

Aspects of Jets from First Principles

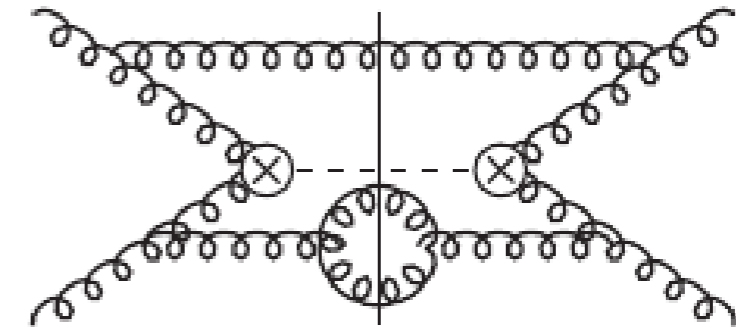
Jesse Thaler



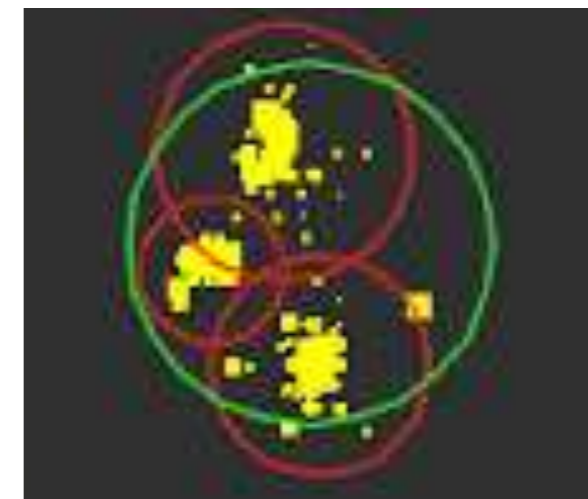
Jets @ LHC, ICTS-TIFR, Bangalore — January 25, 2017

QCD Renaissance

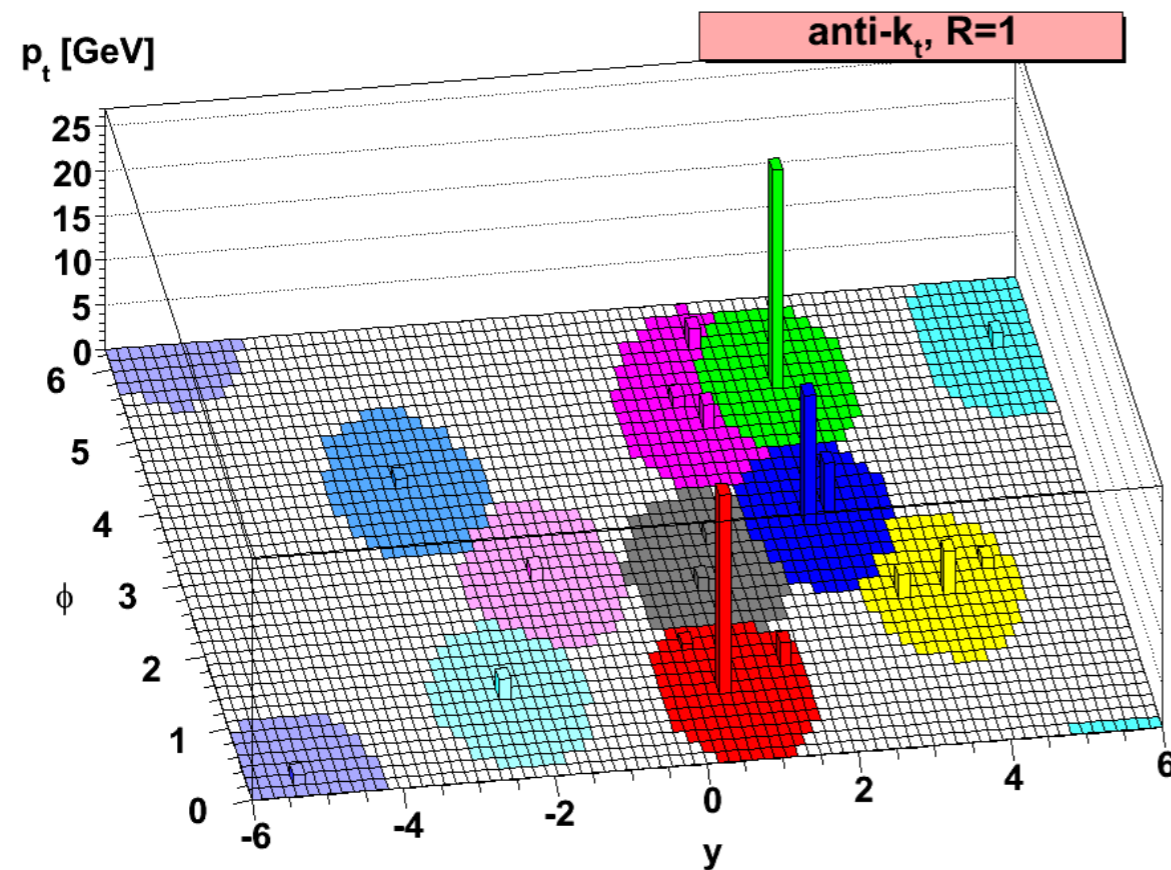
Theory c. 2008–present



Loop/Leg/Log Explosion



Jet Substructure



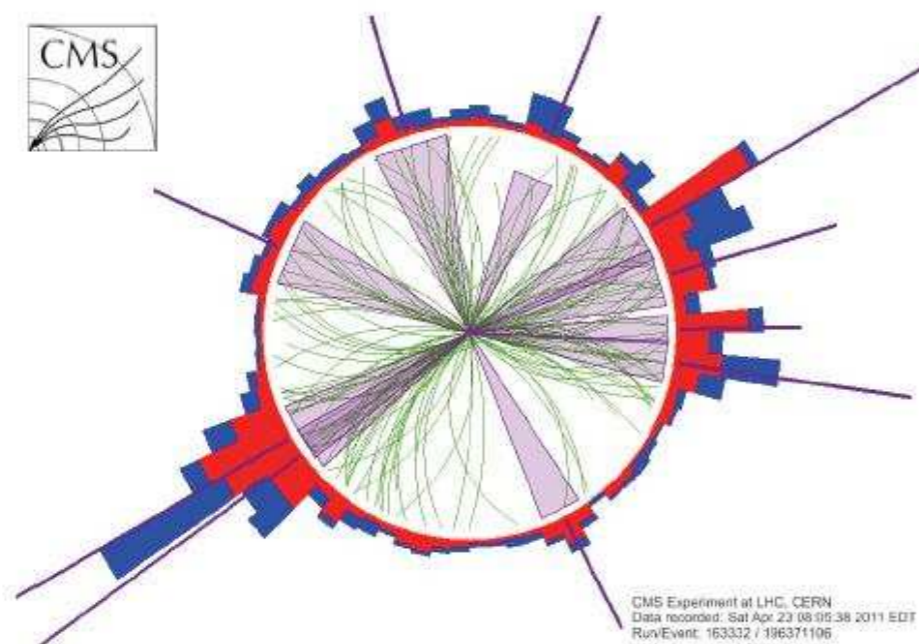
New Jet Algorithms

[Anti- k_T : Cacciari, Salam, Soyez, 2008; see also Delsart, 2006] [N³LO: Anastasiou, Duhr, Dulat, Herzog, Mistlberger, 2015]
[BDRS: Butterworth, Davison, Rubin, Salam, 2008; see also Seymour, 1991, 1994]

What is a Jet?

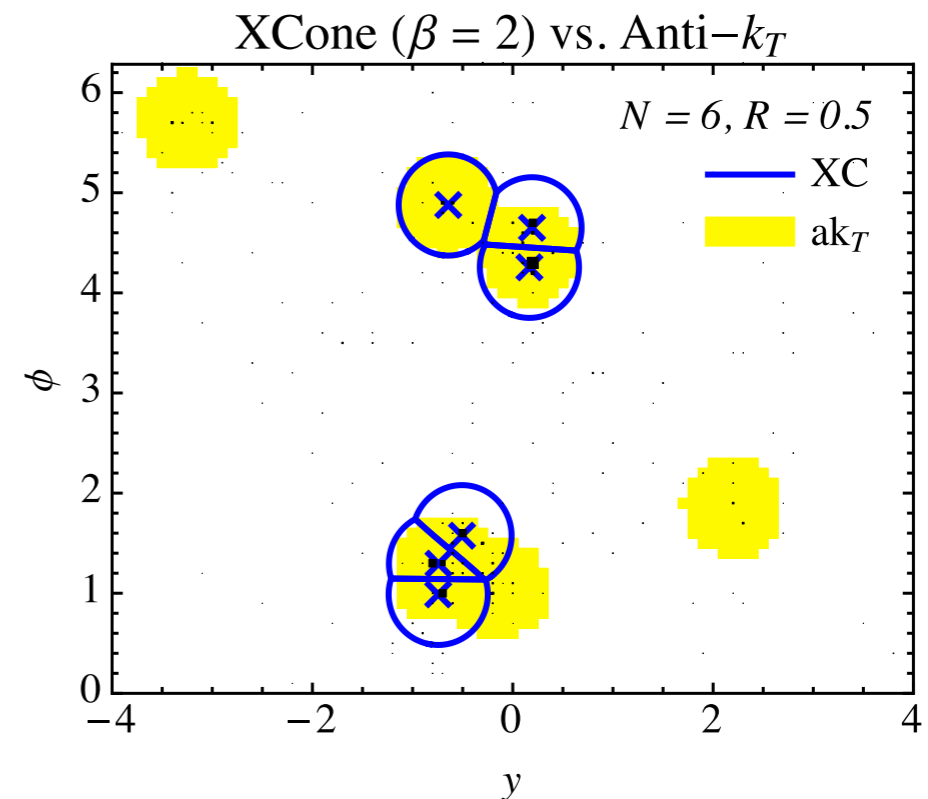
A physical phenomena:

*Emergent feature of
confining gauge theories*

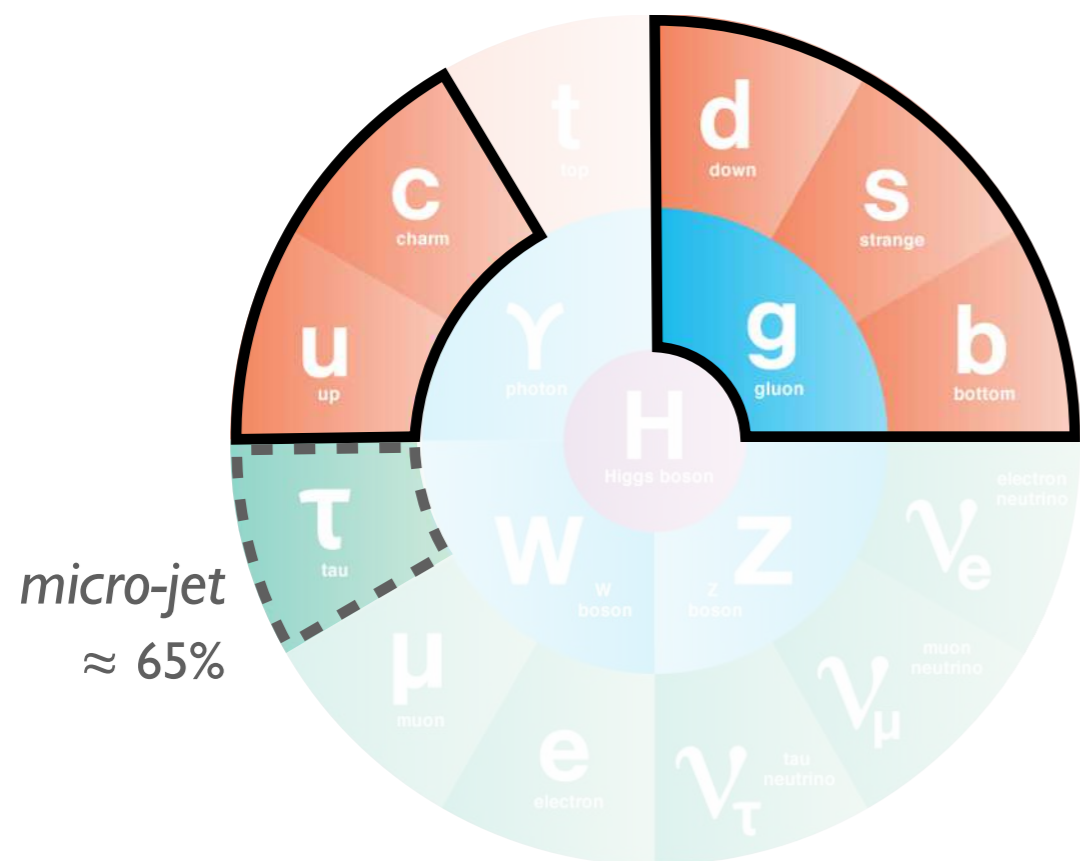


An analysis technique:

*Method to interpret
hadronic final states*

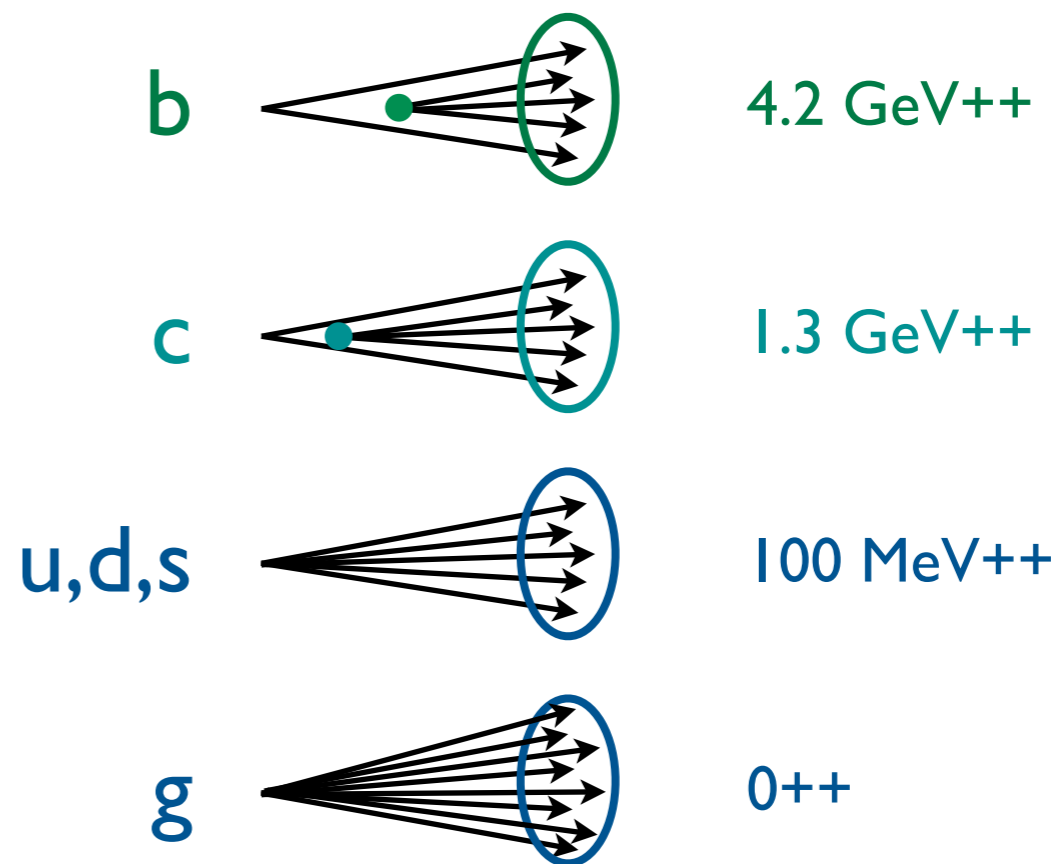


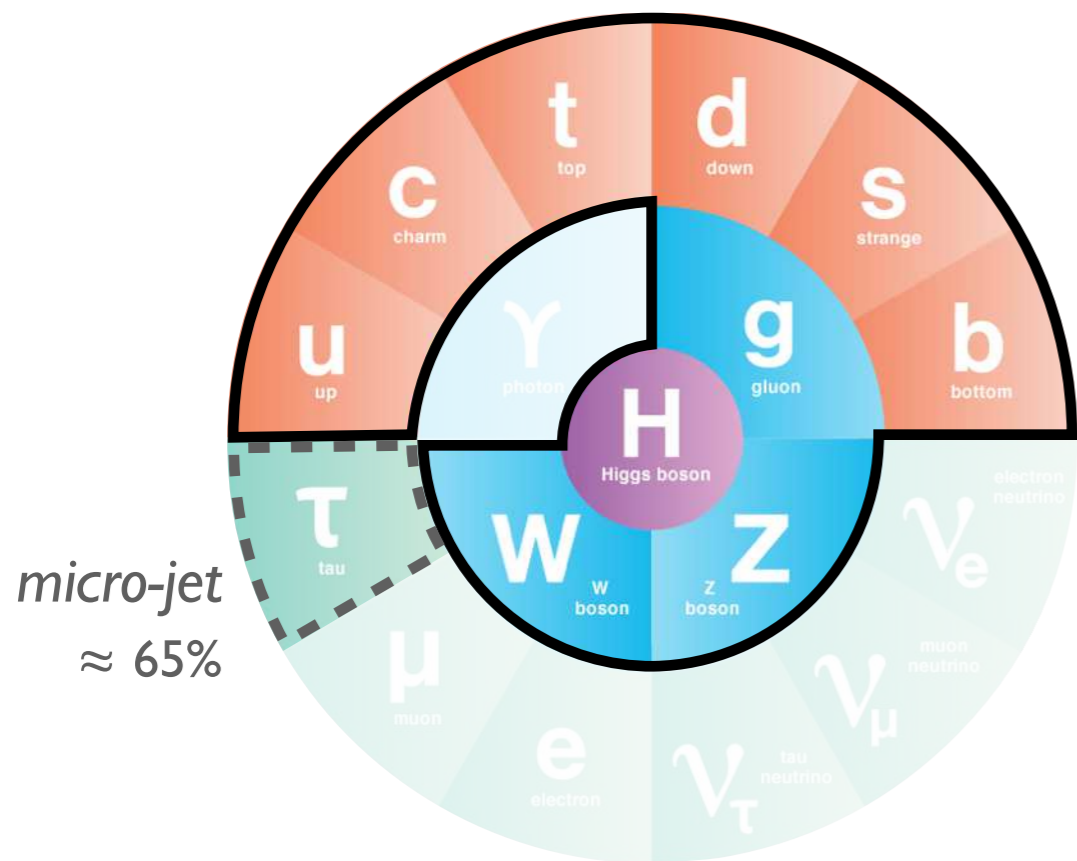
*Freedom to use
different analysis strategies for
different physical questions*



Jets from the Standard Model

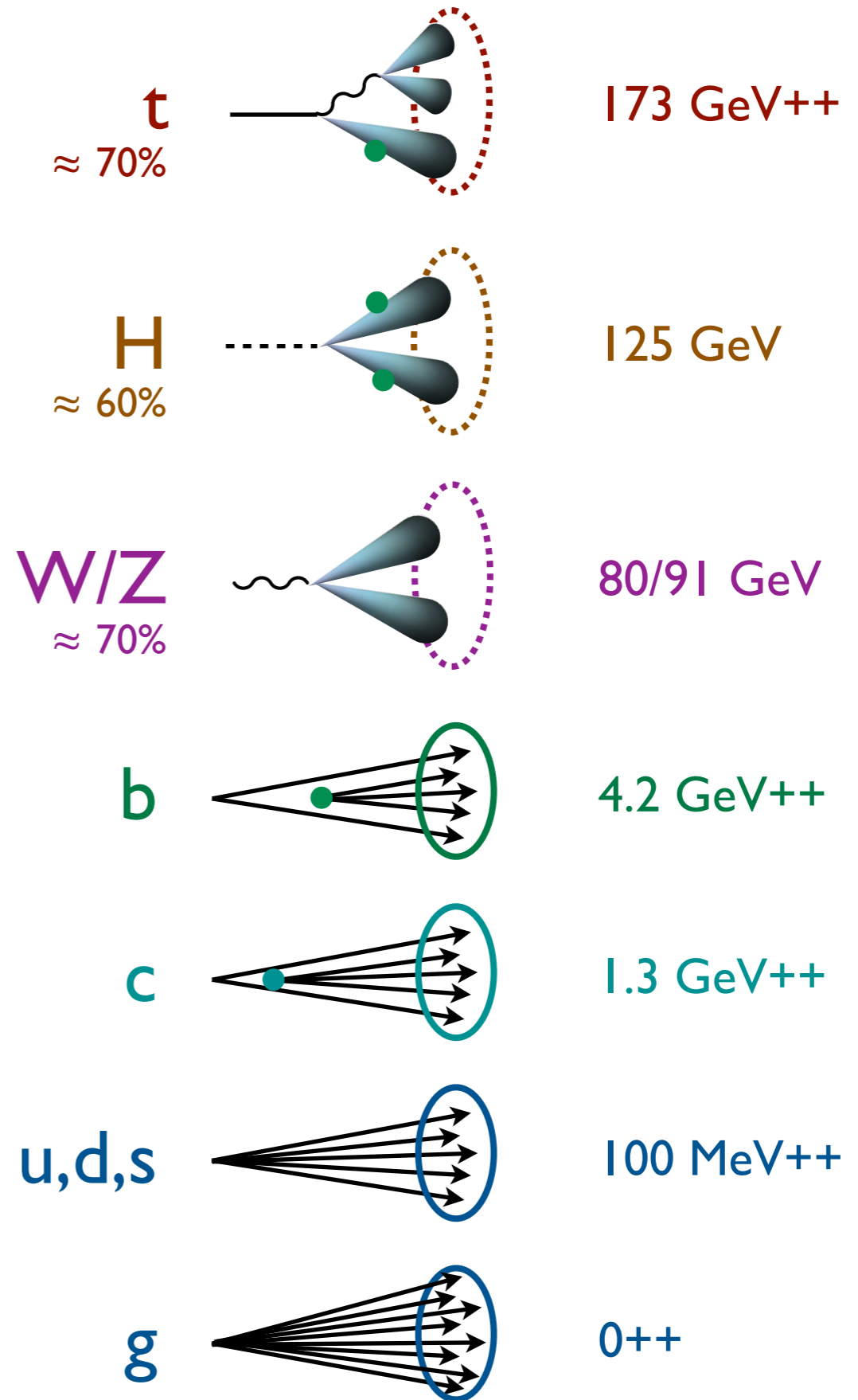
++ = plus gluonic radiation





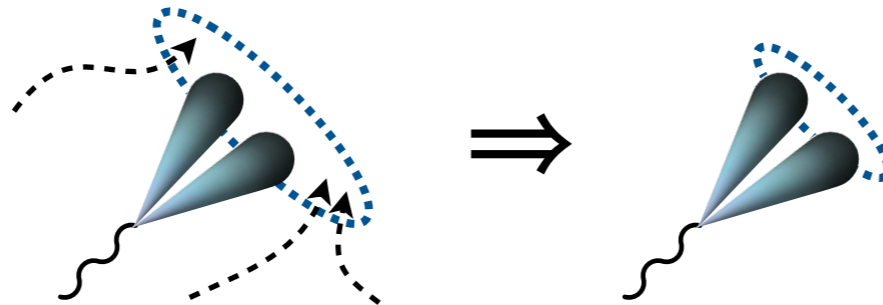
Jets from the Standard Model

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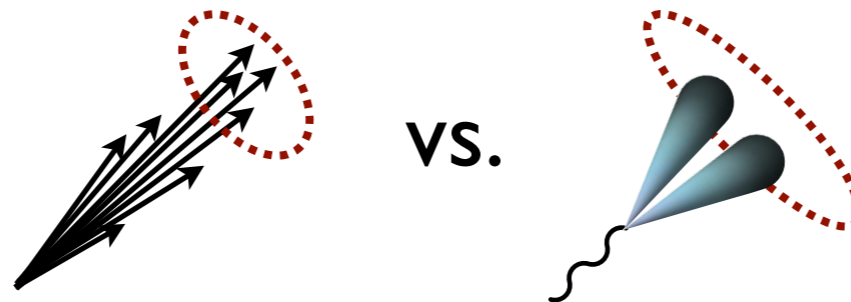
The Rise of Jet Substructure

Grooming:
e.g. ISR/UE/pileup



[Mass Drop/Filtering, Trimming, Pruning, Soft Drop, Jet Reclustering...;
for pileup: Area Subtraction, Jet Cleansing, SoftKiller, PUPPI, Constituent Subtraction...]

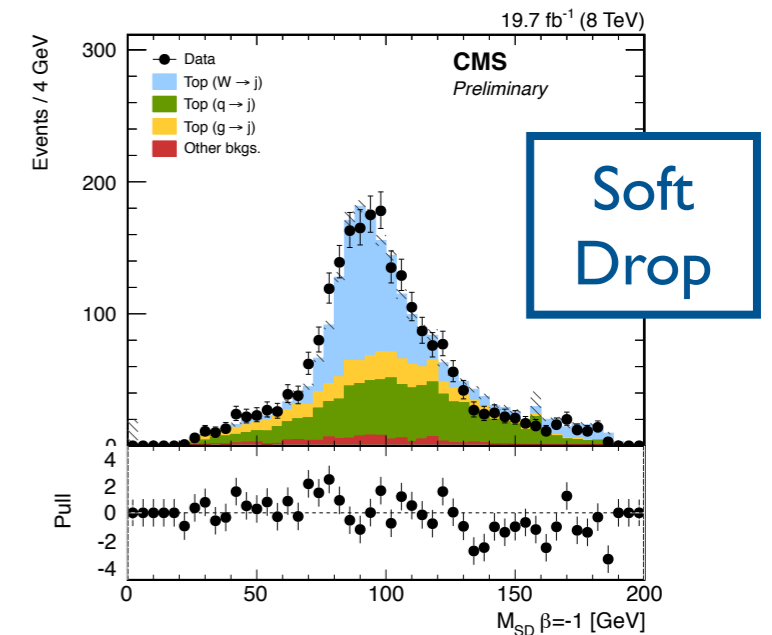
Discrimination:
e.g. 1-prong vs. N-prong



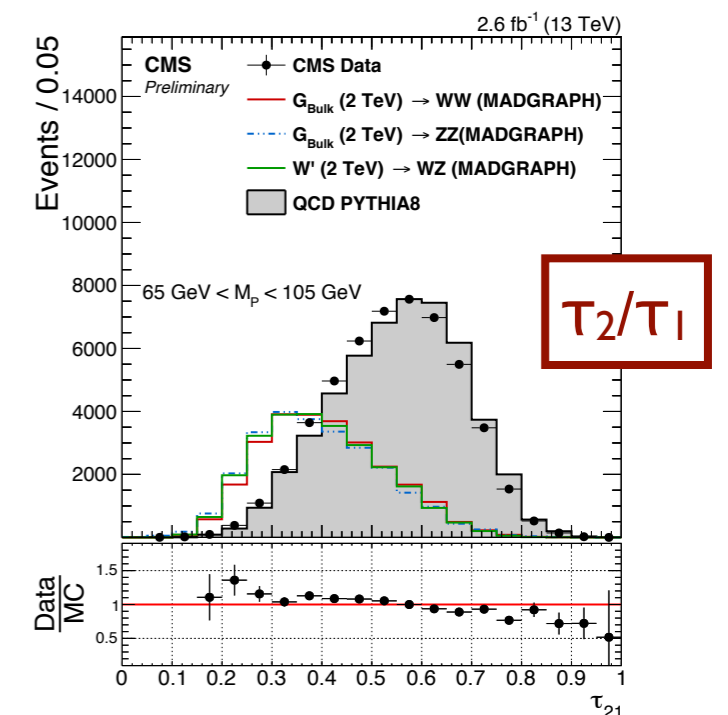
[p_T Balance, Y-splitter, Angularities, Planar Flow, N-subjettiness, Angular Structure Functions, Jet Charge, Jet Pull, Energy Correlation Functions, Dipolarity, p_T^D , Zernike Coefficients, LHA, Fox-Wolfman Moments, JHU/CMSTopTagger, HEPTopTagger, Template Method, Shower Deconstruction, Subjet Counting, Wavelets, Q-Jets, Telescoping Jets...]

