

Jets and QCD — lecture 4

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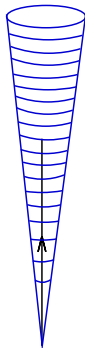
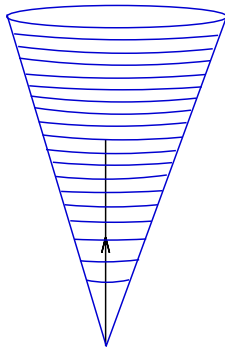
⁴ Princeton University

Focus Week at the GGI Workshop
High energy QCD after the start of the LHC
Florence, Italy, 12-16 September 2011

Towards an understanding of jets

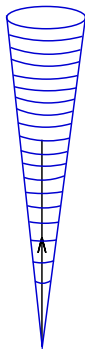
How a jet is and isn't like a parton —
quantitatively

And how this relationship is affected by the jet
radius

Small jet radius**Large jet radius**

single parton @ LO: **jet radius irrelevant**

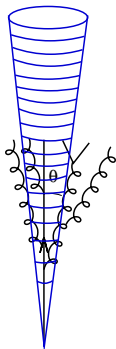
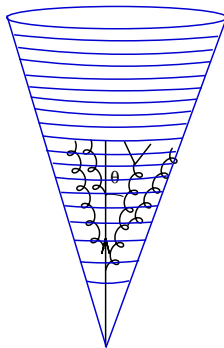
Small jet radius



single part

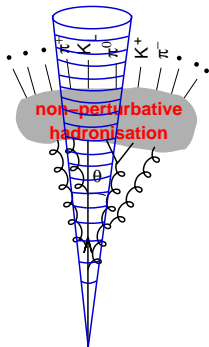
Large jet radius



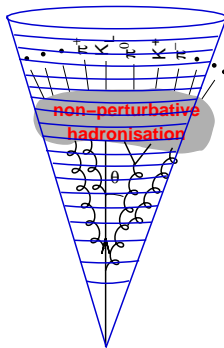
Small jet radius**Large jet radius**

perturbative fragmentation: **large jet radius better**
(it captures more)

Small jet radius

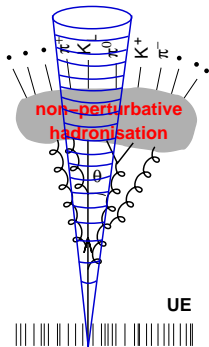


Large jet radius

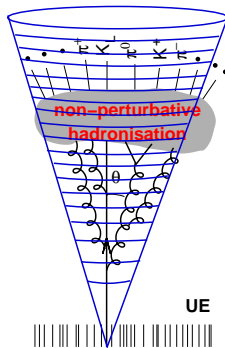


non-perturbative fragmentation: **large jet radius better**
(it captures more)

Small jet radius

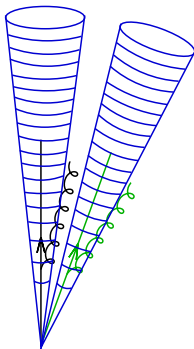


Large jet radius

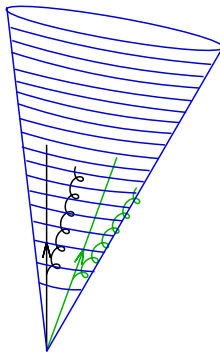


underlying ev. & pileup “noise”: **small jet radius better**
(it captures less)

Small jet radius



Large jet radius



multi-hard-parton events: **small jet radius better**
(it resolves partons more effectively)

Parton p_t v. jet p_t

3 physical effects:

1. Gluon radiation from the parton
2. Hadronisation
3. Underlying Event

One important consideration:

Whether the parton is a quark or a gluon
[quarks radiate with colour factor $C_F = 4/3$
gluons radiate with colour factor $C_A = 3$]

The question's dangerous: a "parton" is an ambiguous concept

Three limits can help you:

- ▶ Threshold limit e.g. de Florian & Vogelsang '07
- ▶ Parton from color-neutral object decay (Z')
- ▶ Small- R (radius) limit for jet

One simple result (small- R limit)

$$\frac{\langle p_{t,jet} - p_{t,parton} \rangle}{p_t} = \frac{\alpha_s}{\pi} \ln R \times \begin{cases} 1.01 C_F & \text{quarks} \\ 0.94 C_A + 0.07 n_f & \text{gluons} \end{cases} + \mathcal{O}(\alpha_s^2)$$

only $\mathcal{O}(\alpha_s^2)$ depends on algorithm & process
cf. Dasgupta, Magnea & GPS '07

Hadronisation: the “parton-shower” → hadrons transition

Method:

- ▶ “infrared finite α_s ” à la Dokshitzer & Webber '95
- ▶ **prediction** based on e^+e^- event shape data
- ▶ could have been deduced from old work Korchemsky & Sterman '95
Seymour '97

Main result

$$\langle p_{t,jet} - p_{t,parton-shower} \rangle \simeq -\frac{0.4 \text{ GeV}}{R} \times \begin{cases} C_F & \text{quarks} \\ C_A & \text{gluons} \end{cases}$$

cf. Dasgupta, Magnea & GPS '07
coefficient holds for anti- k_t ; see Dasgupta & Delenda '09 for k_t alg.

“Naive” prediction (UE \simeq colour dipole between pp):

$$\Delta p_t \simeq 0.4 \text{ GeV} \times \frac{R^2}{2} \times \begin{cases} C_F & q\bar{q} \text{ dipole} \\ C_A & \text{gluon dipole} \end{cases}$$

Modern Monte Carlo tunes tell you ($\sqrt{s} = 7 \text{ TeV}$):

$$\Delta p_t \simeq \mathbf{8 \text{ GeV}} \times \frac{R^2}{2} \simeq 1.2 \text{ GeV} \times (\pi R^2)$$

This big coefficient motivates special effort to understand interplay between jet algorithm and UE: “jet areas”

How does coefficient depend on algorithm?

How does it depend on jet p_t ? How does it fluctuate?

cf. Cacciari, GPS & Soyez '08

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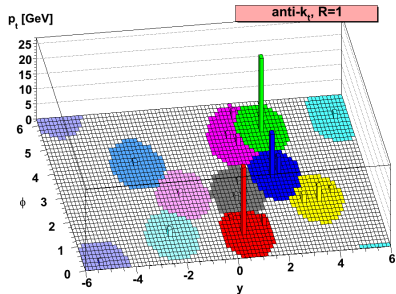
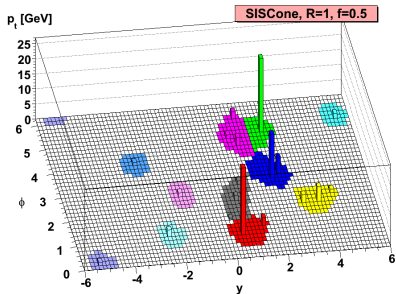
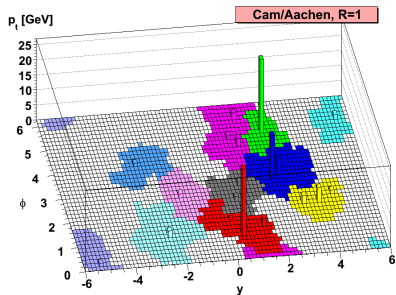
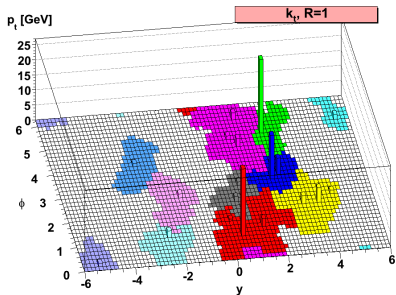
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Using our understanding to help discover a dijet resonance, $q\bar{q} \rightarrow X \rightarrow q\bar{q}$.

E.g. to reconstruct $m_X \sim (p_{tq} + p_{t\bar{q}})$

PT radiation:

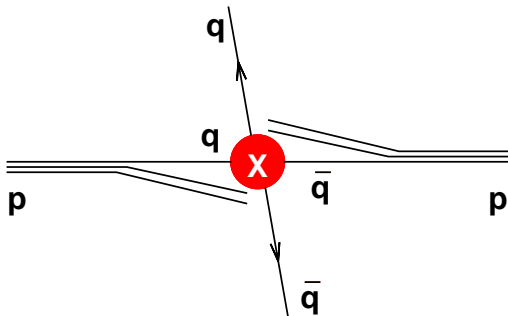
$$q : \quad \langle \Delta p_t \rangle \simeq \frac{\alpha_s C_F}{\pi} p_t \ln R$$

Hadronisation:

$$q : \quad \langle \Delta p_t \rangle \simeq -\frac{C_F}{R} \cdot 0.4 \text{ GeV}$$

Underlying event:

$$q, g : \quad \langle \Delta p_t \rangle \simeq \frac{R^2}{2} \cdot 2.5 - 15 \text{ GeV}$$



Minimise fluctuations in p_t

Use crude approximation:

$$\langle \Delta p_t^2 \rangle \simeq \langle \Delta p_t \rangle^2$$

in small- R limit (!)

NB: full calc, correct fluct: Soyez '10

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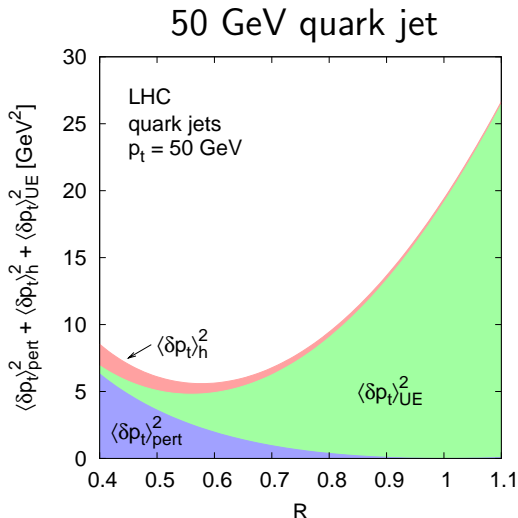
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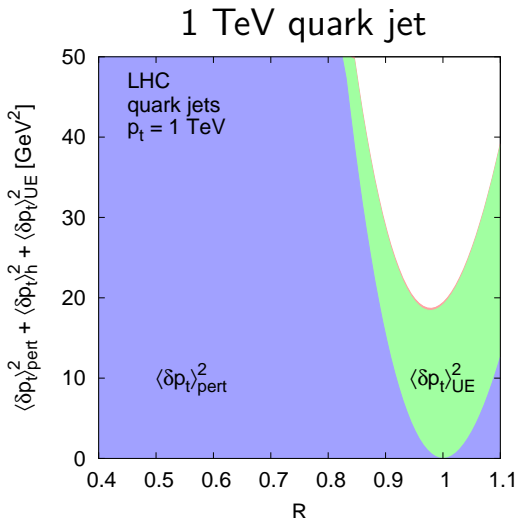
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PT radiation:

$$q : \langle \Delta p_t \rangle \simeq \frac{\alpha_s C_F}{\pi} p_t \ln R$$

Had

$q :$ At low p_t , small R limits relative impact of UE
 At high p_t , perturbative effects dominate over non-perturbative $\rightarrow R_{best} \sim 1$.

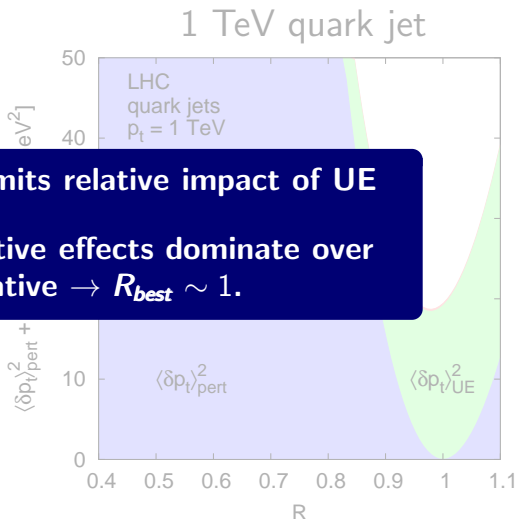
Underlying event:

