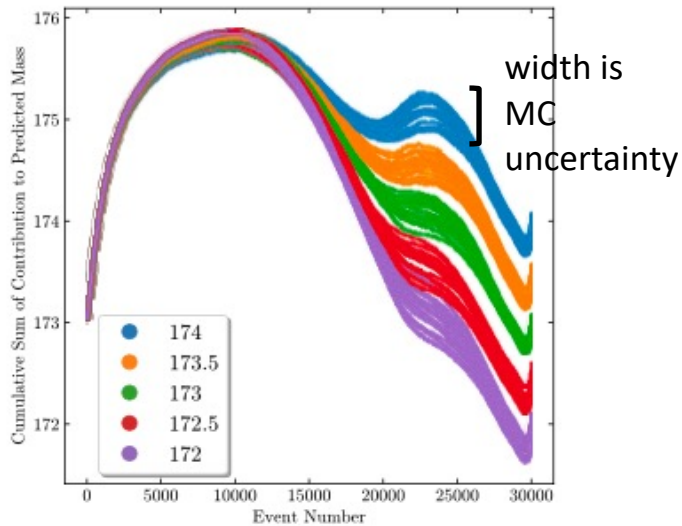


Can use all the information,
not just peak

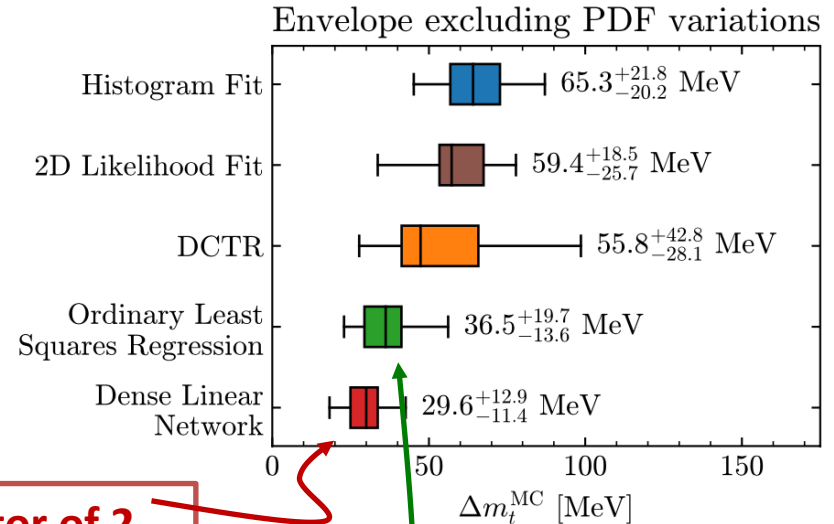
input into regression method

Event ensembles

Predicted mass
multiplied by network weight



- full distribution gives good discrimination

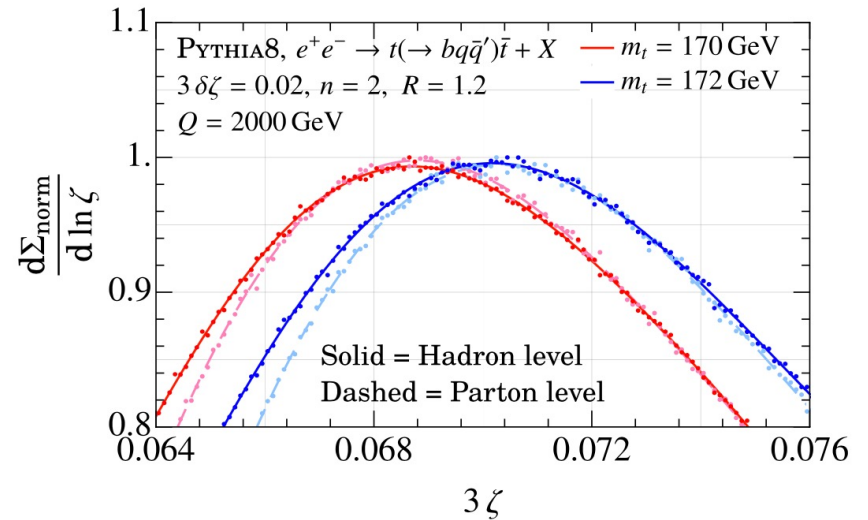
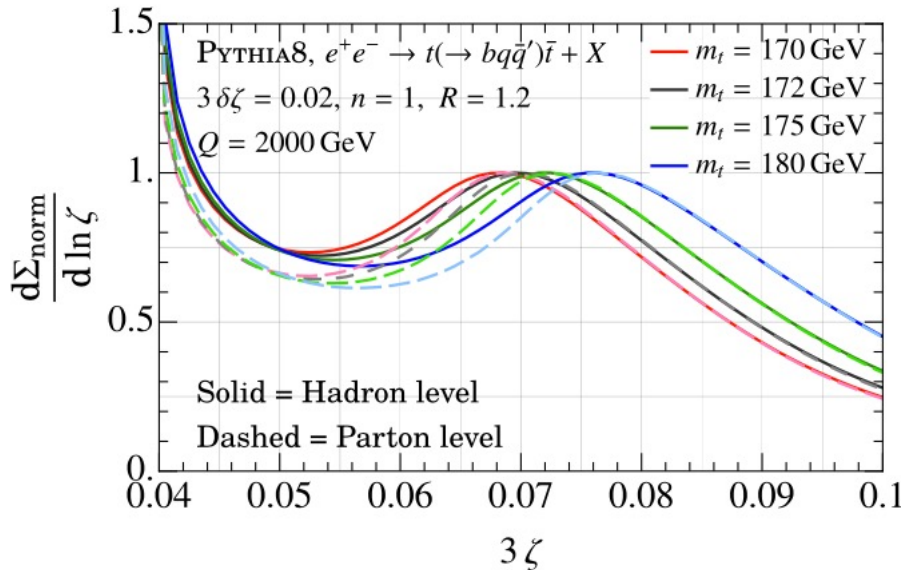


Can reduce Monte-Carlo uncertainty by a factor of 2

- Learn from **ensemble of events**
- Works better than fitting histograms
- Simple and effective way to use all available data
- Can get similar performance with **linear weights = ordinary least-squares regression** (not machine learning)
- super fast, no training

EECs for top mass

Holguin et al. arXiv:2201.08393



Measure the 3-point function in boosted top events

Insensitive to hadronization

$$\widehat{\mathcal{M}}^{(n)}(\zeta_{12}, \zeta_{23}, \zeta_{31}) = \sum_{i,j,k} \frac{E_i^n E_j^n E_k^n}{Q^{3n}} \delta(\zeta_{12} - \hat{\zeta}_{ij}) \delta(\zeta_{23} - \hat{\zeta}_{ik}) \delta(\zeta_{31} - \hat{\zeta}_{jk})$$

Factors of energy in definition suppress soft radiation

Looks promising for m_t measurement

- In principle, direct theory-experiment comparison with short-distance top definition
- Early days, but worth watching

Conclusions

A lot of exciting progress in jet physics

- Machine learning
- Precision measurements
- Top mass determination
- Energy-energy correlators
- Heavy ion physics
- Lund plane kinematics
- Improvements in unfolding
- Jet energy calibration
- Antenna showers
- Hadronization models
- Fixed order matching
- Anomaly detection
- Heavy flavor tagging (e.g. $h \rightarrow cc$)
- ...