W/Z-Tagging @ CMS

[JME-14-002, CMS-PAS-EXO-15-002]

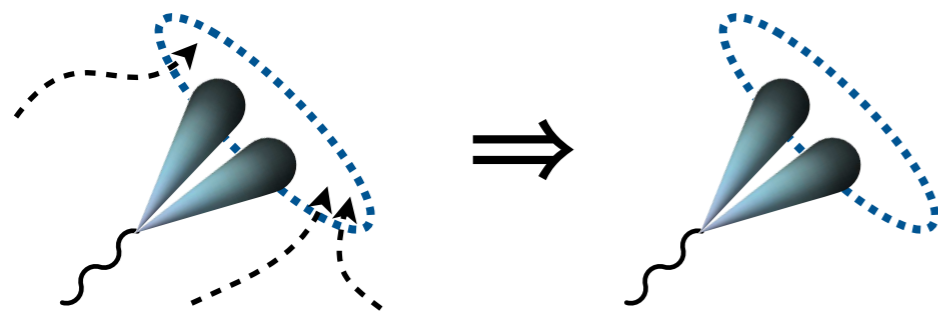


[using Larkoski, Marzani, Soyez, JDT, 1402.2657]



[using JDT, Van Tilburg, 1011.2268, 1108.2701]

W/Z Tagging in 2016

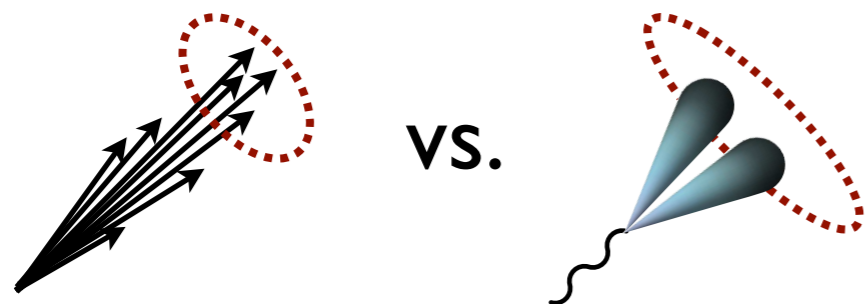


Soft Drop

[Larkoski, Marzani, Soyez, JDT, 2014;
see also Dasgupta, Fregoso, Marzani, Salam, 2013]

Trimming

[Krohn, JDT, Wang, 2009]



N-subjettiness

before grooming, after decorrelation
[JDT, Van Tilburg, 2010, 2011;
Dolen, Harris, Marzani, Rappoccio, Tran, 2016]

D₂

after grooming
[Larkoski, Mout, Neill, 2014;
based on Larkoski, Salam, JDT, 2013]

W/Z Tagging in 2016



*Can we understand
these choices from
first principles QCD?*

*Can we construct
improved algorithms
for 2017?*

Soft Drop

[Larkoski, Marzani, Soyez, JDT, 2014;
see also Dasgupta, Fregoso, Marzani, Salam, 2013]

Trimming

[Krohn, JDT, Wang, 2009]

N-subjettiness

before grooming, after decorrelation
[JDT, Van Tilburg, 2010, 2011;
Dolen, Harris, Marzani, Rappoccio, Tran, 2016]

D_2

after grooming
[Larkoski, Mout, Neill, 2014;
based on Larkoski, Salam, JDT, 2013]

W/Z Tagging in 2016



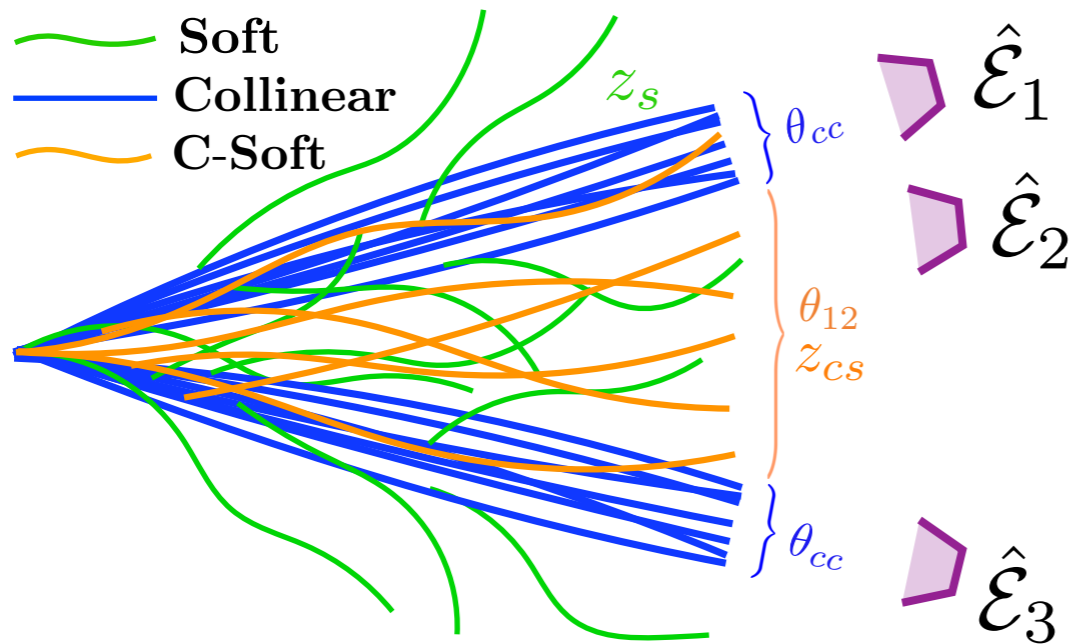
*Can we understand
these choices from
first principles QCD?*

*Can we construct
improved algorithms
for 2017?*

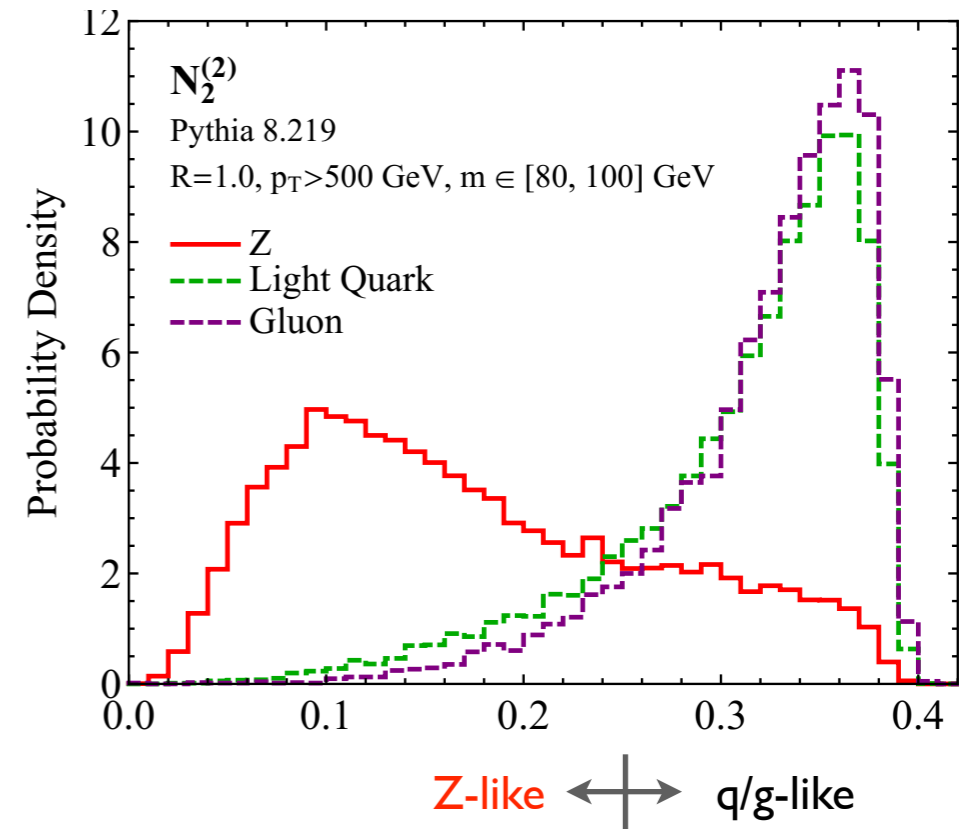
**Yes! Key realization:
Discrimination different
before/after grooming**

**Yes!
Use power counting to
design robust discriminants**

Punchline: W/Z Tagging in 2017?



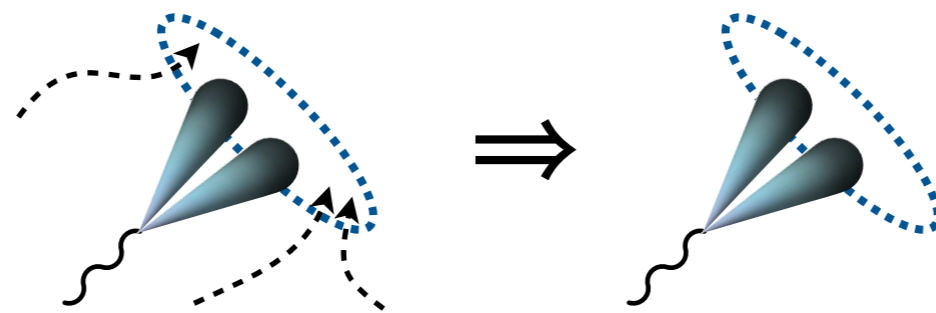
Stable & Performant



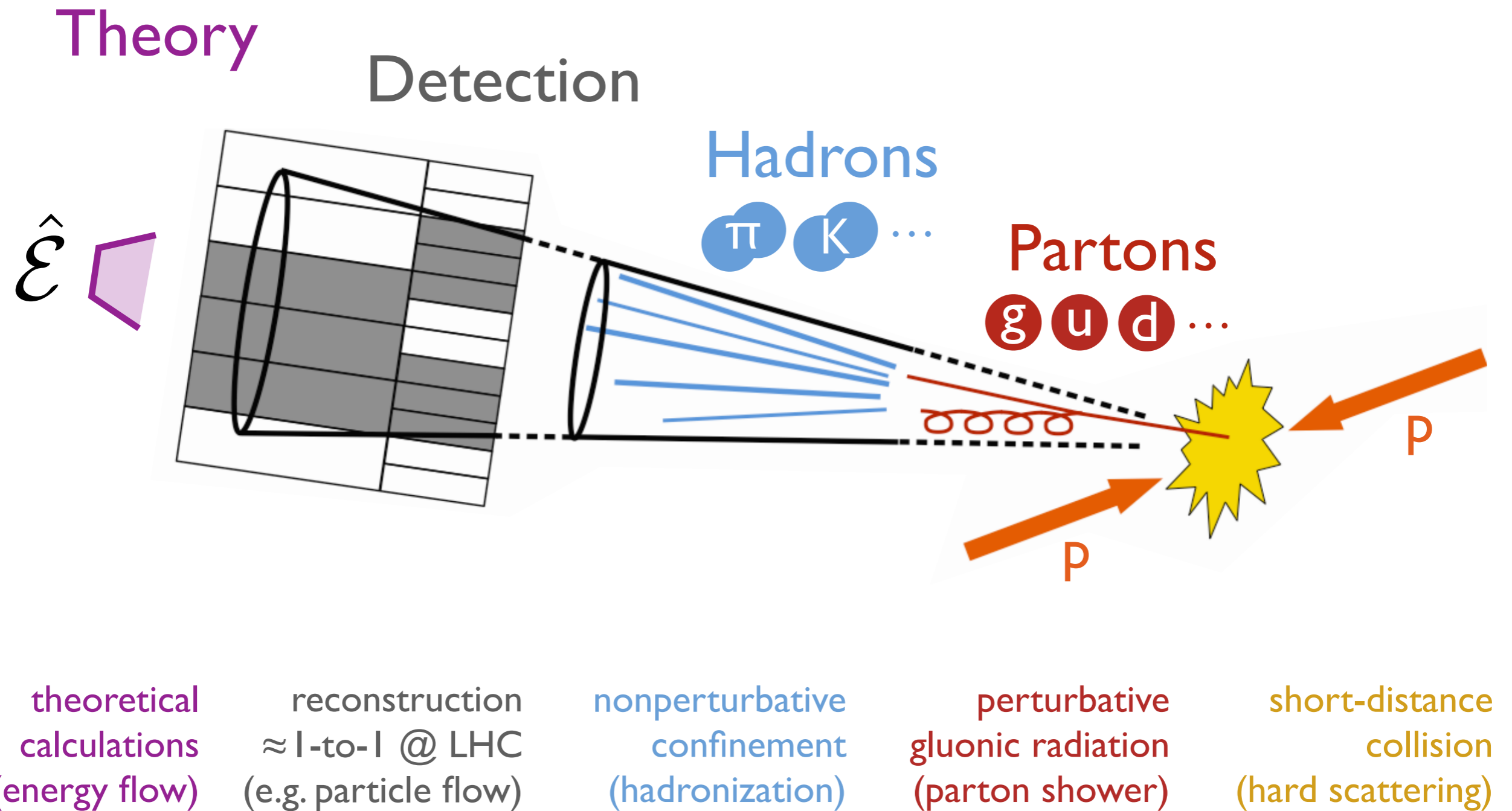
$$\boxed{N_2} = \frac{\sum_{i < j < k} p_{T_i} p_{T_j} p_{T_k} \min \left\{ (R_{ij} R_{jk})^2, (R_{jk} R_{ki})^2, (R_{ki} R_{ij})^2 \right\}}{\left(\sum_{i < j} p_{T_i} p_{T_j} R_{ij}^2 \right)^2 / \sum_i p_{T_i}}$$

[Moult, Necib, JDT, 2016]

Grooming from First Principles



Layers of Jet Understanding



First-Principles Calculations

Jet mass: Dasgupta, Khelifa-Kerfa, Marzani, Spannowsky, 1207.1640; Chien, Kelley, Schwartz, Zhu, 1208.0010; Jouttenus, Stewart, Tackmann, Waalewijn, 1302.0846

Jet shapes: Ellis, Vermilion, Walsh, Hornig, Lee, 1001.0014; Banfi, Dasgupta, Khelifa-Kerfa, Marzani, 1004.3483; Li, Li, Yuan, 1107.4535; Larkoski, Neill, JDT, 1401.2158; Hornig, Makris, Mehen, 1601.01319

Angular scaling: Jankowiak, Larkoski, 1201.2688; Larkoski, 1207.1437

Quarks vs. gluons: Larkoski, Salam, JDT, 1305.0007; Larkoski, JDT, Waalewijn, 1408.3122; Bhattacharjee, Mukhopadhyay, Nojiri, Sakaki, Webber, 1501.04794

QCD grooming: Dasgupta, Fregoso, Marzani, Salam, 1307.0007; Dasgupta, Fregoso, Marzani, Powling, 1307.0013; Larkoski, Marzani, Soyez, JDT, 1402.2657; **Frye, Larkoski, Schwartz, Yan, 1603.06375, 1603.09338**

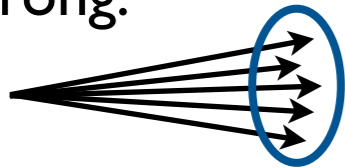
Double differential: Larkoski, JDT, 1307.1699; Larkoski, Moul, Neill, 1401.4458; Procura, Waalewijn, Zeune, 1410.6483

In heavy ions: Chien, Vitev, 1405.4293; Chien, 1411.0741

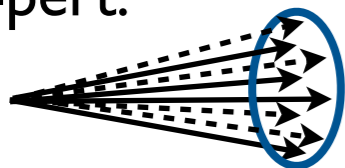
p_T balance: Larkoski, Marzani, JDT, 1502.01719; Chien, Vitev, 1608.07283

Small R jets: Dasgupta, Dreyer, Salam, Soyez, 1411.5182, 1602.01110

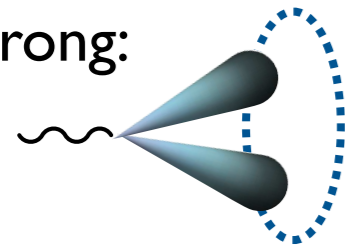
1-prong:



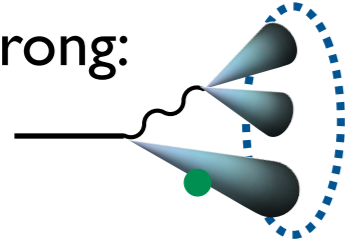
Non-pert:



2-prong:



3-prong:



Jet charge: Krohn, Schwartz, Lin, Waalewijn, 1209.2421; Waalewijn, 1209.3019

Track-only shapes: Chang, Procura, JDT, Waalewijn, 1303.6637, 1306.6630

Signal grooming: Rubin, 1002.4557; Dasgupta, Powling, Siodmok, 1503.01088

2-prong jet shapes: Feige, Schwartz, Stewart, JDT, 1204.3898; Isaacson, Li, Li, Yuan, 1505.06368

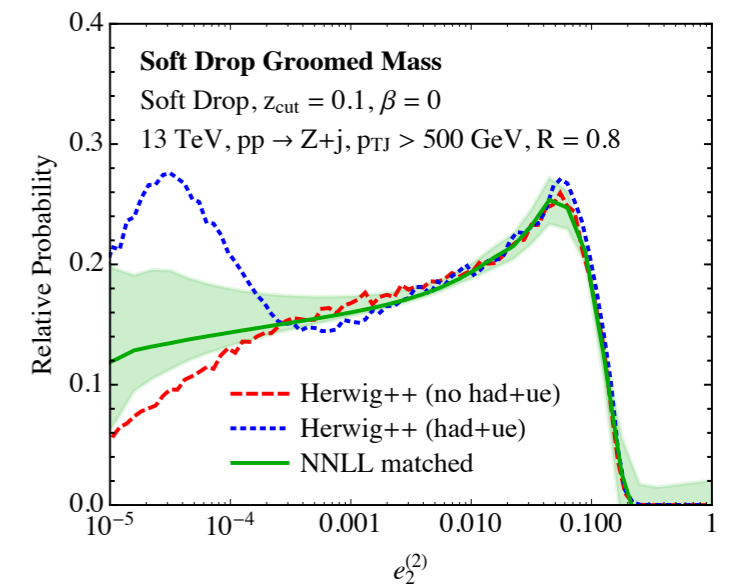
Separation power: Larkoski, Moul, Neill, 1409.6298, 1507.03018; Dasgupta, Schunk, Soyez, 1512.00516; Dasgupta, Powling, Schunk, Soyez, 1609.07149

Planar flow: Field, Gur-Ari, Kosower, Mannelli, Perez, 1212.2106

Fractional jets: Bertolini, JDT, Walsh, 1501.01965

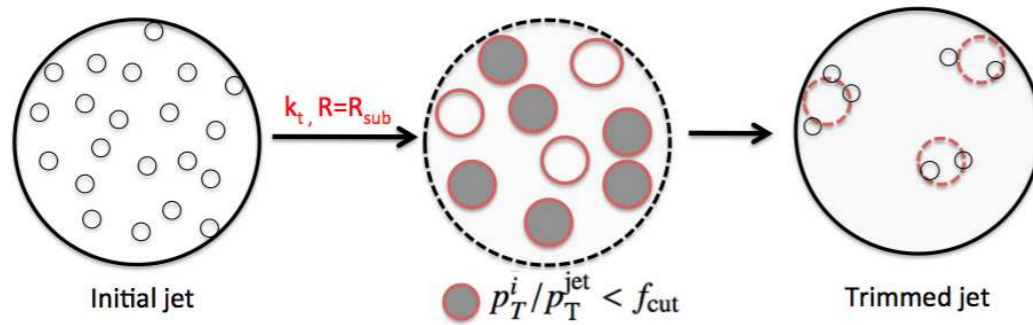
Power counting: Larkoski, Moul, Neill, 1411.0665

First NNLL + $O(\alpha_s^2)$ calculation for substructure in pp



Combination of fixed-order, direct resummation, SCET, RG evolution, and new techniques (e.g. Sudakov safety, multi-differential projections)

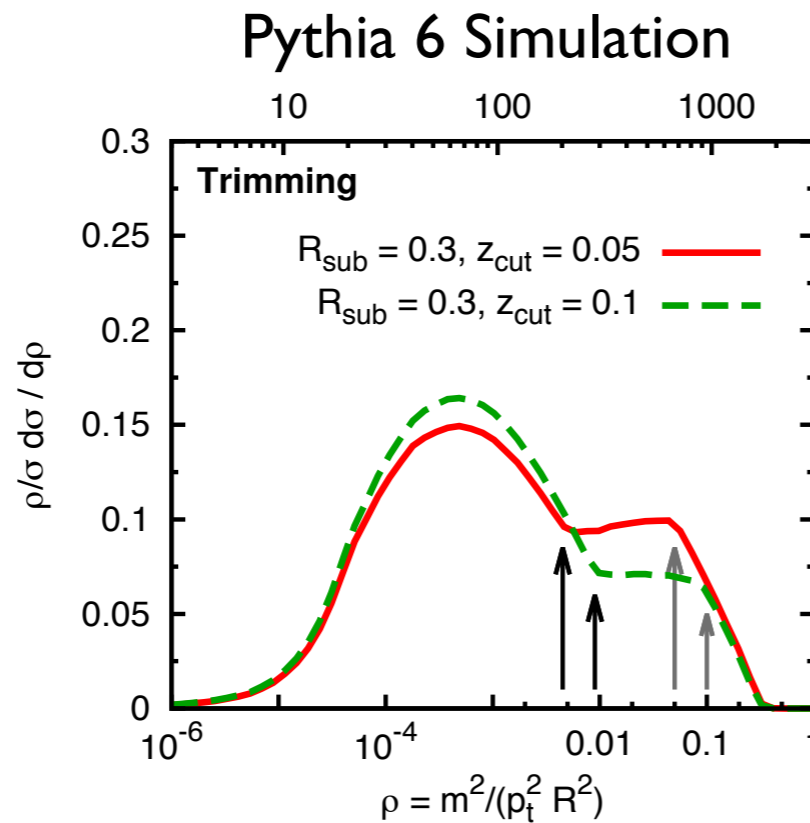
Trimming from First Principles?



R_{sub} : subjet radius
 z_{cut} : fractional energy threshold

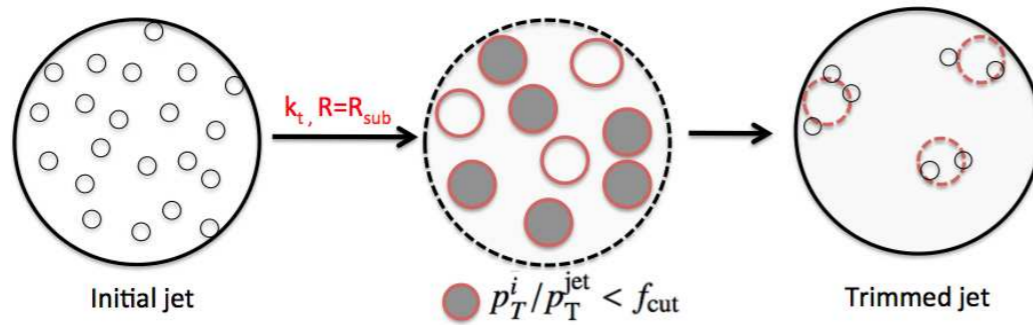
[Krohn, JDT, Wang, 0912.1342; diagram from ATLAS, 1306.4945]

Trimmed
 Jet Mass:
 3 TeV quark jets



[Dasgupta, Fregoso, Marzani, Salam, 1307.0007]

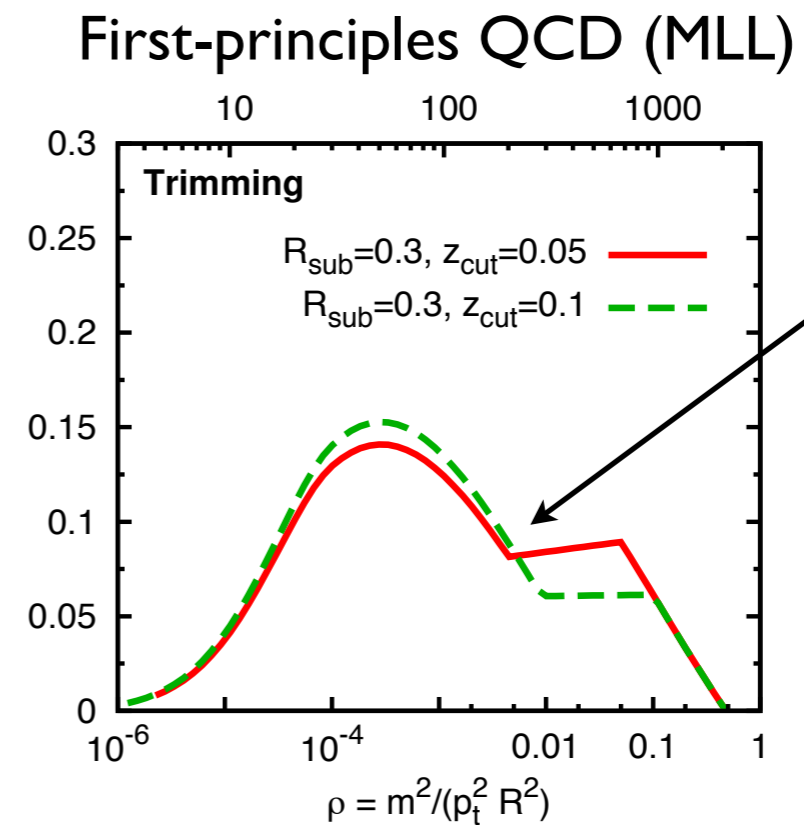
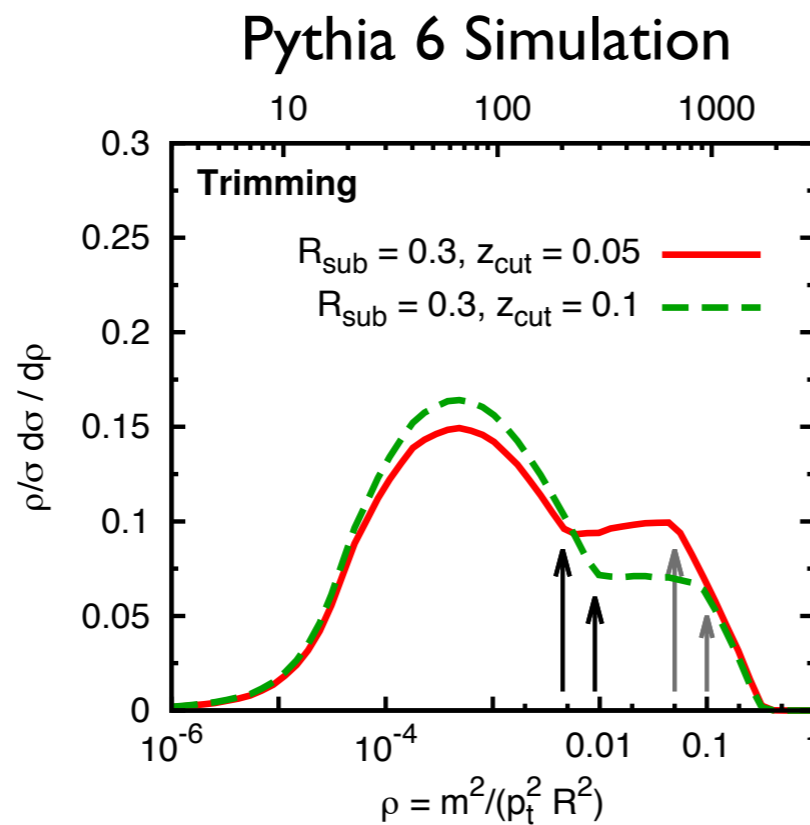
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Trimmed
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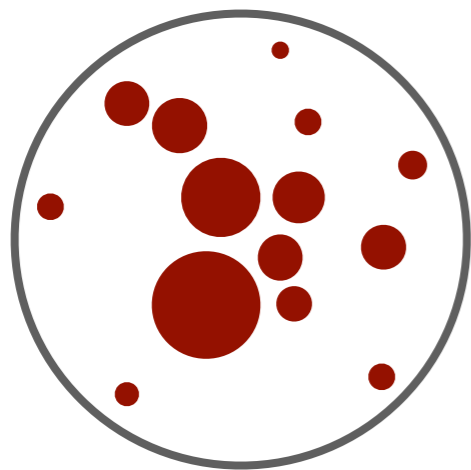
Analytically understand
 Sudakov peak,
 kink, and plateau

[Dasgupta, Fregoso, Marzani, Salam, 1307.0007]

Soft Drop Declustering

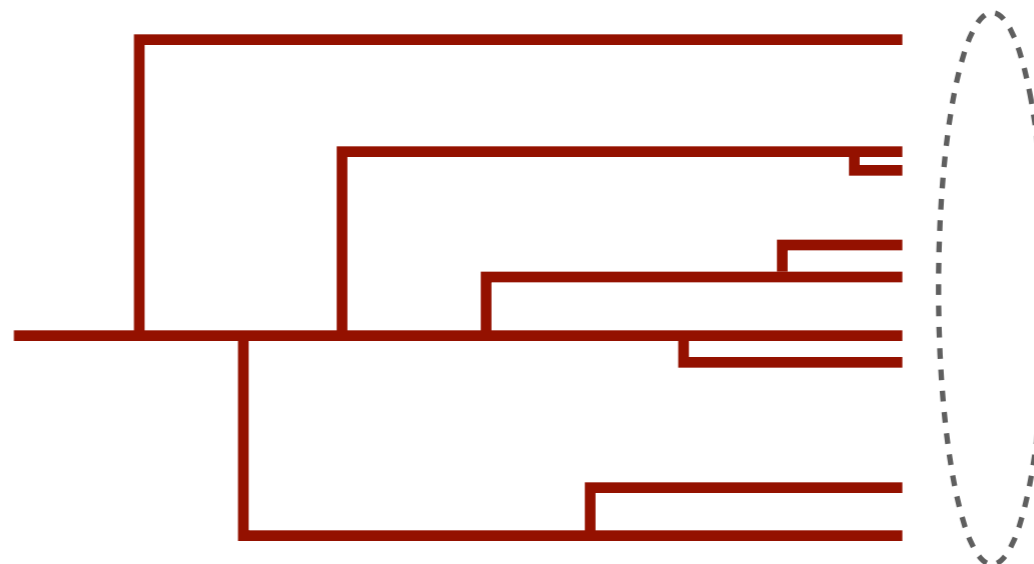


Original Jet



=

Clustering Tree

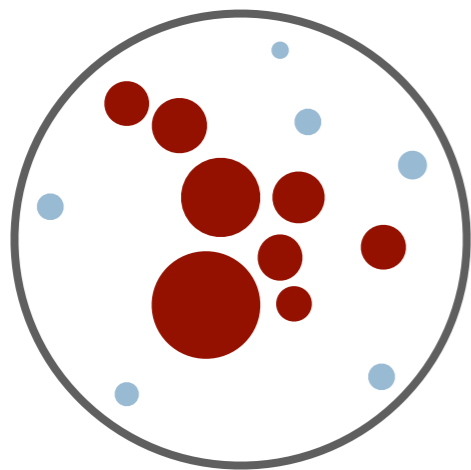


[Larkoski, Marzani, Soyez, JDT, 2014; see also Butterworth, Davison, Rubin, Salam, 2008; Dasgupta, Fregoso, Marzani, Salam/Powling, 2013]

Soft Drop Declustering

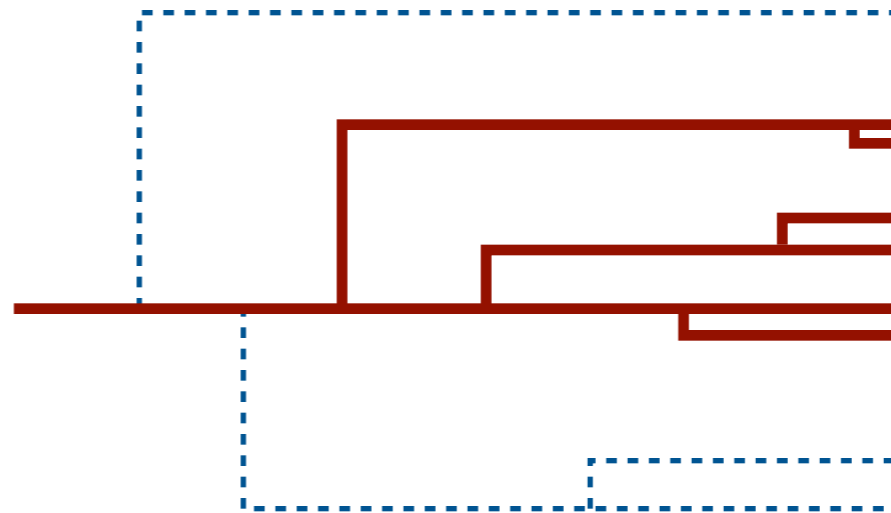


Groomed Jet



=

Groomed
Clustering Tree

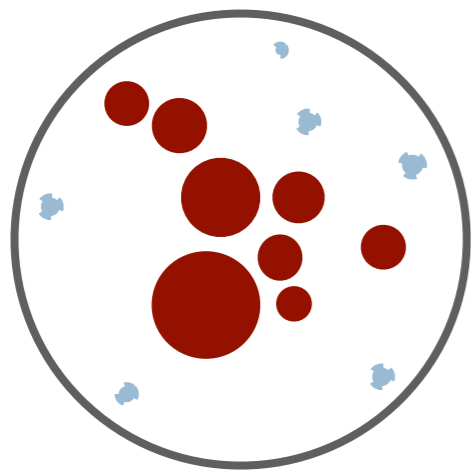


[Larkoski, Marzani, Soyez, JDT, 2014; see also Butterworth, Davison, Rubin, Salam, 2008; Dasgupta, Fregoso, Marzani, Salam/Powling, 2013]