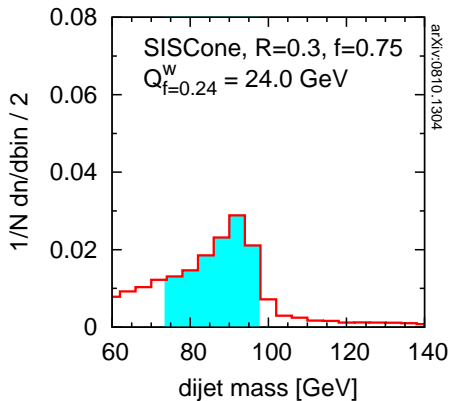
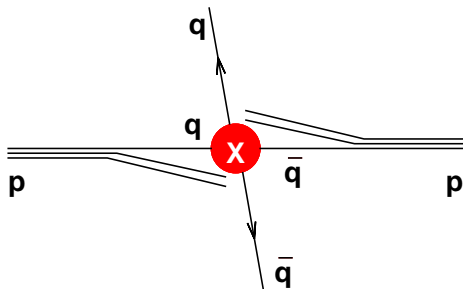
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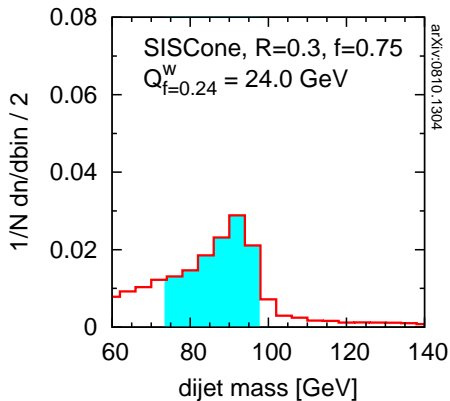
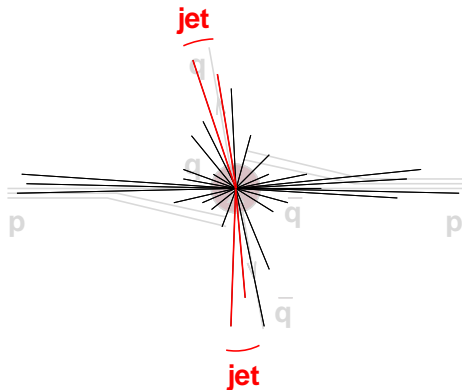
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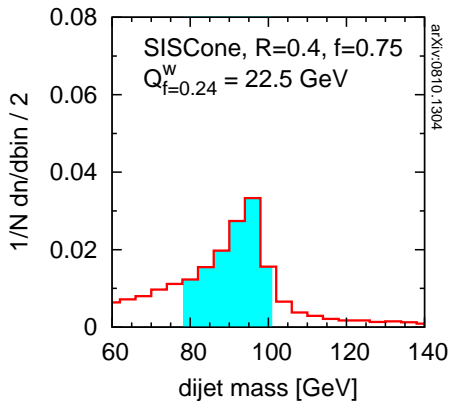
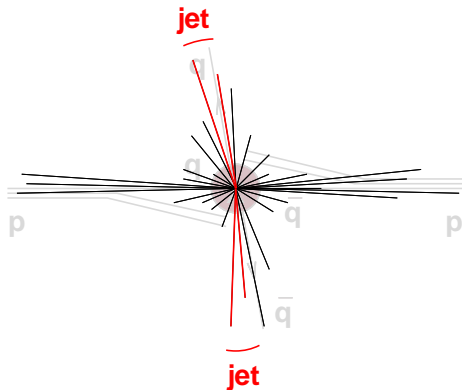
Use crude approximation:

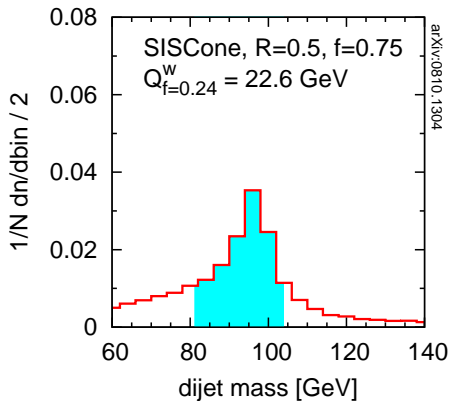
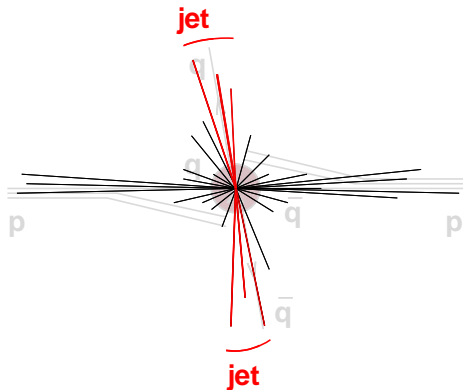
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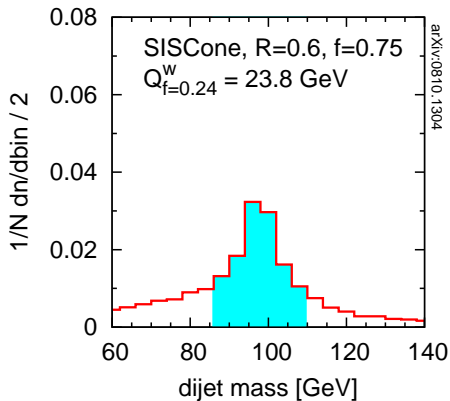
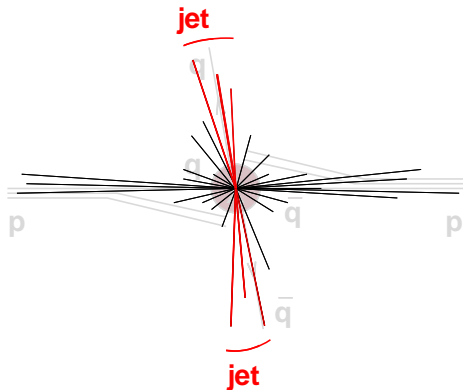


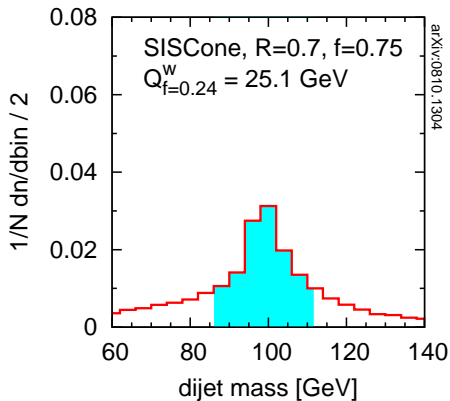
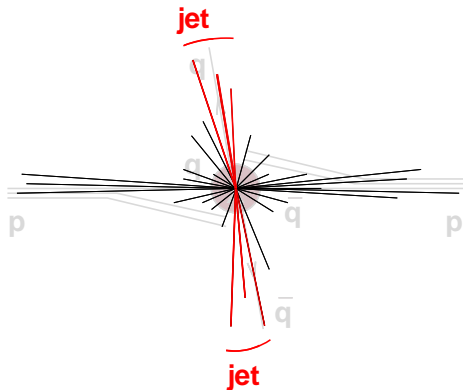
$R = 0.3$ $qq, M = 100 \text{ GeV}$ Resonance $X \rightarrow$ dijets

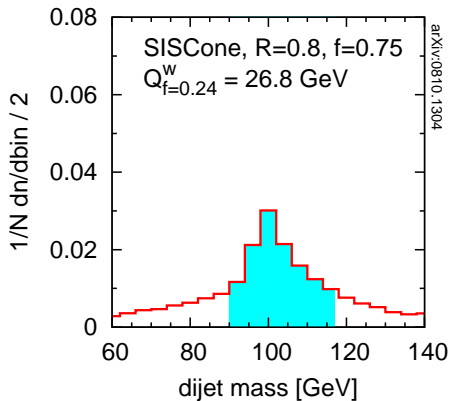
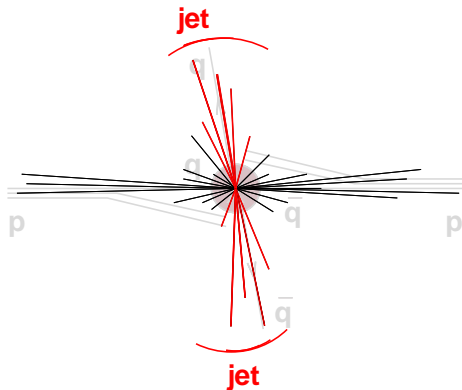
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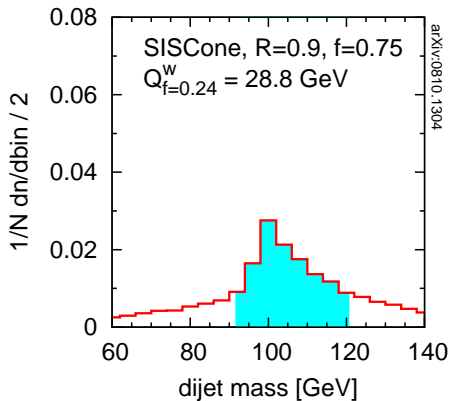
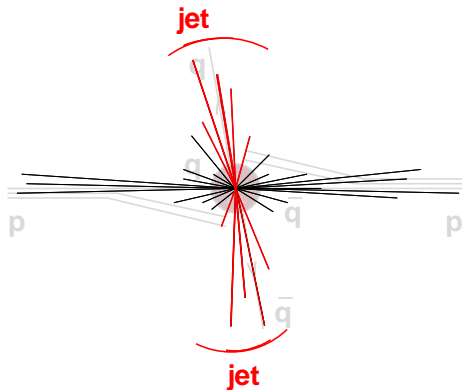
$R = 0.4$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

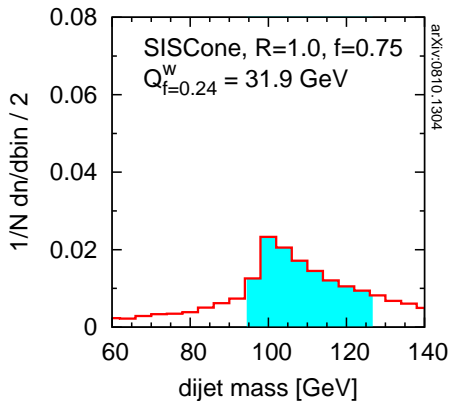
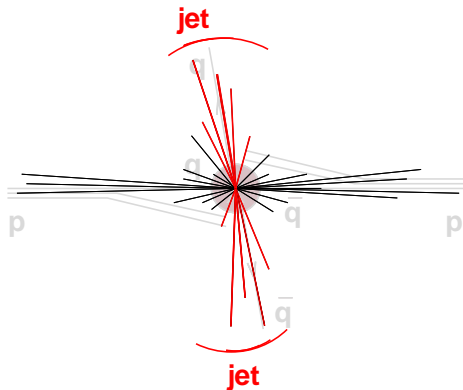
$R = 0.5$ $qq, M = 100 \text{ GeV}$ Resonance $X \rightarrow$ dijets

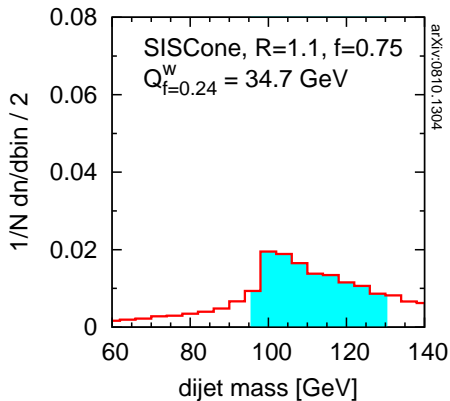
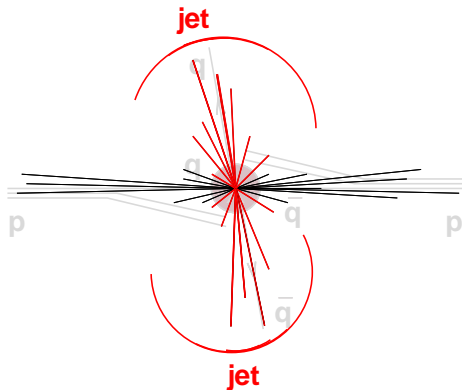
$R = 0.6$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

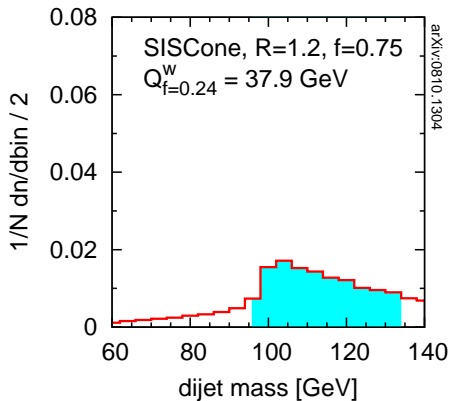
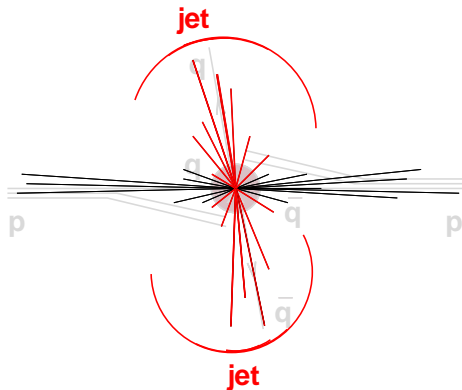
$R = 0.7$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

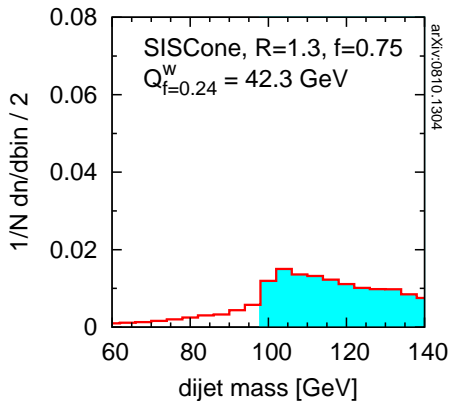
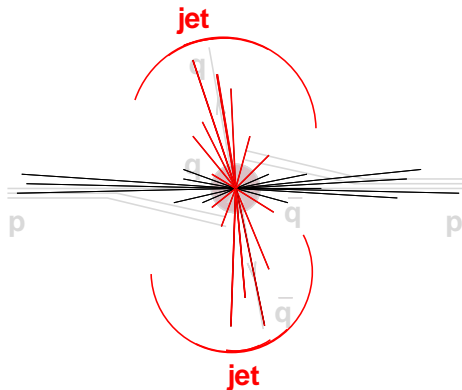
$R = 0.8$ $qq, M = 100 \text{ GeV}$ Resonance $X \rightarrow$ dijets

