



# Jets for LHC Run 3

CERN QCD seminar Friday 11<sup>th</sup> November 2022

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#### Introduction

- LHC Run3 does not give us any substantial increase in energy: but we will have more and more data;
- the theoretical focus is on making our tools better and better
  - NNLO calculations as the new standard (loop-results for at least 2→3 topologies; more flexible subtraction schemes; interface with parton showers and/or resummation);
  - parton-shower simulations upgraded to the precision club (log accuracy, colour, higher-order splitting functions);

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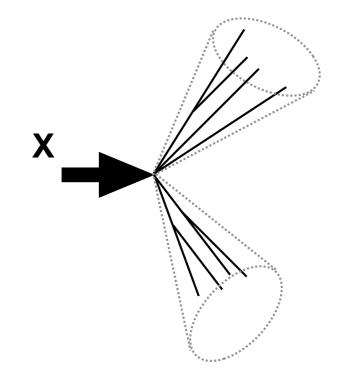
- state-of-the art predictions for standard candles (e.g. N<sup>3</sup>LO, N<sup>3</sup>LL resummation, effects on parton densities);
- one of the challenges ahead is to find new and more efficient ways to interrogate the data and jet physics offers us exciting opportunities

#### **CROSS-POLLINATION**

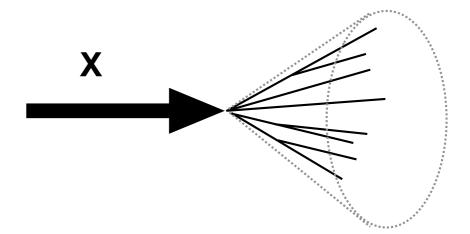
UNDERSTANDING NEW TOOLS

#### An introduction to jet substructure

starting mostly as a tool for new physics searches, jet substructure techniques are now mainstream



X is at rest and its decay products are reconstructed in two jets

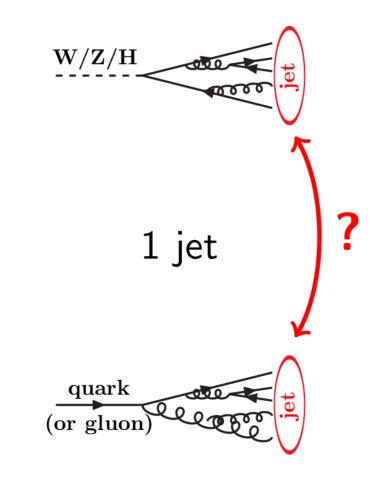


X is boosted and its decay products are reconstructed in one jet

## Looking inside jets

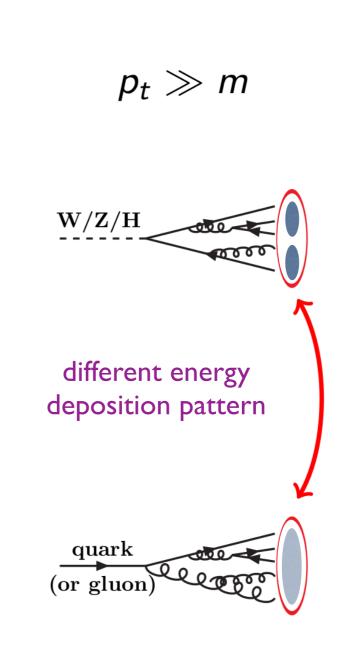
- the two major goals of the LHC
- search for new particles
- characterise the particles we know
- jets can be formed by QCD particles but also by the decay of massive particles (if they are sufficiently boosted)
- how can we distinguish signal jets from background ones?





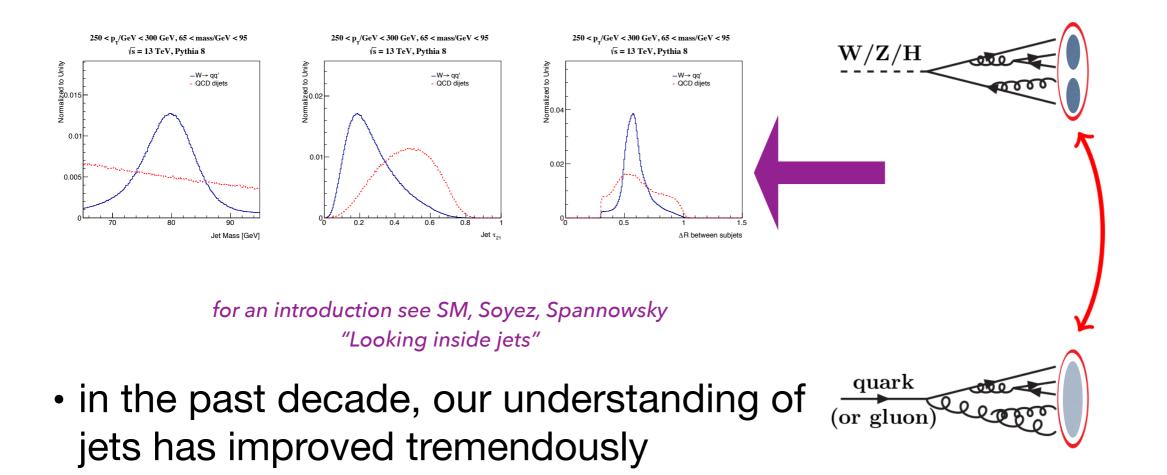
#### Jet substructure in a nutshell

- the final energy deposition pattern is influenced by the originating splitting
- hard vs soft translates into 2-prong vs 1-prong structure
- picture is mudded by many effects (hadronisation, underlying event, pileup)
- two-step procedure:
  - grooming: clean the jets up by removing soft radiation
  - tagging: identify the features of hard decays and cut on them



#### A theorist's job

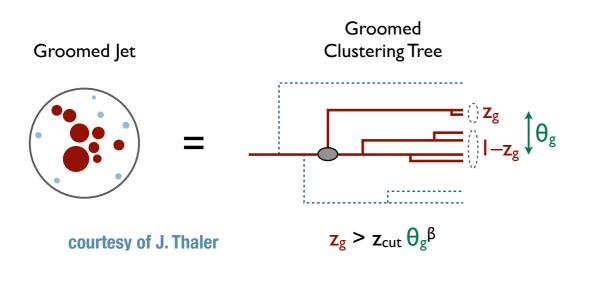
• devise clever ways to project the multi-dimensional parameter space of final-state momenta into suitable lower dimensional (typically 1-D) distributions  $p_t \gg m$ 

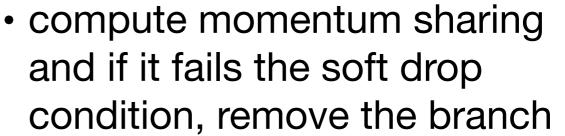


 so it is time to ask ourselves if jet substructure can play a role in the wider LHC physics program

#### An example: Soft Drop

- efficient and robust grooming and tagging a have been developed and exploited at the LHC
- Soft Drop aims to clean up a jet by removing soft radiation



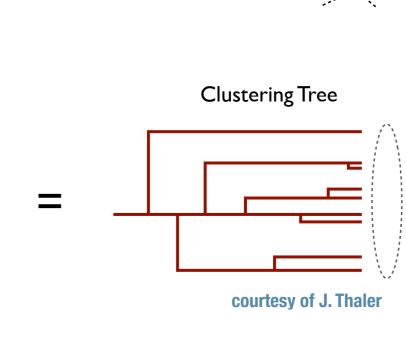


$$z_g = \frac{\min(p_{T1}, p_{T2})}{p_{T1} + p_{T2}} \qquad z_g$$

Original Jet

$$z_g < z_{\rm cut} \theta_g^\beta$$

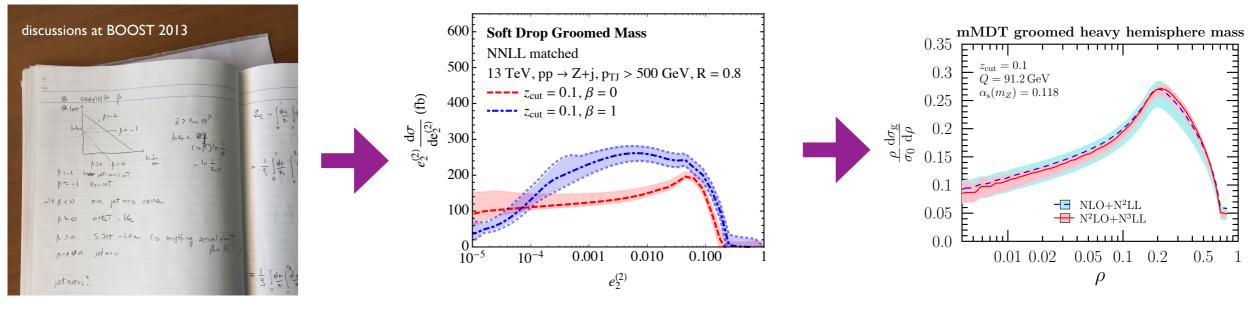
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,....**.** 



#### From ideas to precision



Frye, Larkoski, Schwartz, Yan (2016) Dasgupta, El-Menoufi, Helliwell (2022)

Kardos, Larkoski, Trocsanyi (2020)

- good perturbative properties (convergence, absence of intricate soft effects such as non-global logs)
- what about sensitivity to non-perturbative corrections?

#### Hadronisation corrections

- we have to take into account (at least) two effects:
  - a non-perturbative mass shift

$$\delta m^2 = C_F \Lambda_{\rm NP} p_t R_{\rm eff}$$

• a non-perturbative pt shift (loss)

 $\delta p_t = -C_A \Lambda_{\rm NP} / R_{eff}$ 

 $R_{\text{eff}} = \frac{m}{p_t \sqrt{z(1-z)}}$  • both contribute with the same leading-power correction to the jet mass distribution (assuming

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\frac{d\sigma}{dm}\Big|_{\rm NP} = \frac{d\sigma}{dm}\Big|_{\rm pert} \left(1 + f(z_{\rm cut},\beta) \frac{p_t \Lambda_{\rm NP}}{m^2} \left(\frac{m}{p_t}\right)^{\frac{2}{2+\beta}}\right) \qquad \text{ soft-drop reduction}
Dasgupta, Fregoso, SM, Salam (2013);
SM, Schunk, Soyez (2018)
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detailed field-theory analysis confirms and expands on this picture

 $\mathbf{z}$ 

1 - z

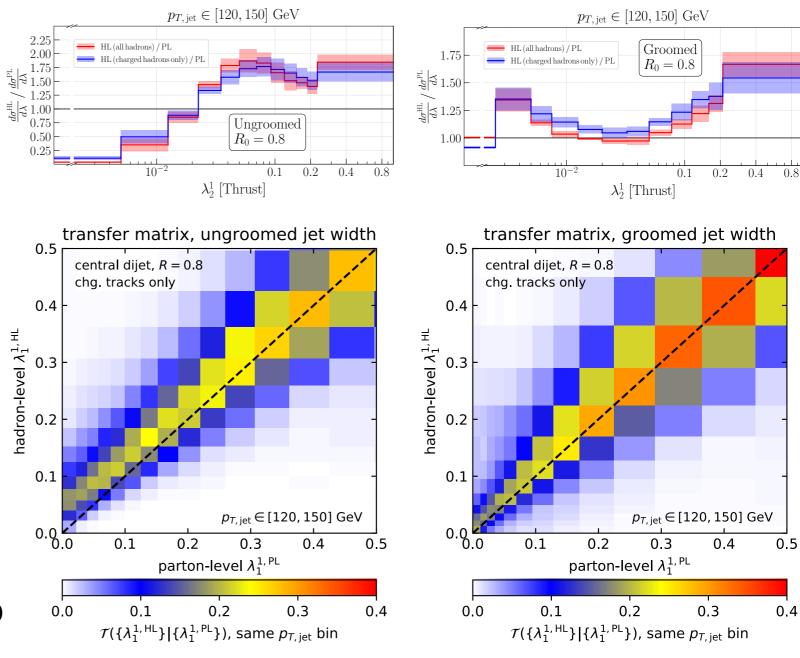
#### Size of the corrections

 simple analytic arguments confirmed by detailed MC simulations (with both hadronisation and underlying event)

$$\lambda_{\alpha}^{\kappa} = \sum_{i \in jet} \left(\frac{p_{ti}}{p_t}\right)^{\kappa} \left(\frac{\Delta_i}{R}\right)^{\alpha}$$

grooming reduces

 onset, size and
 migration effects due to
 non-perturbative
 corrections!