Soft Drop Declustering

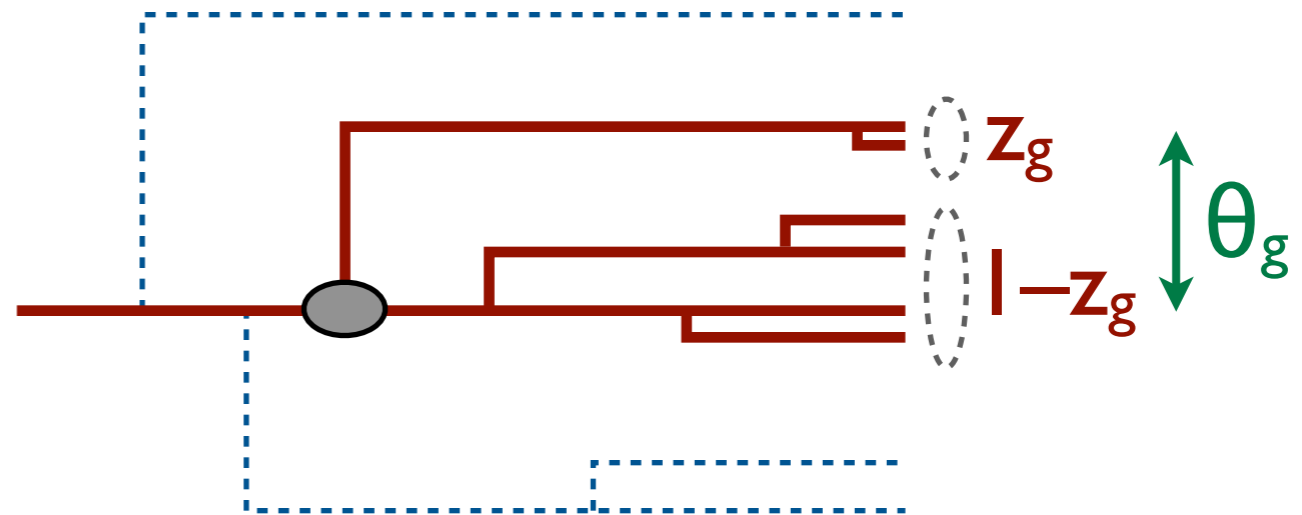


Groomed Jet



=

Groomed Clustering Tree



$$z_g > z_{\text{cut}} \theta_g^\beta$$

More Grooming

Less Grooming

$\beta \rightarrow -\infty$

$\beta < 0$

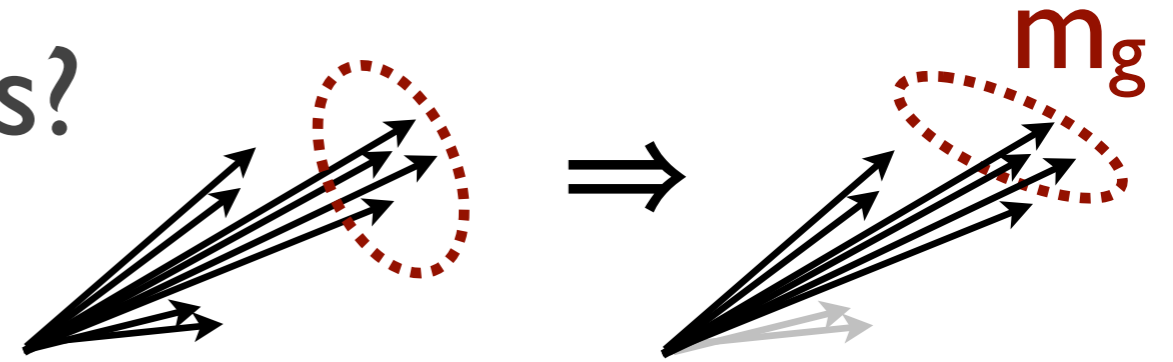
$\beta = 0$

$\beta > 0$

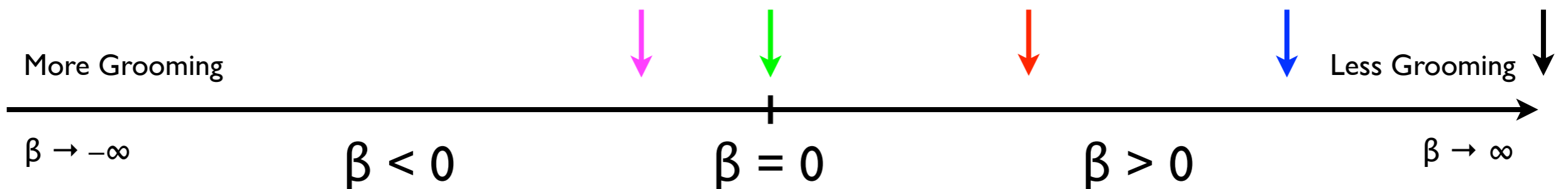
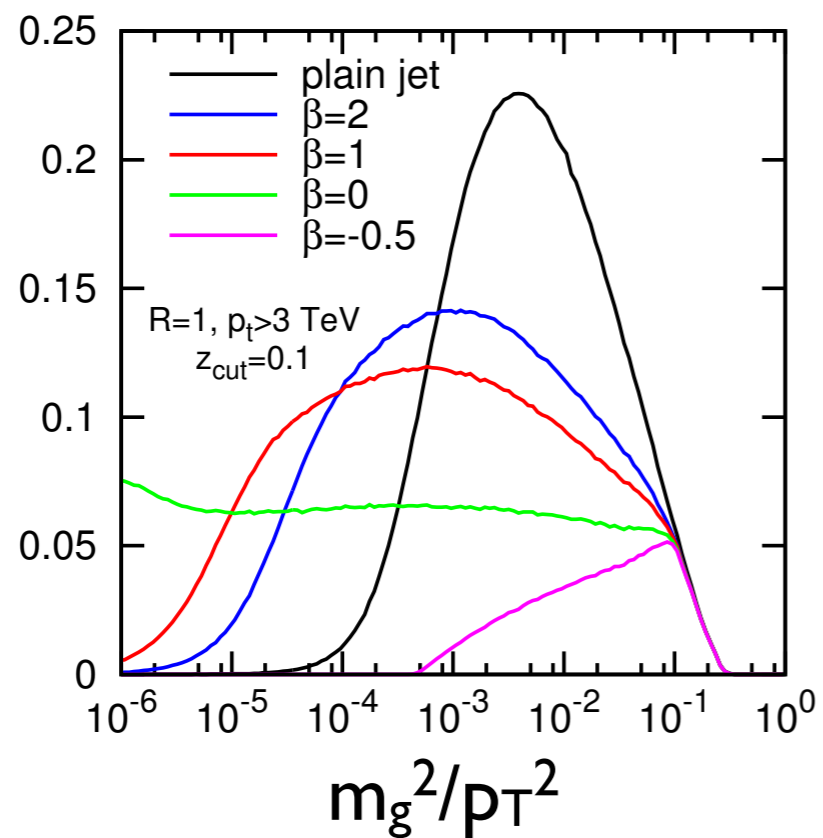
$\beta \rightarrow \infty$

[Larkoski, Marzani, Soyez, JDT, 2014; see also Butterworth, Davison, Rubin, Salam, 2008; Dasgupta, Fregoso, Marzani, Salam/Powling, 2013]

Calculating Groomed Mass?

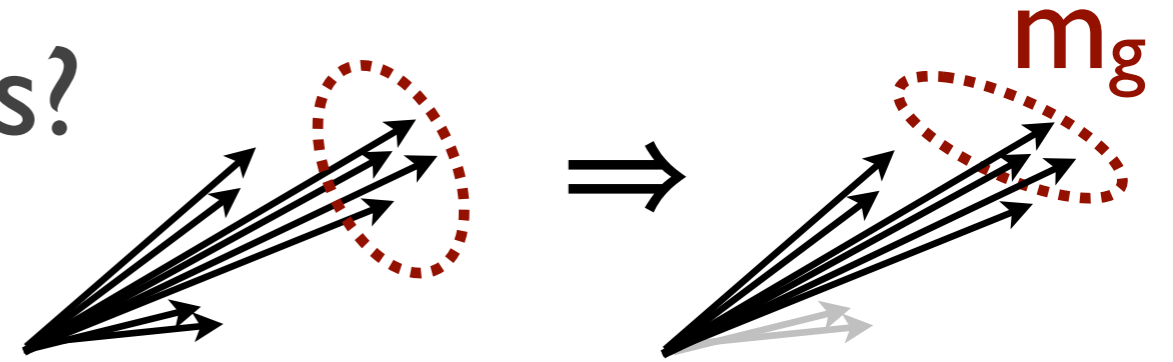


Simulated LHC Data

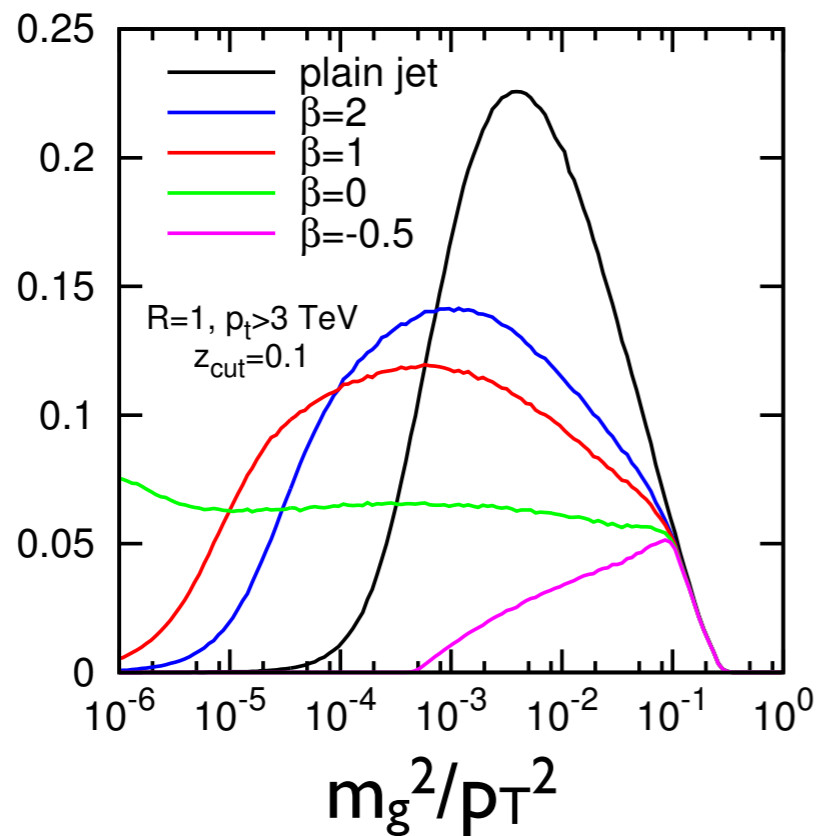


[Larkoski, Marzani, Soyez, JDT, 2014]

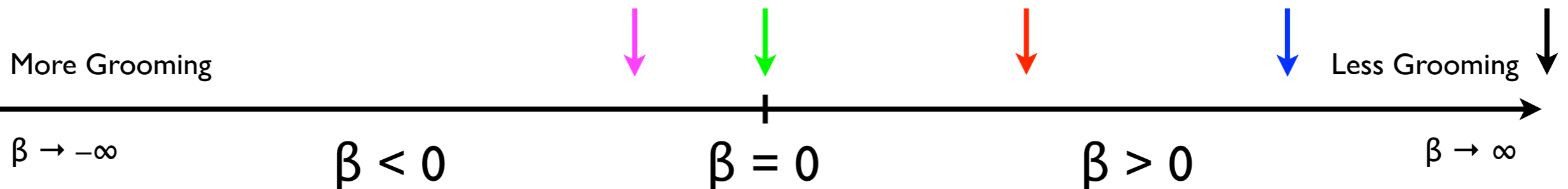
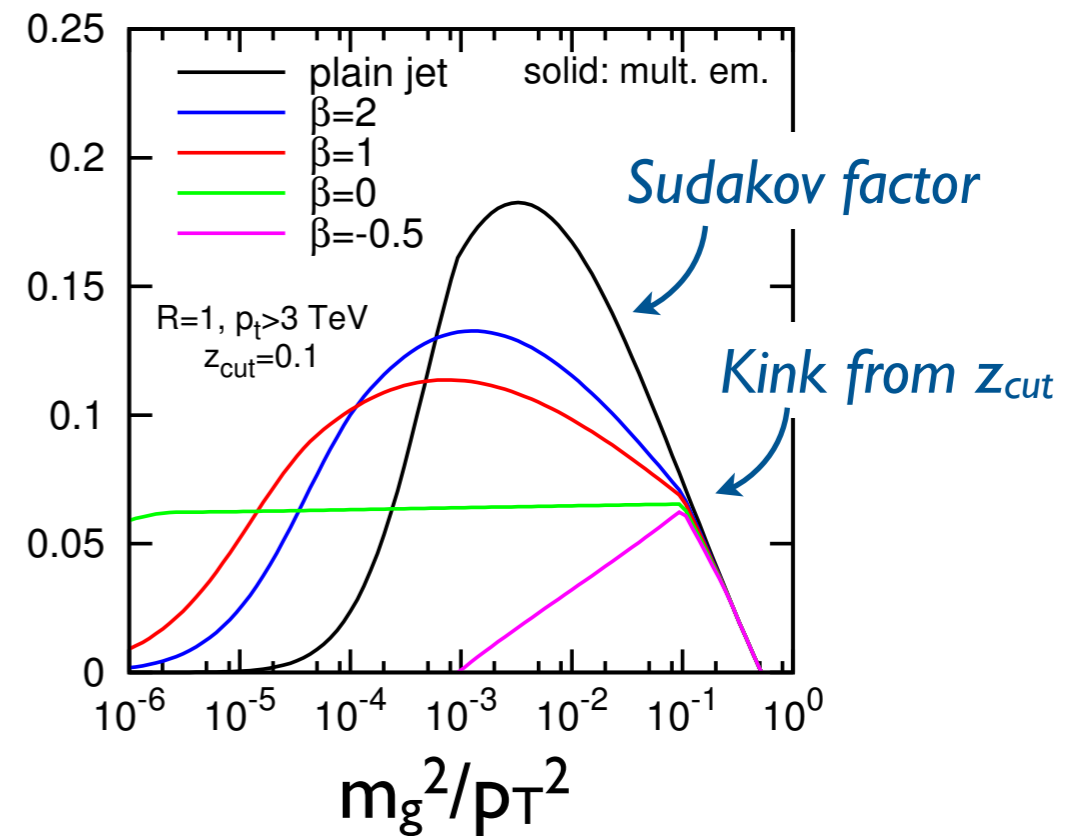
Calculating Groomed Mass?



Simulated LHC Data



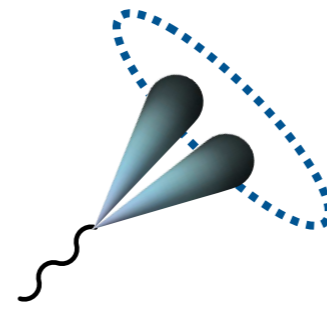
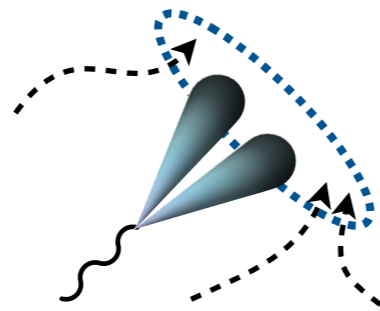
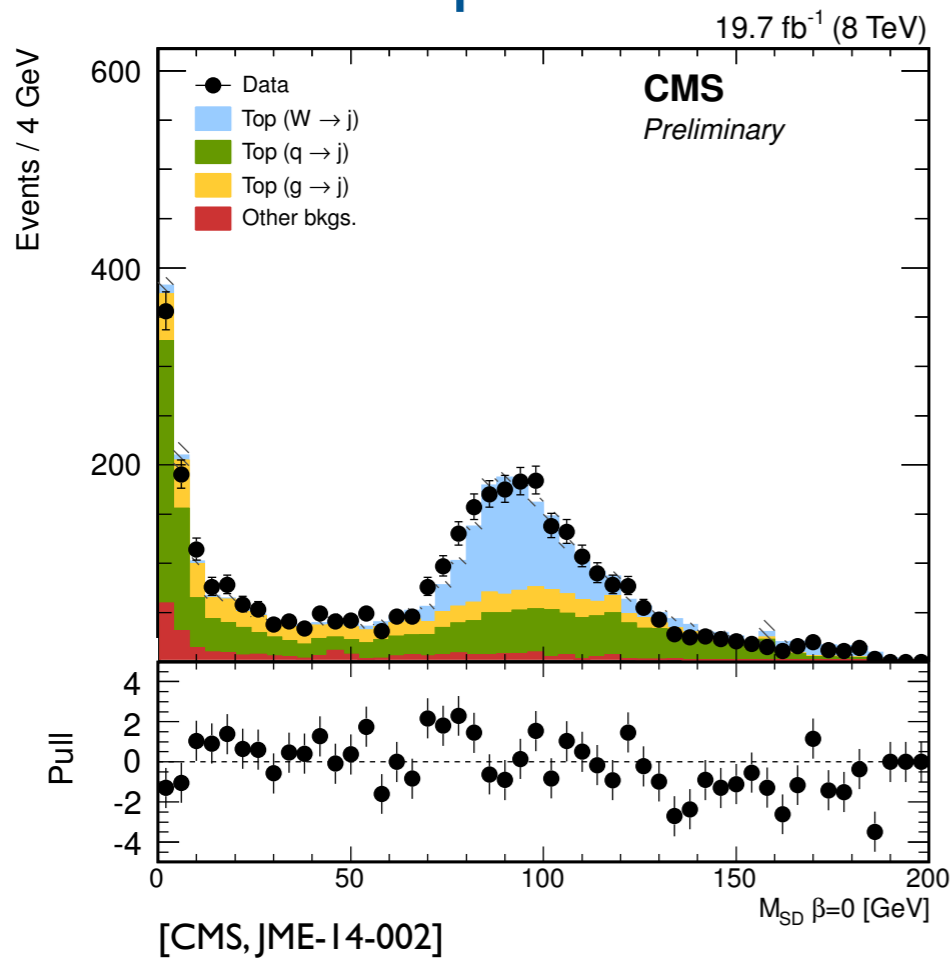
First-principles QCD (MLL)



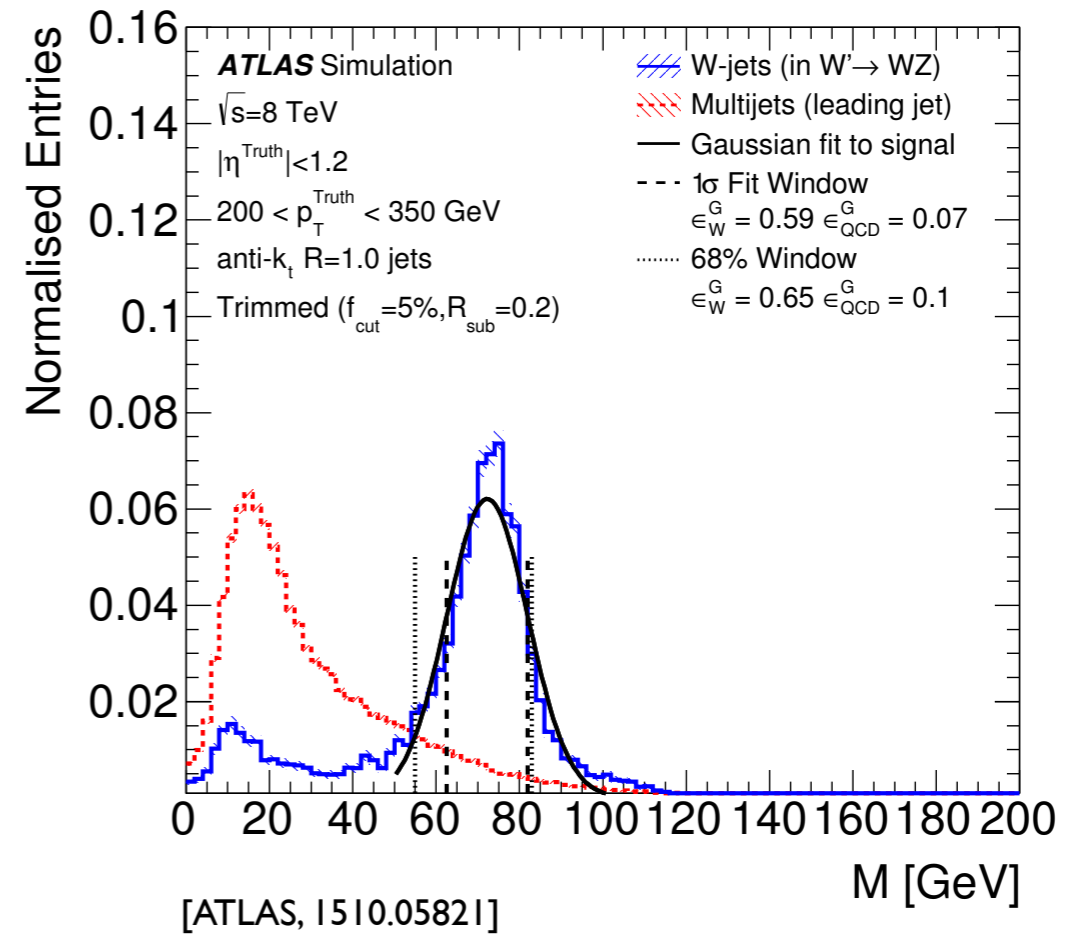
[Larkoski, Marzani, Soyer, JDT, 2014]



Soft Drop



Trimming



Both grooming strategies give good signal isolation

Scorecard

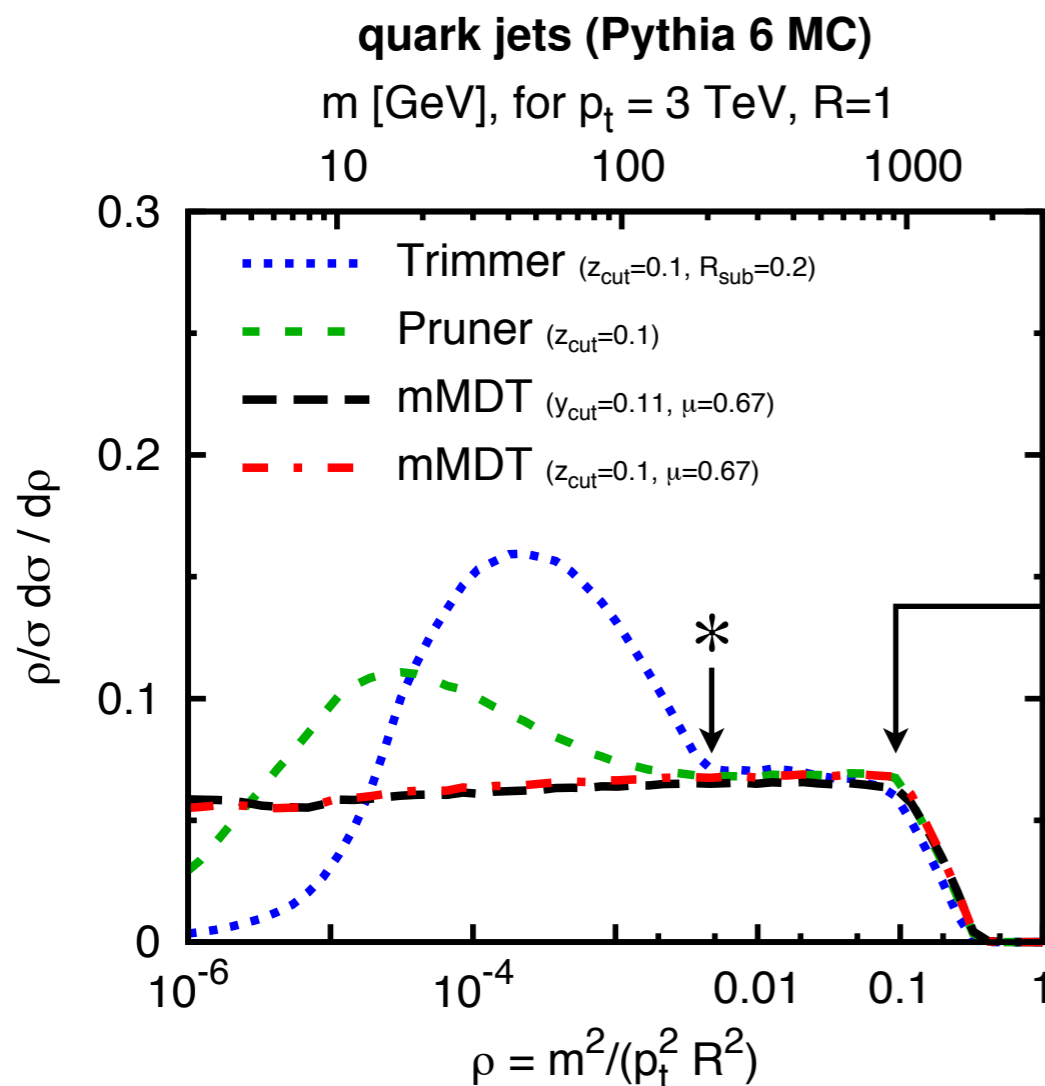


Soft Drop



Soft drop with $\beta = 0$
= mMDT with $\mu = 1$

Trimming



[Dasgupta, Fregoso, Marzani, Salam, 1307.0007]

Both have primary kink at

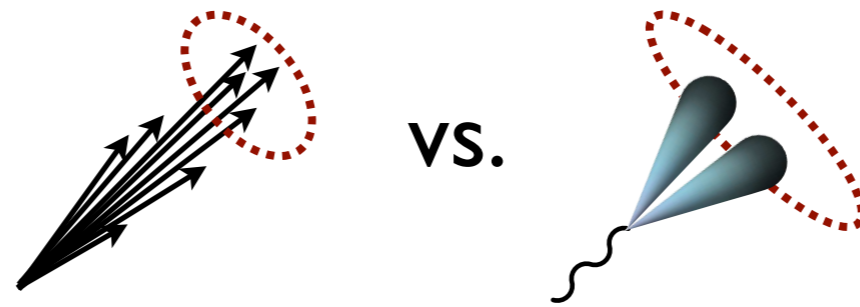
$$m \simeq p_T R \sqrt{z_{\text{cut}}}$$

* Trimming has secondary kink at

$$m \simeq p_T R_{\text{sub}} \sqrt{z_{\text{cut}}}$$

(Pruning also has similar behavior)

Discrimination from First Principles



(aka Axes or Axes-Free?)