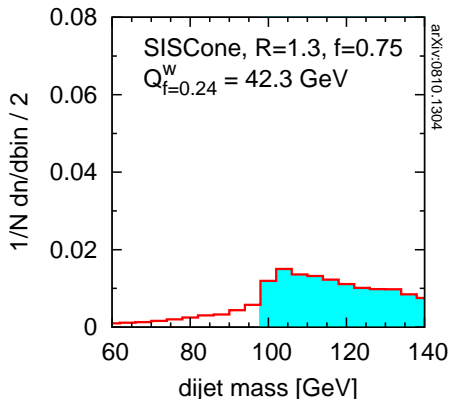
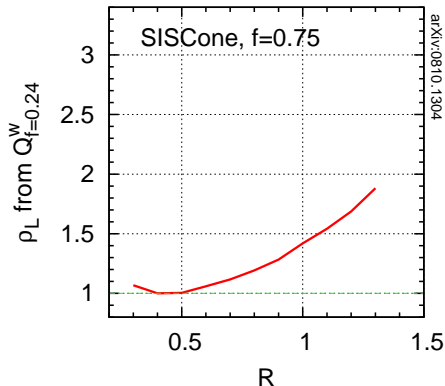
$R = 0.9$ qq, $M = 100$ GeVResonance X \rightarrow dijets

$R = 1.0$ qq, $M = 100$ GeVResonance X \rightarrow dijets

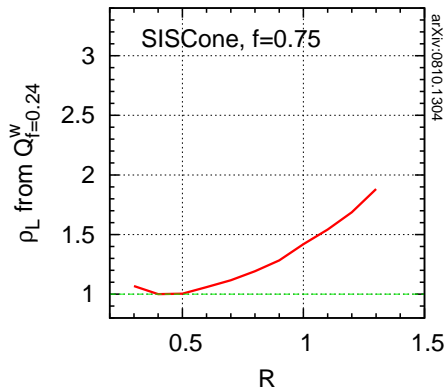
$R = 1.1$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

$R = 1.2$ $qq, M = 100 \text{ GeV}$ Resonance $X \rightarrow$ dijets

$R = 1.3$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

$R = 1.3$ qq, $M = 100$ GeVqq, $M = 100$ GeV

After scanning, summarise “quality” v. R . Minimum \equiv BEST
 picture not so different from crude analytical estimate

$m_{q\bar{q}} = 100 \text{ GeV}$ $q\bar{q}, M = 100 \text{ GeV}$ Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system

- ▶ Increases with mass

- can reproduce this analytically

- Soyez '10

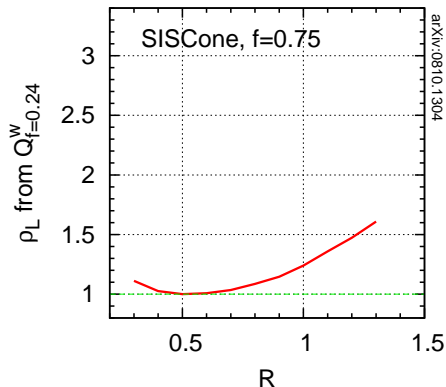
Message received by CMS: they combine all $R = 0.5$ jets ($p_t > 10 \text{ GeV}$) within $\Delta R = 1.1$ of two hardest to improve resolution.

- ATLAS '11 still just use $R = 0.6$

NB: 100,000 plots for various jet algorithms, narrow $q\bar{q}$ and $g\bar{g}$ resonances from <http://quality.fastjet.fr>

- Cacciari, Rojo, GPS & Soyez '08

- Other related work: Krohn, Thaler & Wang '09

$m_{q\bar{q}} = 150 \text{ GeV}$ $q\bar{q}, M = 150 \text{ GeV}$ Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system

- ▶ Increases with mass

- can reproduce this analytically

- Soyez '10

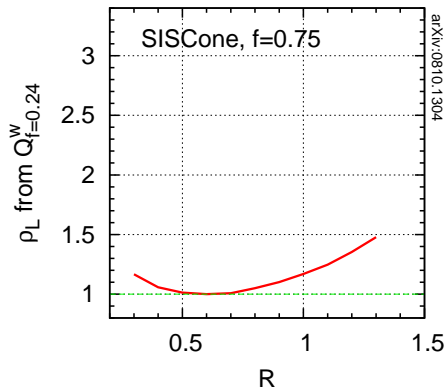
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$m_{q\bar{q}} = 200 \text{ GeV}$ $q\bar{q}, M = 200 \text{ GeV}$ Best R is at minimum of curve

- Best R depends strongly on mass of system

- Increases with mass

- can reproduce this analytically

Soyez '10

Message received by CMS: they combine all $R = 0.5$ jets ($p_t > 10 \text{ GeV}$) within $\Delta R = 1.1$ of two hardest to improve resolution.

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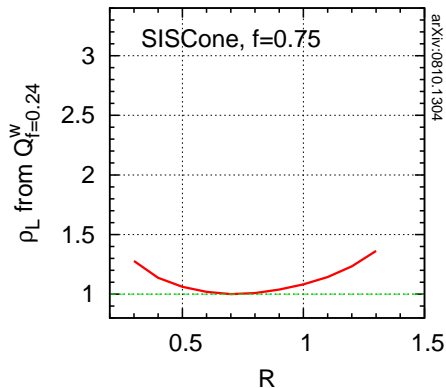
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Cacciari, Rojo, GPS & Soyez '08

Other related work: Krohn, Thaler & Wang '09

$m_{q\bar{q}} = 300 \text{ GeV}$

$q\bar{q}, M = 300 \text{ GeV}$



Best R is at minimum of curve

➤ Best R depends strongly on mass of system

➤ Increases with mass

can reproduce this analytically

Soyez '10

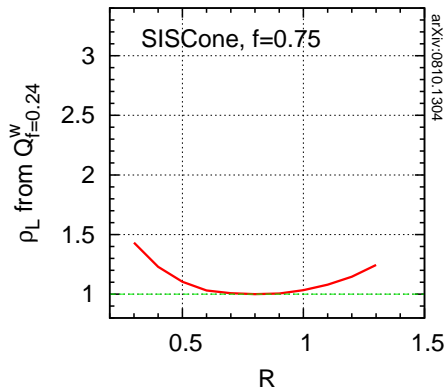
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Cacciari, Rojo, GPS & Soyez '08

Other related work: Krohn, Thaler & Wang '09

$m_{q\bar{q}} = 500 \text{ GeV}$ $q\bar{q}, M = 500 \text{ GeV}$ Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system

- ▶ Increases with mass

- can reproduce this analytically

- Soyez '10

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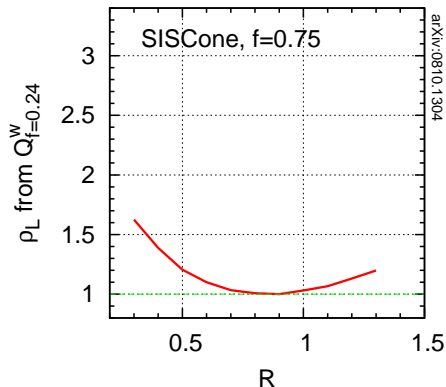
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Cacciari, Rojo, GPS & Soyez '08

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$$m_{q\bar{q}} = 700 \text{ GeV}$$

$$q\bar{q}, M = 700 \text{ GeV}$$



Best R is at minimum of curve

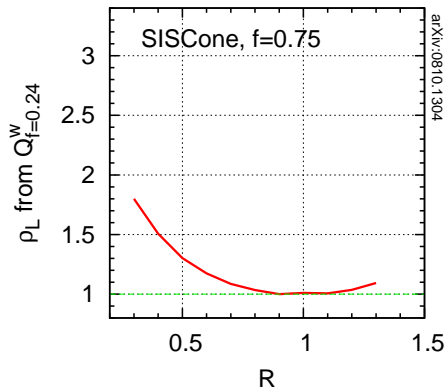
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Other related work: Krohn, Thaler & Wang '09

$m_{q\bar{q}} = 1000 \text{ GeV}$ $q\bar{q}, M = 1000 \text{ GeV}$ Best R is at minimum of curve

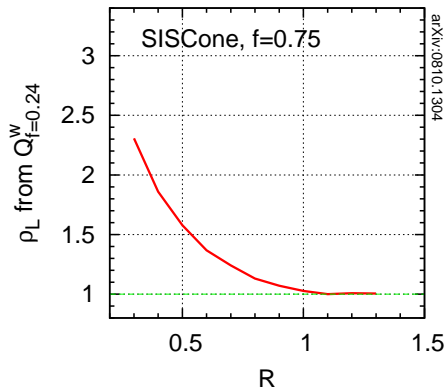
- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass
can reproduce this analytically
Soyez '10

Message received by CMS: they combine all $R = 0.5$ jets ($p_t > 10 \text{ GeV}$) within $\Delta R = 1.1$ of two hardest to improve resolution.

ATLAS '11 still just use $R = 0.6$

NB: 100,000 plots for various jet algorithms, narrow $q\bar{q}$ and $g\bar{g}$ resonances from <http://quality.fastjet.fr> Cacciari, Rojo, GPS & Soyez '08

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$m_{q\bar{q}} = 2000 \text{ GeV}$ qq, $M = 2000 \text{ GeV}$ Best R is at minimum of curve

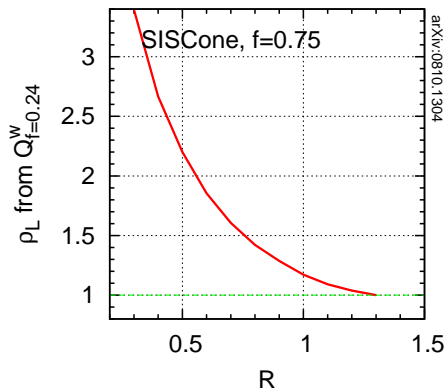
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$m_{q\bar{q}} = 4000 \text{ GeV}$ $q\bar{q}, M = 4000 \text{ GeV}$ Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass

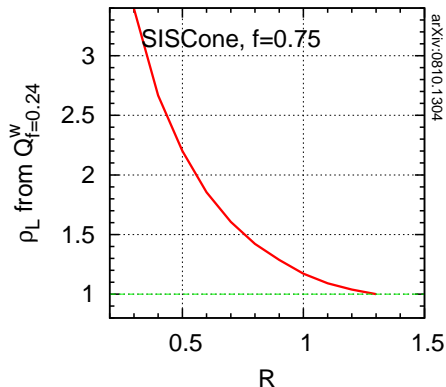
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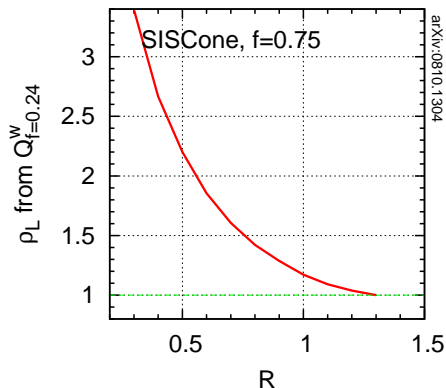
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$q\bar{q}, M = 4000 \text{ GeV}$



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Cacciari, Rojo, GPS & Soyez '08

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File Edit View History Bookmarks Tools Help

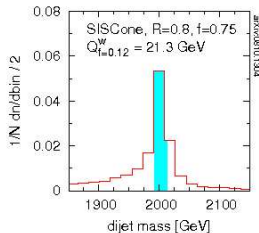
http://www.lpthe.jussieu.fr/~salam/jet-quality/

Testing jet definitions: qq & gg c...

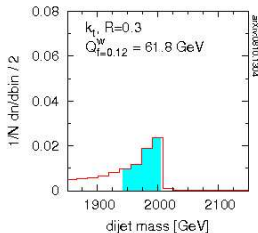
Testing jet definitions: qq & gg cases

by M. Cacciari, J. Rojo, G.P. Salam and G. Soyez, arXiv:0810.1304

qq, M = 2000 GeV



qq, M = 2000 GeV



This page is intended to help visualize how the choice of jet definition impacts a dijet invariant mass reconstruction at LHC.

The controls fall into 4 groups:

- the jet definition
- the binning and quality measures
- the jet-type (quark, gluon) and mass scale
- pileup and subtraction

The events were simulated with Pythia 6.4 (DWT tune) and reconstructed with FastJet 2.3.

For more information, view and listen to the **flash demo**, or click on individual terms.

This page has been tested with Firefox v2 and v3, IE7, Safari v3, Opera v9.5, Chrome 0.2.