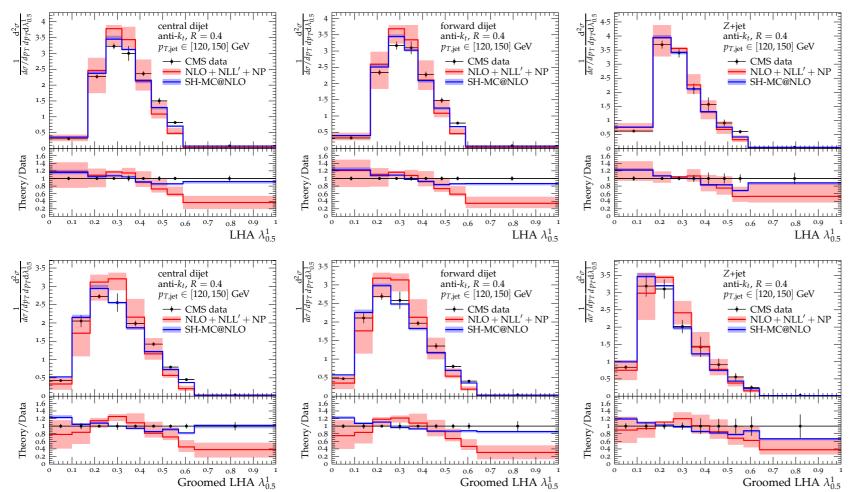


Caletti, Fedkevych, SM, Reichelt, Schumann, Soyez, Theeuwes (2021) Reichelt, Caletti, Fedkevych, SM, Schumann, Soyez (2021)

#### **Comparison to data**

- jet angularities allow us to probe the internal QCD dynamics of jets (set  $\kappa = 1$  for IRC safety)
- multi-scale problem that requires resummation
- NLO+NLL' implemented in the Sherpa-resummation plugin (efficient way to handle kinematics and colour)
- CMS performed detailed studies of jet angularities in Z+jet and dijet events

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CMS collaboration (2021)

Reichelt, Caletti, Fedkevych, SM, Schumann, Soyez (2021)

- a wealth of high-quality data combined with solid theory description
- a lot of phenomenology that can be done!

#### **Quark/gluon enriched samples**

• angularities in Z+jet and dijets: different transverse momentum and rapidity bins allows us to probe samples with rather different quark/gluon components

120, 150] GeV, AK8

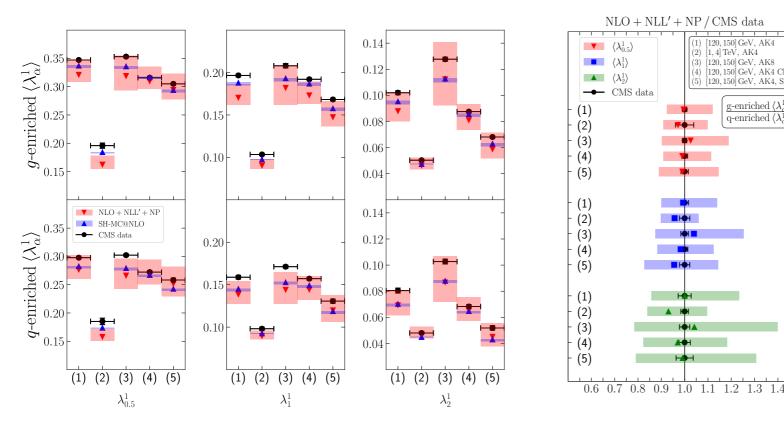
120, 150 GeV, AK4 Ch.

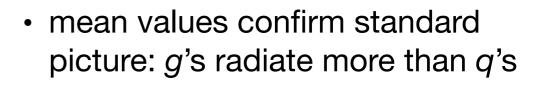
20, 150 GeV, AK4, SD

g-enriched  $\langle \lambda_{\alpha}^{1} \rangle$ 

 $\gamma$ -enriched  $\langle \lambda_{\alpha}^{1} \rangle$ 

configuration	type of jet	$p_{T,\text{jet}}$ [GeV]	g-enriched	q-enriched
(1)	ungroomed $R = 0.4$	[120, 150]	dijet central	Z+jet
(2)	ungroomed $R = 0.4$	[1000, 4000]	dijet central	dijet forward
(3)	ungroomed $R = 0.8$	[120, 150]	dijet central	Z+jet
(4)	ungroomed $R = 0.4$ (tracks only)	[120, 150]	dijet central	Z+jet
(5)	SoftDrop ( $\beta = 0, z_{\rm cut} = 0.1$ ) $R = 0.4$	[120, 150]	dijet central	Z+jet

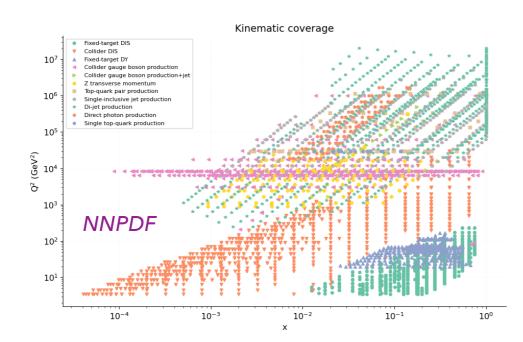




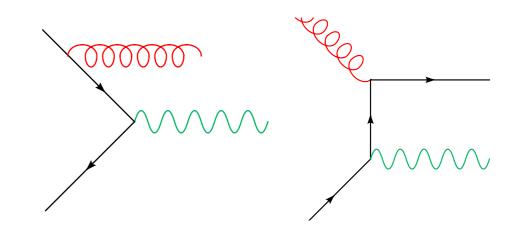
- our calculation tends to underestimate the mean values
- however, it does so democratically for *q*'s and *g*'s: no appreciable bias
- beware! NLO corrections can • significantly alter q/q fractions
- room for improvement: thanks to recent development NNLO+NNLL is not far away

#### Quark/gluon tagging for SM pheno

- q/g tagging traditional important for searches
- can it be useful in Standard Model studies?
- yes, but we need theoretically well-defined (IRC safe) taggers
- Z p<sub>T</sub> studies play a central role in SM phenomenology: probe of initial-state QCD dynamics
- Z p<sub>T</sub> recently entered global PDF fits: they provide info on the gluon PDF

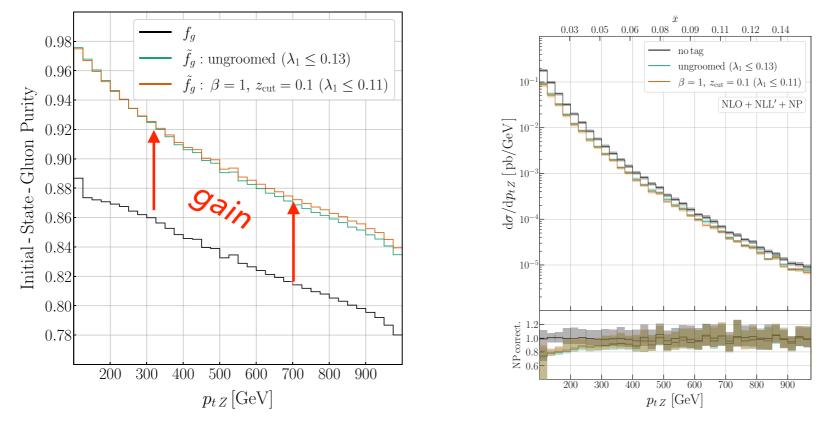


 the Z + jet process features a strong correlation between the flavours of initial- and final-state partons (100% at LO)



## Angularities as q/g taggers

- use jet angularities to select Z + quark-jet events
- this selection ensures a higher initial-state-gluon purity
- measure Z  $p_T$  on those events

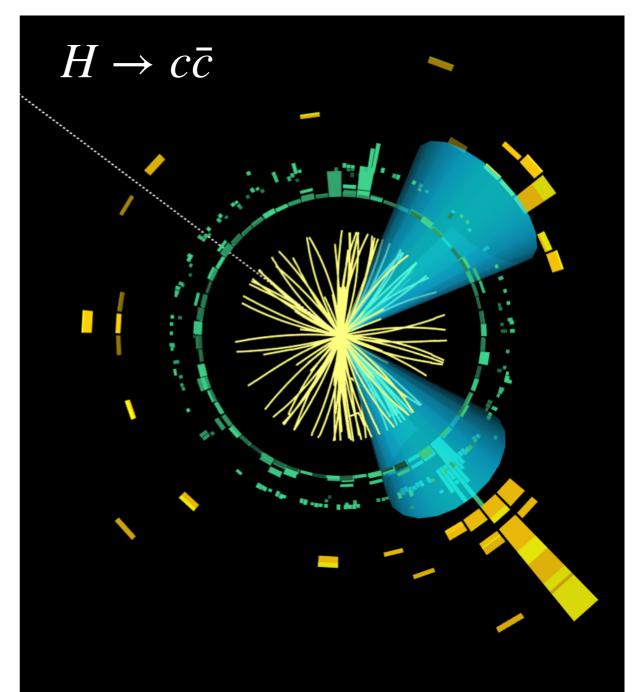


- need to assess the actual impact on the gluon PDF
- plenty of data!

- pro's: good gain, everything is IRC and calculable;
- con's: need to reconstruct the leading jet, acquired sensitivity to non-pert. corrections

#### **Cross**pollination

jet substructure inspires new ideas to study heavy flavours



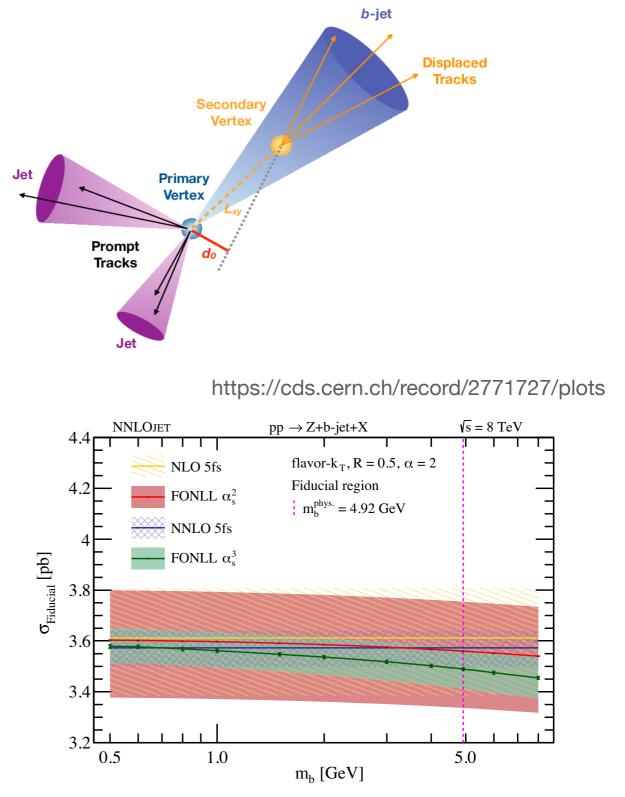


Run: 350440 Event: 1105654304 2018-05-16 23:55:11 CEST

#### Heavy Flavour Jets

- jets containing heavy flavours (charm and beauty) are central to the LHC Higgs program
- important for QCD studies too: PDFs, fragmentation etc.
- they are identified exploiting B hadron lifetime: displaced vertices
- from theory viewpoint, m<sub>b</sub> & m<sub>c</sub> set perturbative scales: high accuracy (NNLO) QCD calculations Z+b/c jet now exist