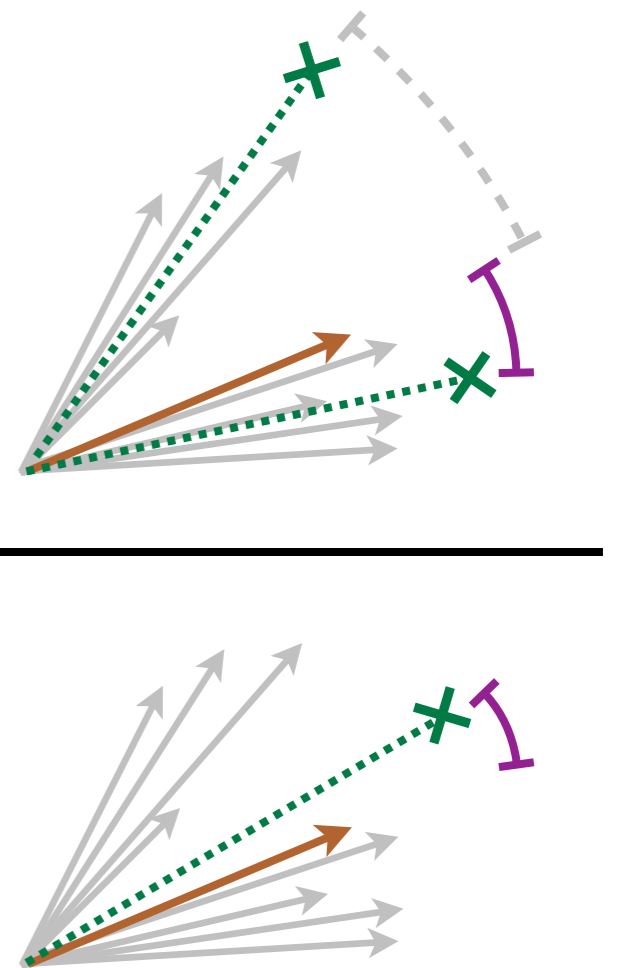
Discrimination with Axes

Dimensionless kinematics:

$$z_i \equiv \frac{p_{Ti}}{\sum_j p_{Tj}} \quad \theta_{ij} \equiv \Delta R_{ij}$$

N-subjettiness

$$\frac{\tau_2}{\tau_1} = \frac{\sum_k z_k \min\{\theta_{k1}, \theta_{k2}\}^\beta}{\sum_k z_k \theta_{k1}^\beta} =$$



Requires a definition of subjet axes
(i.e. minimizing over all axes possibilities)

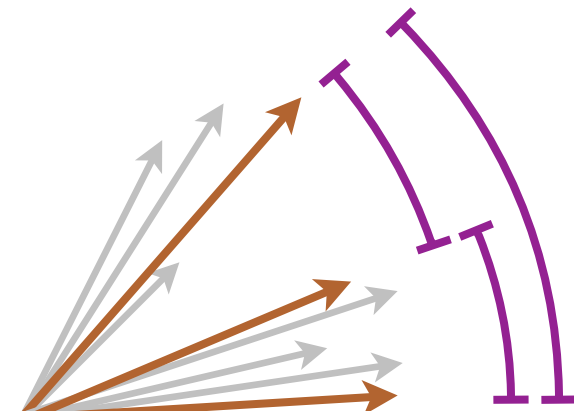
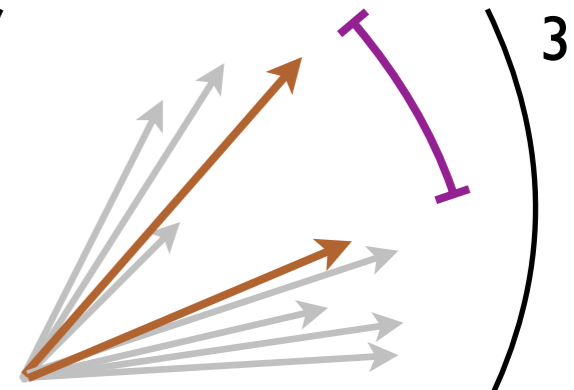
[JDT, Van Tilburg, 1011.2268, 1108.2701; see also Kim, 1011.1493]
[related work in Stewart, Tackmann, Waalewijn, 1004.2489]

Discrimination without Axes

Dimensionless kinematics:

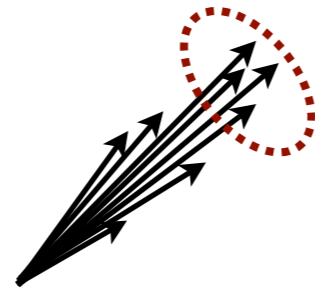
$$z_i \equiv \frac{p_{Ti}}{\sum_j p_{Tj}} \quad \theta_{ij} \equiv \Delta R_{ij}$$

Energy Correlation Functions

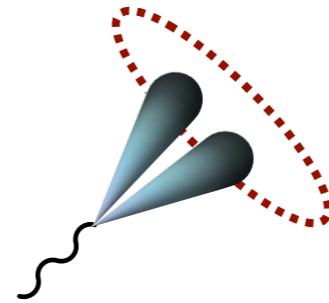
$$D_2 = \frac{e_3}{(e_2)^3} = \frac{\sum_{i < j < k} z_i z_j z_k (\theta_{ij} \theta_{jk} \theta_{ki})^\beta}{\left(\sum_{i < j} z_i z_j \theta_{ij}^\beta \right)^3} = \frac{\text{Diagram 1}}{\text{Diagram 2}^3}$$



Axes-free probe for substructure;
not immediately obvious how it works

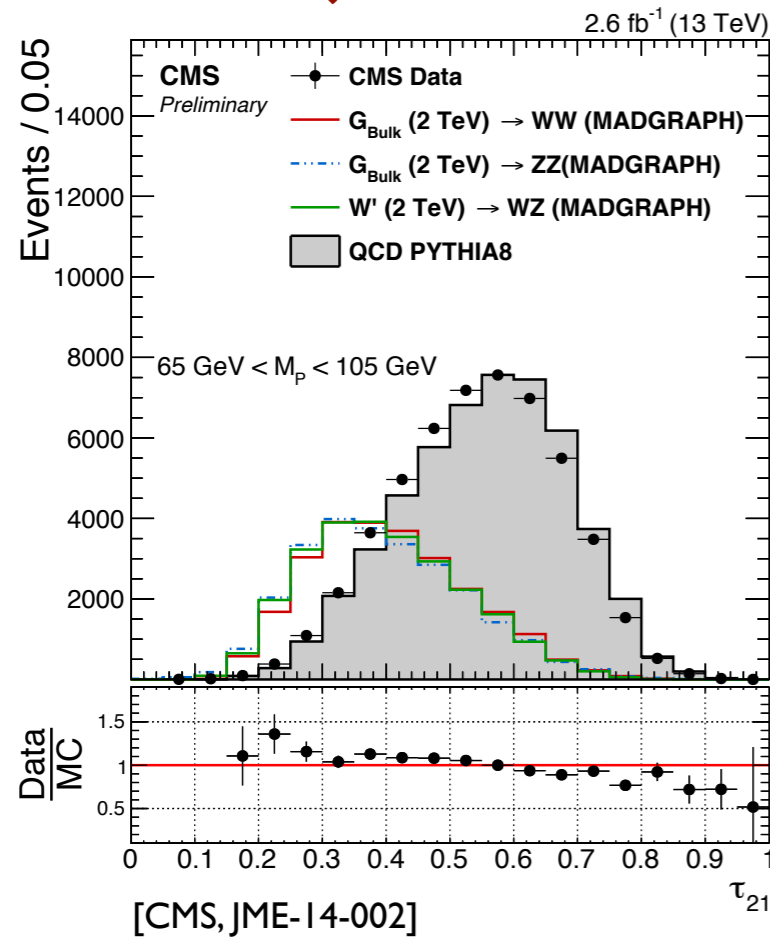
[Larkoski, Salam, JDT, 1305.0007; Larkoski, Moutl, Neill, 1409.6298, 1507.03018;
see also Banfi, Salam, Zanderighi, hep-ph/0407286; Jankowiak, Larkoski, 1104.1646]



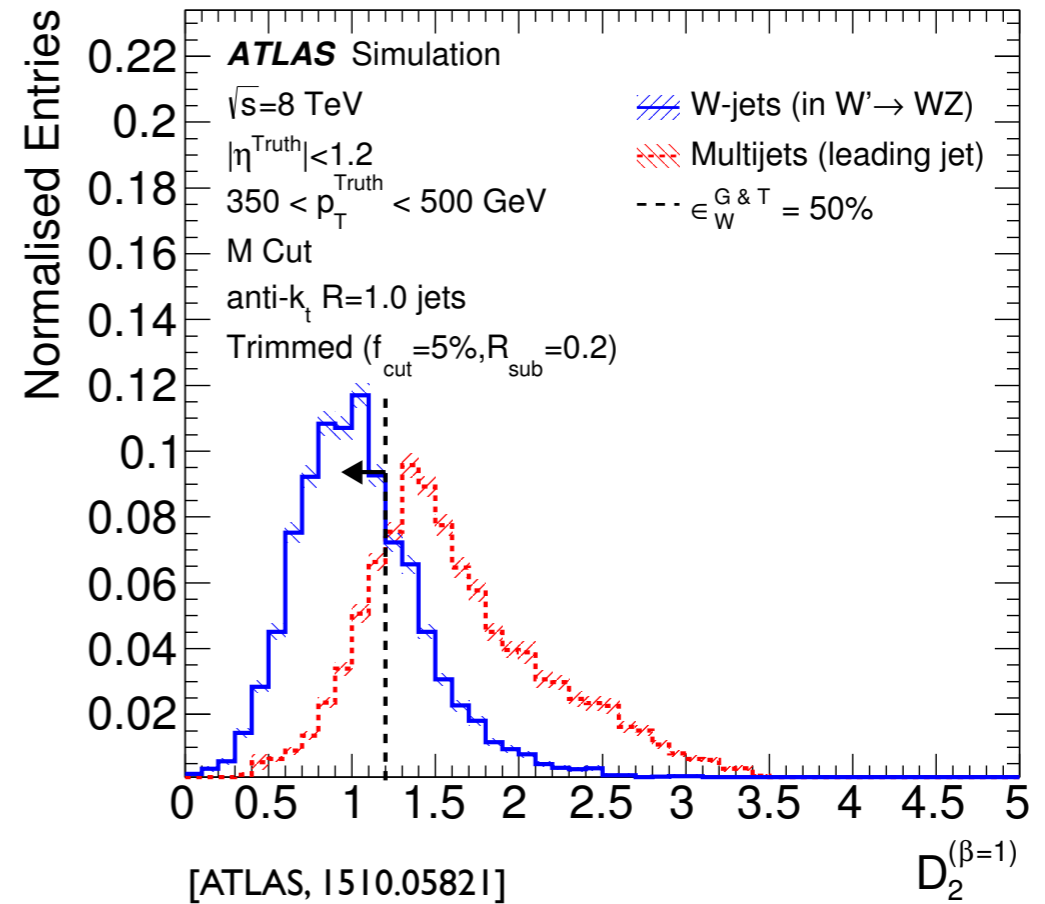
vs.



N-subjettiness

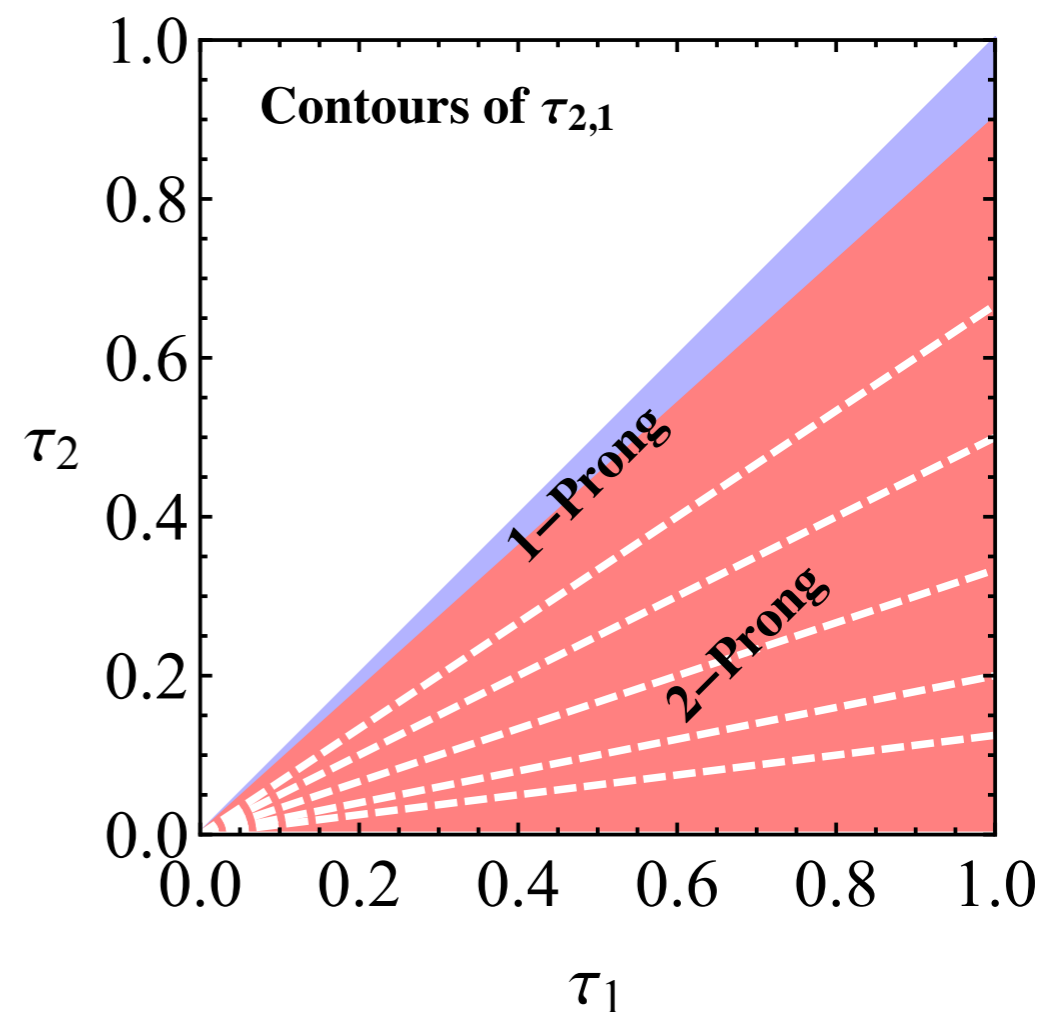
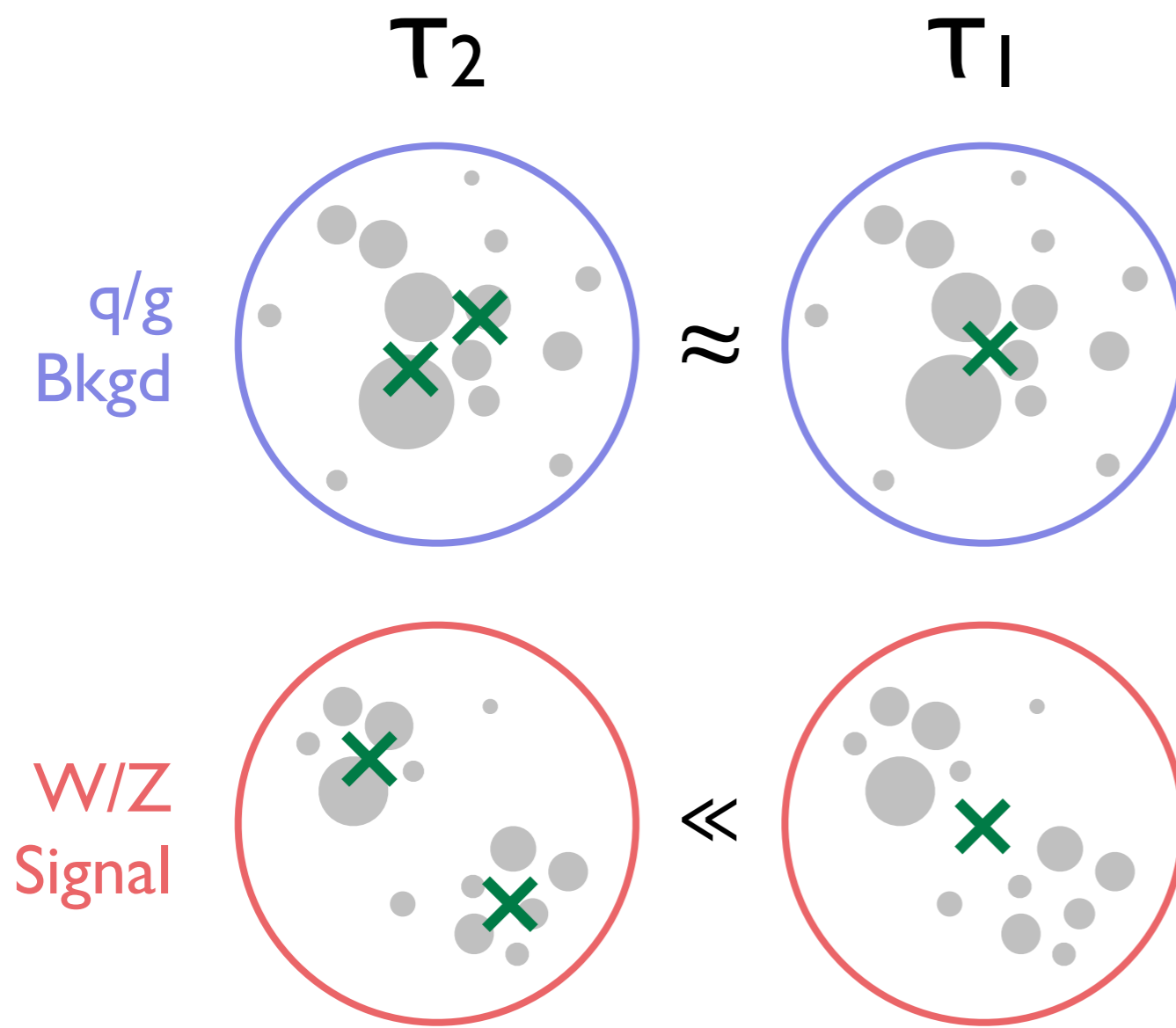


Energy Correlator: D₂



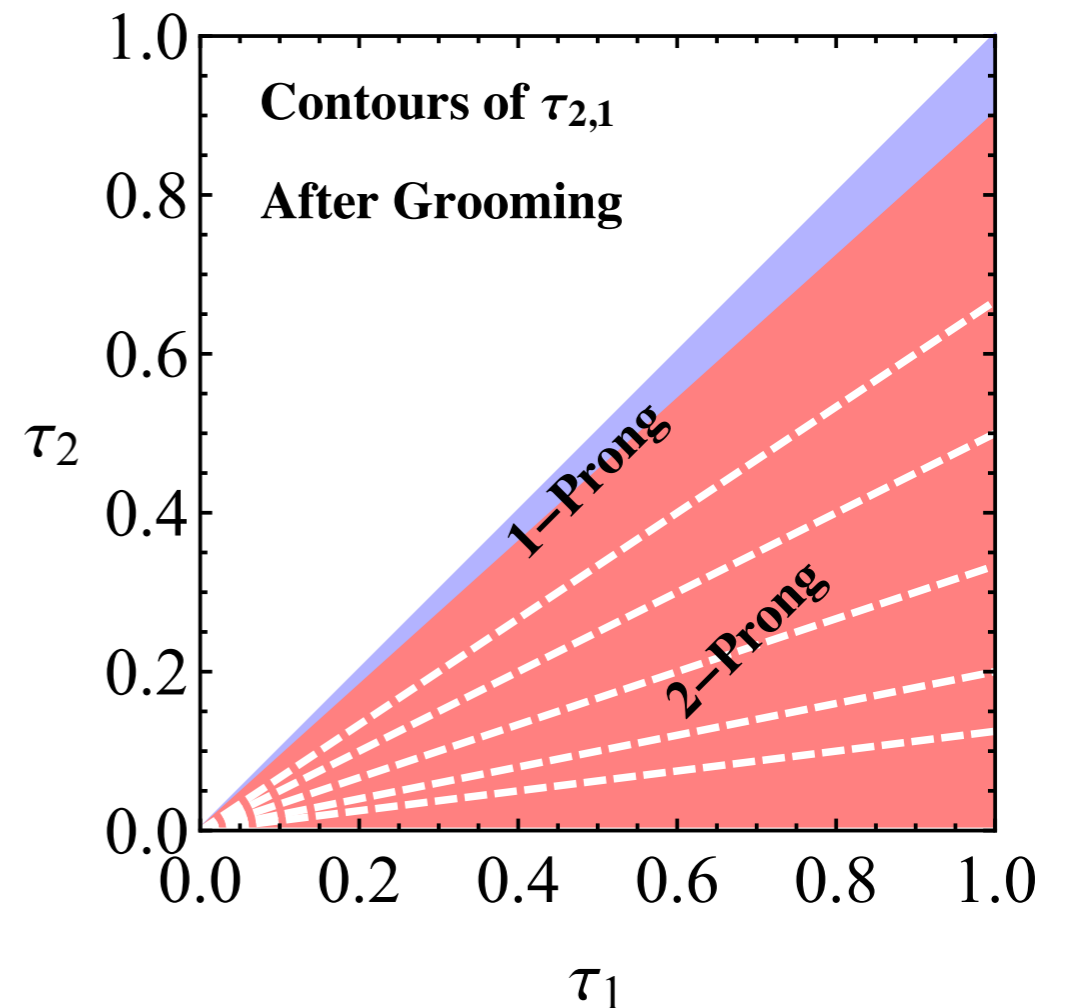
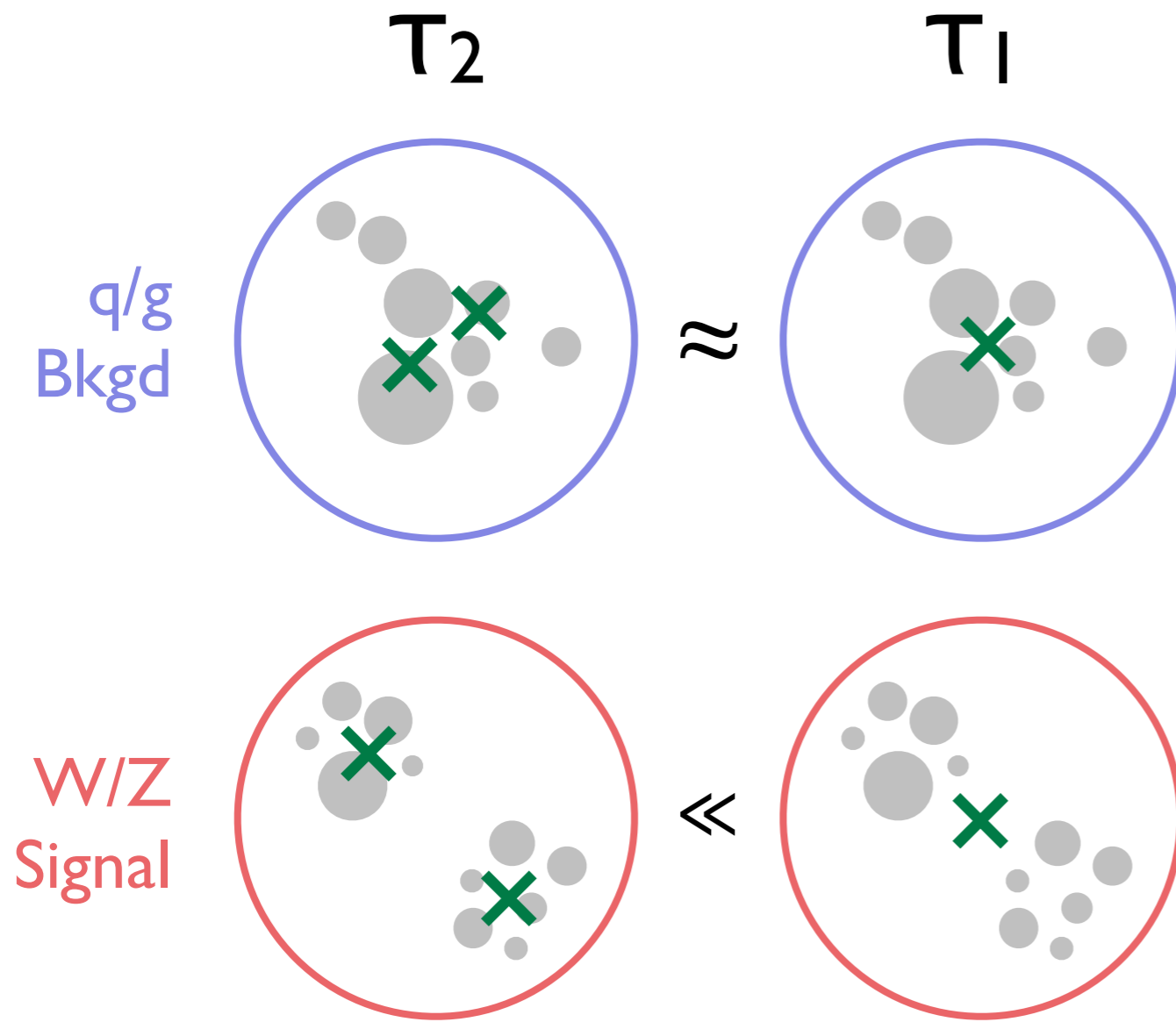
Can we understand this behavior from first principles?

N-subjettiness Behavior



Parametric separation between 1 and 2 prong jets

N-subjettiness Behavior with Grooming



Parametric separation between 1 and 2 prong jets even after removing soft peripheral radiation

Scorecard



Soft Drop



Trimming



N-subjettiness

before grooming



(also works
after grooming)

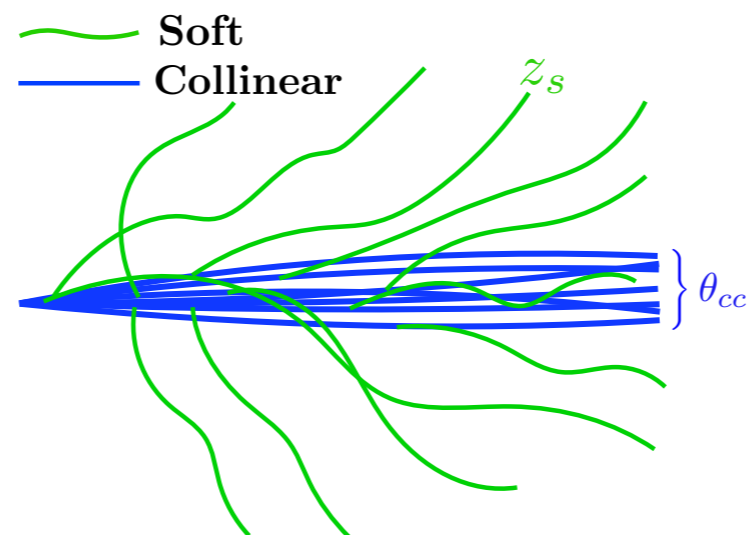
D₂

after grooming

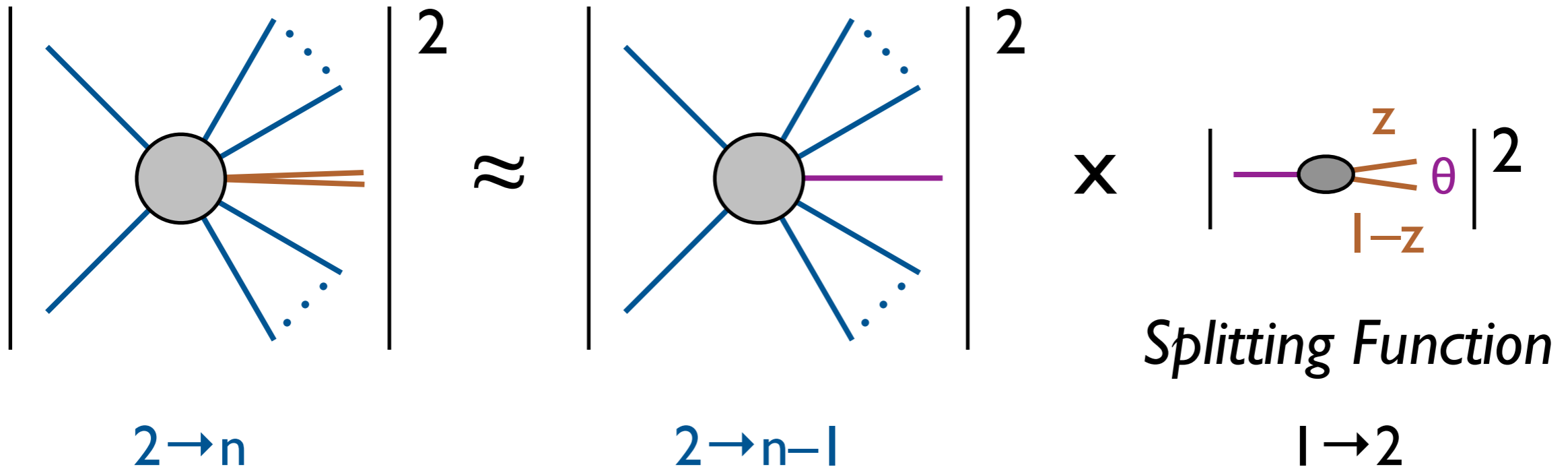


Not just good discrimination,
but stability to choice of cuts

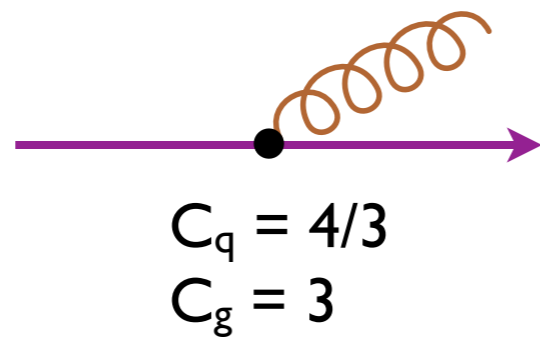
The Power of Power Counting



Textbook QCD



At leading log order:

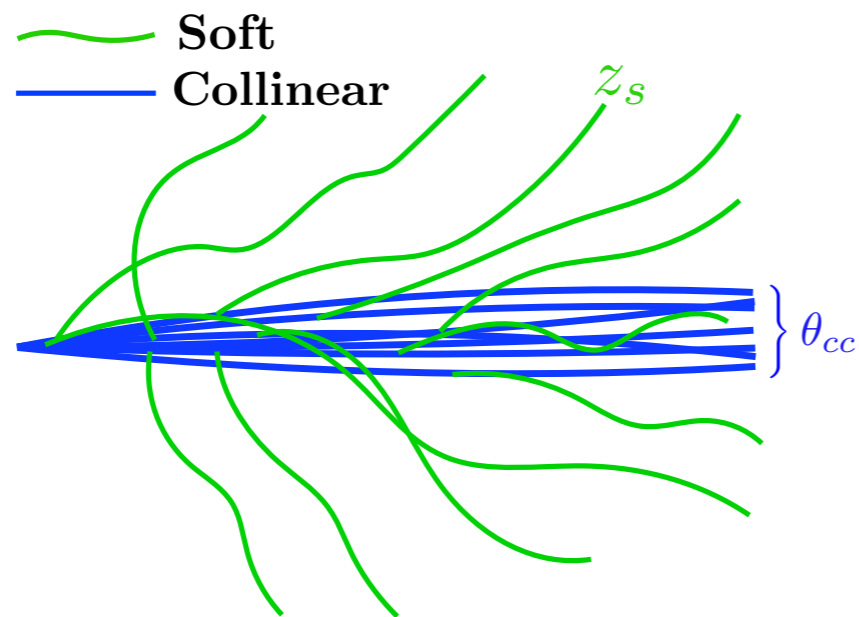


See end of talk for quark/gluon tagging

$$dP_{i \rightarrow ig} \simeq \frac{2\alpha_s}{\pi} C_i \underbrace{\frac{d\theta}{\theta}}_{\text{Collinear singularity}} \underbrace{\frac{dz}{z}}_{\text{Soft singularity}}$$

Power Counting Modes

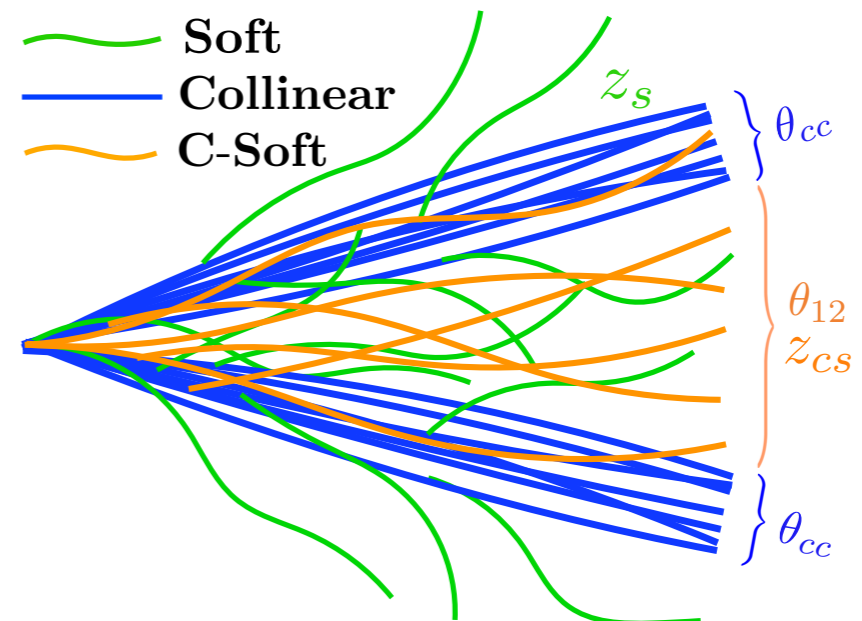
1-prong background
(quark or gluon)