Reset

 k_t C/A anti- k_t SIScone C/A-filt

 R = 0.8
 $Q_{f=z}^W$ $Q_{f=z}^{1/xV_M}$ x 2

 rebin = 2
 qq gg

 mass = 2000

pileup: none 0.05 0.25 mb^{-1}/ev

 k_t C/A anti- k_t SIScone C/A-filt

 R = 0.3
 $Q_{f=z}^W$ $Q_{f=z}^{1/xV_M}$ x 2

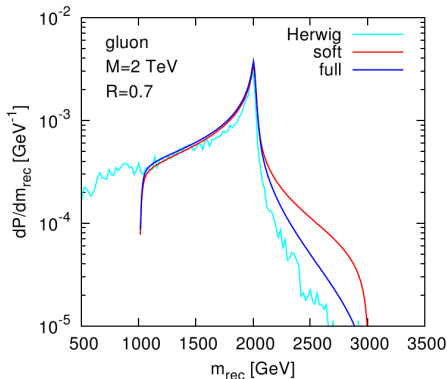
 rebin = 2
 qq gg

 mass = 2000

pileup: none 0.05 0.25 mb^{-1}/ev

Soyez '10

Analytic v. MC lineshape

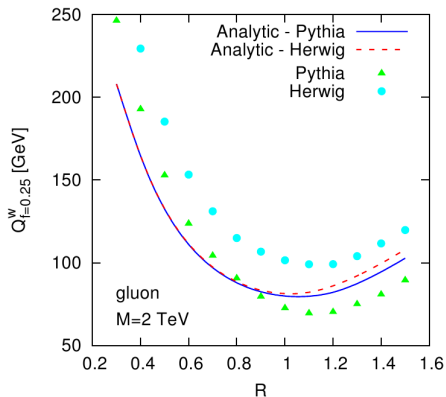


Perturbatively resum resonance “line-shape”, convolute with model for non-perturbative effects.

determine “quality” of line-shape from the analytic results, as a function of jet radius R

Soyez '10

Analytic v. MC quality

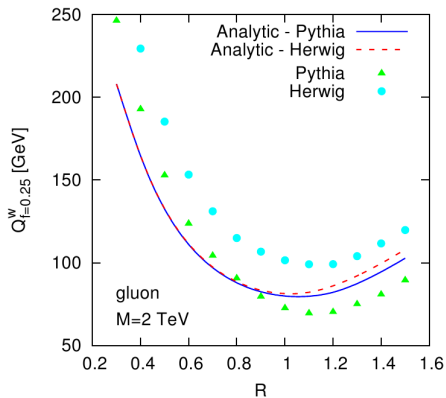


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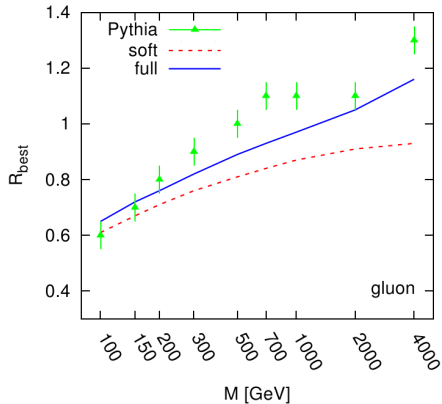
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Soyez '10

Analytic v. MC quality



Best R v. mass scale



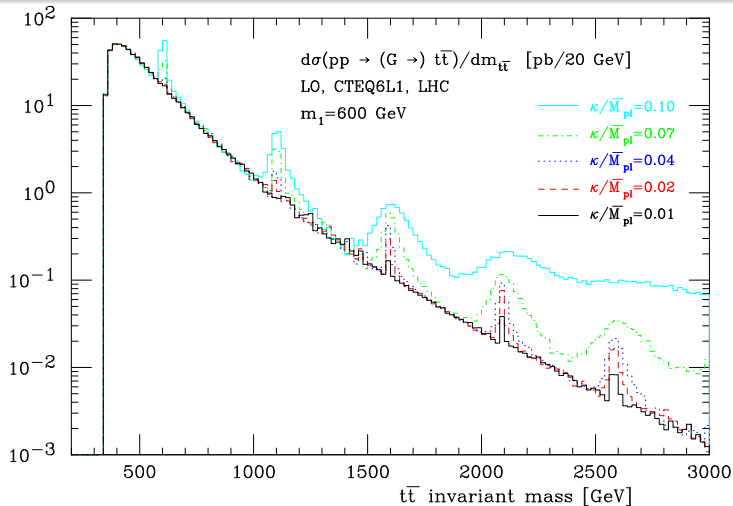
Fat jets

boosted massive hadronically decaying objects

E.g. when a known particle, W , Z or a top \rightarrow a single jet
or a new particle, Higgs, gluino, neutralino \rightarrow a single jet

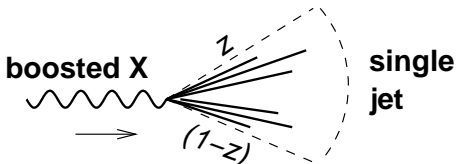
This will be common for electroweak-scale objects at LHC:

$$m_W, m_t \ll 14 \text{ TeV}$$

E.g. $X \rightarrow t\bar{t}$ resonances of varying difficulty

RS KK resonances $\rightarrow t\bar{t}$, from Frederix & Maltoni, 0712.2355

NB: QCD dijet spectrum is $\sim 10^3$ times $t\bar{t}$

Hadronically decaying EW boson at high $p_t \neq$ two jets

$$R \gtrsim \frac{m}{p_t} \frac{1}{\sqrt{z(1-z)}}$$

Rules of thumb:

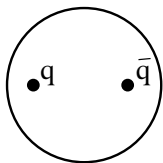
$$m = 100 \text{ GeV}, p_t = 500 \text{ GeV}$$

▶ $R < \frac{2m}{p_t}$: always resolve **two** jets

$$R < 0.4$$

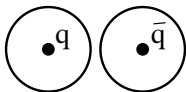
▶ $R \gtrsim \frac{3m}{p_t}$: resolve **one** jet in $\sim 75\%$ of cases ($\frac{1}{8} < z < \frac{7}{8}$)

$$R \gtrsim 0.6$$



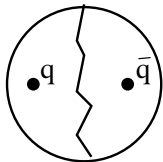
Select on the jet mass with one large (cone) jet

Can be subject to large bkgds
[high- p_t jets have significant masses]



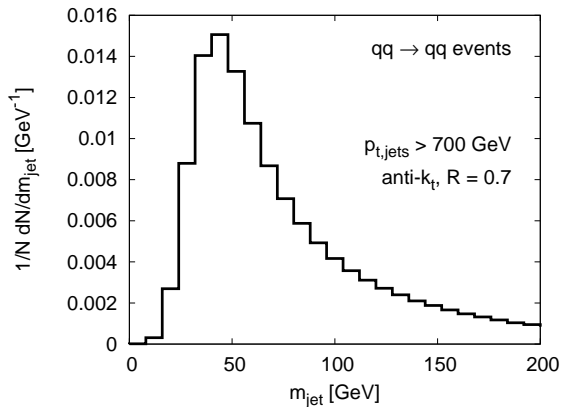
Choose a small jet size (R) so as to resolve two jets

Easier to reject background
if you actually see substructure
[NB: must manually put in “right” radius]



Take a large jet and split it in two

Let jet algorithm establish correct division

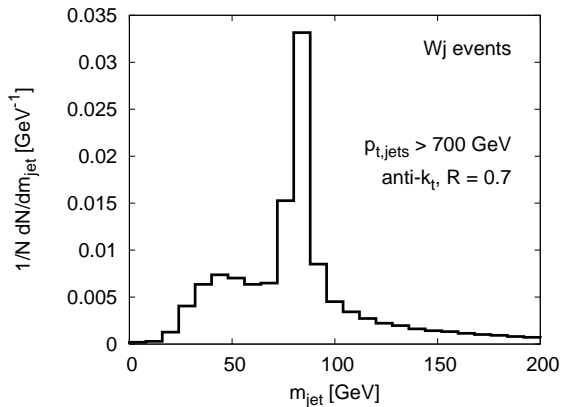


Look at jet mass distribution for two leading jets in

- ▶ $qq \rightarrow qq$ events
- ▶ $pp \rightarrow W + \text{jet}$ events
- ▶ a mixture of the two

In roughly sensible proportions

Jet mass gives clear sign of massive particles inside the jet;

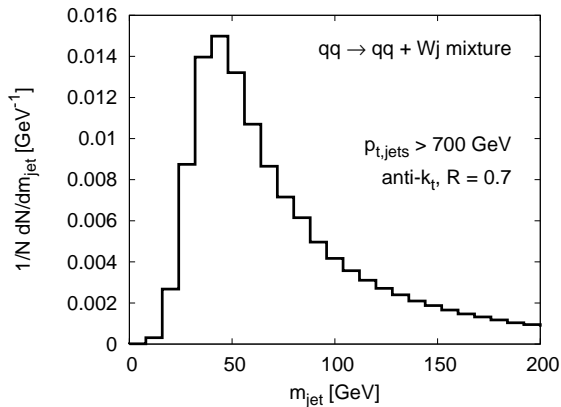


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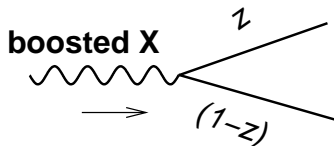


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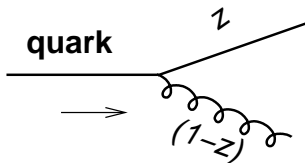
In roughly sensible proportions

Jet mass gives clear sign of massive particles inside the jet; but QCD jets are massive too — must learn to reject them

Signal

Splitting probability for Higgs:

$$P(z) \propto 1$$

Background

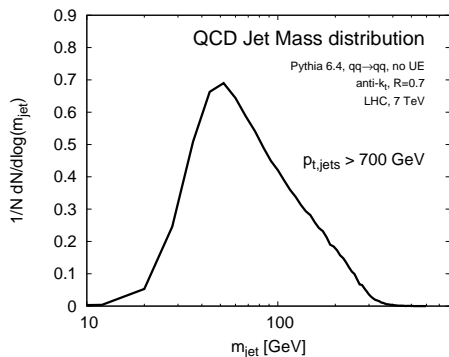
Splitting probability for quark:

$$P(z) \propto \frac{1+z^2}{1-z}$$

$1/(1-z)$ divergence enhances background

Remove divergence in bkdg with cut on z
 Can choose cut analytically so as to maximise S/\sqrt{B}

Originally: cut on (related) k_t -distance
 Butterworth, Cox & Forshaw '02



QCD jet mass distribution has the approximate

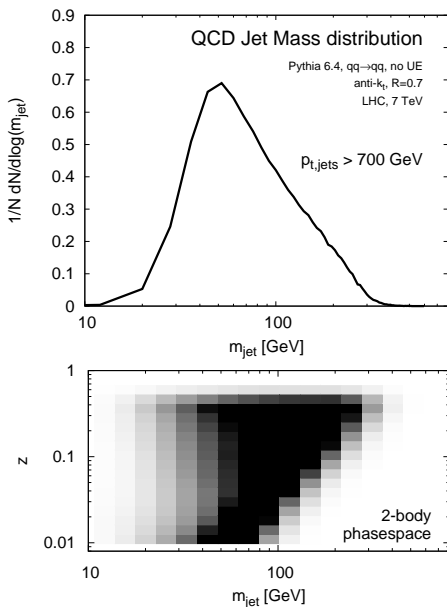
$$\frac{dN}{d \ln m} \sim \alpha_s \ln \frac{p_t R}{m} \times \text{Sudakov}$$

Work from '80s and '90s
+ Almeida et al '08

The logarithm comes from integral over soft divergence of QCD:

$$\int \frac{1}{2} \frac{dz}{z} \frac{m^2}{p_t^2 R^2}$$

A hard cut on z reduces QCD background & simplifies its shape



QCD jet mass distribution has the approximate

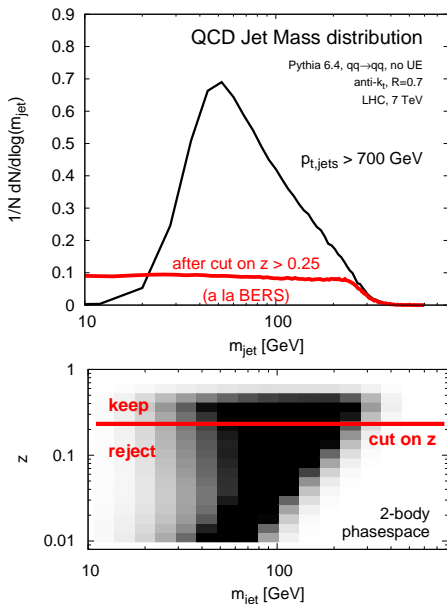
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QCD jet mass distribution has the approximate