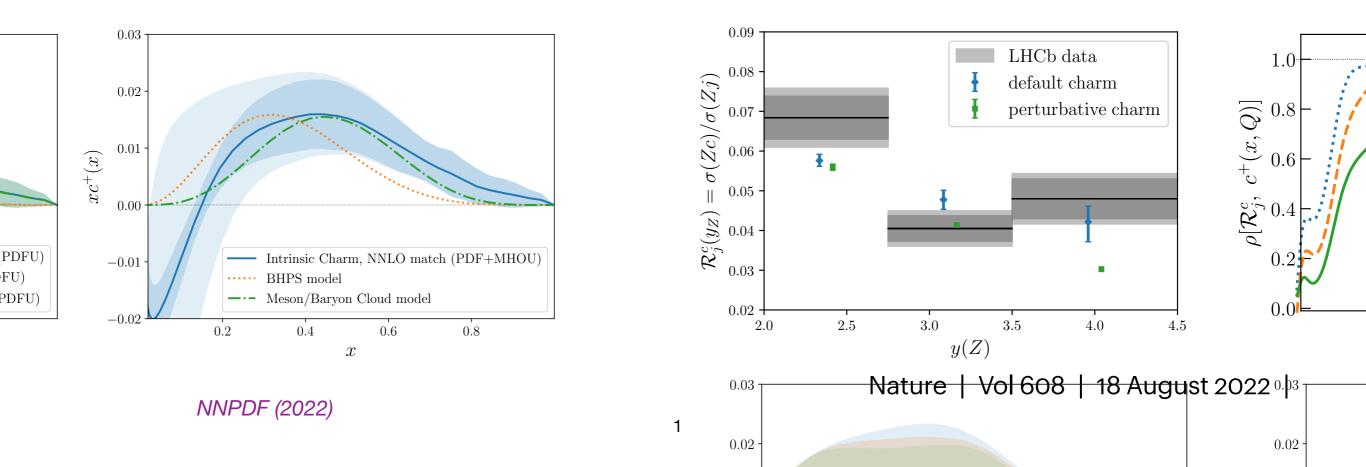
#### two recent highlights...



Gauld et a. (2020)

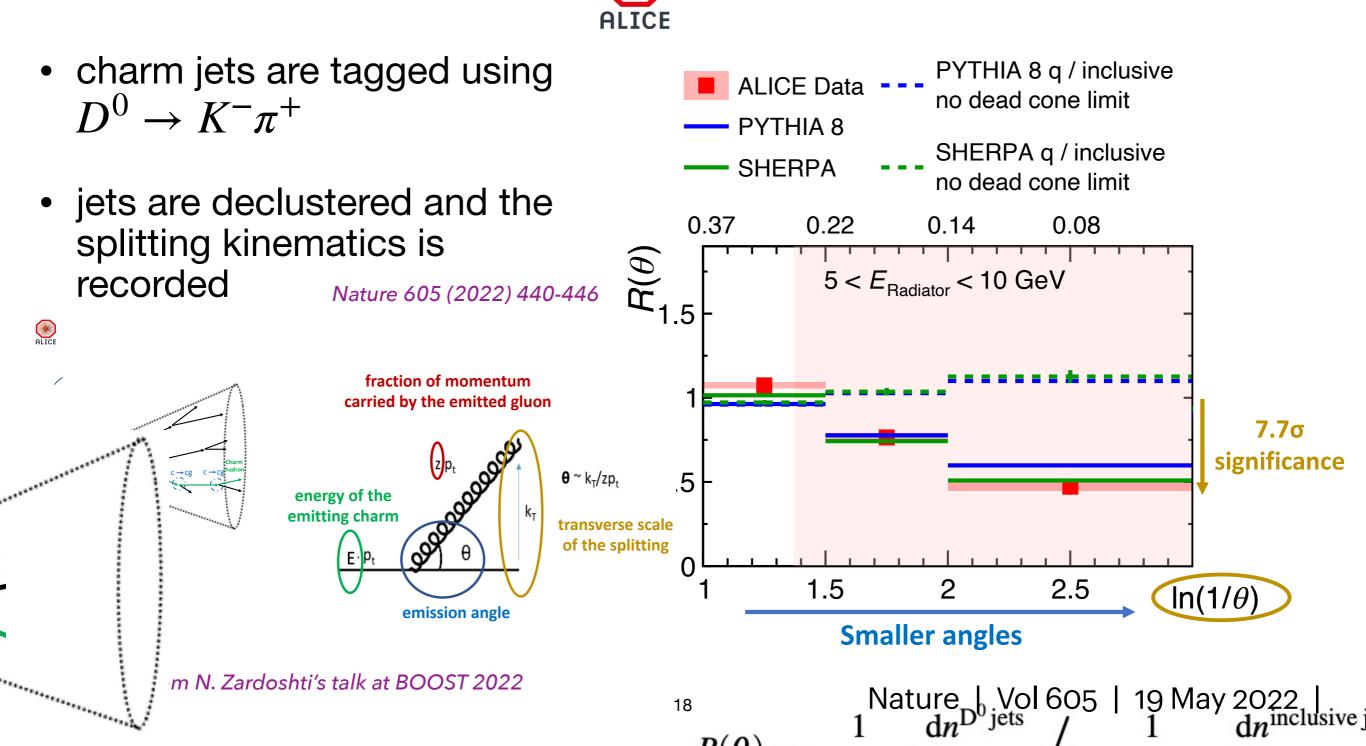
#### There's charm in the proton!

- NNPDF collaboration has recently shown a  $3\sigma$  evidence of intrinsic charm in the proton
  - they fit the charm PDF in the 4-flavour scheme: charm is both radiative and intrinsic
  - they match to the 3-flavour scheme to extract the (only) intrinsic
- good agreement with theory models and and visible in Z+c data!



#### ALICE and the dead cone

 ALICE recently exploited ideas from modern jet physics (e.g. reclustering) to perform the first prect measurement of the dead cone



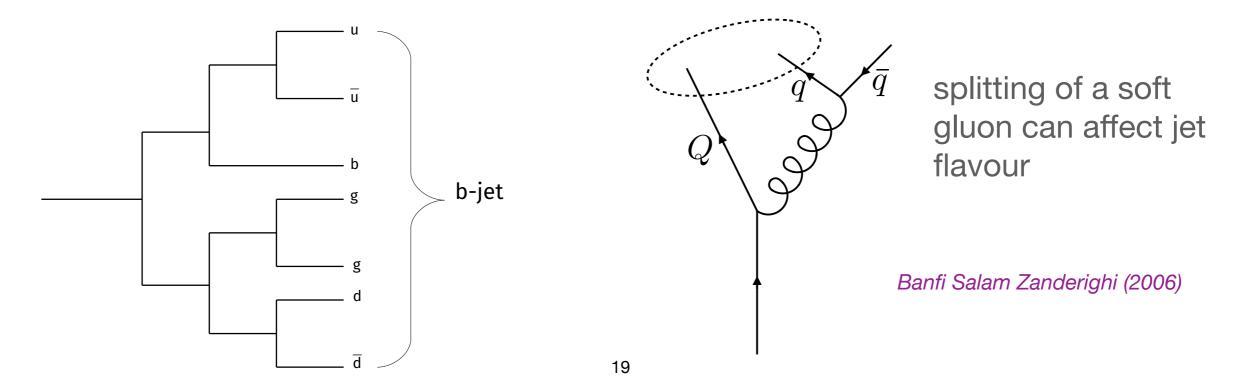
### **Experiment vs Theory (I)**

- Experimental procedure:
  - cluster jets using the anti-kt algorithm
  - run b (c)-tagging

- Theory calculation
  - compute real and virtual
  - cluster jets using an IRC safe (flavour) algorithm

 $\theta$ 

BUT counting the flavour of an anti-k<sub>t</sub> jet is NOT IRC Safe beyond NLO!



#### **BSZ flavour algorithm**

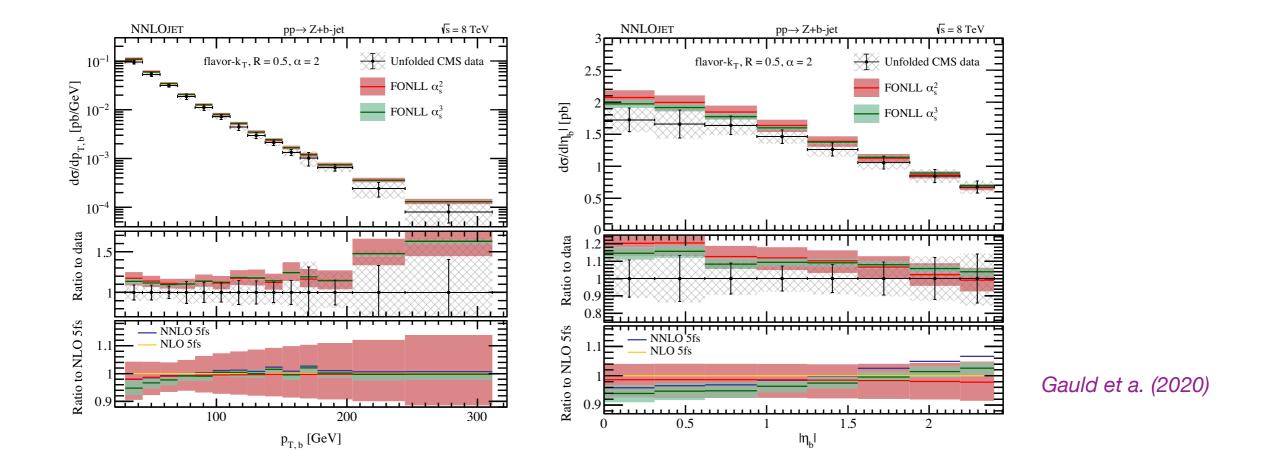
 the flavour-sensitive metric reflects the absence of soft quark singularities:

 $d_{ij}^{(F)} = (\Delta \eta_{ij}^2 + \Delta \phi_{ij}^2) \times \begin{cases} \max(k_{ti}^2, k_{tj}^2), & \text{softer of } i, j \text{ is flavoured}, \\ \min(k_{ti}^2, k_{tj}^2), & \text{softer of } i, j \text{ is flavourless}, \end{cases}$ 

- it is IRC safe because it tends to recombine together the problematic soft  $q\bar{q}$  pair
- however the use of BSZ in experimental analysis is far from straightforward:
  - obviously, it's not anti-kt
  - it requires knowledge of the flavour at each step of the clustering

### **Experiment vs Theory (II)**

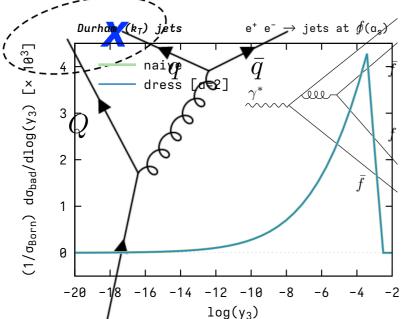
• Comparison between theory and experiments requires to unfold the experimental data to the theory calculation performed with BSZ



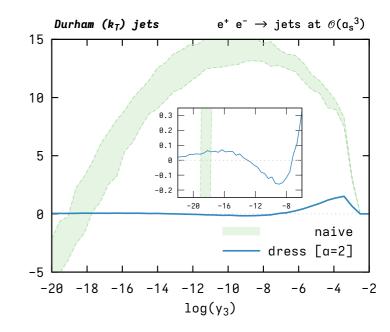
 it would be better to identify a common procedure in order to avoid this unfolding step

#### 3\* new ideas in the past 2 months!

- use Soft Drop to remove soft quarks
- define a flavour algorithm  $\bullet$  **jets** that resembles anti-k<sub>t</sub> dressing
  - → JETS
     construct a flavour dressing for a given jet



 $\begin{bmatrix} f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{O}(a_s^2) \\ f_{ij}^{(F)} = d_{ij} \times P_{urham} (k_T) jets & e^+ e^- \rightarrow jets at \mathcal{$ 



- needs JADE as reclusters, Any gen-k, algo is safe! know to fail at three loops
- flavour-dependent metric, still needs some (smail) emfolding
- n**beidsoffæriog**upnsafer, iffrormationstillsafely (all?) particles in an event

Gauld, Huss, Stagnitto (2022)

Caletti, Larkoski, SM, Reichelt (2022)

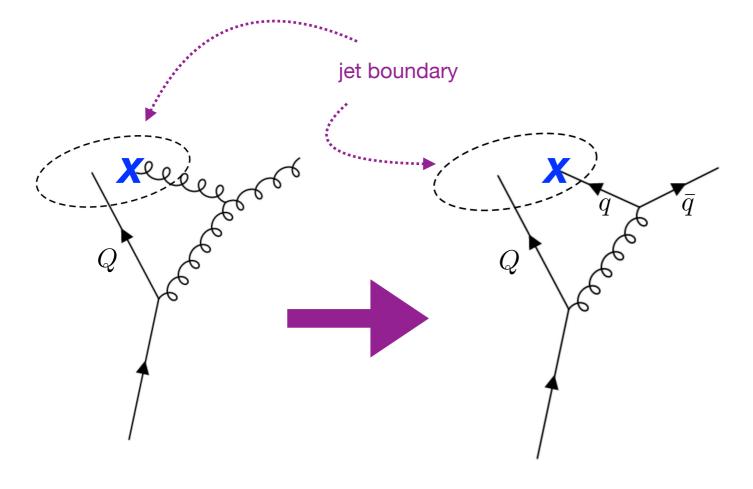
Czakon, Mitov, Poncelet (2022)

it would be interesting to do a dedicated comparison!