	C	S
energies	1	z_s

	CC	SX
angles	θ_{cc}	1

2-prong signal
(W or Z)



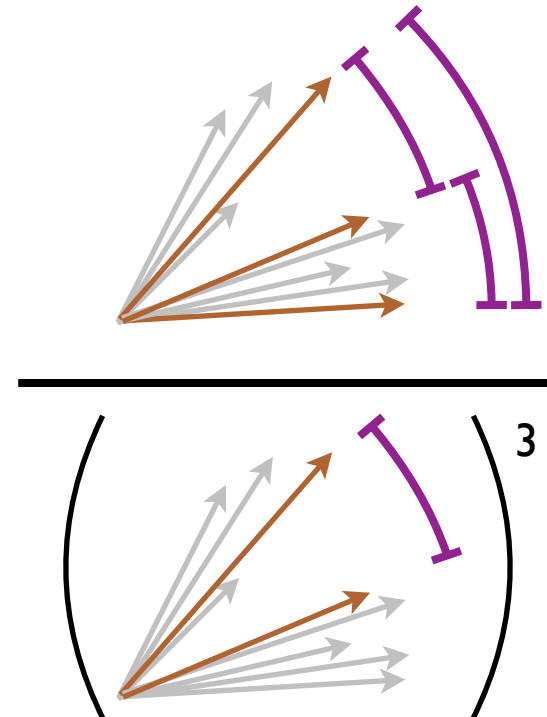
	C	C_s	S
energies	1	z_{cs}	z_s

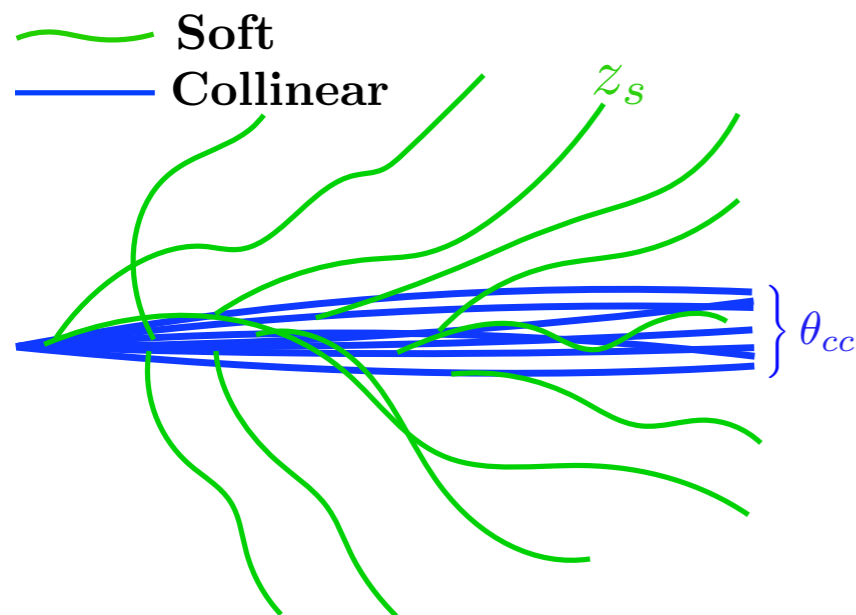
	CC	C₁C₂ or CC_s	SX
angles	θ_{cc}	θ_{12}	1

Power Counting D_2 : Background

$$D_2 = \frac{e_3}{(e_2)^3} = \frac{\sum_{i < j < k} z_i z_j z_k (\theta_{ij} \theta_{jk} \theta_{ki})^\beta}{\left(\sum_{i < j} z_i z_j \theta_{ij}^\beta \right)^3} = \frac{\text{Diagram 1}}{\left(\text{Diagram 2} \right)^3}$$

Setting $\beta = 1$



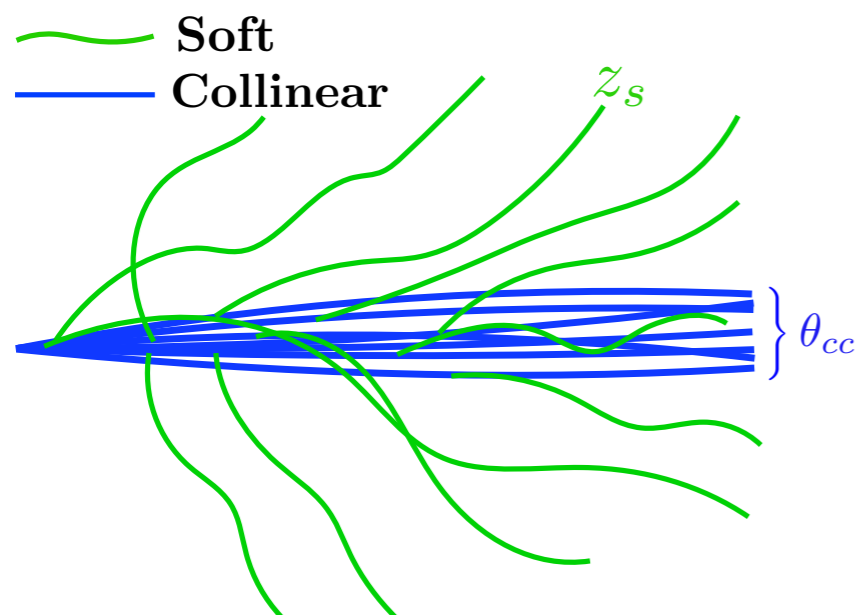


$$\sim \frac{\begin{matrix} \text{CCC} & \text{CCS} & \text{CSS} \\ \theta_{cc}^3 + \theta_{cc} z_s + z_s^2 \\ \text{CC} & \text{CS} \end{matrix}}{\left(\theta_{cc} + z_s \right)^3}$$

Power Counting D_2 : Background

$$D_2 = \frac{e_3}{(e_2)^3} = \frac{\sum_{i < j < k} z_i z_j z_k (\theta_{ij} \theta_{jk} \theta_{ki})^\beta}{\left(\sum_{i < j} z_i z_j \theta_{ij}^\beta \right)^3} = \frac{\text{Diagram 1}}{\left(\text{Diagram 2} \right)^3}$$

Setting $\beta = 1$



$$\sim \frac{\begin{matrix} \text{CCC} & \text{CCS} & \text{CSS} \\ \theta_{cc}^3 + \theta_{cc} z_s + z_s^2 \end{matrix}}{\left(\theta_{cc} + z_s \right)^3} \Rightarrow \frac{1}{z_s} \sim \frac{p_T}{m}$$

$\text{CC} \quad \text{CS}$

if soft dominates

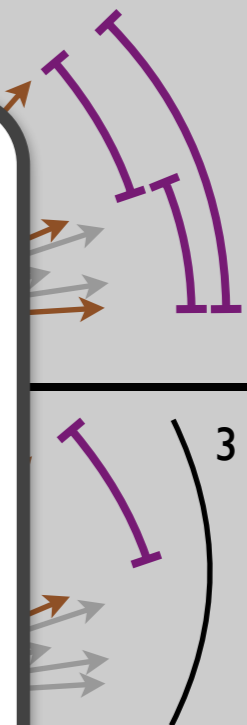
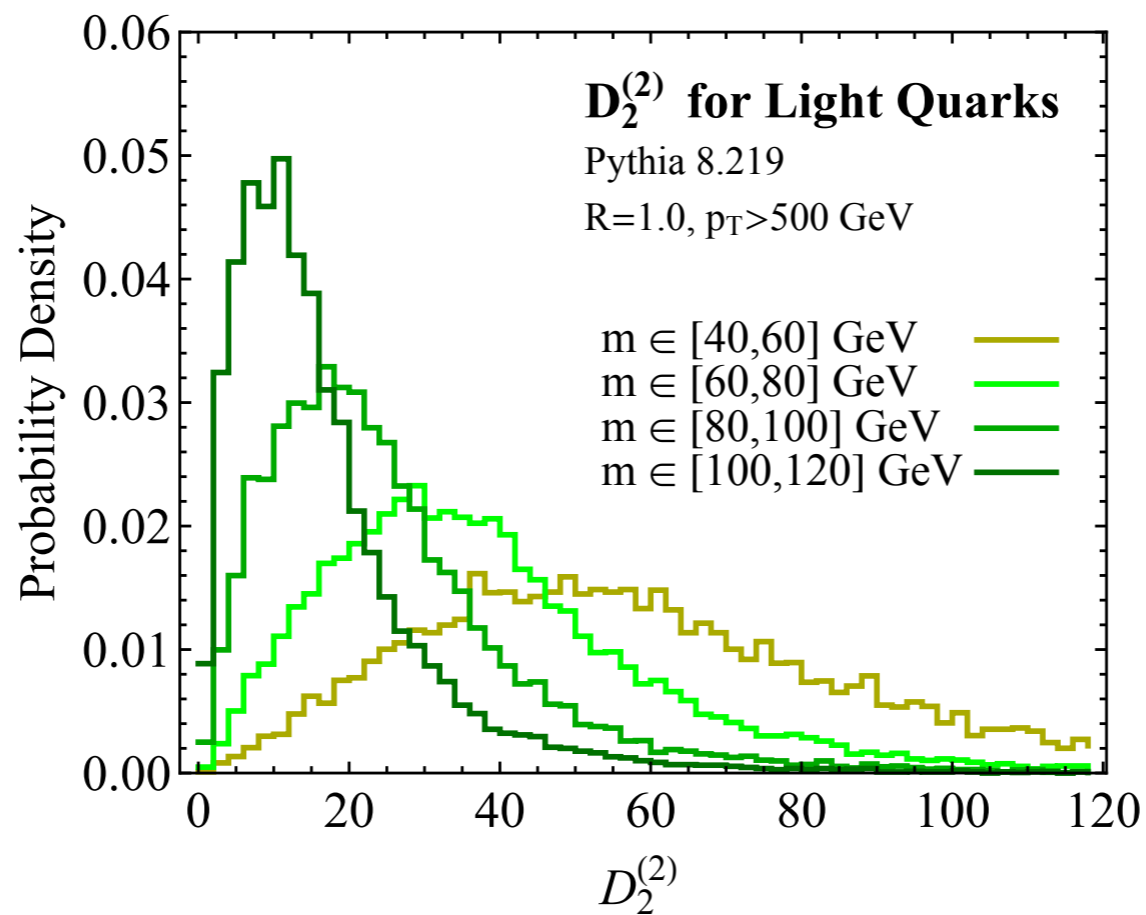
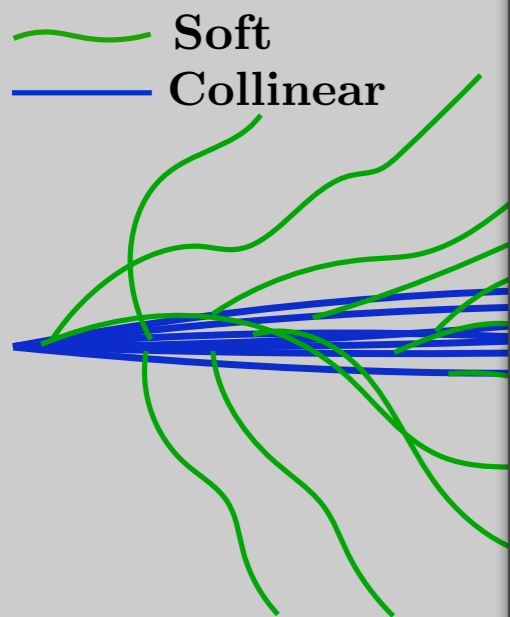
Power Counting D_2 : Background

Without grooming,
 D_2 unstable to mass cut
 (hard to do sideband analysis)

$D_2 =$

Setting f

Soft
 Collinear

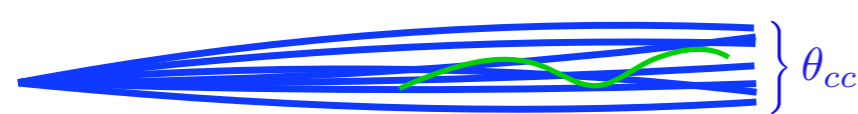


$$\frac{1}{z_s} \approx \frac{p_T}{m}$$

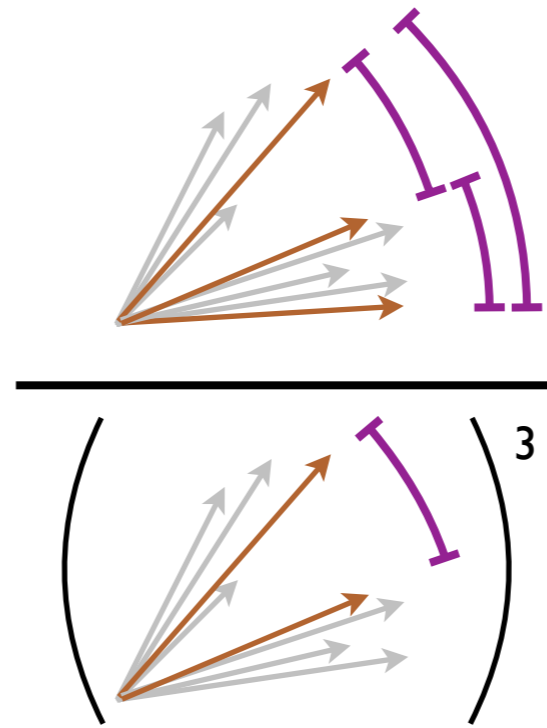
if soft
 dominates

Background after Grooming

— Collinear

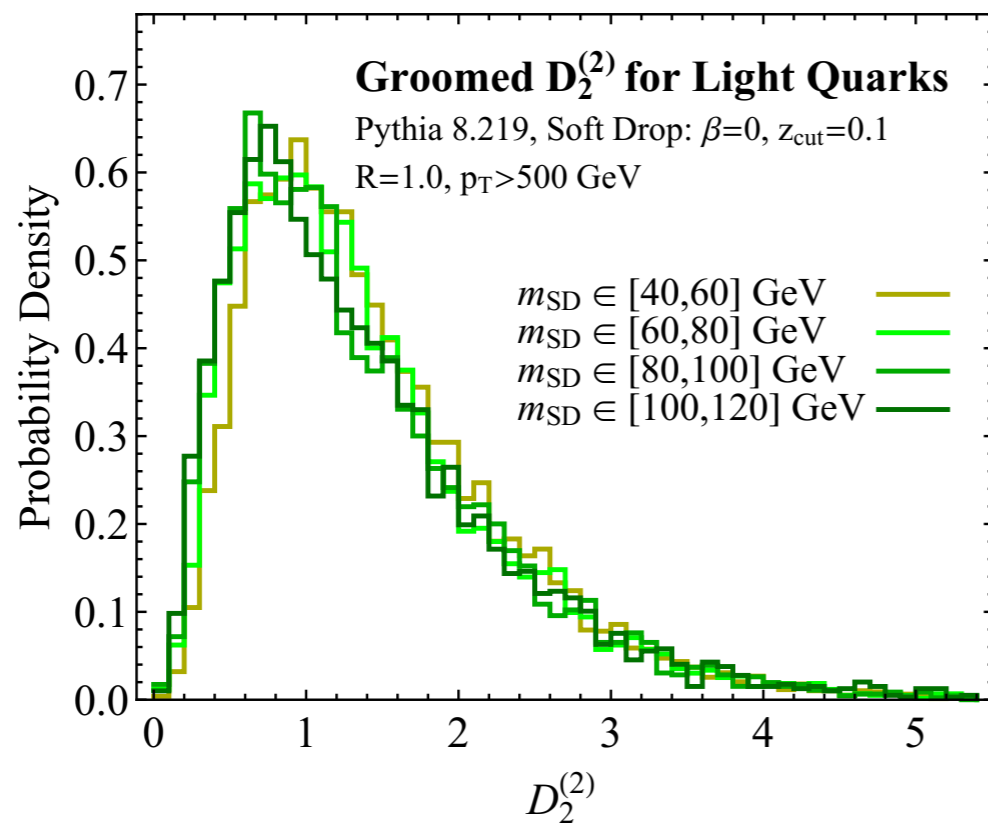


$$D_2 =$$



$$\approx \frac{\text{CCC}}{\left(\theta_{cc}\right)^3} \Rightarrow 1$$

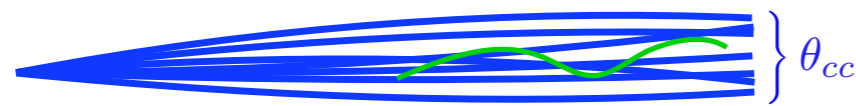
CC



After grooming,
 D_2 background distribution
 is **parametrically stable**

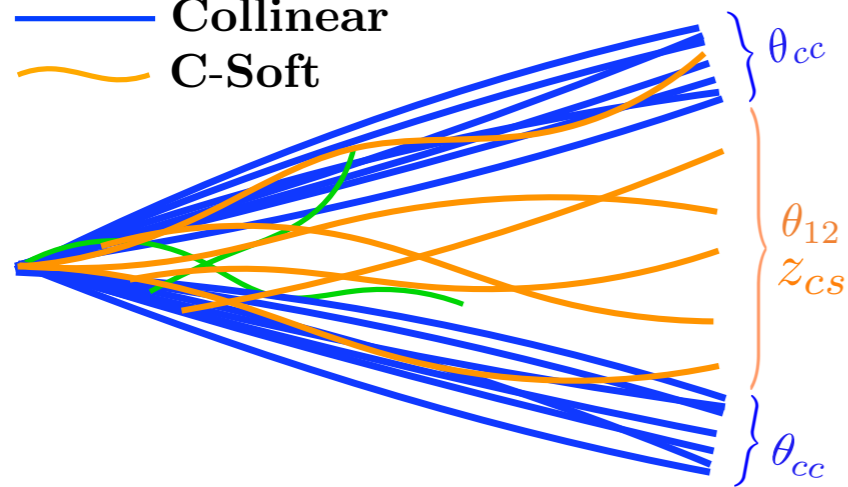
Signal/Background Separation

— Collinear



$$D_2 = \frac{\left(\begin{array}{c} \text{CCC} \\ \theta_{cc}^3 \end{array} \right)}{\left(\begin{array}{c} \text{CC} \\ \theta_{cc} \end{array} \right)^3} \approx \frac{\text{CCC}}{(\theta_{cc})^3} \Rightarrow 1$$

— Collinear
— C-Soft



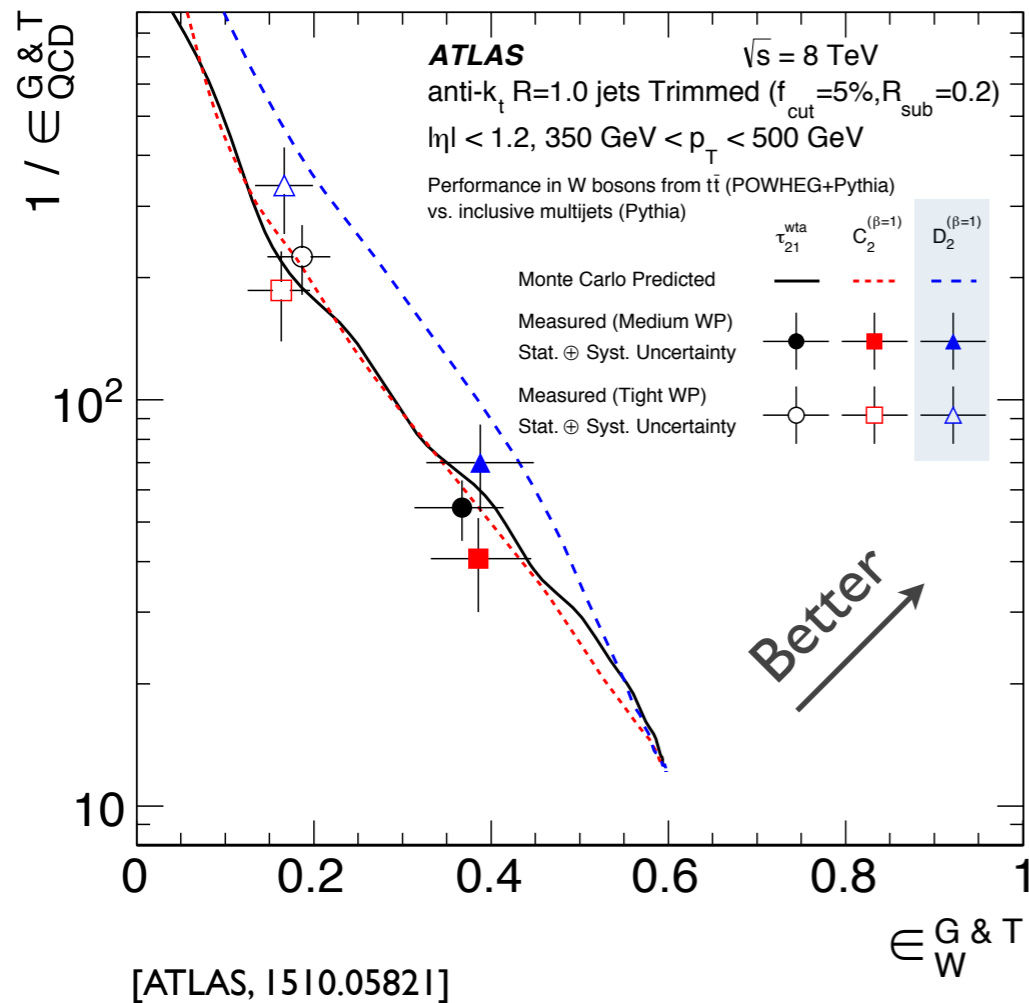
$$D_2 \approx \frac{\begin{array}{c} C_1 C_2 C_s \\ \theta_{12}^3 z_{cs} + \theta_{12}^2 \theta_{cc} \end{array}}{\left(\begin{array}{c} C_1 C_2 \\ \theta_{12} \end{array} \right)^3} \ll 1$$

Signal parametrically distinct

Scorecard



ATLAS “R2D2” Tagger



Soft Drop



Trimming



N-subjettiness

before grooming



(also works after grooming)

D₂

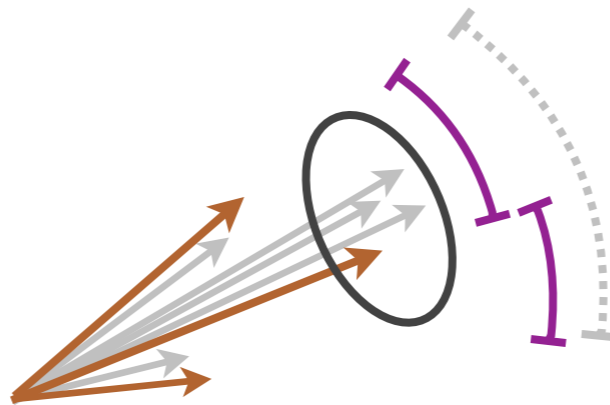
after grooming



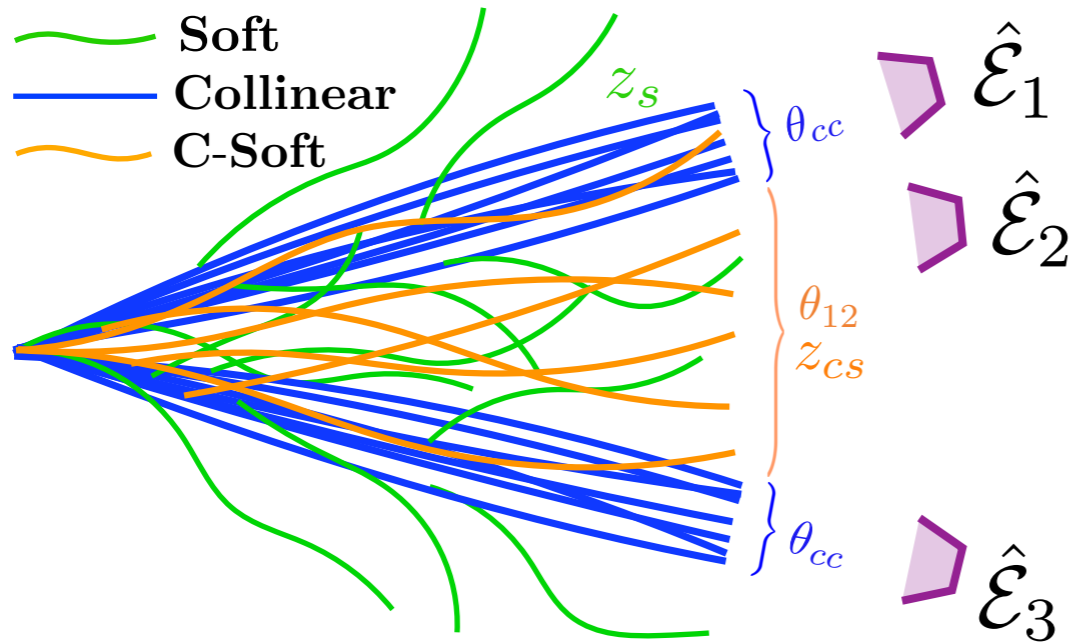
*unstable before grooming

Theoretically sound W/Z tagging strategies for the LHC

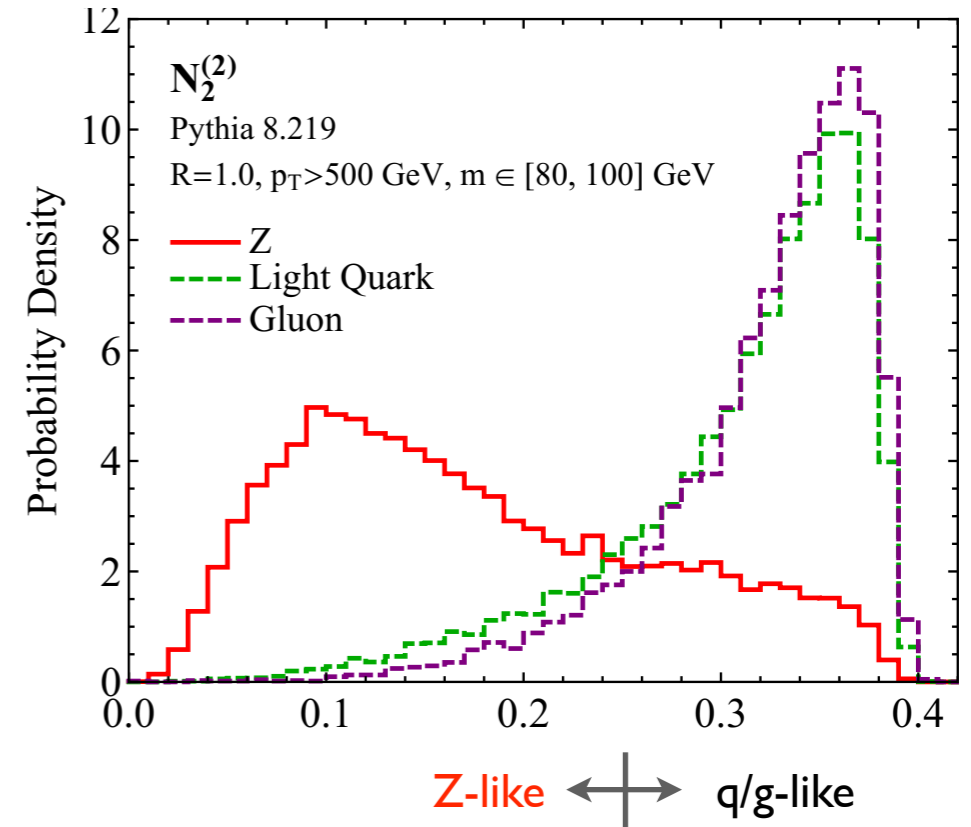
A New Angle on Jet Substructure



Punchline: W/Z Tagging in 2017?