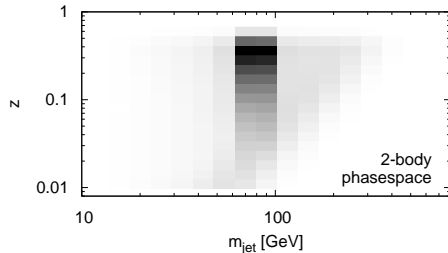
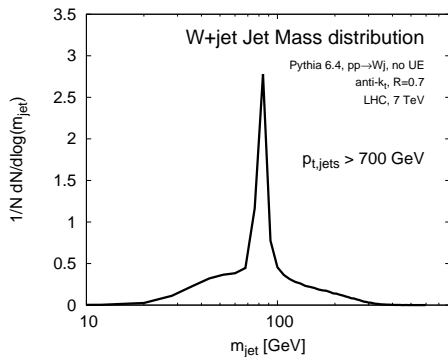
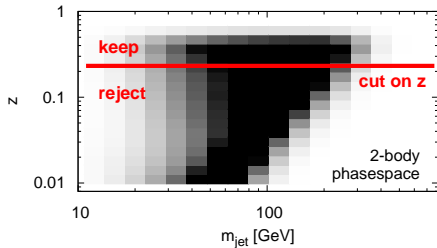
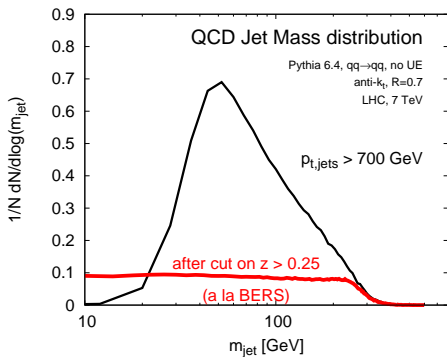
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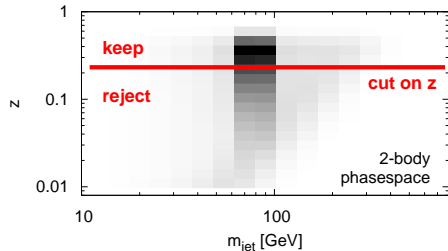
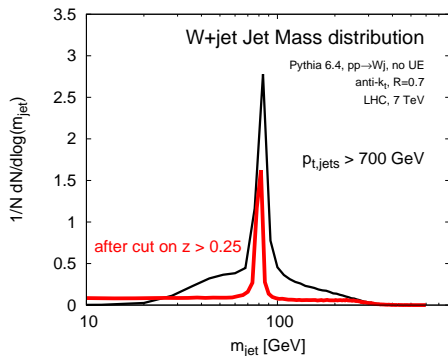
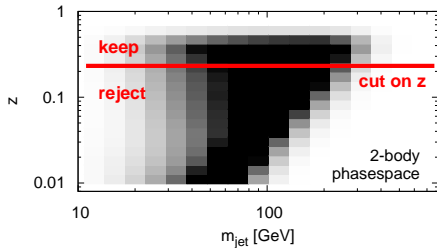
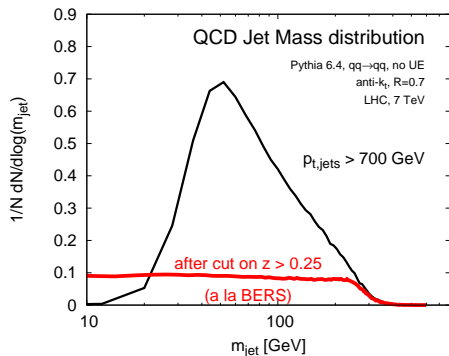
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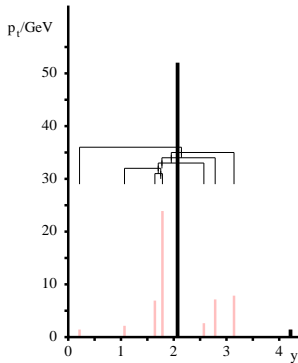
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A hard cut on z reduces QCD background & simplifies its shape

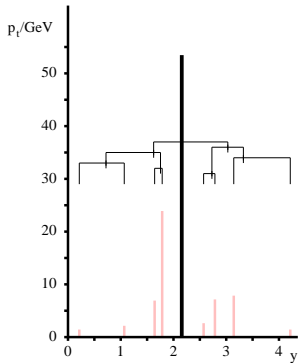




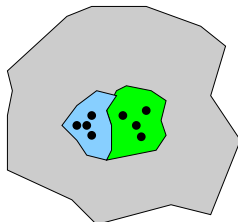
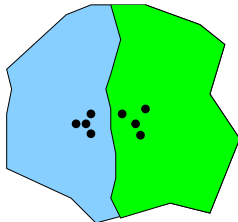
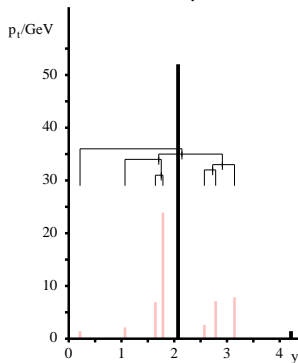
anti- k_t algorithm

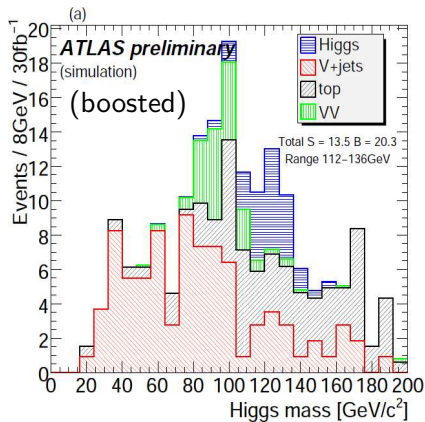
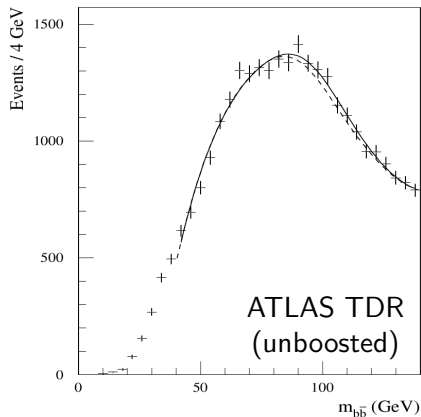


k_t algorithm



Cambridge/Aachen



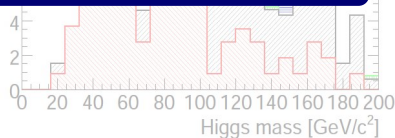
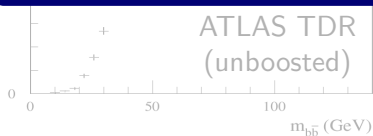
Search for main decay of light Higgs boson in $W/Z+H$, $H \rightarrow b\bar{b}$ 

restricting search to $p_{tH} > 200$ GeV
using the method from Butterworth, Davison, Rubin & GPS '08

Search for main decay of light Higgs boson in $W/Z+H$, $H \rightarrow b\bar{b}$ 

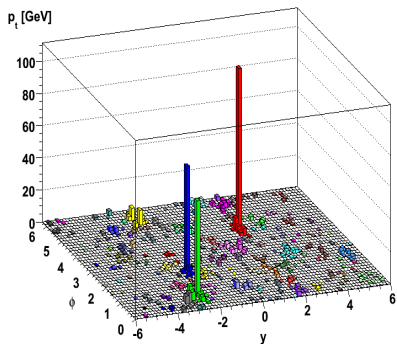
One of many applications of “boosted” searches, using variety of techniques, many involving jet substructure

See proceedings proceedings of Boost 2010 and talks at <http://boost2011.org>

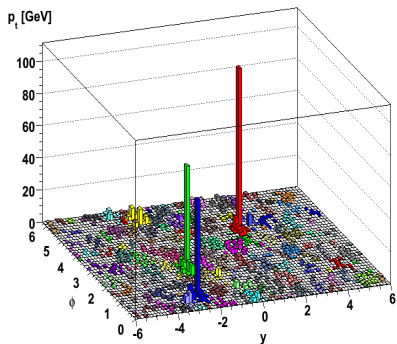


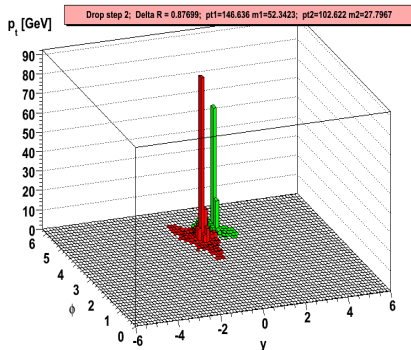
restricting search to $p_{tH} > 200$ GeV using the method from Butterworth, Davison, Rubin & GPS '08

Plain pythia event



Plain pythia event + 10 pileup





Key idea:

- ▶ Look at jet on smaller angular scale
- ▶ Discard its softer parts

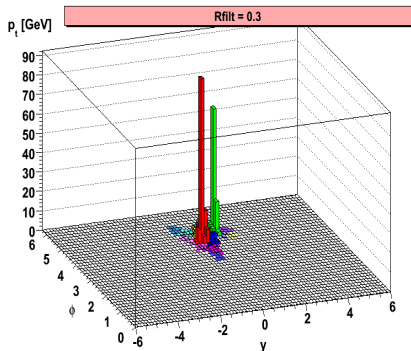
- ▶ Filtering
- ▶ Pruning
- ▶ Trimming

Butterworth et al '08

Ellis, Vermillion and Walsh '09

Krohn, Thaler & Wang '09

*[With earlier methods by Seymour '93 and Kodolova et al '07;
Rubin '10 for filtering optimisation; also Soper & Spannowsky '10, '11]*



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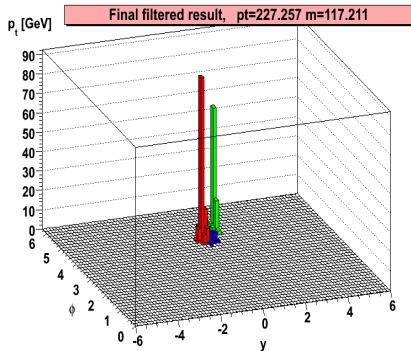
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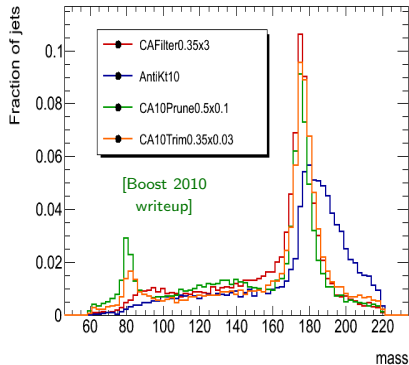
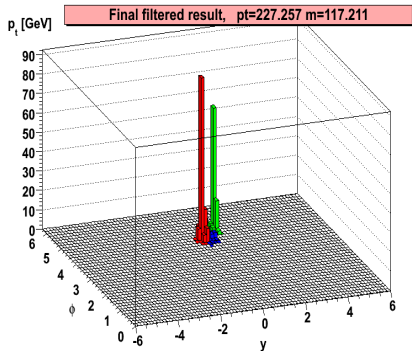
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- ▶ Filtering
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- ▶ Trimming

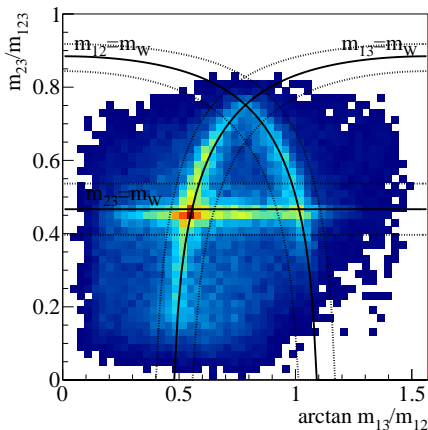
Butterworth et al '08

Ellis, Vermillion and Walsh '09

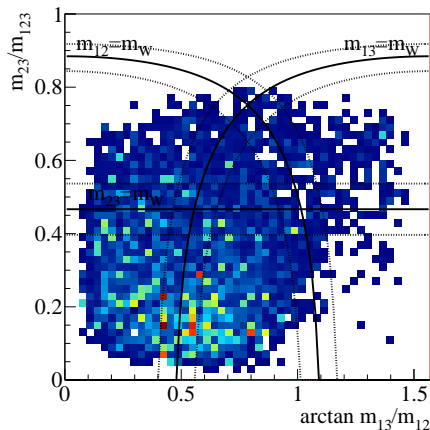
Krohn, Thaler & Wang '09

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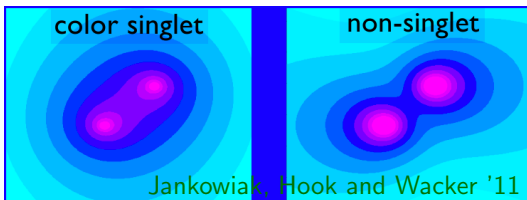
Top signal



QCD background



Using Dalitz-like plots to pull out the W **without using b tagging**
Plehn, Spannowsky, Takeuchi & Zerwas '10



Background (e.g. $g \rightarrow gg$) and signal (e.g. $W \rightarrow q\bar{q}$) often have different colour structure \rightarrow **different radiation patterns**.

- ▶ Pull (non-boosted context)
- ▶ N-subjettiness
- ▶ “Buried Higgs” light singlets
- ▶ Boosted decision trees
- ▶ Dipolarity, applied to HEPTopTagger
- ▶ Jet deconstruction
- ▶ Template method beyond LO
- ▶ ...

Gallicchio & Schwartz '10

Jihun Kim '10; Thaler & Van Tilburg '10

Falkowski et al '10; Chen et al '10

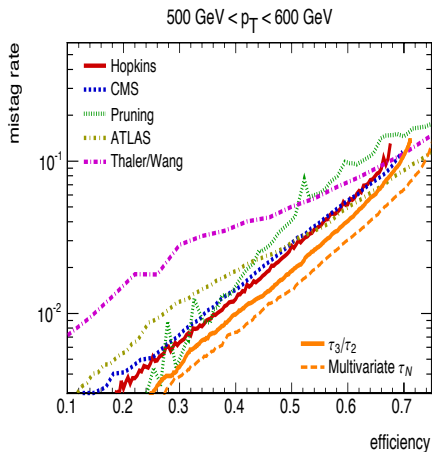
Cui & Schwartz '10

Jankowiak, Hook and Wacker '11

Soper & Spannowsky '11

Almeida et al '11

Jet shape variables (here for top tagging)



Thaler & van Tilburg '11
 cf. also J.-H. Kim '10 for Higgs

Early proposals include planar flow
 (3- v. 2-body structure of top decay)

Thaler & Wang '08

Almeida et al '08

Recent try: N -subjettiness. Break jet into subjets $1, \dots, N$

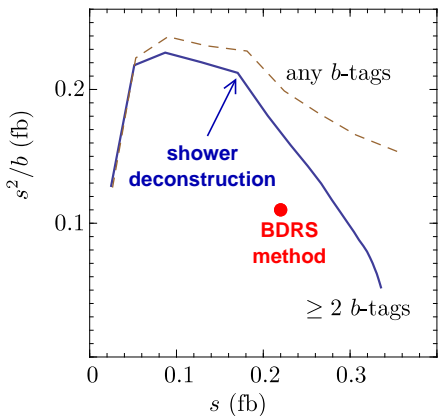
$$\tau_N = \frac{1}{p_{t,jet}} \sum_i p_{ti} \min(R_{i1}, \dots, R_{iN})$$

N -pronged decay: cut on mass &

$$\frac{\tau_N}{\tau_{N-1}}$$

Combines constraints on LO structure (energy sharing among prongs) and higher-order rad^n (from quarks in signal v. gluons in bkgd)

Matrix-element method on steroids



For each event estimate the probability that event is signal-like or background like.