#### \*4<sup>th</sup> proposal soon to appear

#### what about Soft Drop jets?

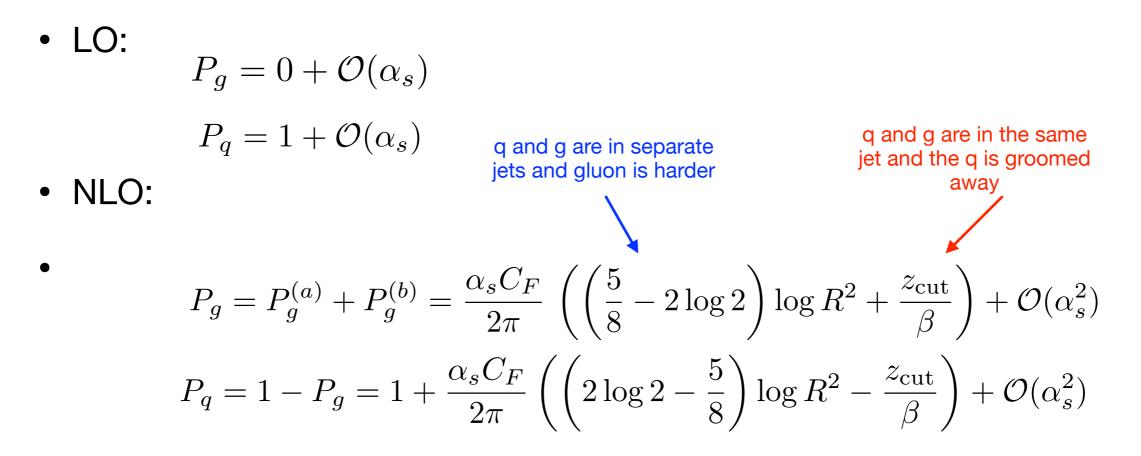
- grooming algorithms remove soft radiation from jets
- in particular is Soft Drop is beneficial in the context of non-global logarithms
- the problematic configurations are similar



- so the idea is:
  - cluster jets with any algorithm you wish
  - apply Soft Drop and measure the flavour
- this is experimentally viable, it is IRC safe?

#### Soft Drop flavour at NLO

- Through NLO even anti-kt flavour is IRC safe. Is there any subtlety with Soft Drop flavour?
- We concentrate on quark vs gluon, in e<sup>+</sup>e<sup>-</sup>



• so even at NLO, we must use Soft Drop with  $\beta > 0$  (perhaps not a surprise)

#### Soft Drop flavour at NNLO

• if the dashed oval represents the if the dashed oval ulleteffective grooming boundary represents the jet boundary (clustering log configuration), Soft (NGLs configuration), Soft Drop fails to screen the singularity Drop screens the singularity Why is that?  $\Theta_{\rm SD}^{\rm C/A} = \Theta(\theta_{Q\bar{q}}^2 - \theta_{Qq}^2)\Theta(\theta_{q\bar{q}}^2 - \theta_{Qq}^2)\Theta\left(z_q - z_{\rm cut}\left(\frac{\theta_{Qq}^2}{R^2}\right)^\beta\right)\Theta\left(z_{\rm cut}\left(\frac{\theta_{Q\bar{q}}^2}{R^2}\right)^\beta - z_{\bar{q}}\right)$ C/A clustering quark passes Soft Drop antiquark fails Soft Drop  $d\Pi_3 |\mathcal{M}(z_q, z_{\bar{q}})|^2 \Theta_{\mathrm{SD}}$  $\simeq d\Pi_3 |\mathcal{M}(x_q, x_{\bar{q}})|^2 \Theta(\theta_{Q\bar{q}}^2 - \theta_{Qq}^2) \Theta(\theta_{q\bar{q}}^2 - \theta_{Qq}^2) \Theta(x_q - 1) \Theta(1 - x_{\bar{q}})$ 

rescaling:  $z_q = x_q z_{\text{cut}} \left( \frac{\theta_{Qq}^2}{R^2} \right)$ 

#### which is singular in the collinear limit

#### **Jade Soft Drop**

- Can we modify Soft Drop to save the day?
- we can change the algorithm used for reclustering
- gen-kt algorithms do not cluster two soft particles together, if there is a hard particle around at smaller angle, but Jade does
- let's look at the problematic configuration with Jade reclustering

$$\Theta_{\mathrm{SD}}^{\mathrm{JADE}} = \Theta(m_{Q\bar{q}}^2 - m_{Qq}^2)\Theta(m_{q\bar{q}}^2 - m_{Qq}^2) \ \Theta\left(z_q - z_{\mathrm{cut}}\left(\frac{\theta_{Qq}^2}{R^2}\right)^{\beta}\right) \Theta\left(z_{\mathrm{cut}}\left(\frac{\theta_{Q\bar{q}}^2}{R^2}\right)^{\beta} - z_{\bar{q}}\right)$$
$$= \Theta\left(x_{\bar{q}}\theta_{Q\bar{q}}^{2(\beta+1)} - x_q\theta_{Qq}^{2(\beta+1)}\right) \Theta\left(x_{\bar{q}}z_{\mathrm{cut}}\left(\frac{\theta_{Q\bar{q}}^2}{R^2}\right)^{\beta}\theta_{q\bar{q}}^2 - \theta_{Qq}^2\right) \Theta\left(x_q - 1\right) \Theta\left(1 - x_{\bar{q}}\right)$$

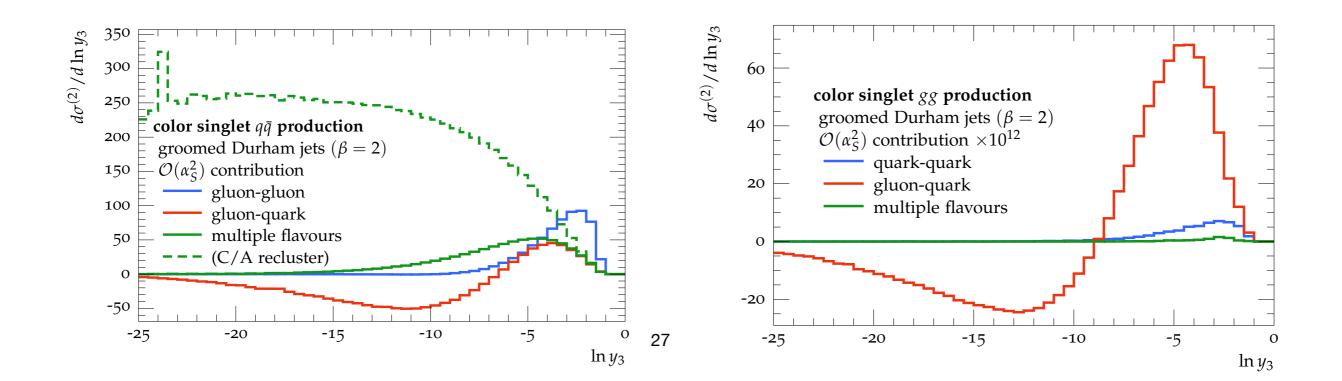
 with Jade reclustering energies and angles are coupled even after rescaling: the singularity is successfully screened

#### **Numerical checks**

• introduce a resolution parameters to separate

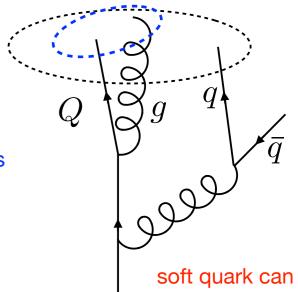
$$\sigma^{\text{NNLO}} = \int_0^{y_3} dy'_3 \, \frac{d\sigma}{dy'_3} + \int_{y_3}^{y_{\text{max}}} dy'_3 \, \frac{d\sigma}{dy'_3}$$

- following BSZ, we use the 3-jet resolution parameter
- distributions of non-Born configurations should be integrable as  $y_3 \rightarrow 0$



#### Problems at N<sup>3</sup>LO and beyond

- Jade Soft Drop allows us to formulate a definition of flavour which is
  - viable from an experimental view point (original jets can be anti-kt and the flavour algorithm is applied after jet clustering)
  - IRC safe through NNLO so that it can be used with stateof-the-art calculation
- however, the algorithm is unsafe at N<sup>3</sup>LO: maybe one can think of applying recursive/iterative Soft Drop?

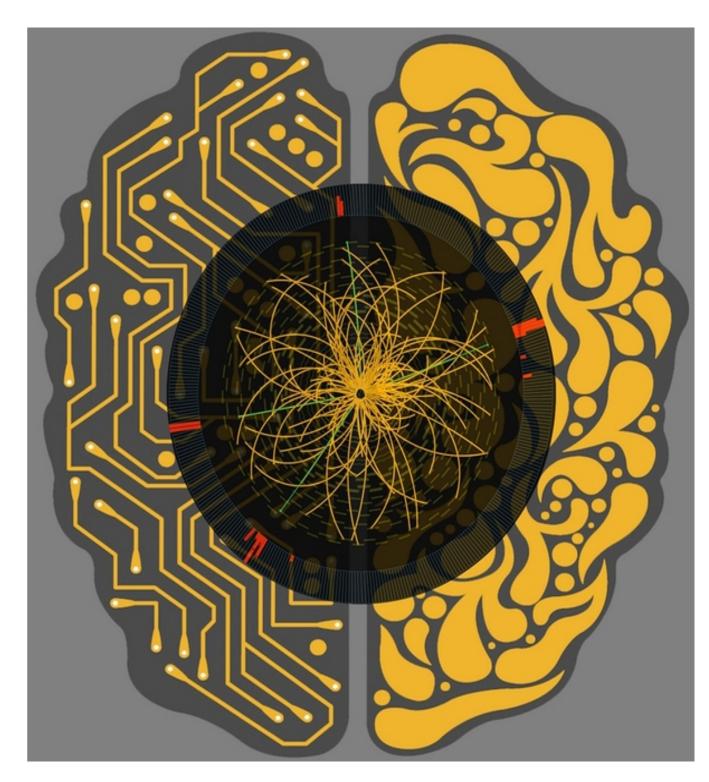


and passes SD

soft quark can alter the flavour

# Understanding new tools

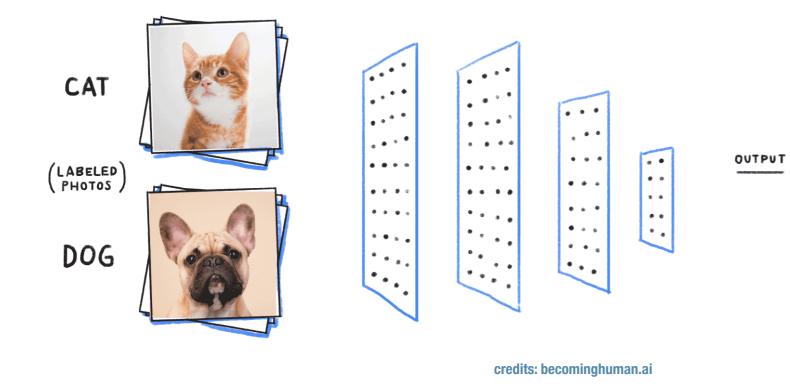
machine learning is reshaping the way we think analyses and searches



https://news.mit.edu/2019/boosting-computing-power-forfuture-particle-physics-mit-lns-0819