Stable & Performant

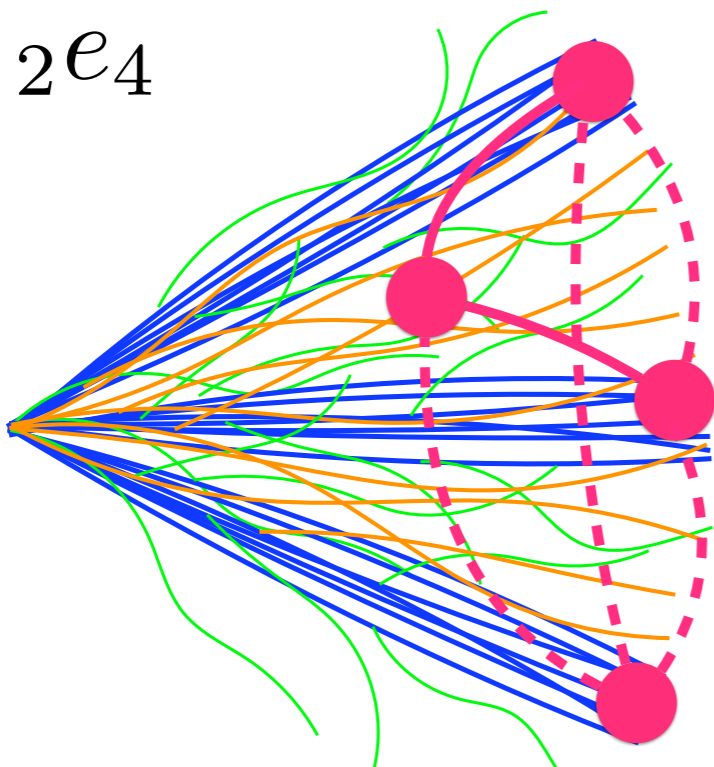


$$\boxed{N_2} = \frac{\sum_{i < j < k} p_{Ti} p_{Tj} p_{Tk} \min \left\{ (R_{ij} R_{jk})^2, (R_{jk} R_{ki})^2, (R_{ki} R_{ij})^2 \right\}}{\left(\sum_{i < j} p_{Ti} p_{Tj} R_{ij}^2 \right)^2 / \sum_i p_{Ti}}$$

[Moult, Necib, JDT, 2016]

A New Angle on Energy Correlators

$$v e_n^{(\beta)} = \sum_{\text{all } n\text{-tuples}} (n \text{ energies}) (v \text{ smallest angles})^\beta$$



Systematic jet dissection

$$n = 2, 3, 4, \dots$$

$$v = 1, 2, 3, \dots, n \text{ choose } 2$$

$$\beta = \dots, 0.5, 1, 2, \dots$$

collinear
dominated

soft
dominated

3-point Correlators

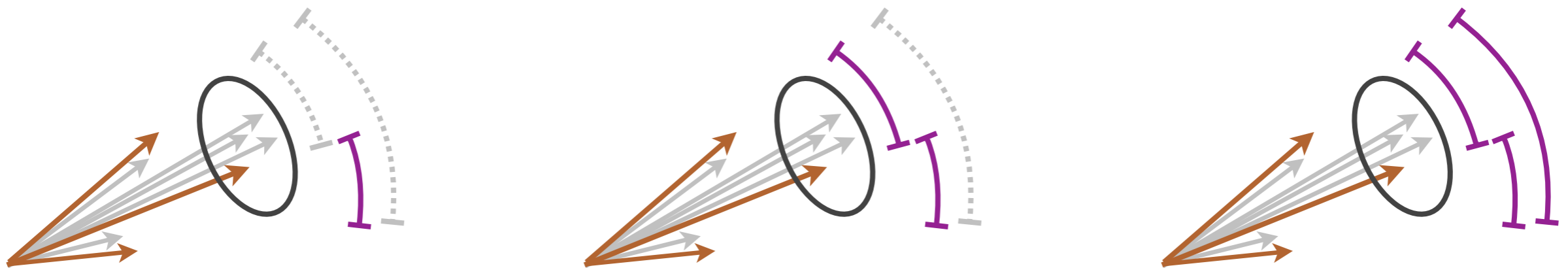
3-point:

$${}_1e_3^{(\beta)} = \sum_{i < j < k} z_i z_j z_k \min\{\theta_{ij}, \theta_{jk}, \theta_{ki}\}^\beta$$

$${}_2e_3^{(\beta)} = \sum_{i < j < k} z_i z_j z_k \min\{\theta_{ij}\theta_{jk}, \theta_{jk}\theta_{ki}, \theta_{ki}\theta_{ij}\}^\beta$$

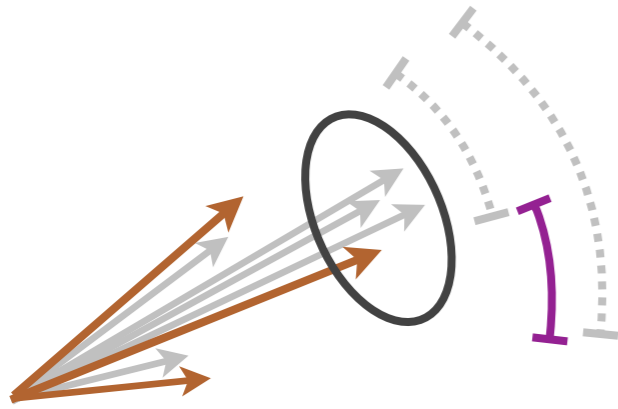
$${}_3e_3^{(\beta)} = \sum_{i < j < k} z_i z_j z_k (\theta_{ij}\theta_{jk}\theta_{ki})^\beta$$

used
for D_2

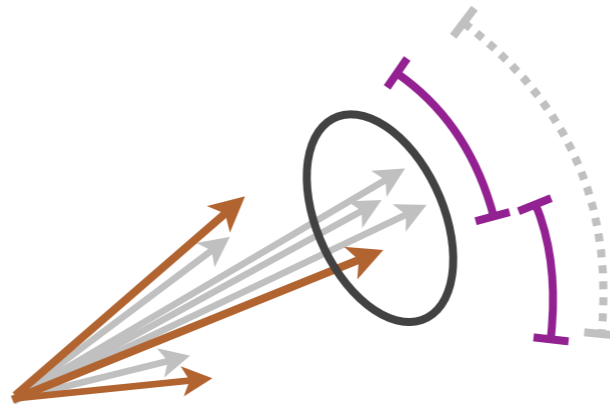


Probe of hierarchical jet substructure

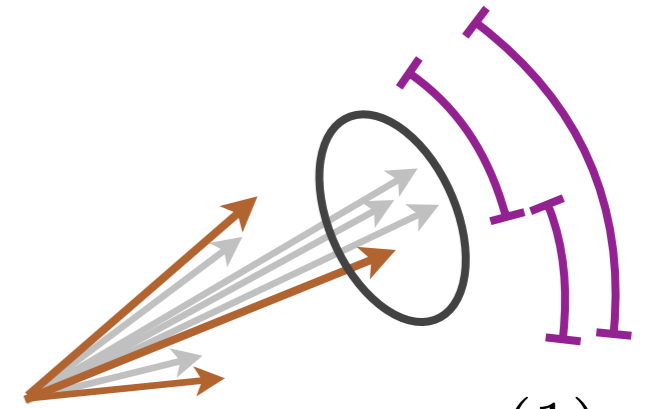
New Boosted W/Z Discriminants



$$M_2 = \frac{1e_3}{1e_2}$$



$$N_2 = \frac{2e_3}{(1e_2)^2}$$



$$D_2^{(1,2)} = \frac{3e_3^{(1)}}{(1e_2^{(2)})^{\frac{3}{2}}}$$

Rule of thumb: boost invariance along jet axis

$$z_i \rightarrow z_i \quad \theta_{ij} \rightarrow \gamma^{-1} \theta_{ij}$$

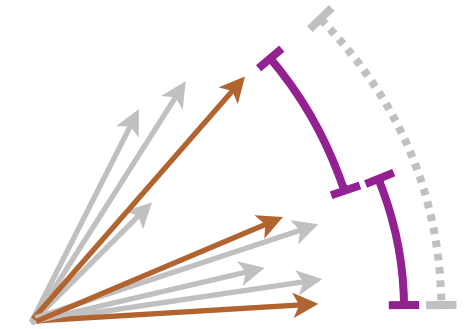
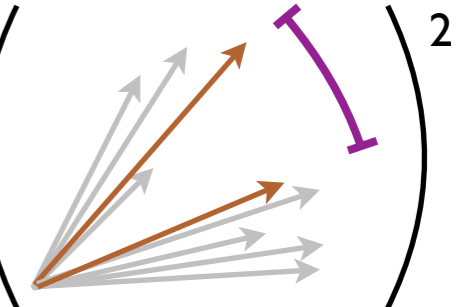
(i.e. same angular scaling in numerator and denominator)

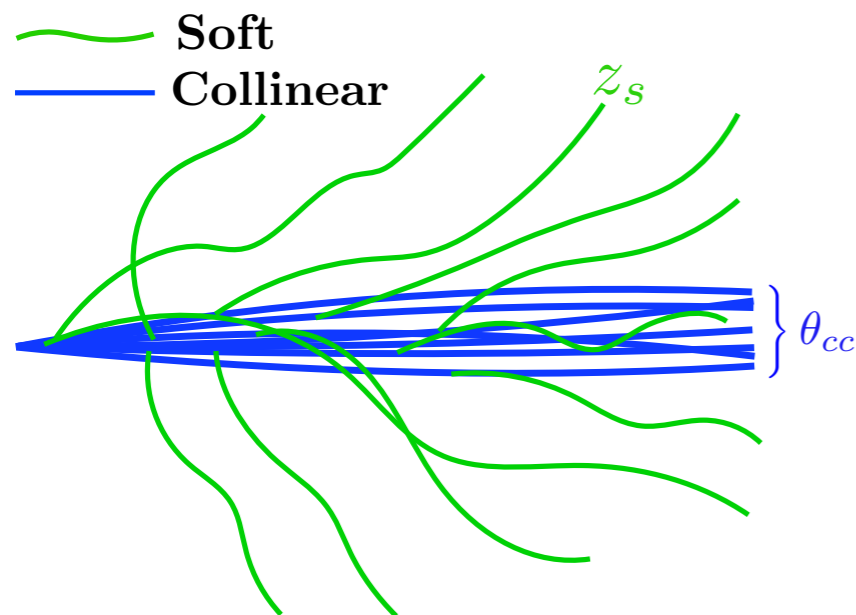
See also N_3 in the case of boosted top quarks

Power Counting N_2 : Background

$$N_2 = \frac{2e_3}{(e_2)^2} = \frac{\sum_{i < j < k} z_i z_j z_k \min\{\theta_{ij}\theta_{jk}, \theta_{jk}\theta_{ki}, \theta_{ki}\theta_{ij}\}^\beta}{\left(\sum_{i < j} z_i z_j \theta_{ij}^\beta\right)^2} = \frac{\text{Diagram 1}}{\left(\text{Diagram 2}\right)^2}$$

Setting $\beta = 1$

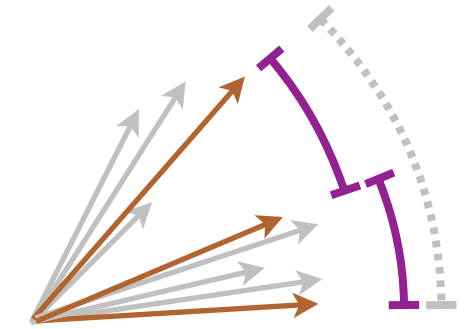
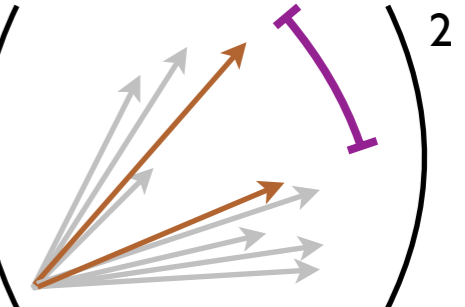


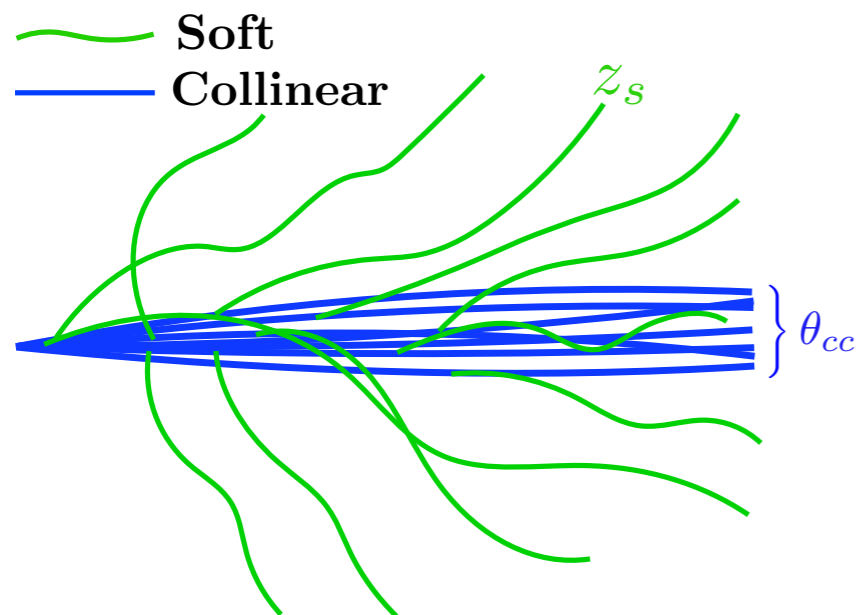
$$\sim \frac{\begin{matrix} \text{CCC} & \text{CCS} & \text{CSS} \\ \theta_{cc}^2 + \theta_{cc} z_s + z_s^2 \end{matrix}}{\left(\theta_{cc} + z_s\right)^2} \begin{matrix} \text{CC} & \text{CS} \end{matrix}$$

Power Counting N_2 : Background

$$N_2 = \frac{2e_3}{(e_2)^2} = \frac{\sum_{i < j < k} z_i z_j z_k \min\{\theta_{ij}\theta_{jk}, \theta_{jk}\theta_{ki}, \theta_{ki}\theta_{ij}\}^\beta}{\left(\sum_{i < j} z_i z_j \theta_{ij}^\beta\right)^2} = \frac{\text{Diagram 1}}{\left(\text{Diagram 2}\right)^2}$$

Setting $\beta = 1$



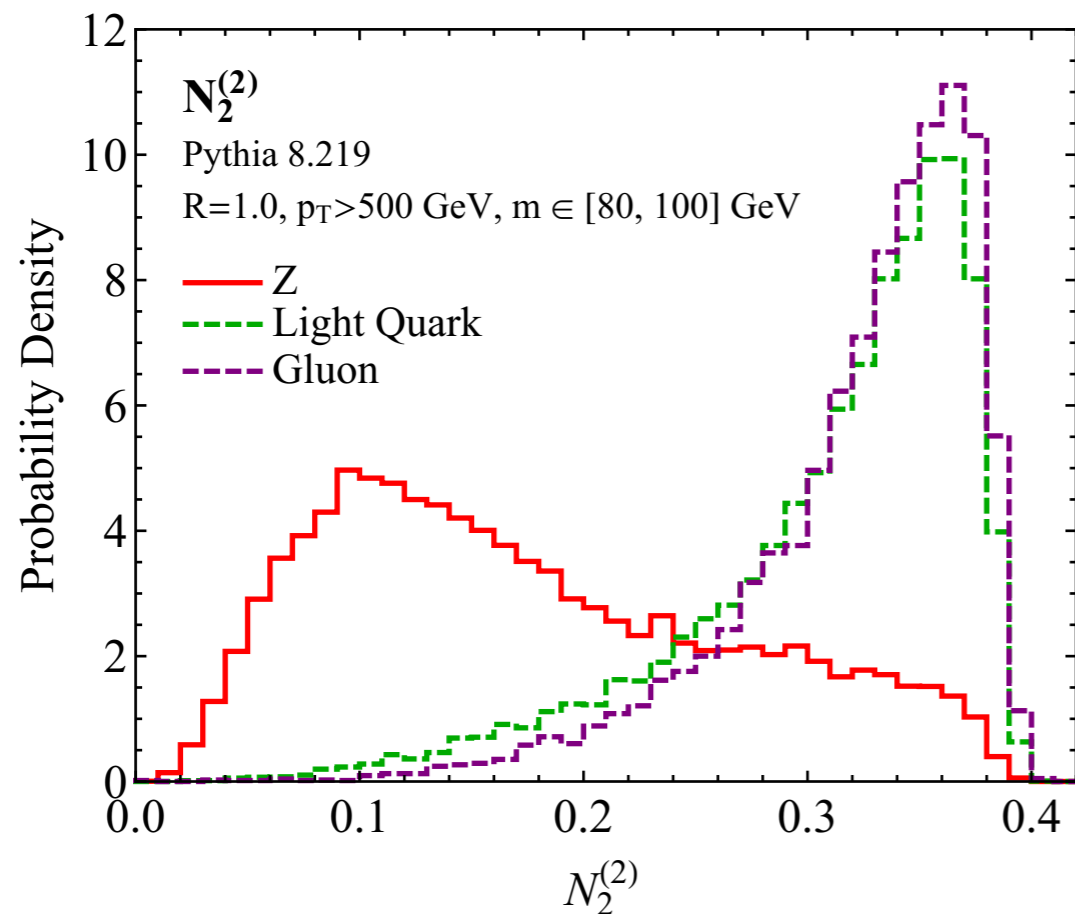
$$\approx \frac{\begin{matrix} \text{CCC} & \text{CCS} & \text{CSS} \\ \theta_{cc}^2 + \theta_{cc}z_s + z_s^2 \\ \text{CC} & \text{CS} \end{matrix}}{\left(\theta_{cc} + z_s\right)^2} \Rightarrow 1$$

Parametrically stable
in both **soft** and
collinear regimes

N_2 : Before Grooming

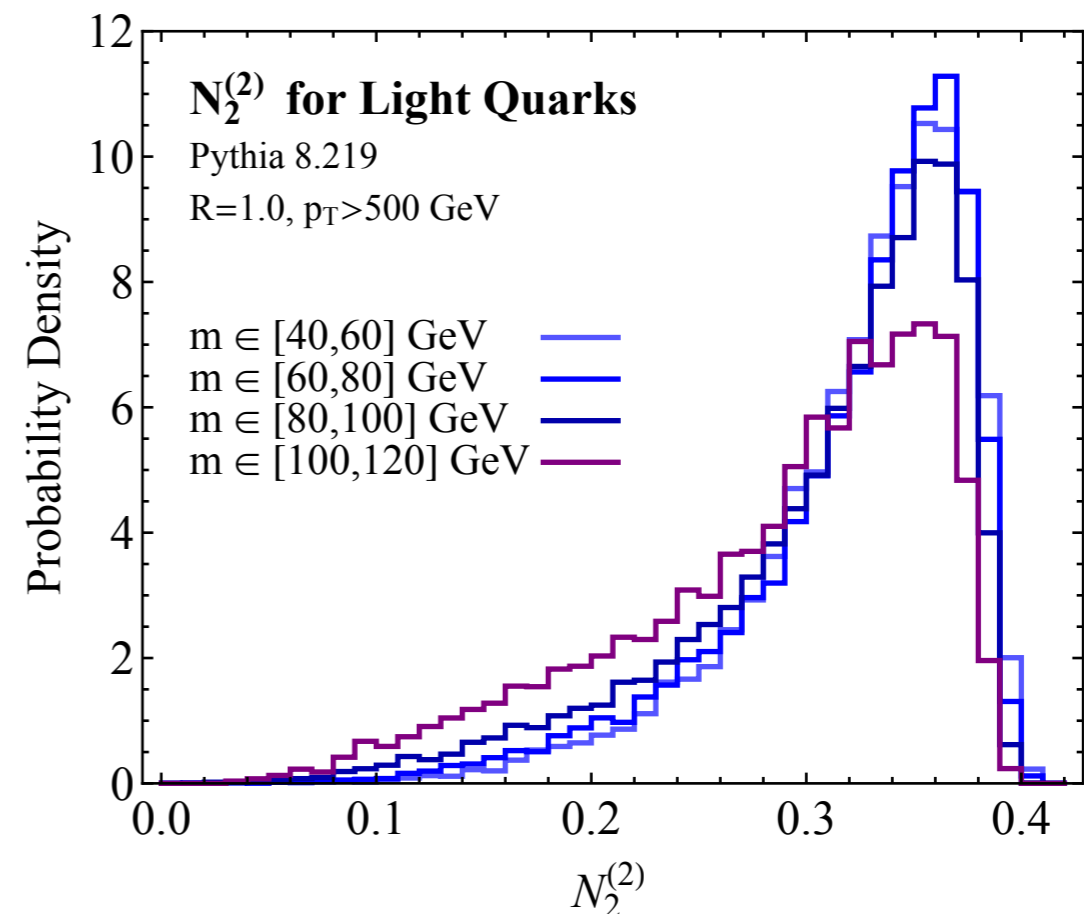
Derived using EFT power counting for both...

Performance



&

Robustness

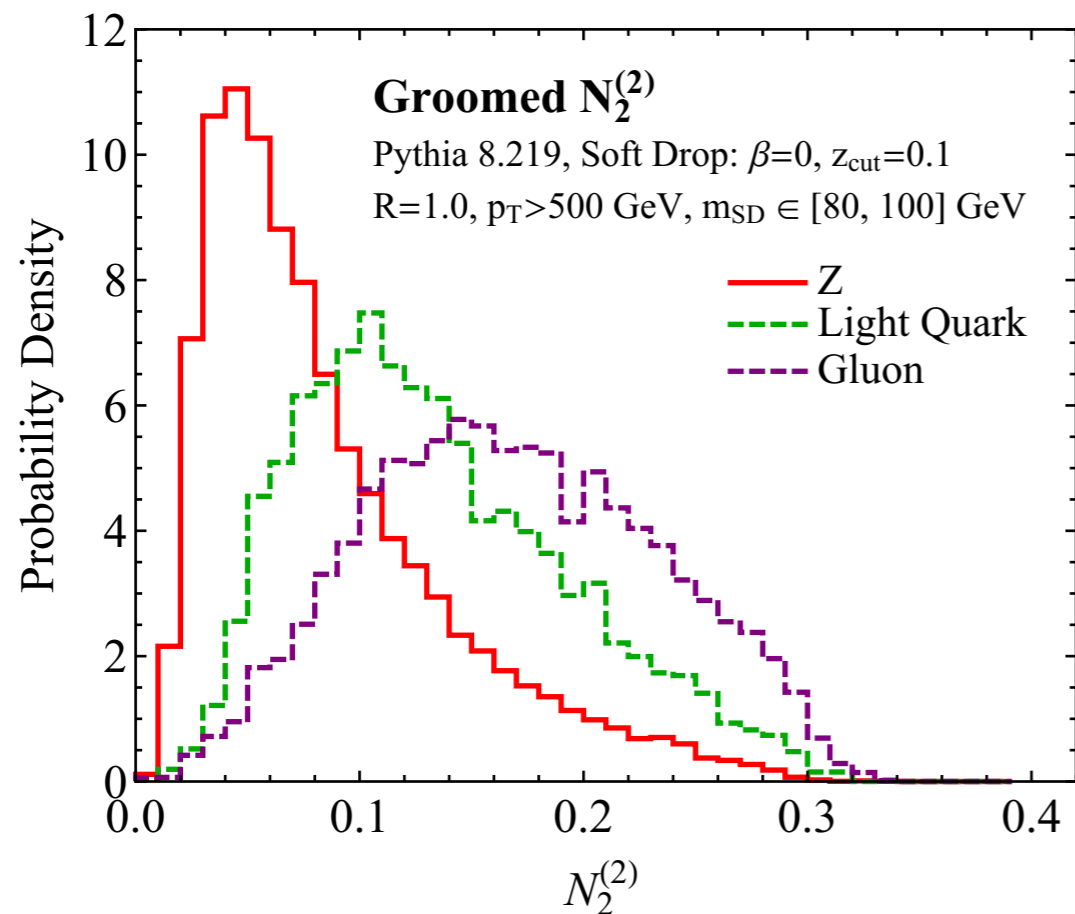


[Moult, Necib, JDT, 1609.07483]

N_2 : After Grooming

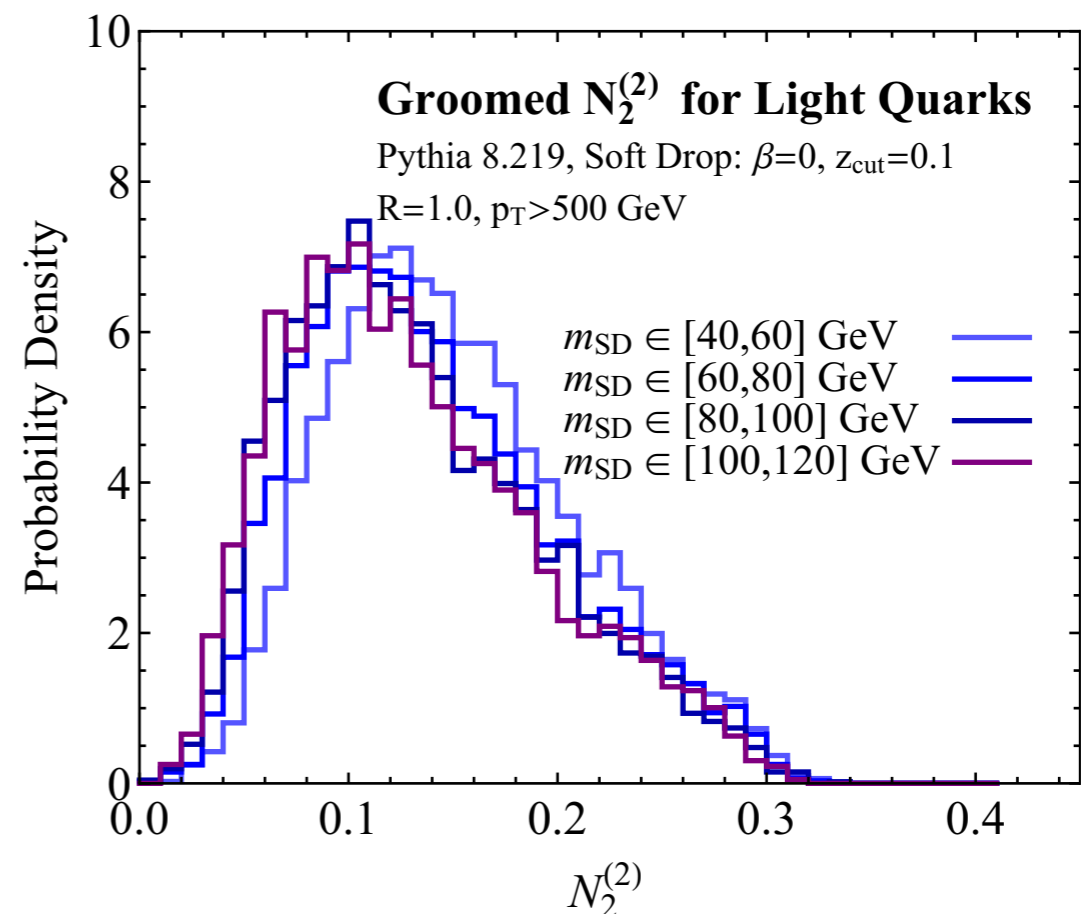
Derived using EFT power counting for both...

Performance



&

Robustness

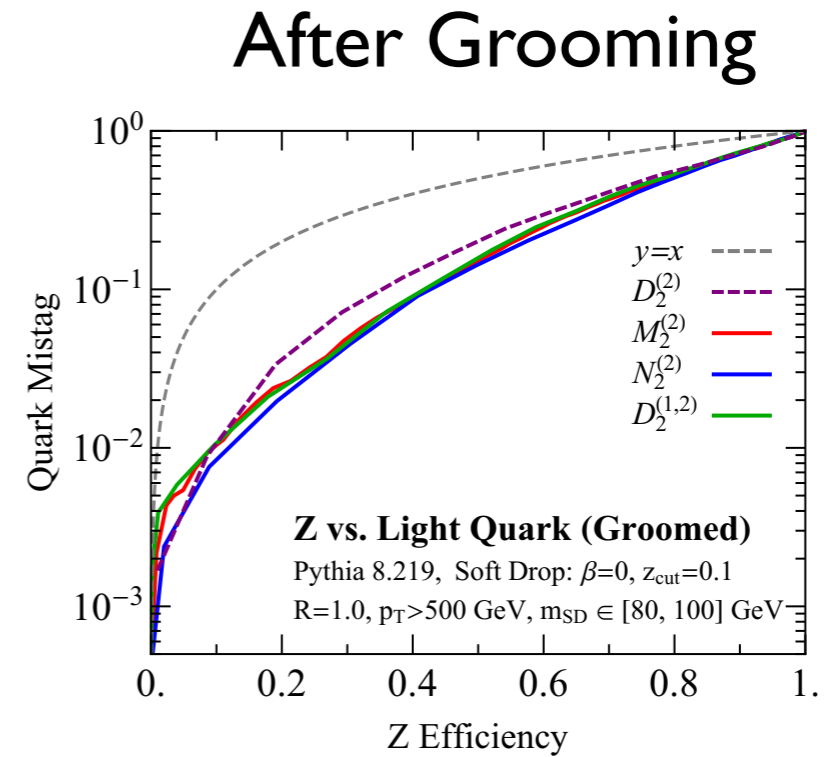
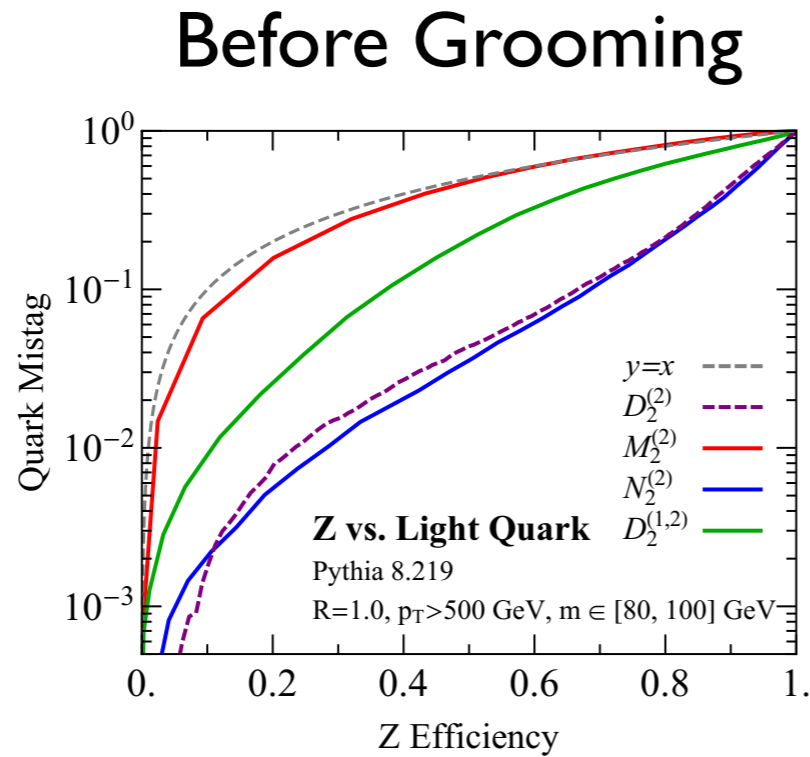


[Moult, Necib, JDT, 1609.07483]