Break event into many mini-jets; use Monte-Carlo type Sudakovs and splitting functions to get estimate of multi-parton matrix element for S & B hypotheses.

Intelligently combines full info about LO splitting, radiation, b-tags, etc.

Soper & Spannowsky '11

cf. also multivariate (BDT) type methods
from Cui & Schwartz '10

The study of jets is an increasingly broad subject.

Driven by the ubiquity of jets in LHC studies, the greater flexibility of LHC detectors relative to Tevatron ones, and by the broad dynamic range of LHC.

[jets from 20 GeV to several TeV — from below to far above EW scale;
UE \sim 1 GeV per unit area, pileup 10 – 20 GeV, HIC 100 – 300 GeV]

Many basic ideas ideas have been worked out.

But intense recent activity suggests that there is still scope for significant further progress.

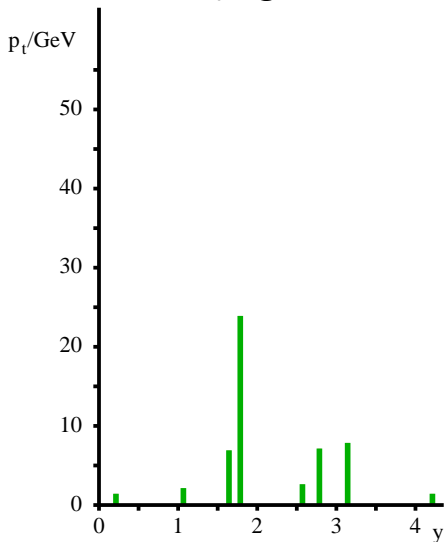
Supplementary material

	k_t	Cam/Aachen	anti- k_t	SISCone
reach	R	R	R	$(1 + \frac{p_{t2}}{p_{t1}})R$
$\Delta p_{t,PT} \simeq \frac{\alpha_s C_i}{\pi} \times$	$\ln R$	$\ln R$	$\ln R$	$\ln 1.35R$
$\Delta p_{t,hadr} \simeq -\frac{0.4 \text{ GeV} C_i}{R} \times$	0.7	?	1	?
area = $\pi R^2 \times$	0.81 ± 0.28	0.81 ± 0.26	1	0.25
$+ \pi R^2 \frac{C_i}{\pi b_0} \ln \frac{\alpha_s(Q_0)}{\alpha_s(Rp_t)} \times$	0.52 ± 0.41	0.08 ± 0.19	0	0.12 ± 0.07

In words:

- ▶ k_t : area fluctuates a lot, depends on p_t (bad for UE)
- ▶ Cam/Aachen: area fluctuates somewhat, depends less on p_t
- ▶ anti- k_t : area is constant (circular jets)
- ▶ SISCone: reaches far for hard radiation (good for resolution, bad for multijets), area is smaller (good for UE)

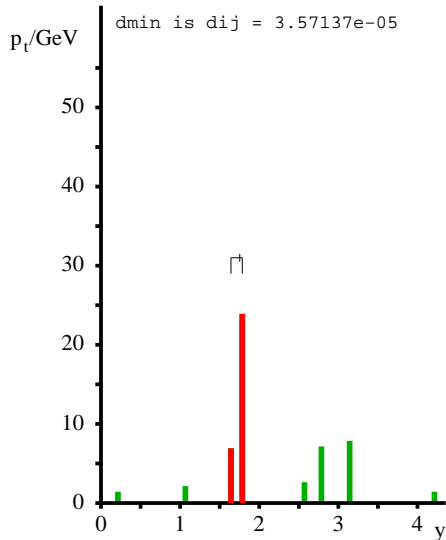
anti- k_t algorithm



How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

This is crucial for identifying the kinematic variables of the partons in the jet (e.g. z).

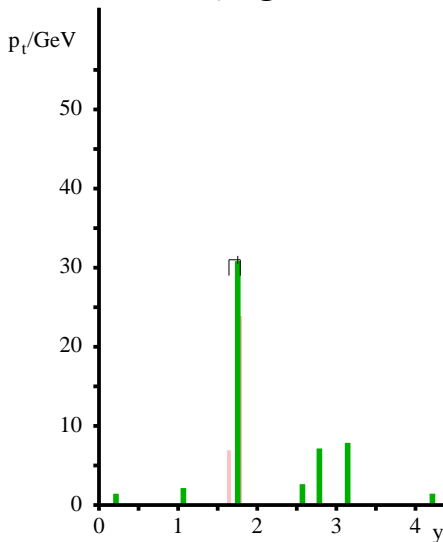
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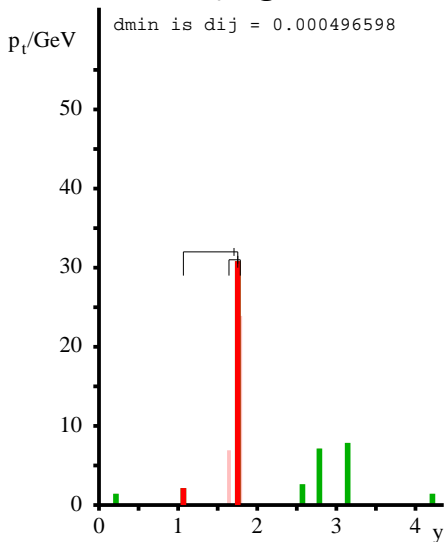
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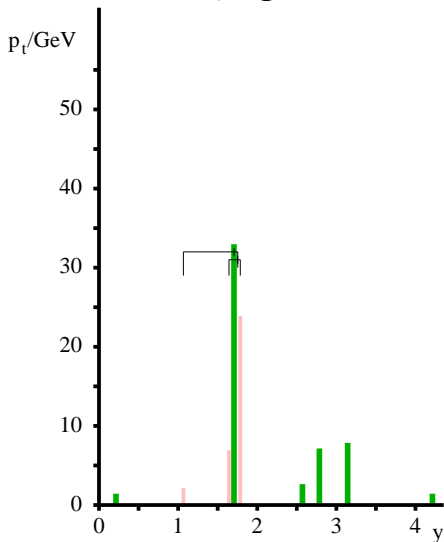
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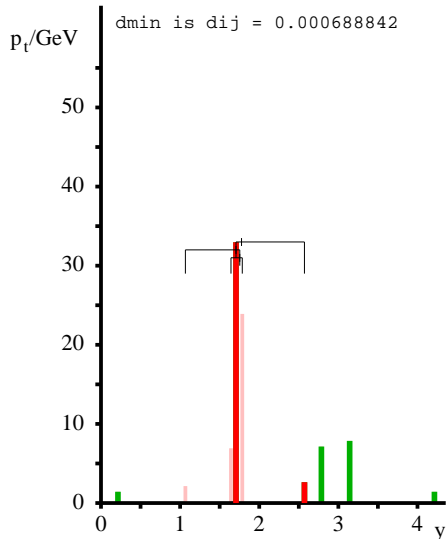
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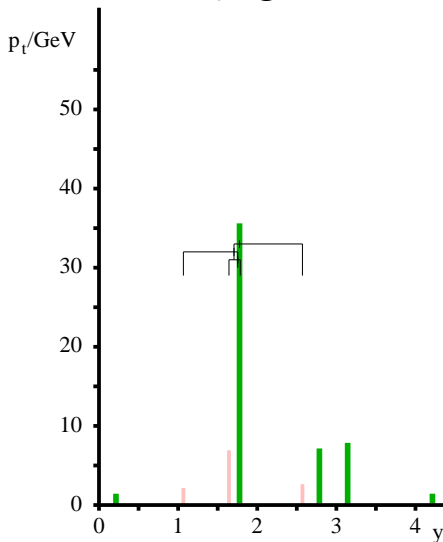
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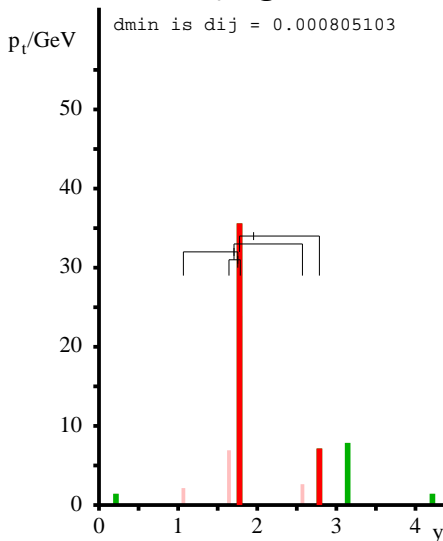
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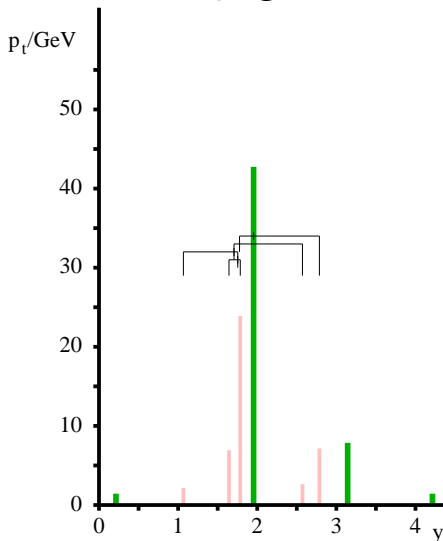


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Anti- k_t gradually makes its way through the secondary blob \rightarrow no clear identification of substructure associated with 2nd parton.

anti- k_t algorithm

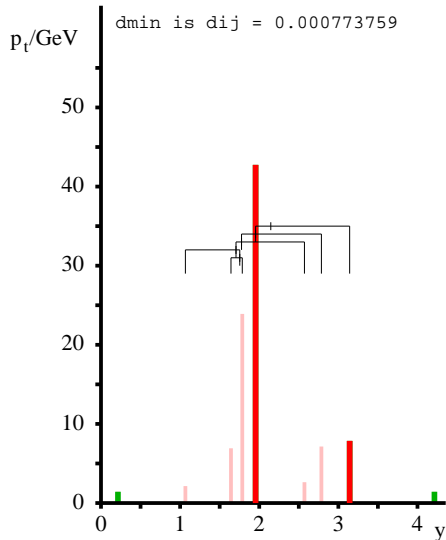


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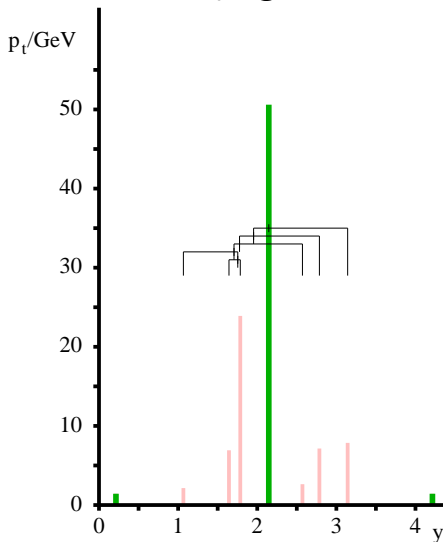


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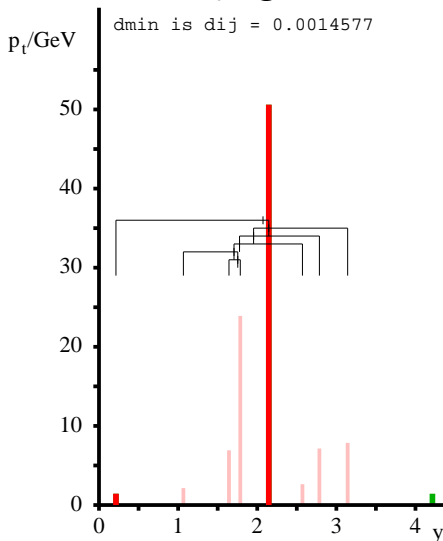


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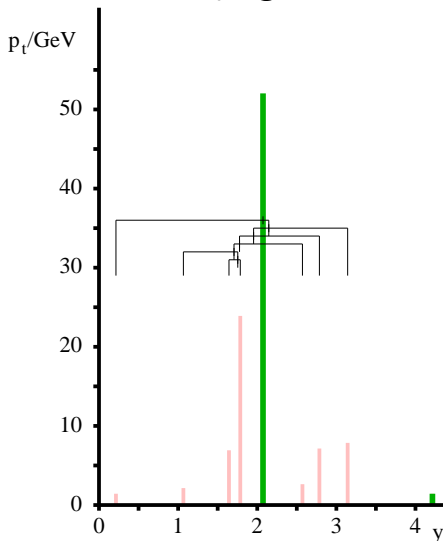