#### **Deep learning revolution**

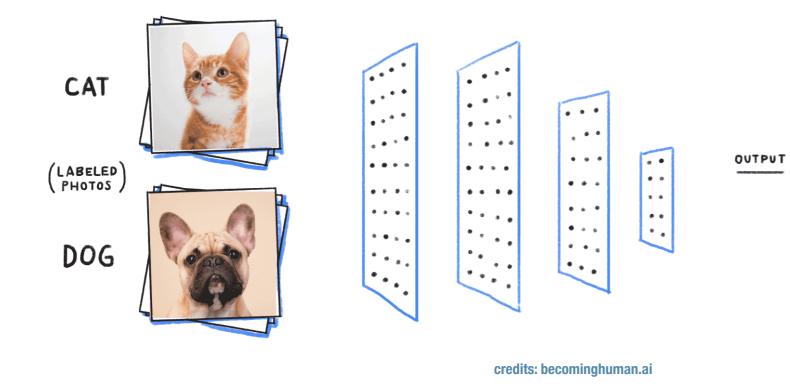
- a wave of machine learning algorithms has hit HEP in the recent past
- ML algorithms are powerful tools for classification, and they have successfully applied to our tasks



- if an algorithm can distinguish pictures of cats and dogs, can it also distinguish QCD jets from boosted-objects?
- very active and fastdeveloping field

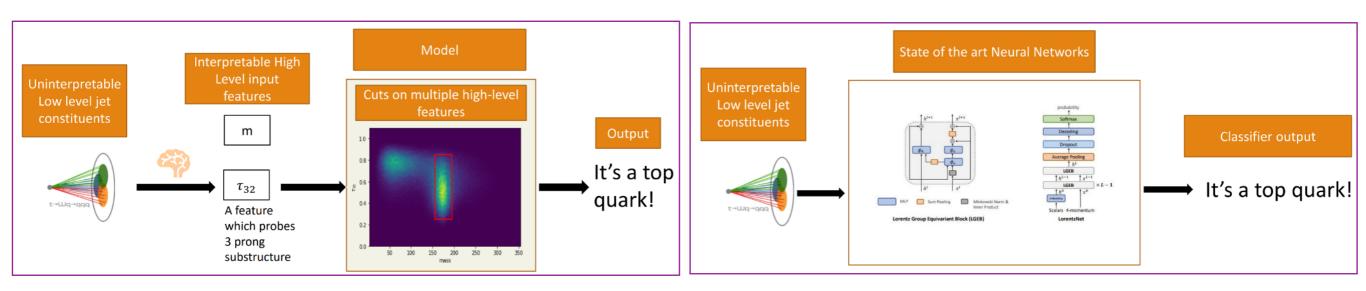
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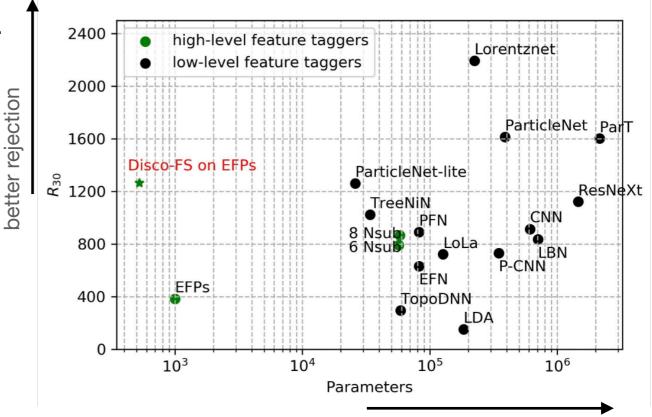


- if an algorithm can distinguish pictures of cats and dogs, can it also distinguish QCD jets from boosted-objects?
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#### **High-level vs low-level**



- traditionally, phenomenologists build "clever observables" that are able to capture the desired features of particle collisions
- neural networks and computational advances allow us to exploit low-level information (cal cells, momenta etc)
- what are pro's and con's of the two approaches?
- can we define a metric to measure robustness ?



# roduction images

RN is the largest and most powerful particle accelerator in on-proton collision data every year. A true instance of Big r rare-event detection, and hope to catch glimpses of new ed collision energies.

#### Deep Convolutional Networks

Physic

Our analysis sh

new physics p

enhancing the

suggests that t

physics-motivat

140

120

100

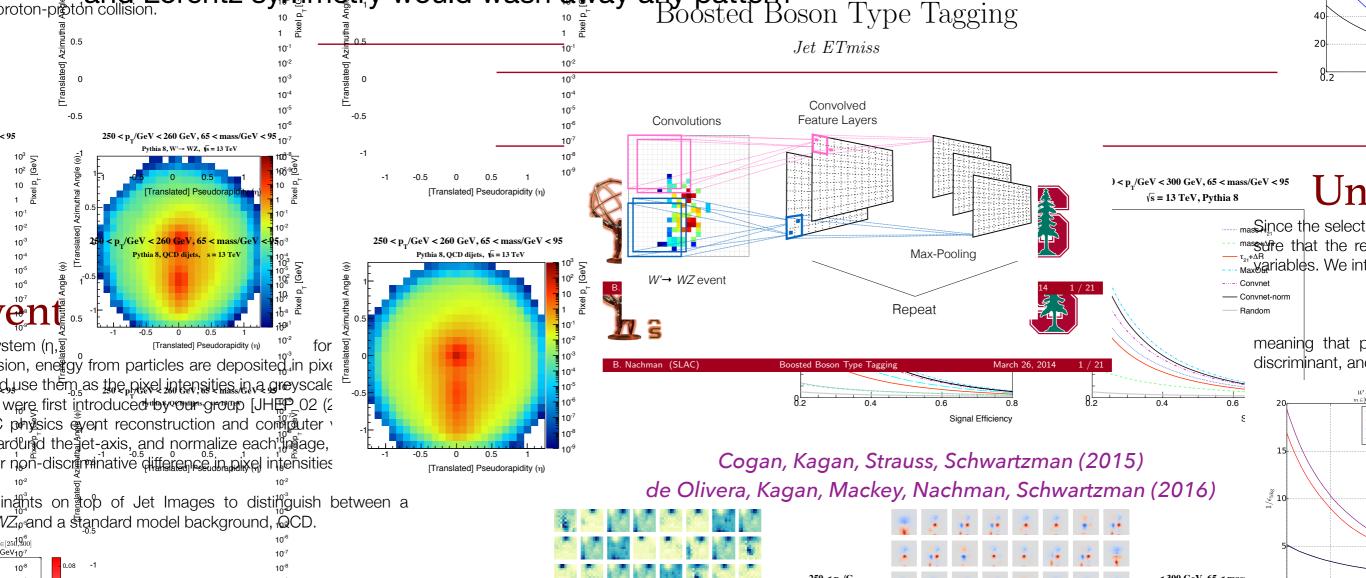
80

Deep Learning — convolutional networks in particular — currently represent the state of the art in most image recognition tasks. We apply a deep convolutional architecture to Jet Images, and perform model selection. Below, we visualize a simple architecture used to great success.

• jet images do what they say: project the project of agentitics approved a project of accuracy. The learned filters from the convolutional layers exhibit a two prong and location based success of ICLE DISHLY IS a GIVERS by concerning the provide a project of the project of the project of the provide a project of the pr

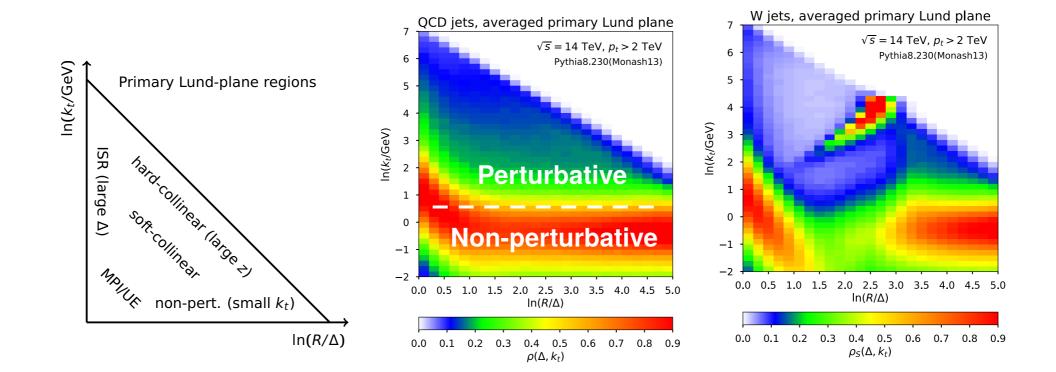
#### LAS detector

general-purpose experiments at the LHC. The 100 million repeated and reasons: we average over many reasons: we average over ma



#### **Theory inputs**

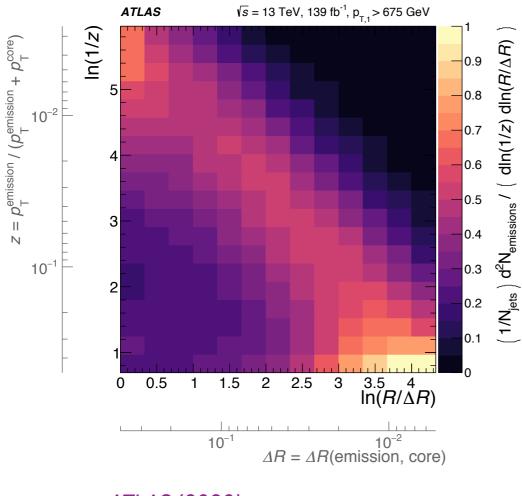
- physics intuition can lead us to construct better representations of a jet: the Lund jet plane
- the primary Lund jet plane is constructed by de-clustering the jet following the hard branch and record ( $k_t$ ,  $\Delta$ ) at each step



Dryer, Salam, Soyez (2018)

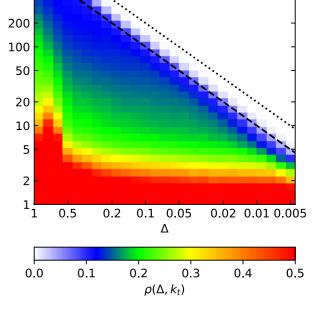
#### Mapping out the Lund plane

 ATLAS performed an unfolded measurement of the primary Lund plane density



ATLAS (2020)