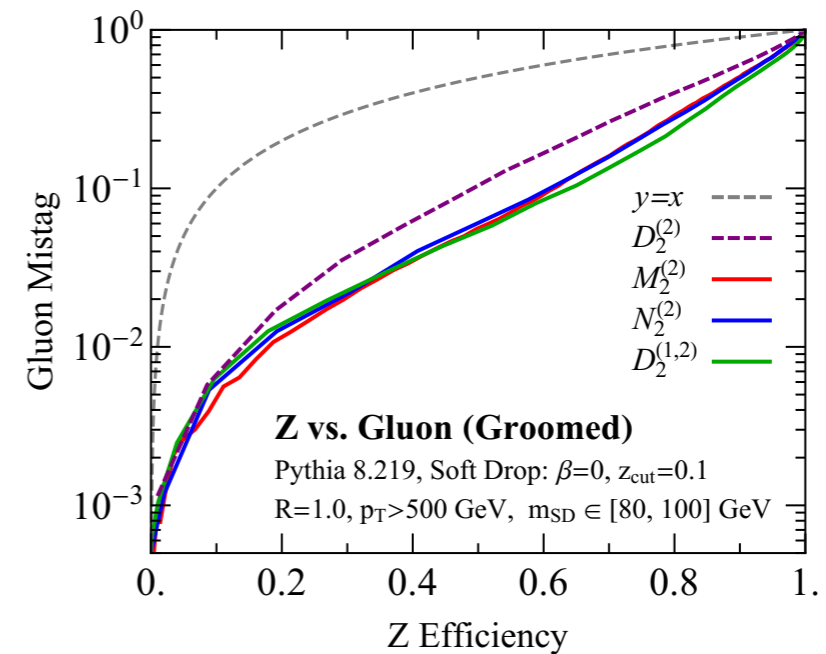
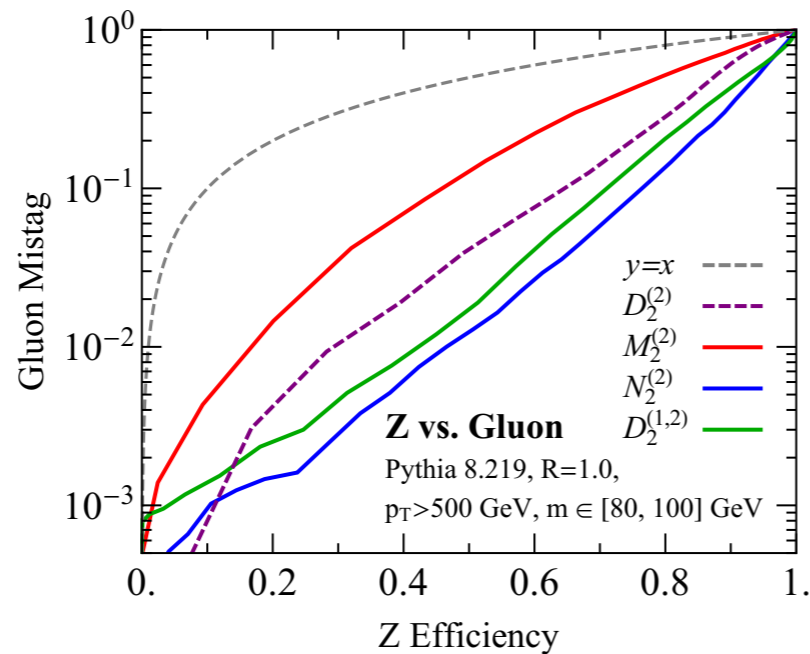
Performance Follows Power Counting

Better

Z vs. q

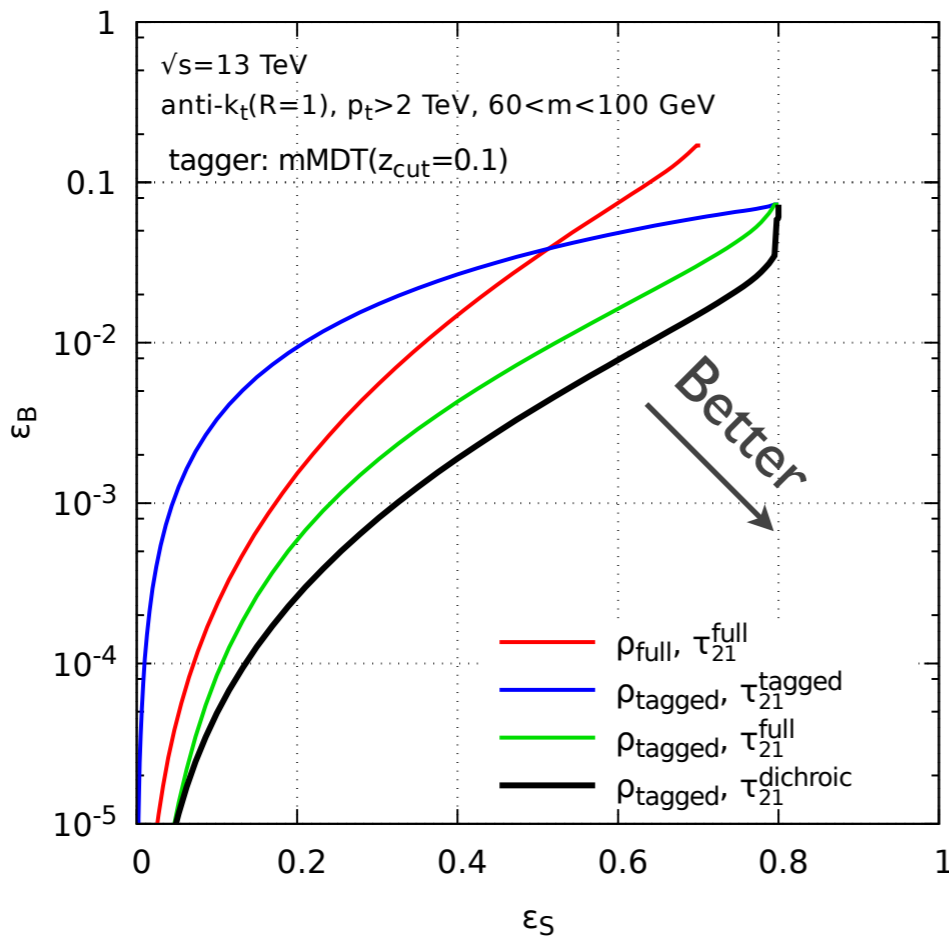


Z vs. g



(Note, groomed mass gives x3-5 better background rejection by itself)

More Grooming/Discrimination Interplay



no grooming

(N.B. for this talk,
tagged \Rightarrow groomed)

groomed mass
groomed τ_2/τ_1

groomed mass
ungroomed τ_2/τ_1

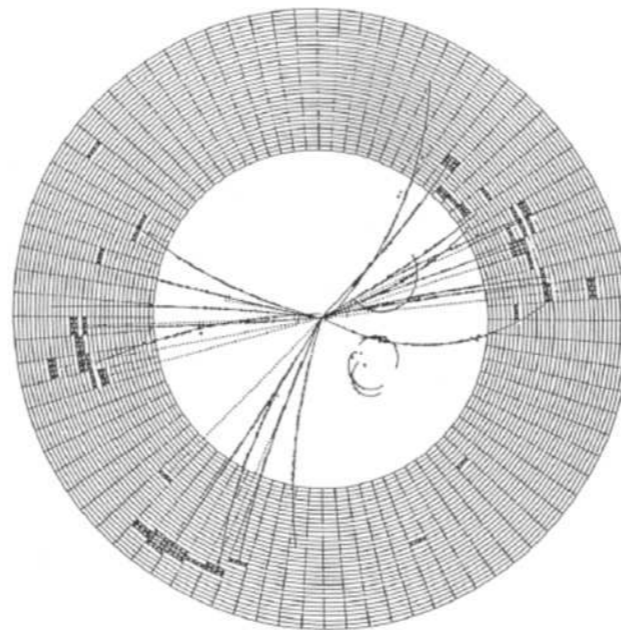


groomed mass
“dichroic” $\tau_2/\tau_1 = \frac{\text{ungroomed } \tau_2}{\text{groomed } \tau_1}$

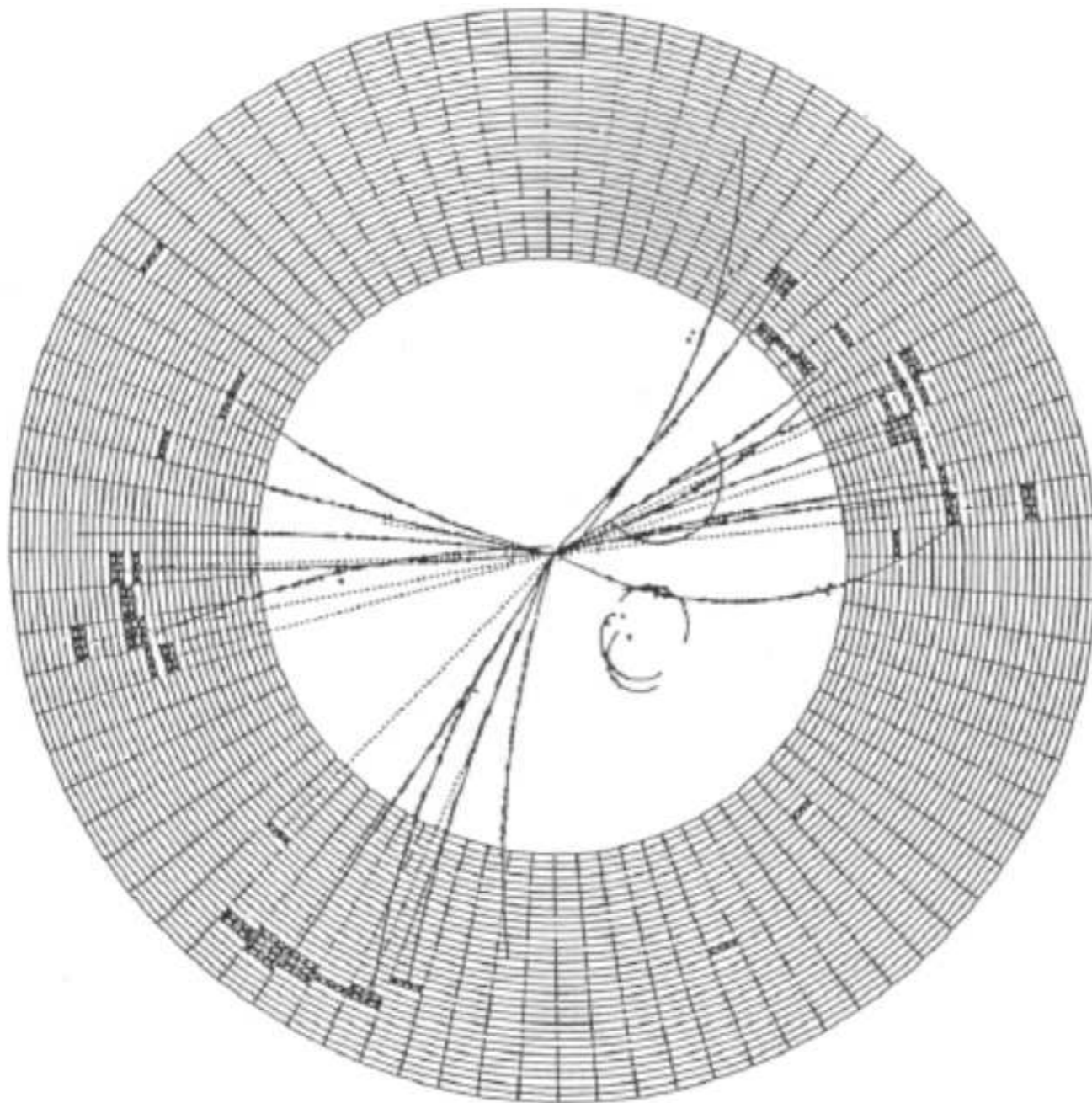
Analytic calculations to identify optimal use of substructure information

[Salam, Schunk, Soyez, 1612.03917]

Back to the Future: Quarks vs. Gluons



Quark vs. Gluon?

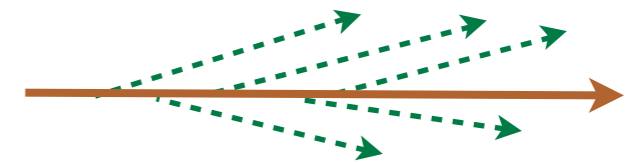


$$e^+e^- \rightarrow q\bar{q}g \quad [\text{JADE, 1979}]$$

Cartoon:

Quark: $C_F = 4/3$

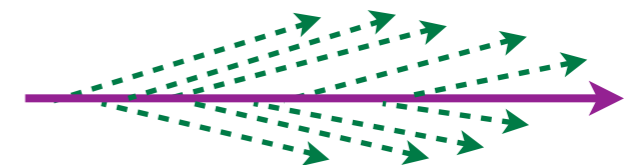
usually signals



vs.

Gluon: $C_A = 3$

usually backgrounds



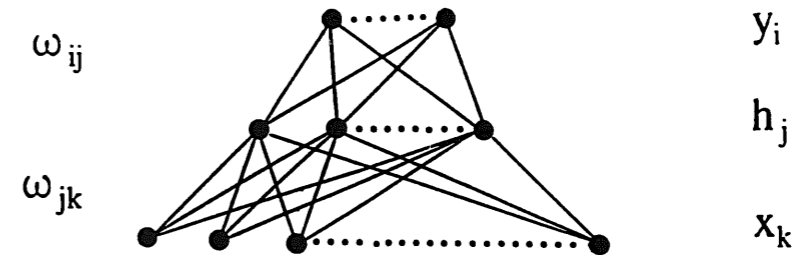
$$\text{gluon mistag} \approx \left(\text{quark efficiency} \right)^{C_A/C_F}$$

barely improves S/\sqrt{B}

An Old Idea...

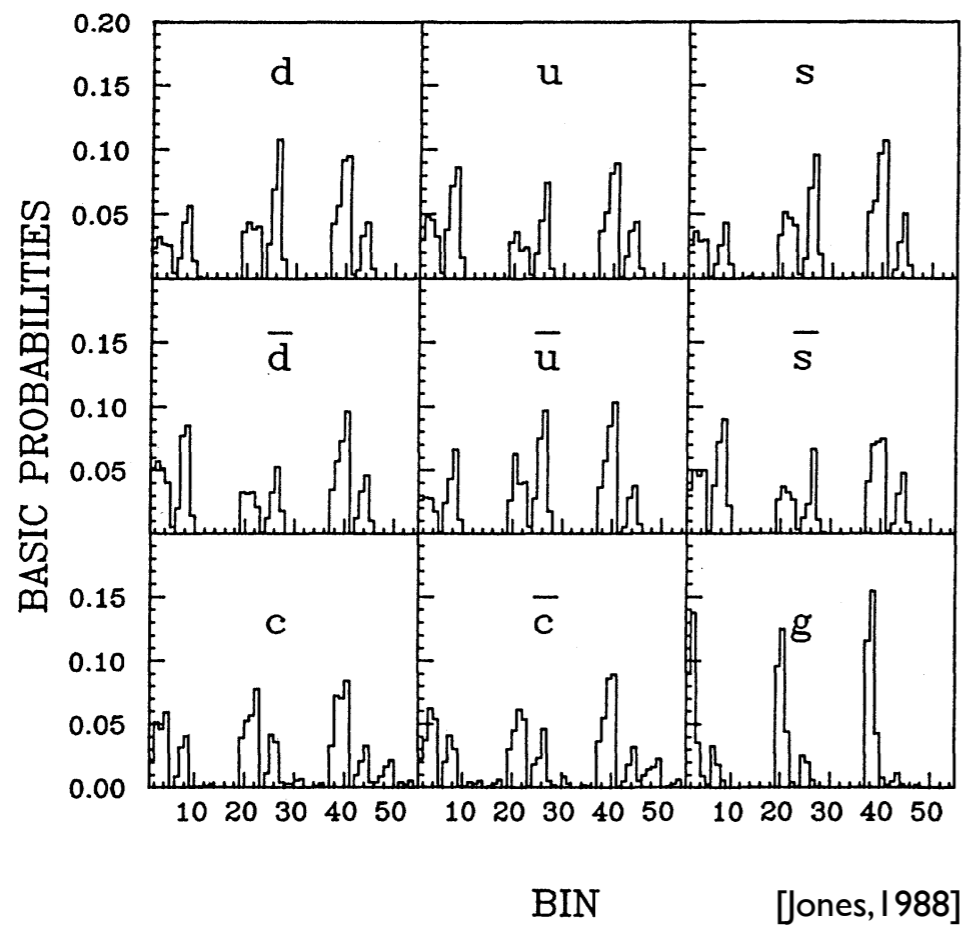
Neural Network Classification

Binned hadron flavor/kinematics

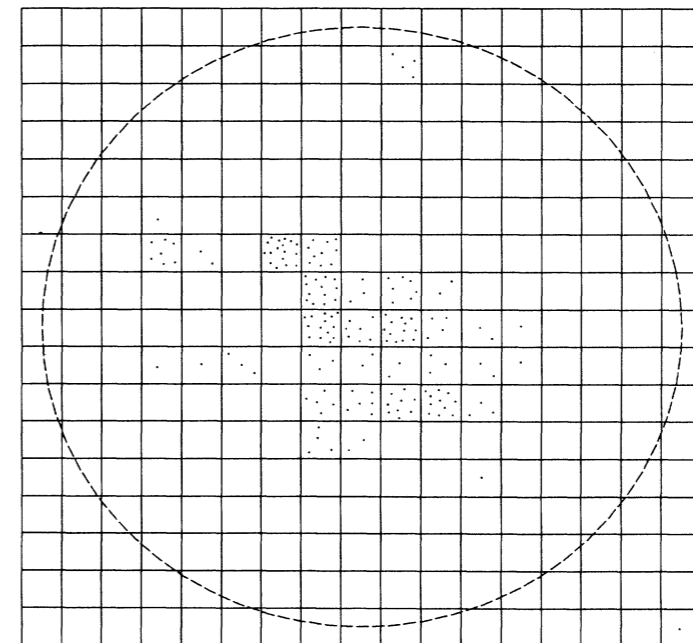


[Lönblad, Peterson, Rönvaldsson, 1991]

N_{90} = Cell count for 90% of jet p_T



(d) GLUON $N_{90} = 21.41$ $E_{2,min} = 0.0222$

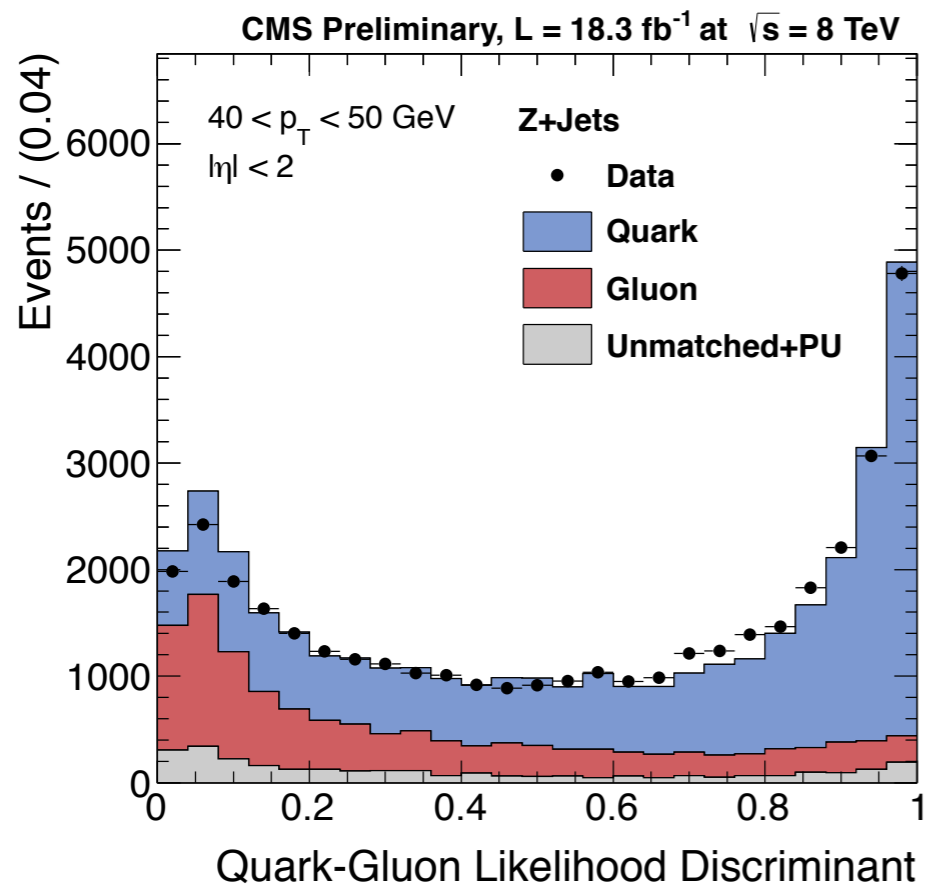


[Pumplin, 1991]

[see also Nilles, Streng, 1981; Fodor, 1990; ...]

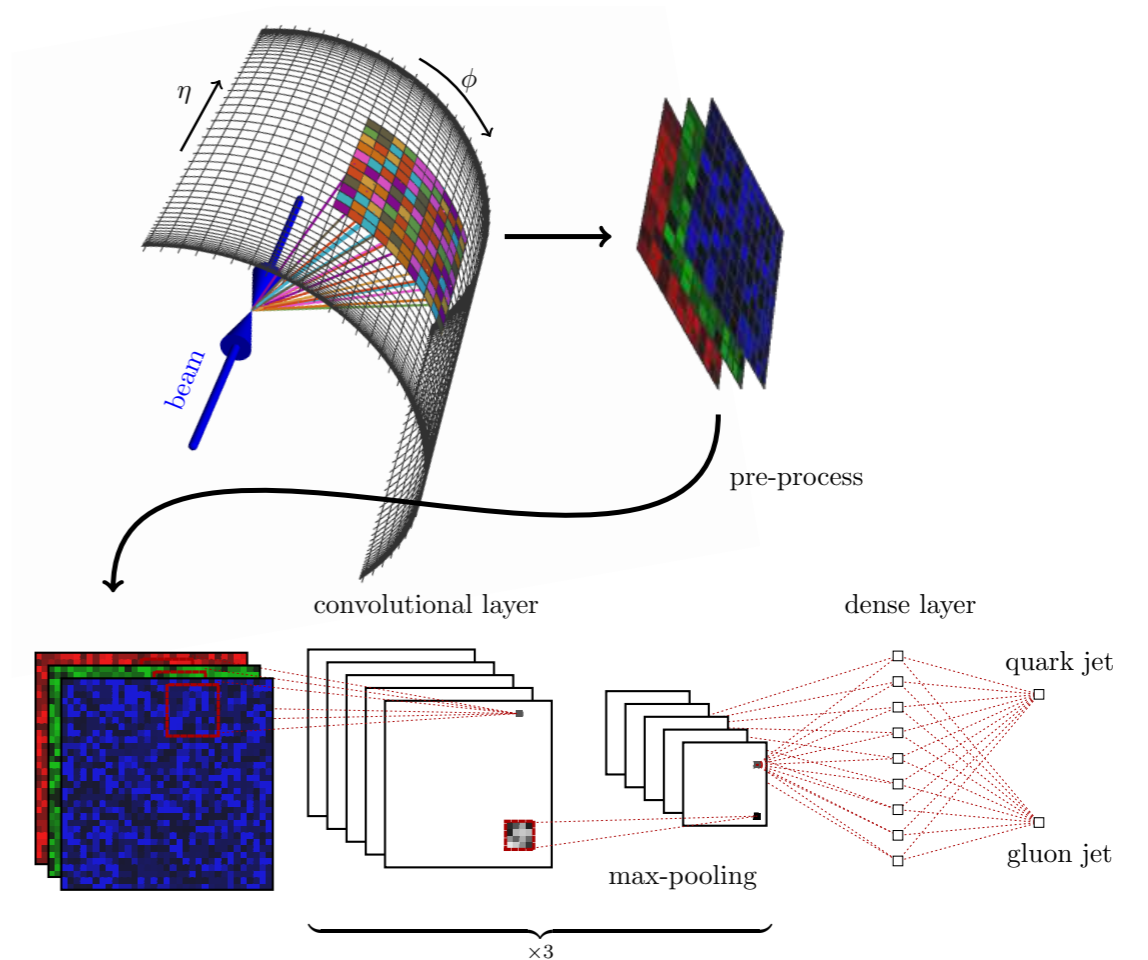
...with Renewed LHC Interest

Three Observable Discriminant ($n_{\text{had}}, p_{\text{T}}^{\text{D}}, \sigma_2$)



[CMS, JME-13-002]

Convolutional (“Deep”) Network ($p_{\text{T}}^{\text{charged}}, p_{\text{T}}^{\text{neutral}}, n_{\text{charged}}$)

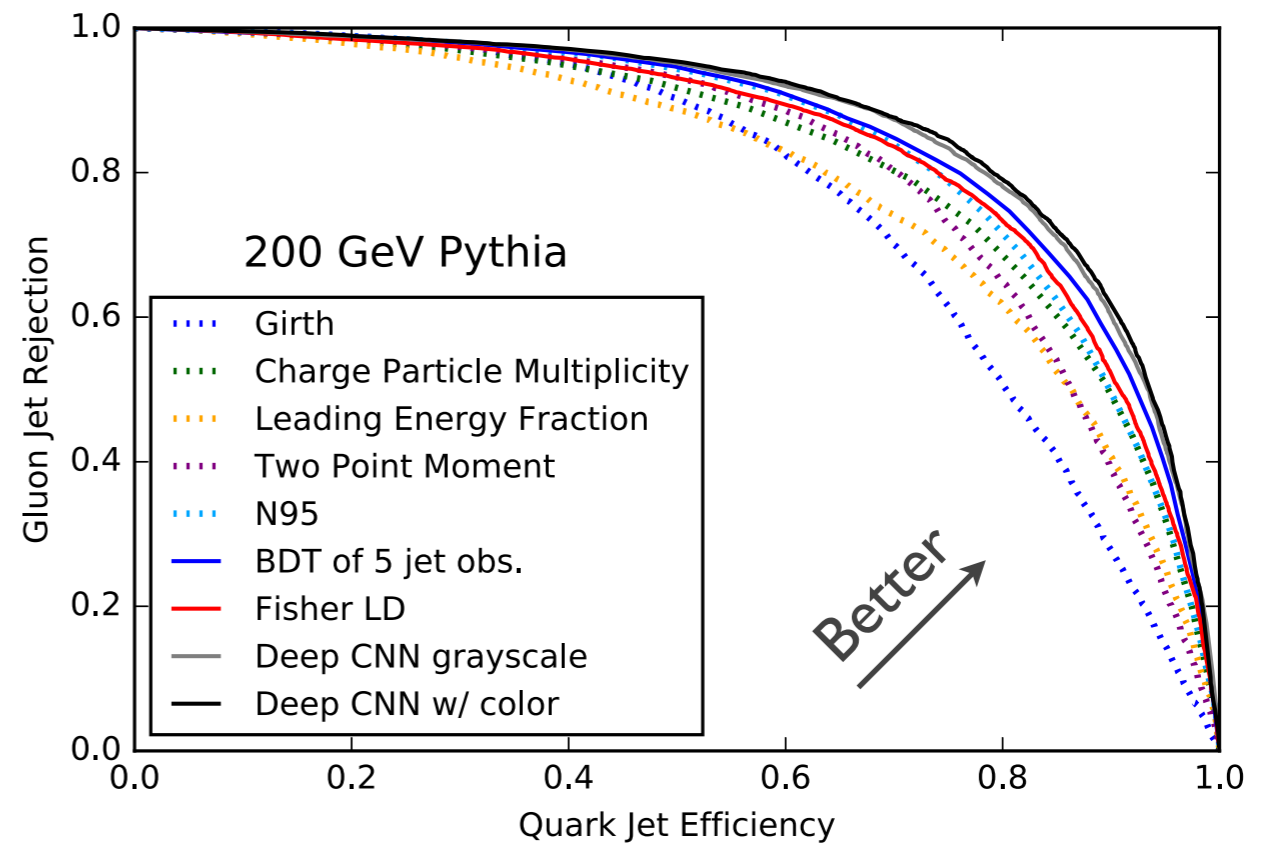
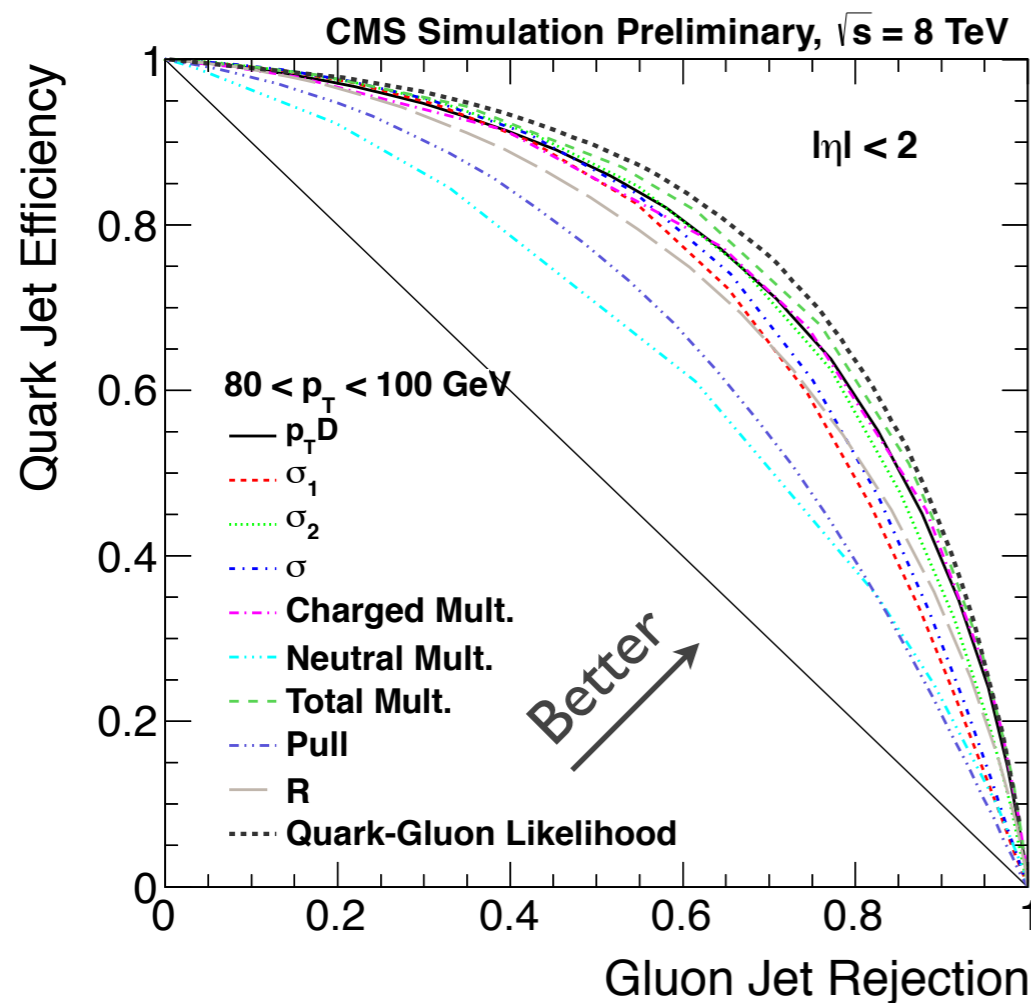