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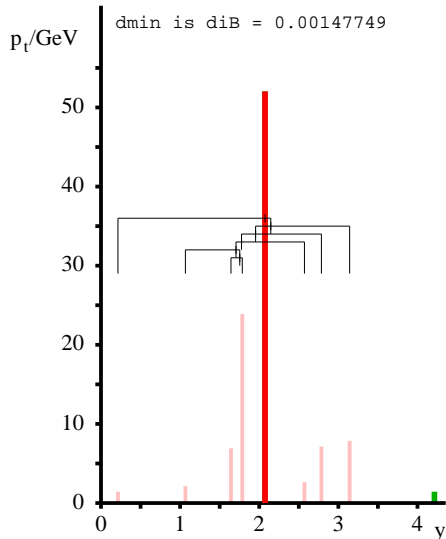


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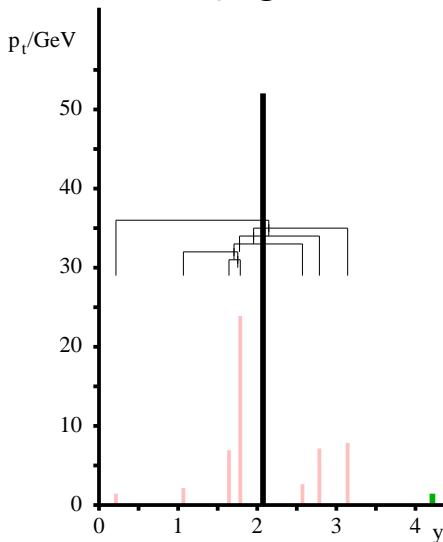


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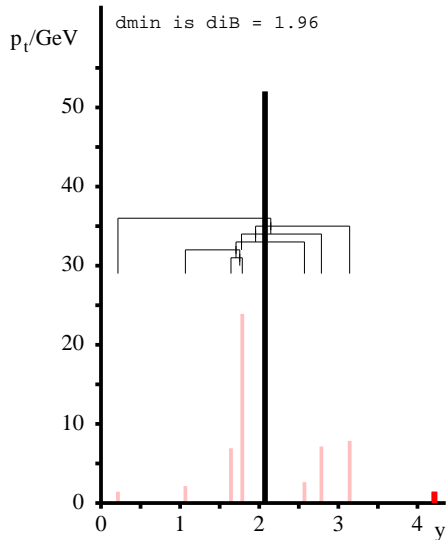


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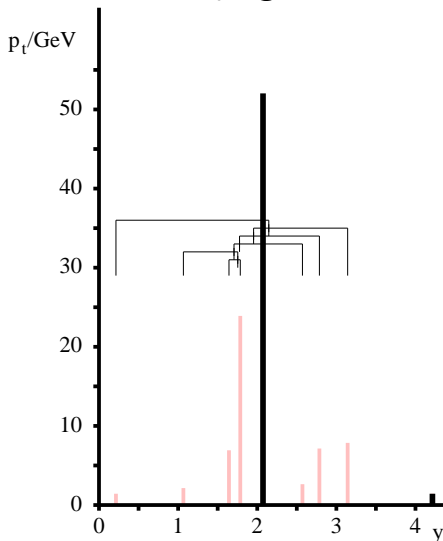


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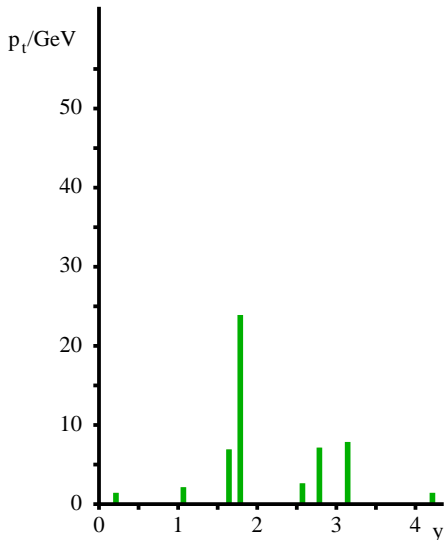


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k_t algorithm

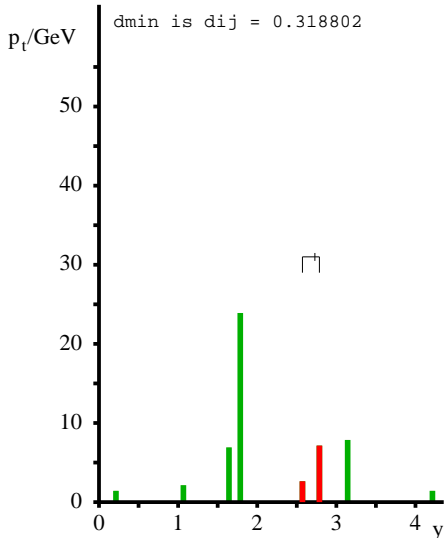


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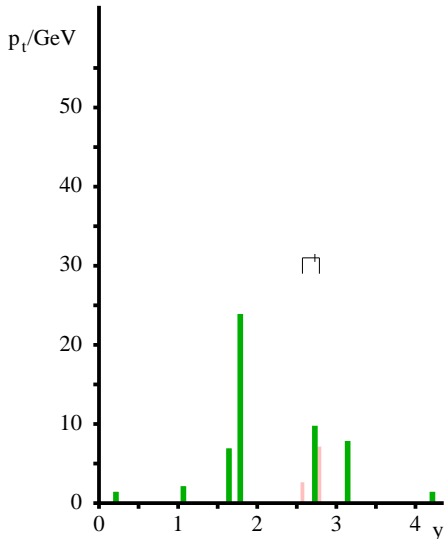
d_{\min} is $d_{ij} = 0.318802$



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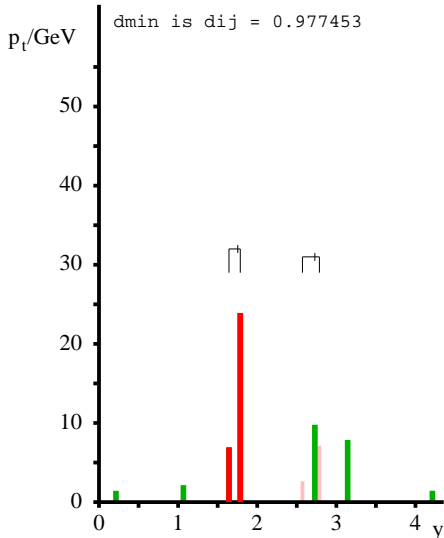


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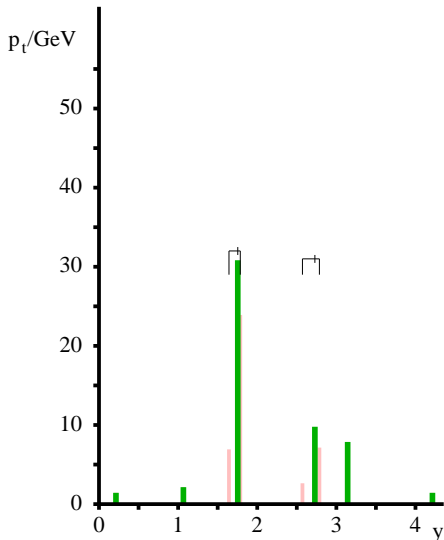
d_{\min} is $d_{ij} = 0.977453$



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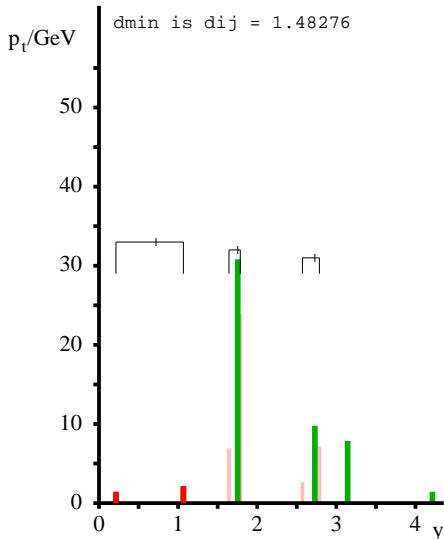


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k_t algorithm

d_{\min} is $d_{ij} = 1.48276$

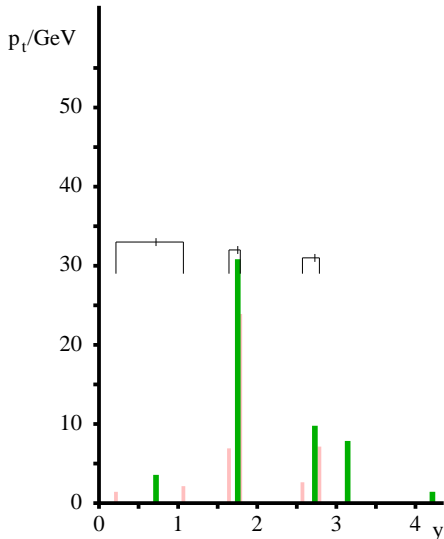


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k_t clusters soft “junk” early on in the clustering

k_t algorithm



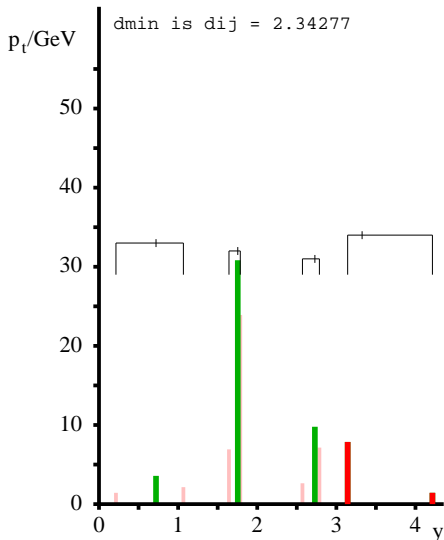
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d_{\min} is $d_{ij} = 2.34277$

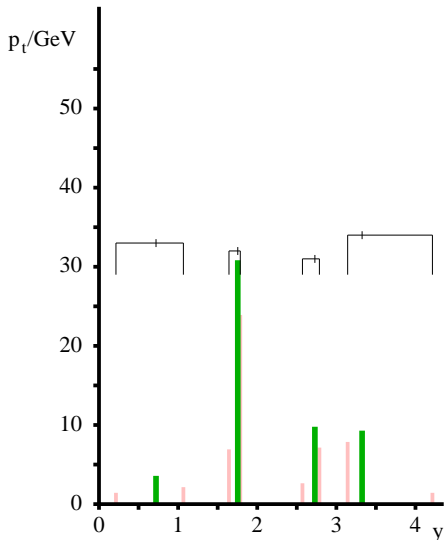


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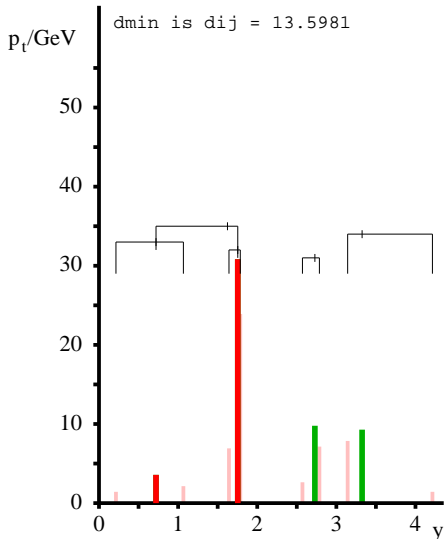
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d_{\min} is $d_{ij} = 13.5981$

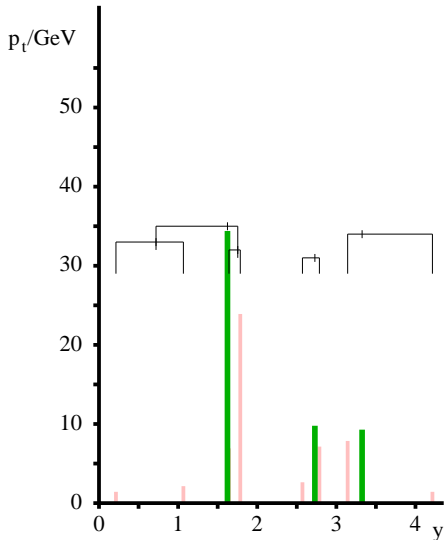


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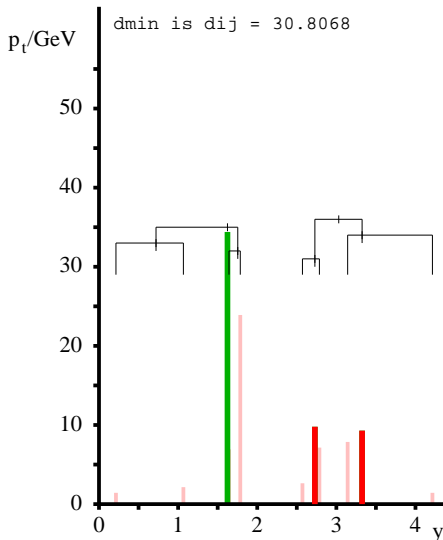
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d_{\min} is $d_{ij} = 30.8068$