1000 • First-principle calculation of  $k_t$  [GeV] the Lund plane density

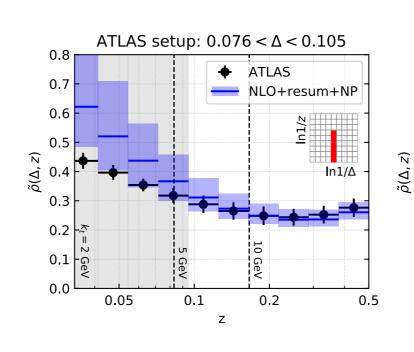


High- $p_{\perp}$  setup: NLO+resum+NP

 $p_{\perp} > 2 \text{ TeV}, R = 1$ 

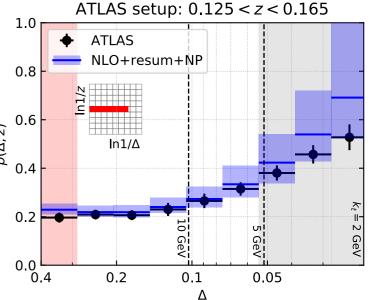
2000

500



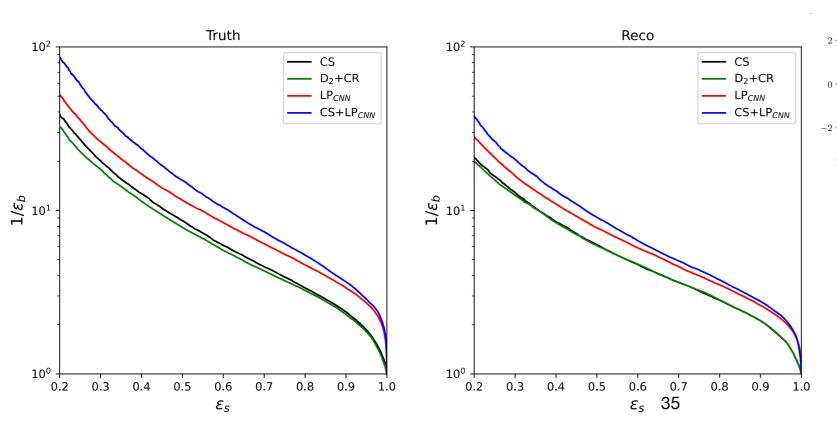
Lifson, Salam, Soyez

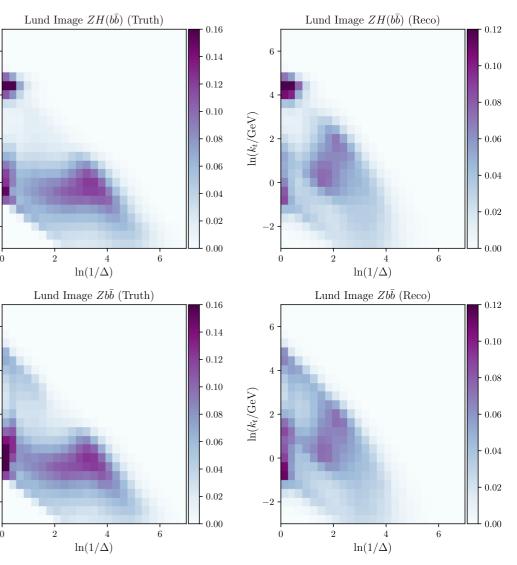
(2020)



## Lund plane images: Higgs

- primary Lund plane provides us with an alternative jet image
- used as input to CNN to built taggers
- Hbb tagger that exploits different colour correlations between  $H\to b\bar{b}$  vs  $g\to b\bar{b}$
- improved performance wrt simpler colour-sensitive variables, such as the colour ring



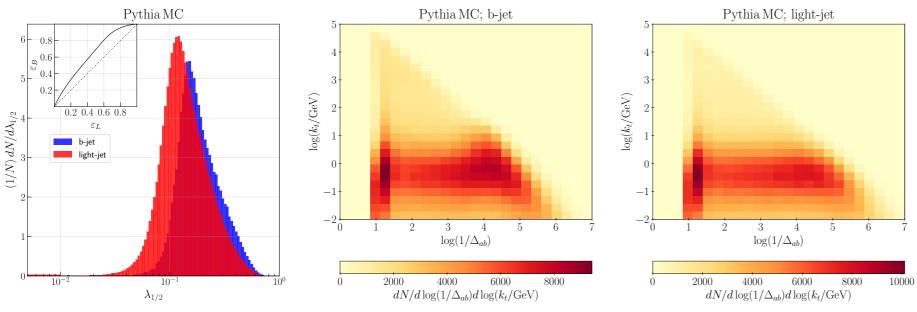


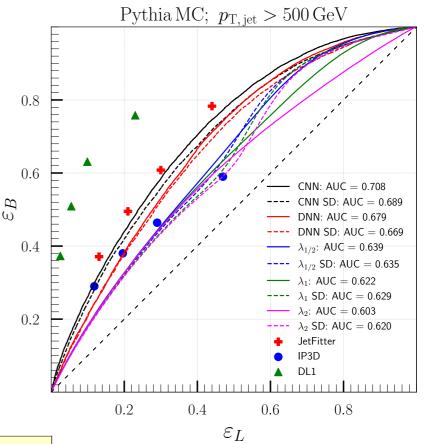
 $n(k_t/\text{GeV})$ 

 good performance also for Higgs decay into light jets, where colour ring fails

# Lund plane images: heavy quarks

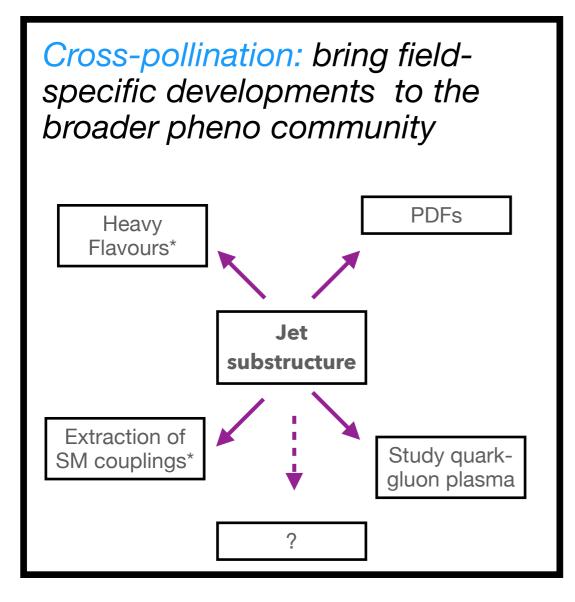
- the presence of massive quarks alters the QCD radiation pattern (so-called dead-cone effect)
- we build a *b*-tagger which exploits orthogonal information to standard approach
- again we compare to simpler variables (here jet angularities)





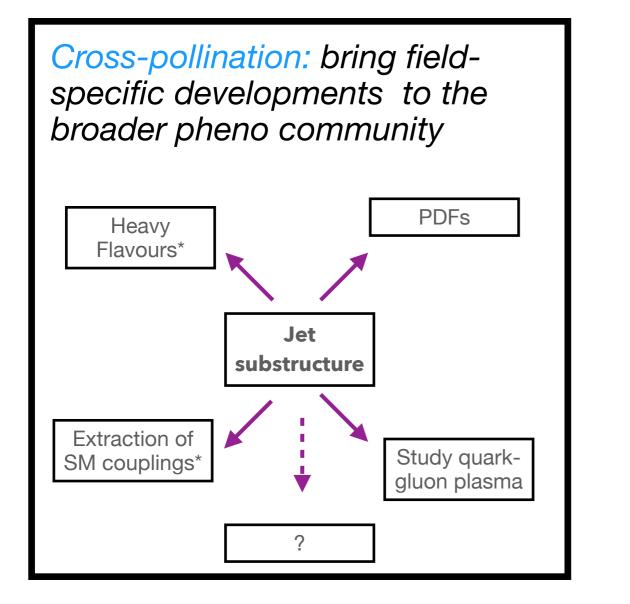
- work in progress to actually compute these distributions from firstprinciples
- resummed calculations with masses leads to nontrivial effects

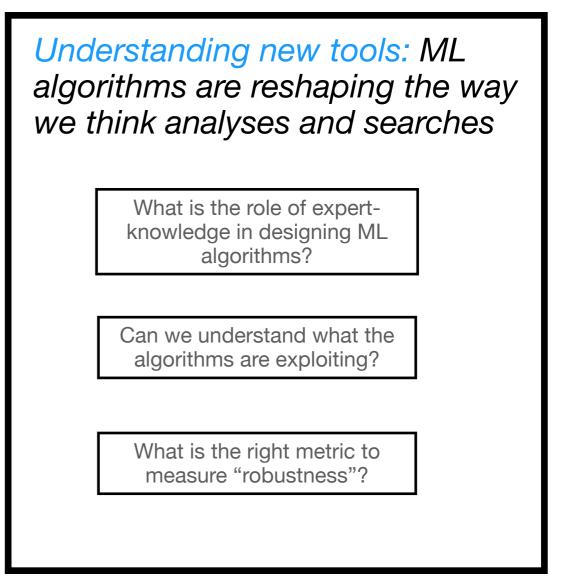
# **Conclusions and Outlook**



Understanding new tools: ML algorithms are reshaping the way we think analyses and searches What is the role of expertknowledge in designing ML algorithms? Can we understand what the algorithms are exploiting? What is the right metric to measure "robustness"?

# **Conclusions and Outlook**



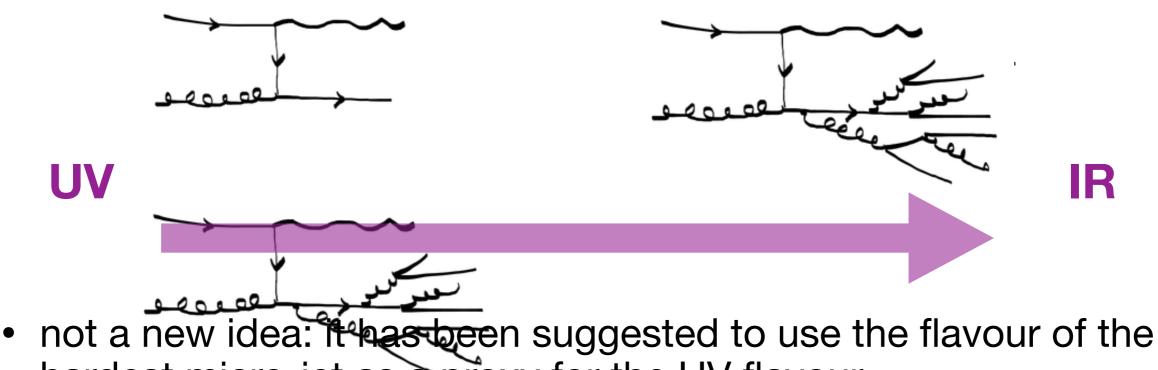


#### Thank you very much for your attention !

#### **BACKUP SLIDES**

# A more physical approach

- "counting" (heavy) flavours may obscure the physical picture
- a more physical approach should try to link the flavour of the UV parton to the IR jet



hardest micro-jet as a proxy for the UV flavour

Dasgupta, Dryer, Salam, Soyez (2014)

# Winner Take All axis

- Cluster and find jets in your collision event with any desired jet algorithm.
- On a given jet, recluster its constituents with a pairwise, IRC safe, algorithm, using the WTA recombination scheme. Specifically:
  - 1. For all pairs i and j of particles in your jet, calculate the pairwise metric  $d_{ij}$ .
  - 2. For the pair i and j that corresponds to the smallest dij, recombine their momenta into a new massless particle ij such that  $E_{ij}=E_i+E_j$ , and the direction of ij is along the direction of the harder between i and j.
  - 3. Replace particles i and j with their combination ij in the collection of particles in the jet.
  - 4. Repeat clustering until there is a single, combined particle that remains. The direction of this particle corresponds to the WTA axis of the jet.
- The sum of the flavour of all particles in the jet whose momenta lie exactly along the WTA axis is defined to be the flavour of the jet.