[Komiske, Metodiev, Schwartz, 1612.01551]

[see also Gallicchio, Schwartz, 1106.3076, 1211.7038; Krohn, Schwartz, Lin, Waalewijn, 1209.2421; Larkoski, Salam, JDT, 1305.0007; Larkoski, JDT, Waalewijn, 1408.3122; Bhattacharjee, Mukhopadhyay, Nojiri, Sakaki, Webber, 1501.04794]

...with Renewed LHC Interest

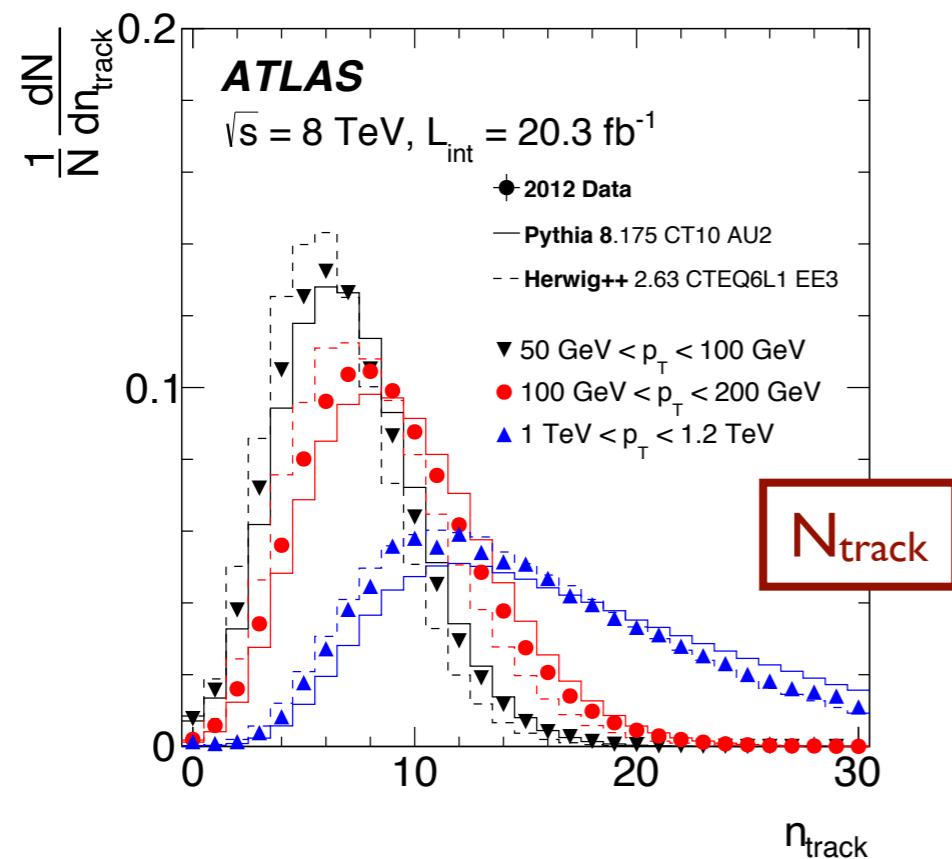
Performance gains from multivariate information



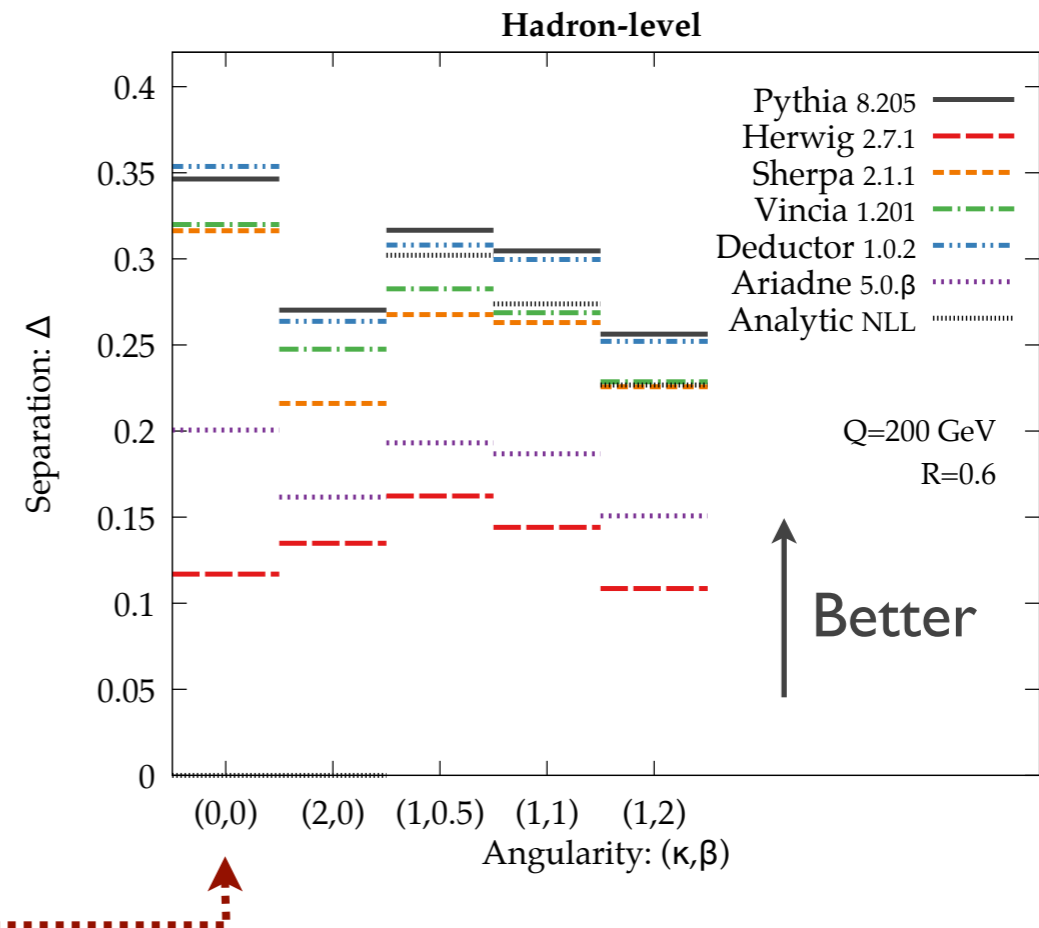
[see also Gallicchio, Schwartz, I 106.3076, I 21 I.7038; Krohn, Schwartz, Lin, Waalewijn, I 209.242 I;
Larkoski, Salam, JDT, I 305.0007; Larkoski, JDT, Waalewijn, I 408.3 I 22; Bhattacharjee, Mukhopadhyay, Nojiri, Sakaki, Webber, I 50 I.04794]

Challenge: Theoretical Uncertainties

“Pythia/Herwig Sandwich”



Classifier Separation



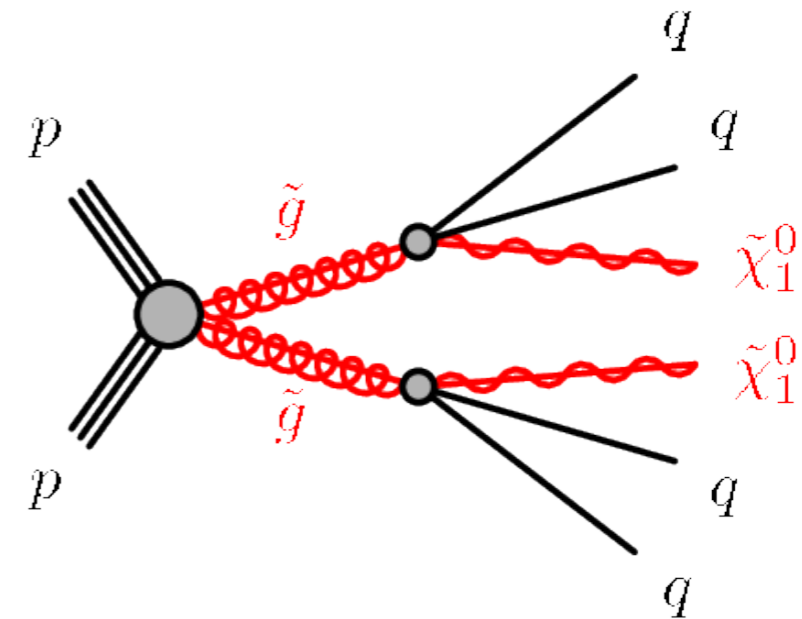
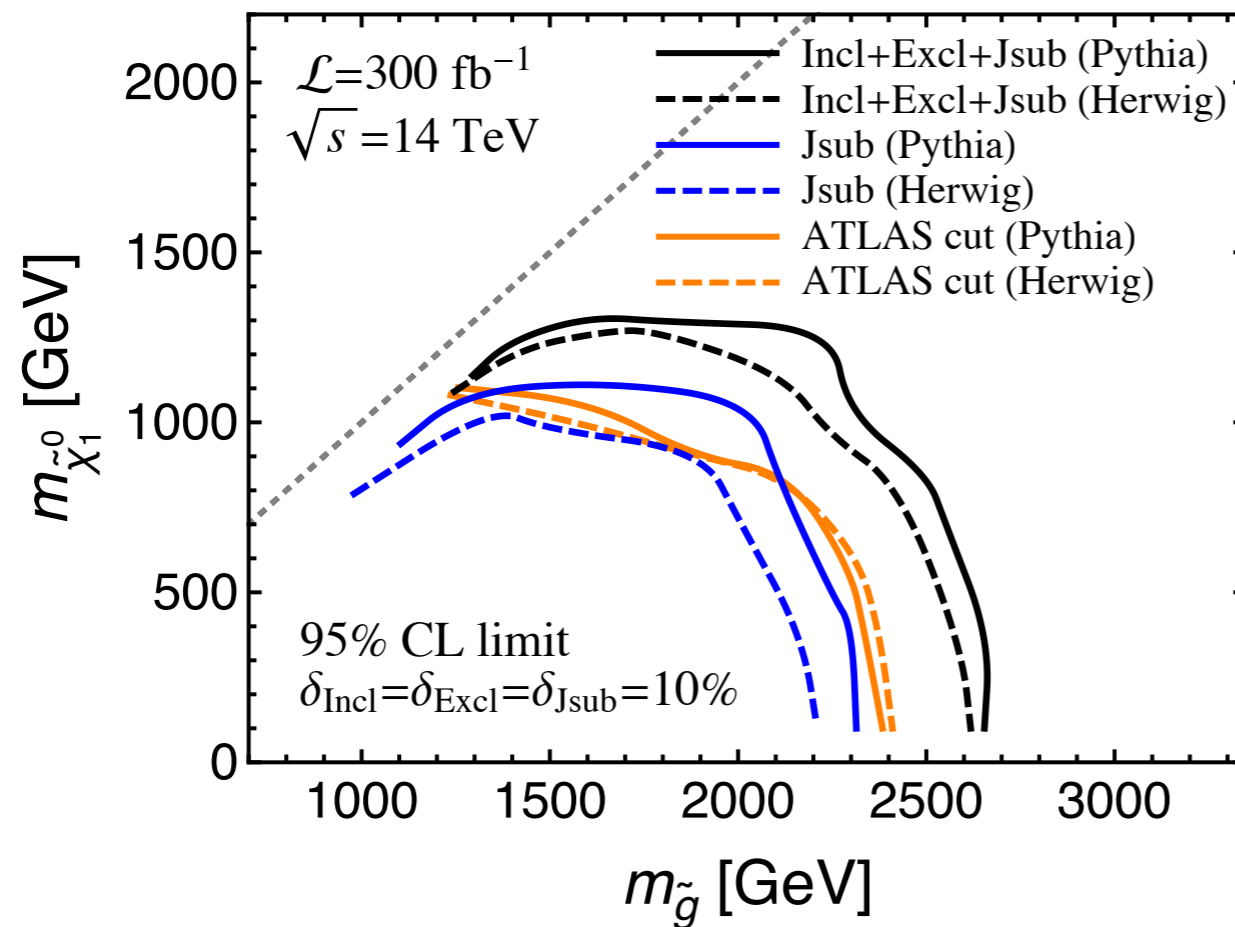
Small uncertainties in distributions yield large uncertainties in discrimination power

[ATLAS, I 602.00988]

[Soyez, JDT, Freytsis, Gras, Kar, Lönnblad, Plätzer, Siodmok, Skands, Soper, I 605.04692]

Challenge: Theoretical Uncertainties

Opportunity: New Physics Applications



[Bhattacharjee, Mukhopadhyay, Nojiri, Sakakie, Webber, I 609.08781]

[ATLAS, I 602.00988]

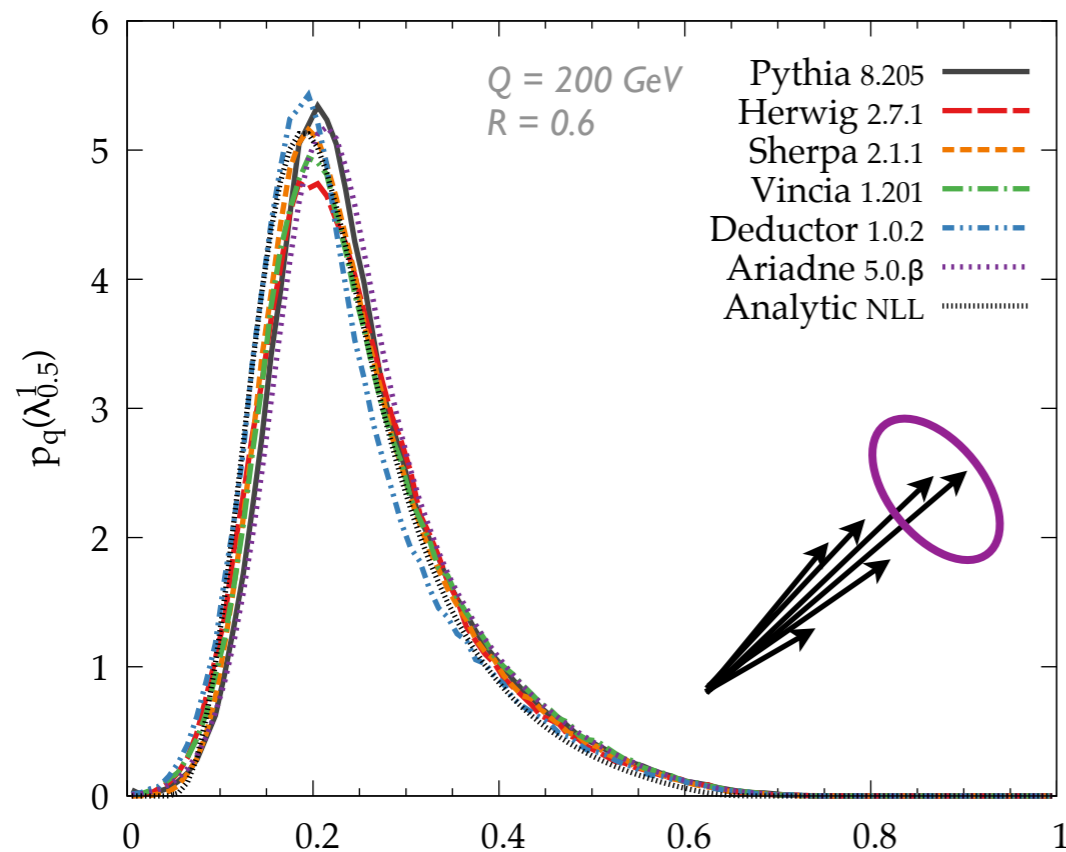
[Soyez, JDT, Freytsis, Gras, Kar, Lönnblad, Plätzer, Siodmok, Skands, Soper, I 605.04692]

Key Task for Jet Substructure

$e^+e^- \rightarrow$ quarks ($C_F = 4/3$)

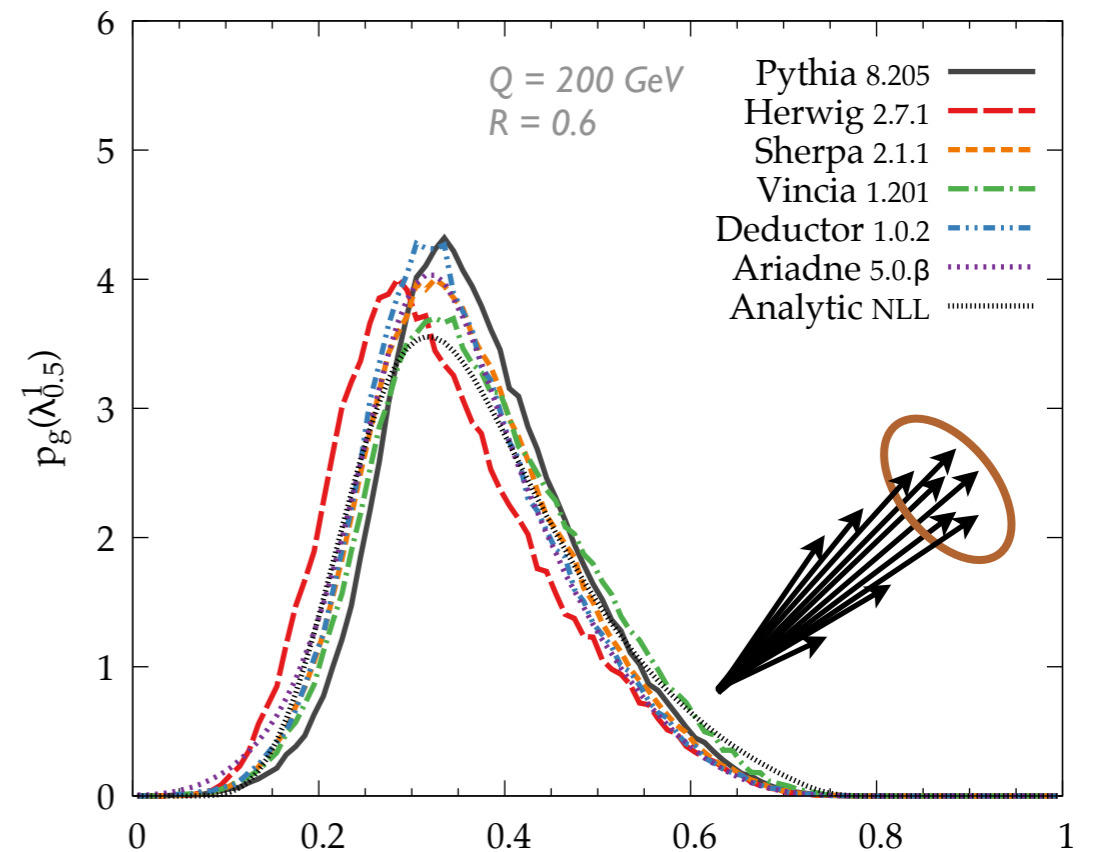
VS.

$e^+e^- \rightarrow$ gluons ($C_A = 3$)



$$\text{LHA} = \sum_i z_i \sqrt{\theta_i}$$

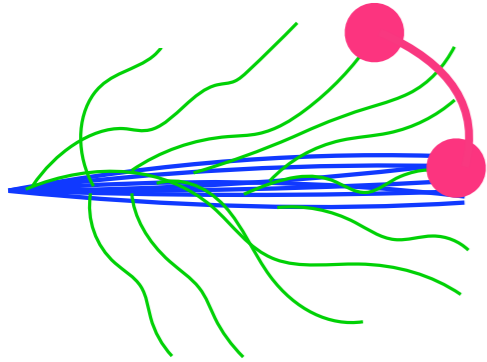
Well-constrained by
LEP measurements



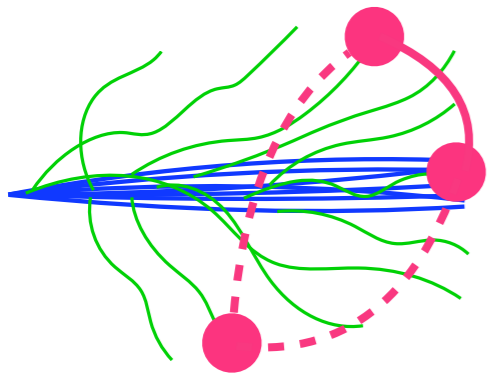
$$\text{LHA} = \sum_i z_i \sqrt{\theta_i}$$

Needs more input from
experiment (and theory)

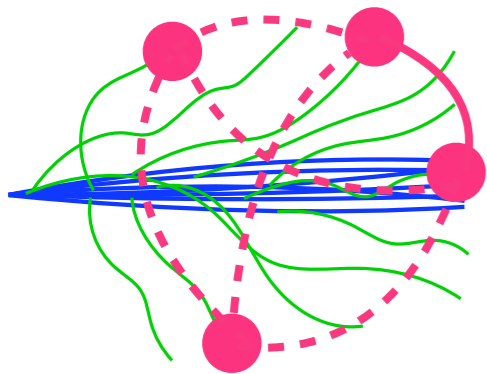
The Next Frontier: Information Beyond C_A/C_F



$$U_1 = \frac{\sum_{i < j} p_{Ti} p_{Tj} R_{ij}^\beta}{\left(\sum_i p_{Ti} \right)^2} \quad \leftarrow \text{Typical Quark/Gluon Discriminants}$$



$$U_2 = \frac{\sum_{i < j < k} p_{Ti} p_{Tj} p_{Tk} \min\{R_{ij}, R_{jk}, R_{ki}\}^\beta}{\left(\sum_i p_{Ti} \right)^3}$$

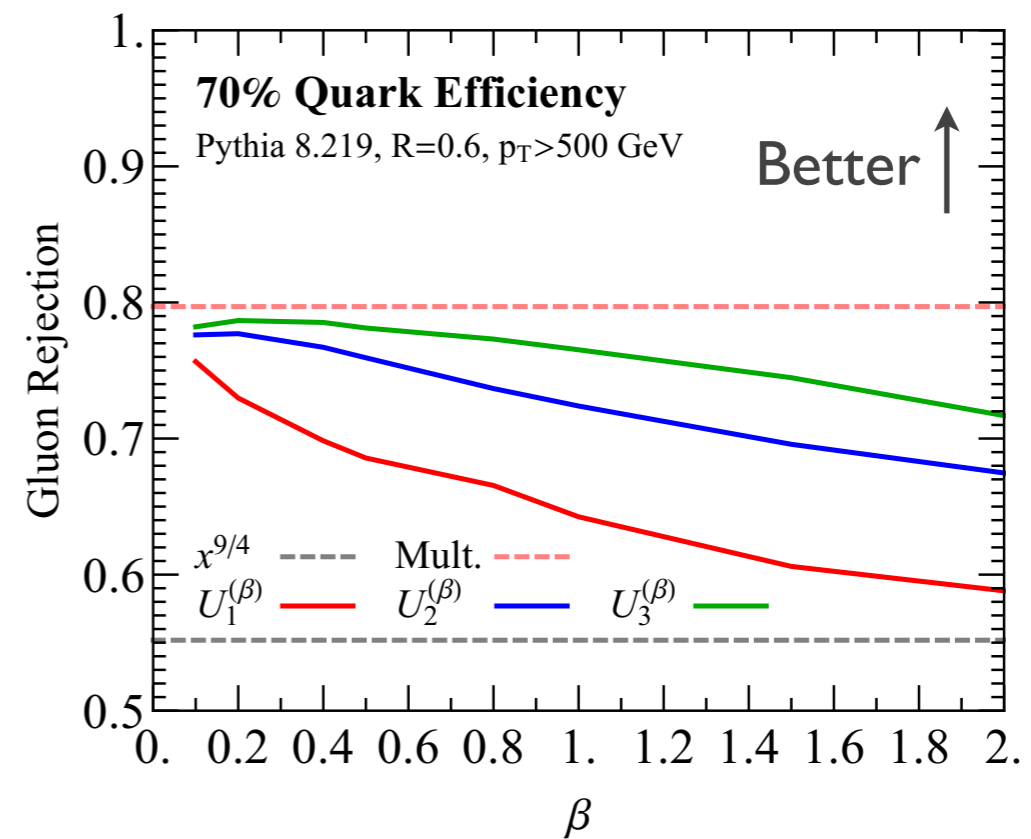
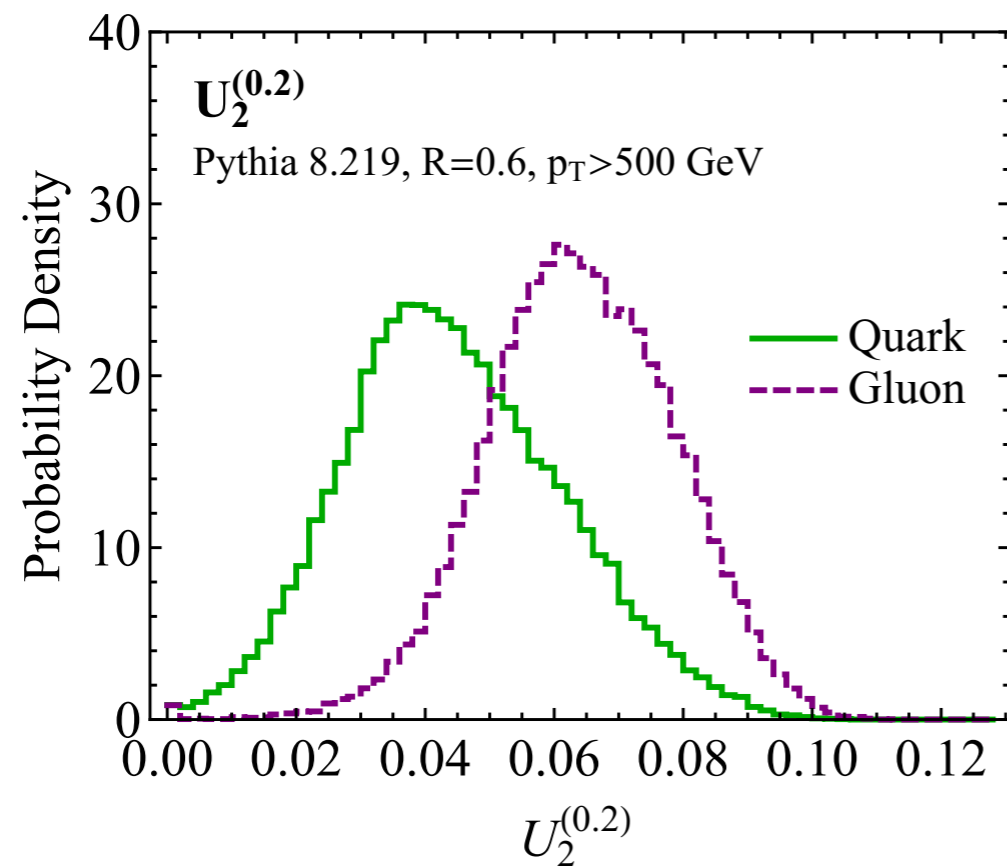


$$U_3 = \frac{\sum_{i < j < k < l} p_{Ti} p_{Tj} p_{Tk} p_{Tl} \min\{R_{ij}, R_{jk}, R_{kl}, R_{ik}, R_{jl}, R_{il}\}^\beta}{\left(\sum_i p_{Ti} \right)^4}$$

“Deep Learning” inspires “Deep Thinking”

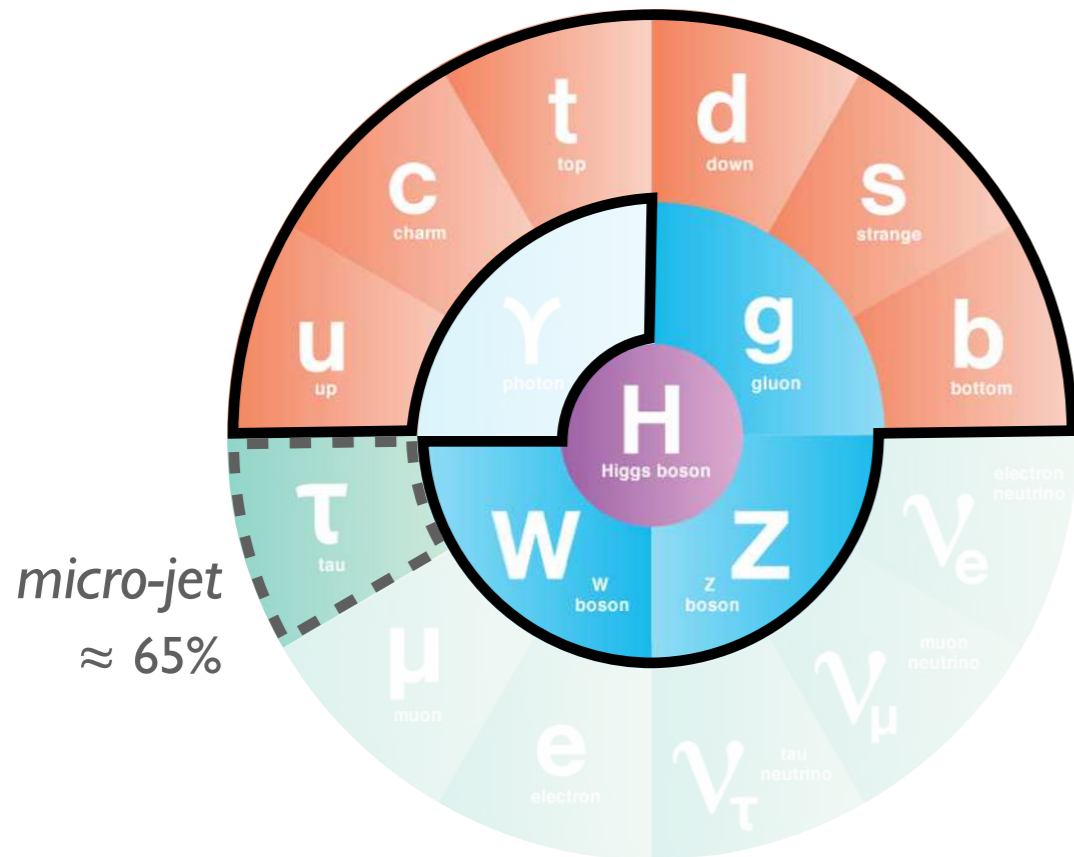
The Next Frontier: Information Beyond C_A/C_F

Derived using EFT power counting to probe perturbative multi-point soft-gluon phase space



“Deep Learning” inspires “Deep Thinking”

Jet Substructure in 2017



*Goal: Systematic
Jet Classification
from First Principles*