# Winner-Take-All Flavour

- instead of the hardest micro-jet, assign the flavour of the jet to be the flavour of the particle(s) lying along the Winner-Take-All axis (see backup for WTA definition)
- this definition is soft-safe, but collinear unsafe: we make use of WTA fragmentation functions
- unlike the micro-jet approach, the evolution equations are DGLAP-like and, hence, linear

 $Q^{2} \frac{df_{q}(x,Q^{2})}{dQ^{2}} = \frac{\alpha_{S}}{2\pi} \int_{x}^{\min[1,2x]} \frac{dz}{z} \left[ P_{qg \leftarrow q} \left(\frac{x}{z}\right) f_{q}(z,Q^{2}) + P_{q\bar{q}\leftarrow g} \left(\frac{x}{z}\right) f_{g}(z,Q^{2}) \right]$  $Q^{2} \frac{df_{g}(x,Q^{2})}{dQ^{2}} = \frac{\alpha_{s}}{2\pi} \int_{x}^{\min[1,2x]} \frac{dz}{z} \left[ P_{gq\leftarrow q} \left(\frac{x}{z}\right) \sum_{i=1}^{n_{f}} \left( f_{q_{i}}(z,Q^{2}) + f_{\bar{q}_{i}}(z,Q^{2}) \right) + P_{gg\leftarrow g} \left(\frac{x}{z}\right) f_{g}(z,Q^{2}) \right]$ 

• interesting properties, e.g. IR fixed-point

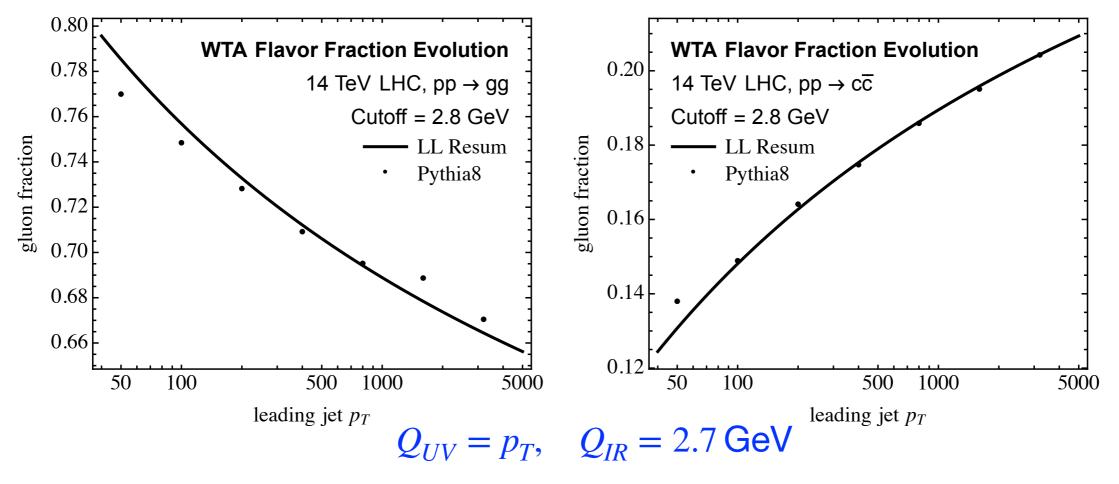
### **Comparison to Parton Shower**

good agreement between LL evolution and Pythia parton shower

$$f_g(Q_{UV}) = 1,$$

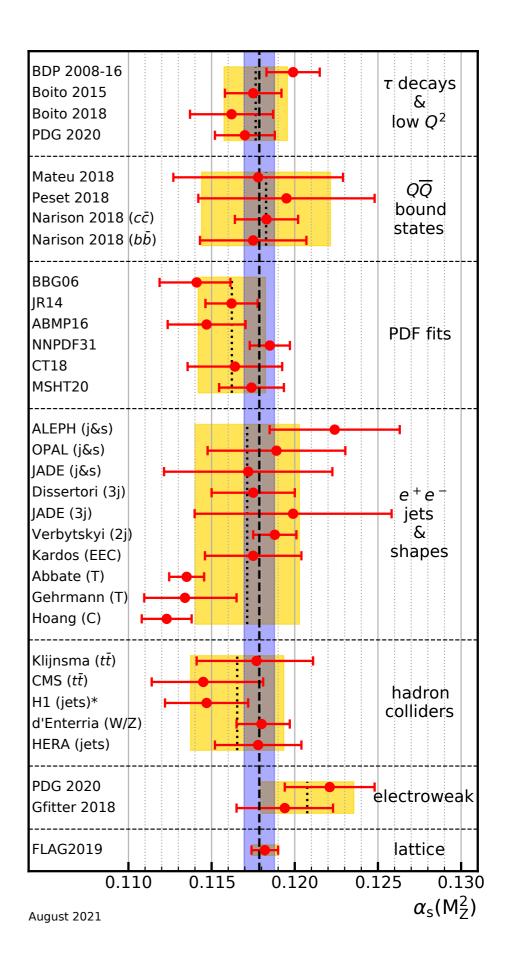
 $f_q(Q_{UV}) = 0, \forall q$ 

 $f_c(Q_{UV}) = 1,$  $f_g(Q_{UV}) = f_q(Q_{UV}) = 0, \forall q \neq c$ 

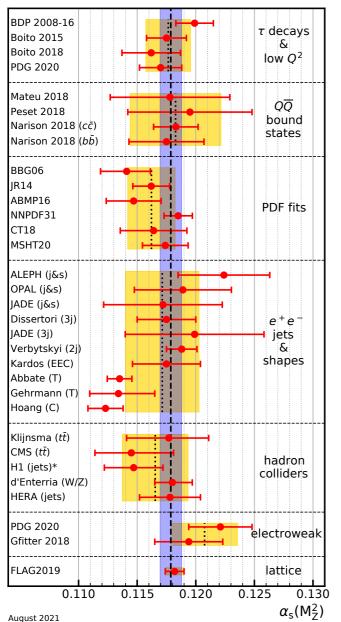


## Crosspollination (II)

#### the strong coupling

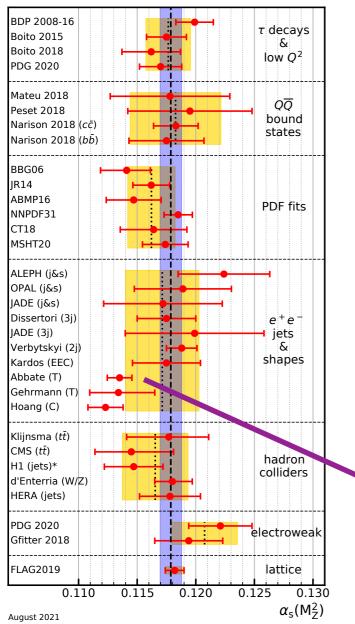


#### $\alpha_s$ from event shapes

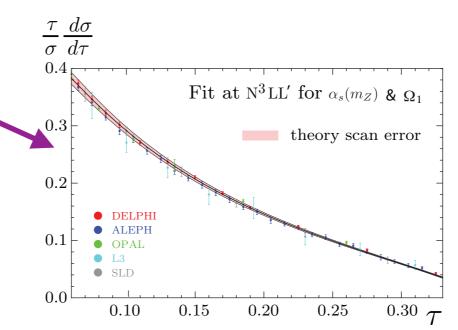


- current precision below 1%, dominated by lattice extractions
- LEP event shapes also very precise (5%)
- however they are in tension with the world average
- thrust (and C parameter) known with outstanding accuracy
  - strong coupling correlated with non-perturbative parameters
  - need better understanding of these effects
  - or... we can try and reduce them

#### $\alpha_s$ from event shapes

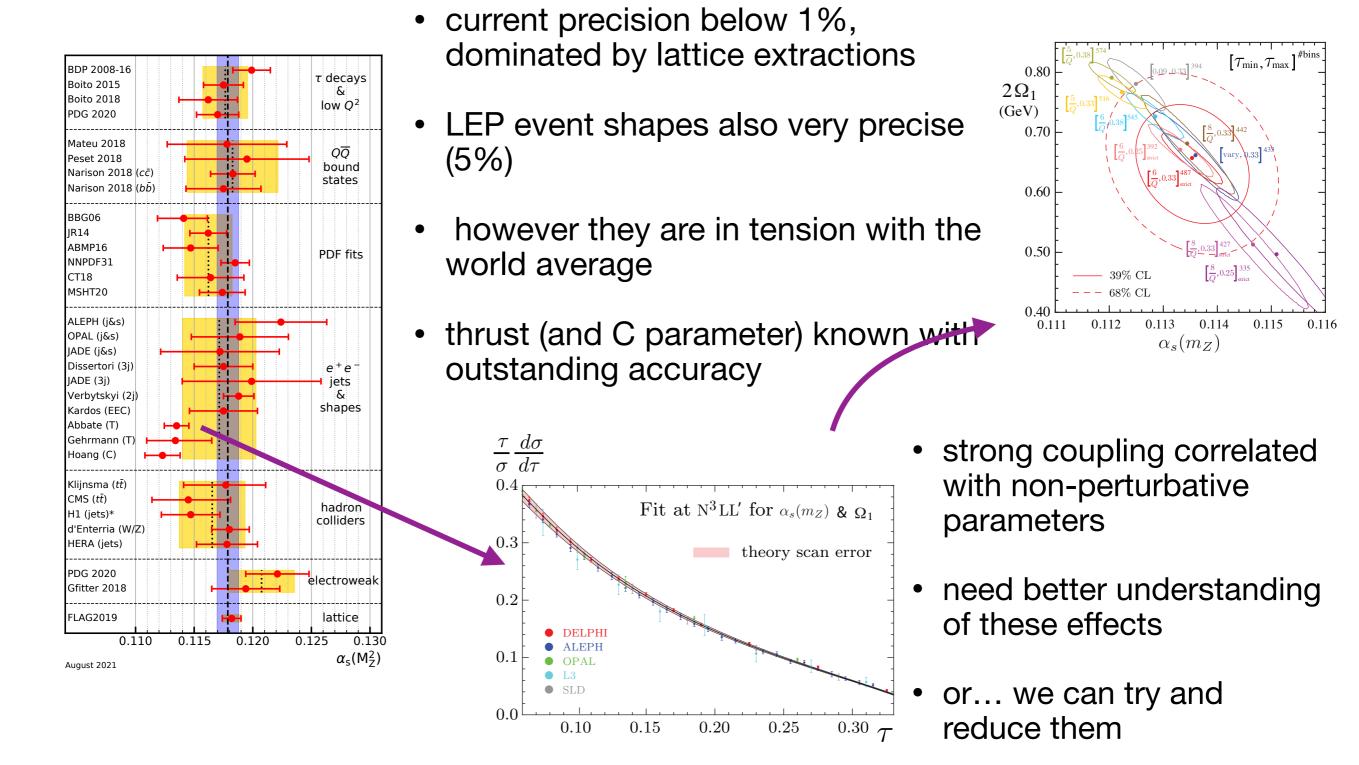


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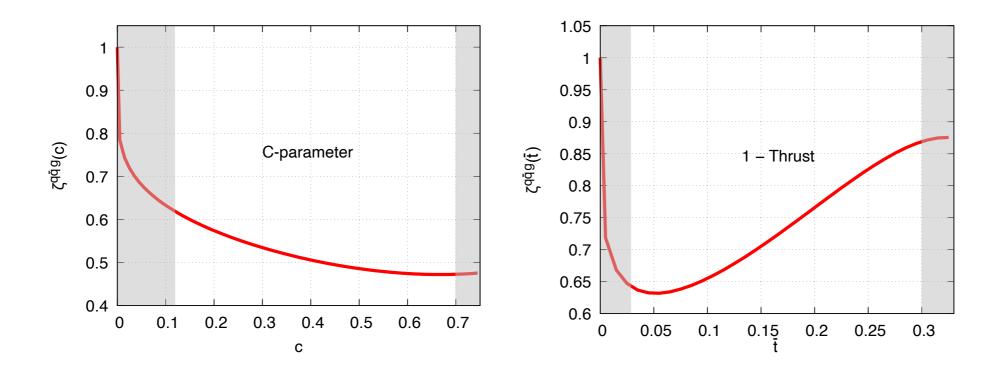
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### $\alpha_s$ from event shapes



# A better understanding

- MC models of non-pert corrections are tuned with parton showers of limited accuracy
- analytic models of non-pert corrections are usually derived in the two-jet limit
- recently a full calculation of the leading non-pert corrections has been performed
- can these improvements alleviate the tension in strong coupling determinations using event shapes?



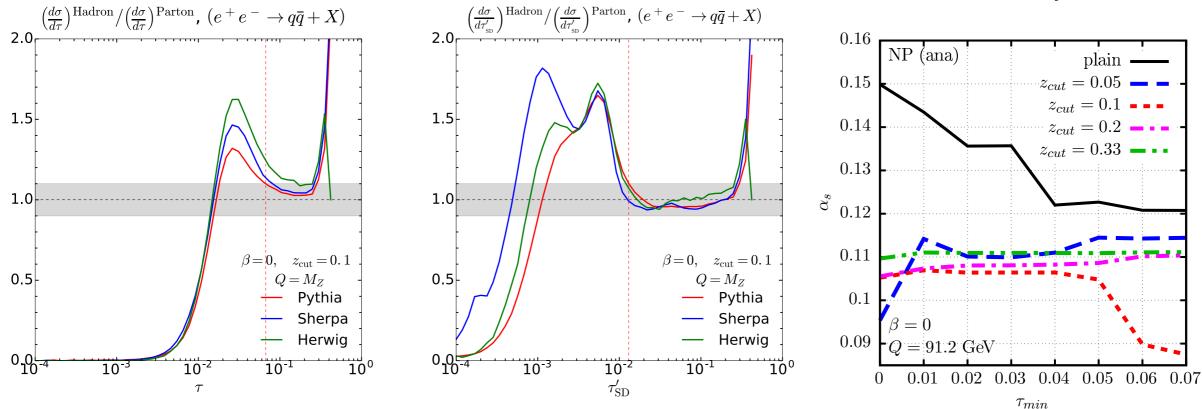
45

Luisoni, Monni, Salam (2021)

Caola, Ferrario Ravasio, Limatola, Melnikov, Nason (2021)

# **Groomed event shapes**

- in the past decade, our understanding of jets has improved tremendously
- efficient and robust grooming and tagging algorithms have been developed and exploited at the LHC
- Soft Drop aims to clean up a jet by removing soft radiation

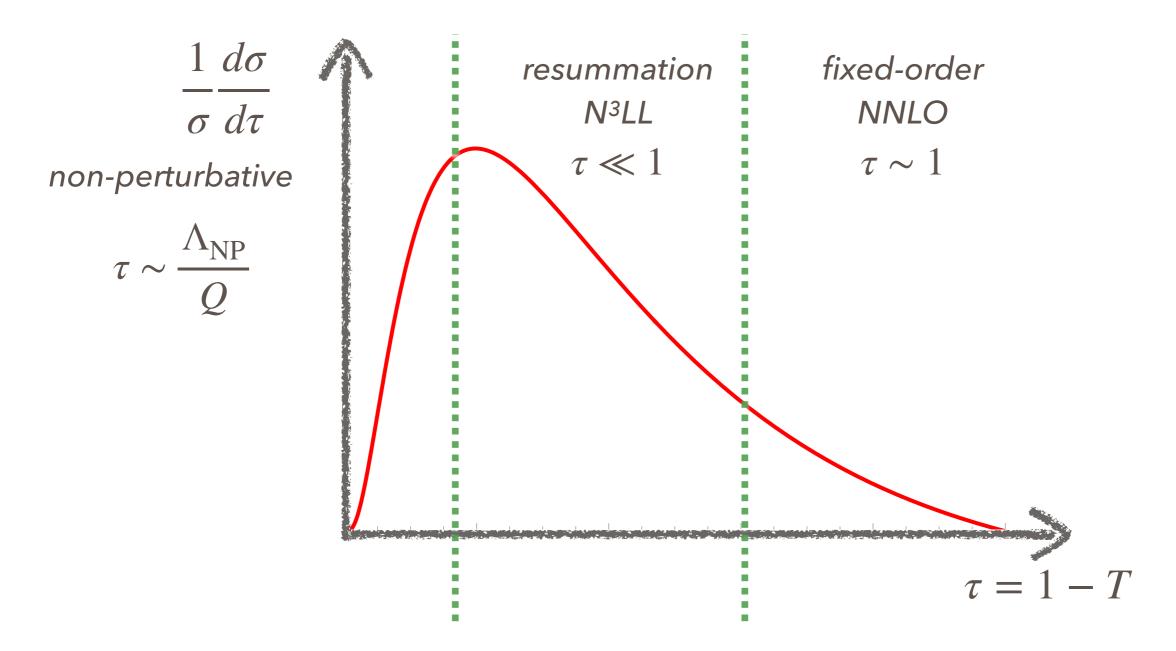


Larkoski, SM, Soyez, Thaler (2014)

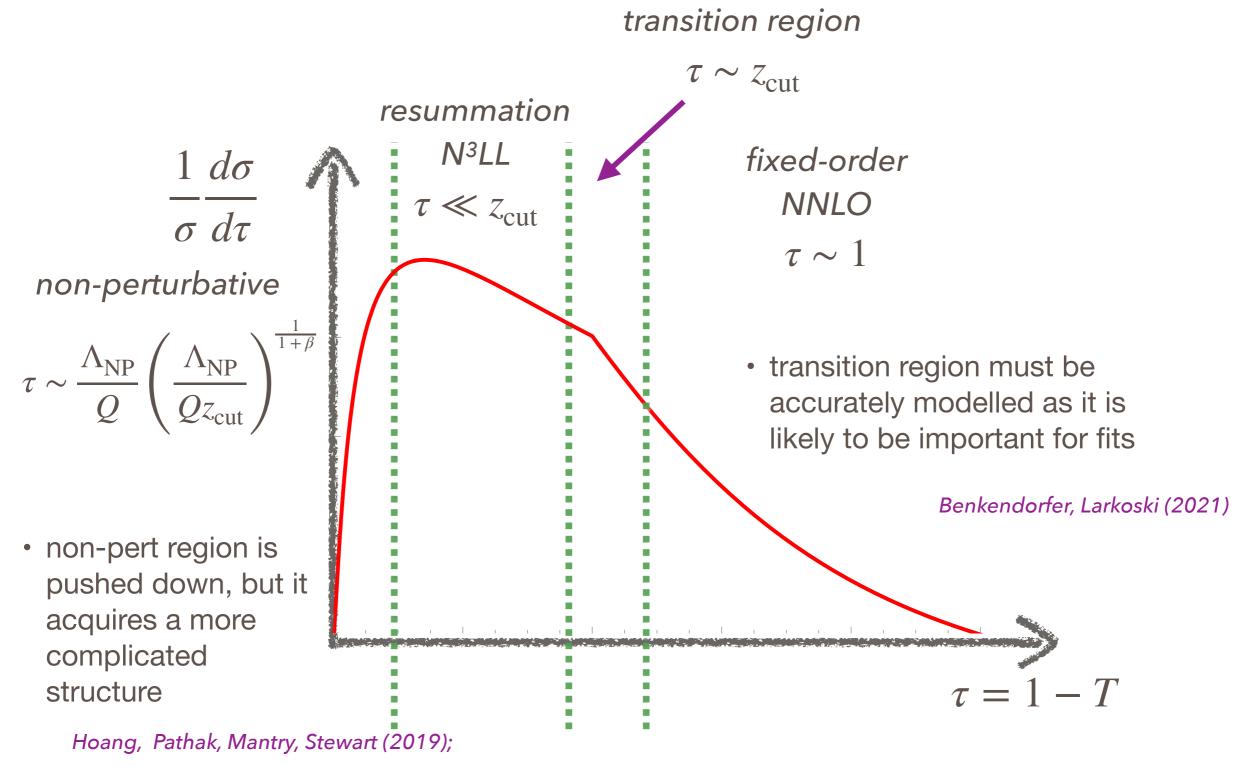
Baron, SM, Theeuwes (2018)

#### SM, Reichelt, Schumann, Soyez, Theeuwes (2019)

## **Challenges for Soft Drop thrust**



# **Challenges for Soft Drop thrust**



Pathak, Vaida, Stewart, Zoppi (2020)

#### we compare the sensitivity to $\alpha_s$ with that of $\delta$ the parameters. We will state ways of normalizing the cross section. 4.1 Quark- and gluon-jet dependence on $\alpha_s$ Fitting for Before analyzing the theoretica Oncertainty on as measurement o inclusive x-sec in the $p_T - \eta$ biause to discuss how the jet mass spectrum values with $\alpha_s$ . A le $\beta = 1$ $\beta = 0$ gives, cf. Eq. (C.18) $d^3\sigma$ $- \operatorname{NP va} \frac{\mathrm{d} \, \sigma}{\mathrm{d} \, \rho_T \, \mathrm{d} \, \eta \, \mathrm{d} \, \xi} \qquad 20 \left[ \frac{d \, \sigma_{\mathsf{ts}}^{\kappa}}{d \, \log \, \xi} \exp \left[ - \frac{\alpha_s(\mu) C_{\kappa}}{\pi} - \operatorname{NP variation} \xi \right] = \xi^{-1} \right]$ o the fit range: 600 GeV $z_{\rm cut} = 0.1, \ R = 0.8$ pert variation ange was defined in Eq. (3.3) where $a_{k}$ is a $\xi$ -independent constant. This estimate shows the ang $g_{-10}^{\circ}$ uon-jet dependence on $\alpha_s$ solution is solution to $\alpha_s$ . One can be proportional to $\alpha_s$ . ng the theoretical uncertainty on $\alpha_{T}$ manalyze how the slope of the jet mass spectrum chang as how $^{20}$ he jet mass spectrum varies with an estimate of a similar variation of the slo $\beta = 0$ $\beta = 1$ C.18),with $\beta = 0$ , we notice a striking feature: while the estimate fr $d\sigma_{sd}^{\mathcal{H}annesdottir, Pathak, Schwartz Stepht (2930)d}$ for the LL curves (varying the slope of the spectron $\alpha_s(\mu) \mathcal{O}_{\kappa}^{\mathcal{H}}$ as jet, $p_T = 600 \text{ GeV}$ (4.2) • Soft Ogo mass more sensitive to pert. effects than non-pert ones (but choice interval approximation)of normalisation is important) beyond, since the jet mass spectrum flattens out. Thus, $\beta$ -independent constant. This estimate shows that the LL formula predicts the with $\beta > \beta$ masinspearture essentially only have quark jets while limportant limitation for op isl NLL and NNEL predictions.

e how the about estimate of a similar variation of the slope, see Fig. 5. For mark jets  $2^{20}$  and  $2^{-30}$  by  $\pm 10\%$ , we must are obtained as the sensitivity to  $a_{3,2}^{-30}$  by  $\pm 10\%$ .

#### It's all very nice but we have no data

- groomed event shapes or other substructure variables can be used as highprecision observables for future lepton colliders
- reduced sensitivity to non-perturbative physics will allow for cleaner extractions of Standard Model parameters, including the strong coupling
- what can we do now? use LEP archived data!
- there is an MIT led collaboration using ALPEH data, what about data from the other LEP experiments?

PHYSICAL REVIEW LETTERS 123, 212002 (2019)

Measurements of Two-Particle Correlations in  $e^+e^-$  Collisions at 91 GeV with ALEPH Archived Data

Anthony Badea,<sup>1</sup> Austin Baty<sup>1</sup>, Paoti Chang,<sup>2</sup> Gian Michele Innocenti,<sup>1</sup> Marcello Maggi,<sup>3</sup> Christopher McGinn,<sup>1</sup> Michael Peters,<sup>1</sup> Tzu-An Sheng,<sup>2</sup> Jesse Thaler<sup>1</sup>, and Yen-Jie Lee<sup>1</sup>,<sup>\*</sup> <sup>1</sup>Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, USA <sup>2</sup>National Taiwan University, Taipei 10617, Taiwan <sup>3</sup>INFN Sezione di Bari, Bari, Italy Jet energy spectrum and substructure in  $e^+e^-$  collisions at 91.2 GeV with ALEPH Archived Data

Yi Chen,<sup>1,\*</sup> Anthony Badea,<sup>2</sup> Austin Baty,<sup>3</sup> Paoti Chang,<sup>4</sup> Yang-Ting Chien,<sup>5</sup> Gian Michele Innocenti,<sup>6</sup> Marcello Maggi,<sup>7</sup> Christopher McGinn,<sup>8</sup> Dennis V. Perepelitsa,<sup>8</sup> Michael Peters,<sup>1</sup> Tzu-An Sheng,<sup>1</sup> Jesse Thaler,<sup>1</sup> and Yen-Jie Lee<sup>1,†</sup>

Measurements of jet rates with the anti- $\mathbf{k}_t$  and SISCone algorithms at LEP with the OPAL detector

Stefan Kluth<sup>1,a</sup> and Andrii Verbytskyi<sup>1</sup> <sup>1</sup>MPI für Physik, Föhringer Ring 6, 80805 München, Germany

Abstract. We study jet production in e<sup>+</sup>e<sup>-</sup> annihilation to hadrons with data recorded by the OPAL experiment at LEP at centre-of-mass energies between 90 GeV and 207 GeV. The jet production rates were measured for the first time with the anti-k<sub>t</sub> and SISCone jet clustering algorithms. We compare the data with predictions by modern Monte Carlo event generators.