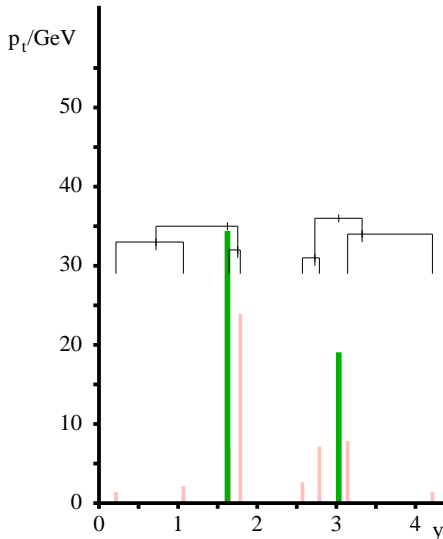


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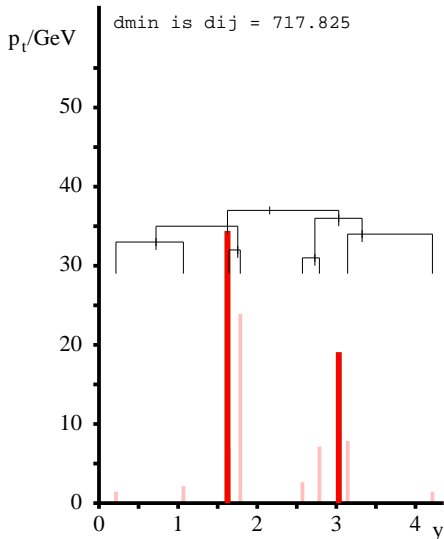
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d_{\min} is $d_{ij} = 717.825$



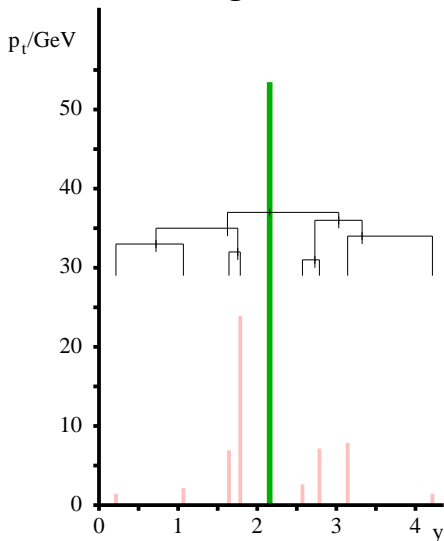
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Its last step is to merge two hard pieces. Easily undone to identify underlying kinematics

k_t algorithm

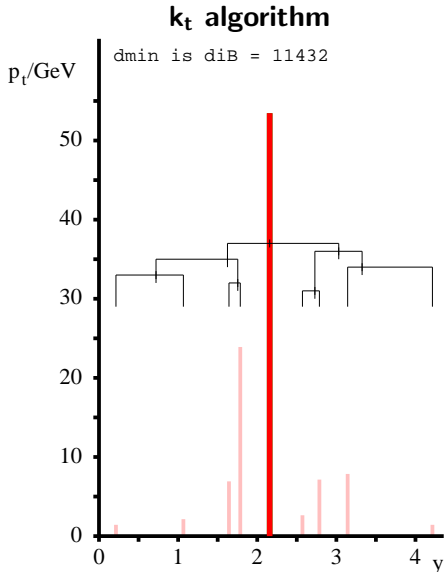


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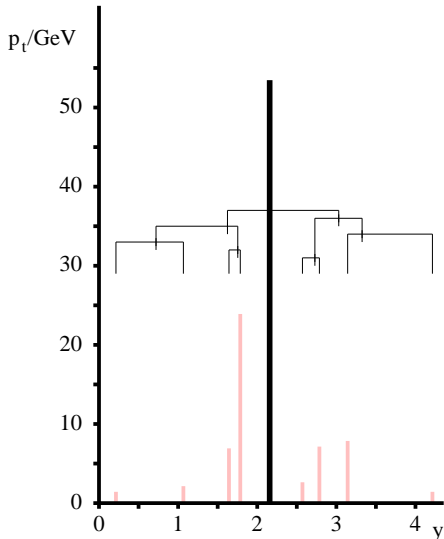
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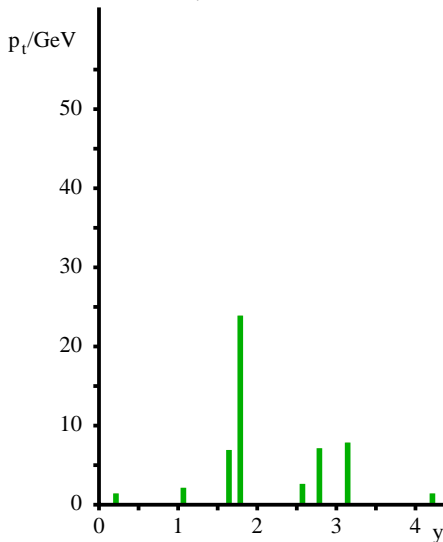
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This meant it was the first algorithm to be used for jet substructure.

Seymour '93

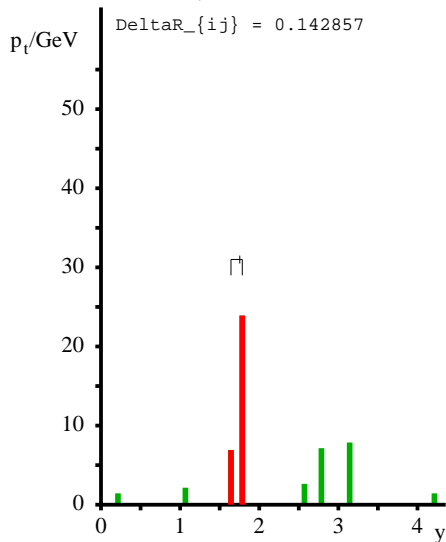
Butterworth, Cox & Forshaw '02

Cambridge/Aachen algorithm



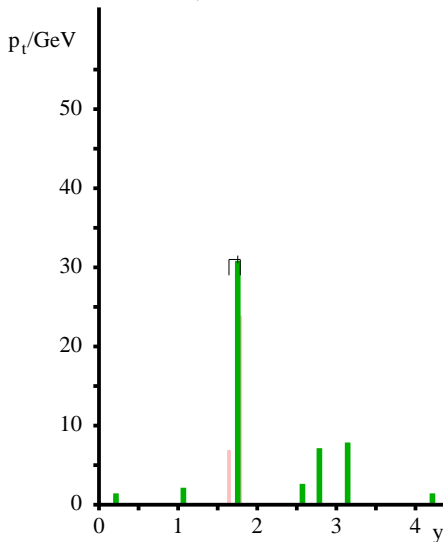
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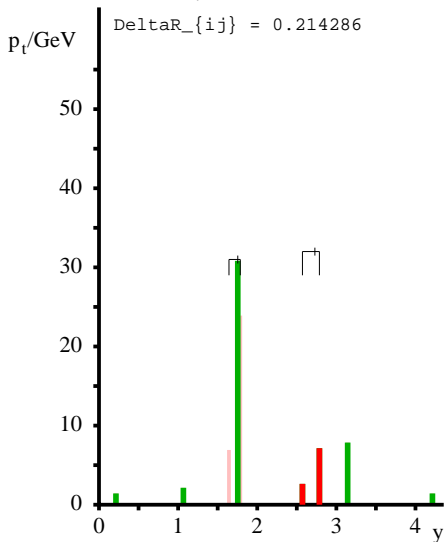
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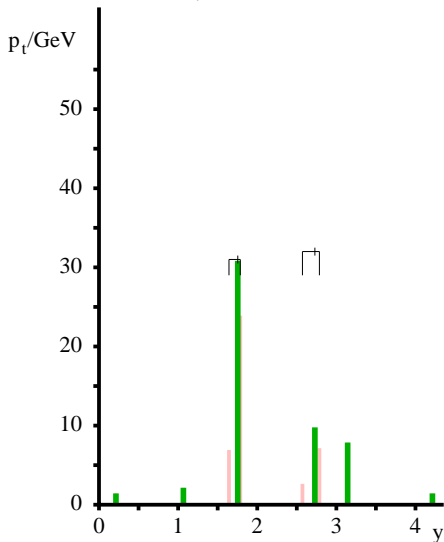
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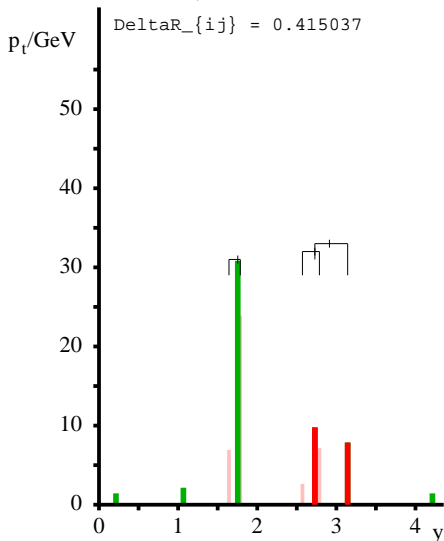
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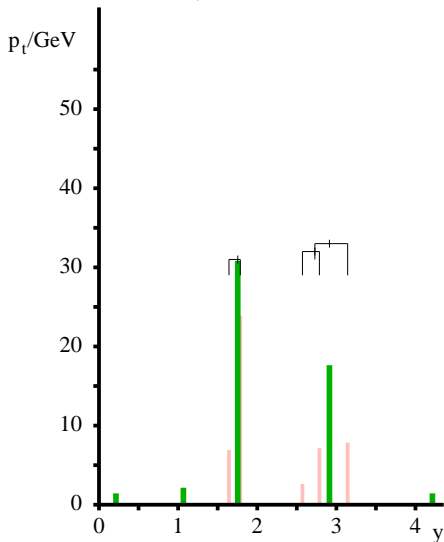
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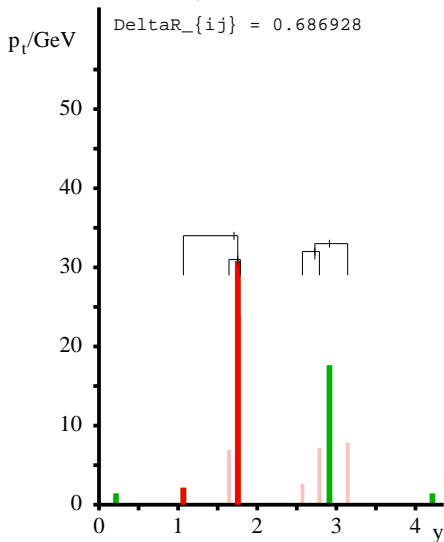
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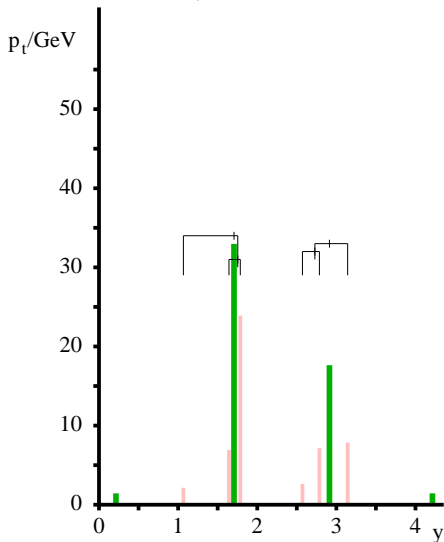
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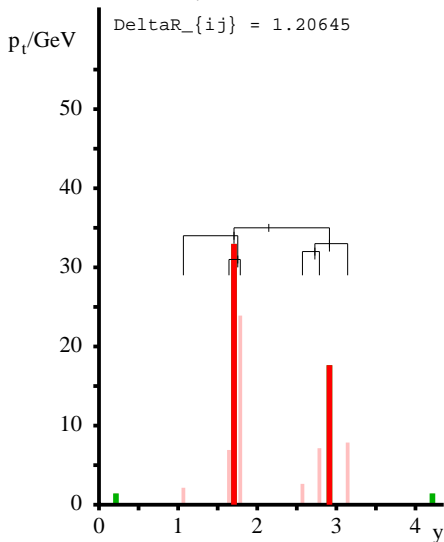
Cambridge/Aachen algorithm



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C/A identifies two hard blobs with limited soft contamination

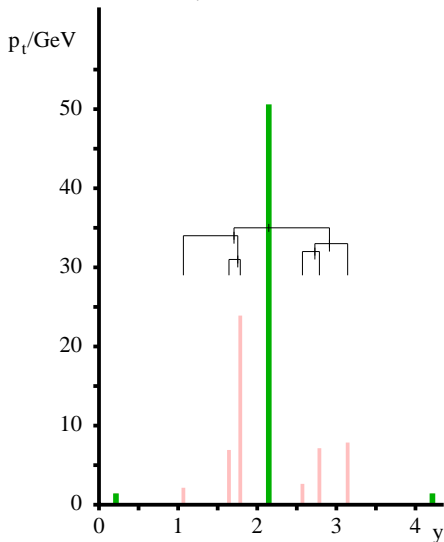
Cambridge/Aachen algorithm



How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

C/A identifies two hard blobs with limited soft contamination, **joins them**

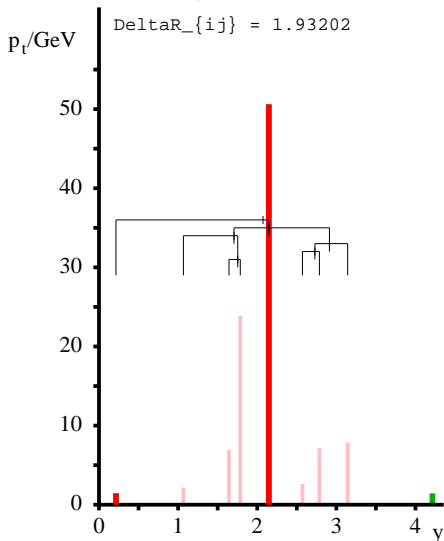
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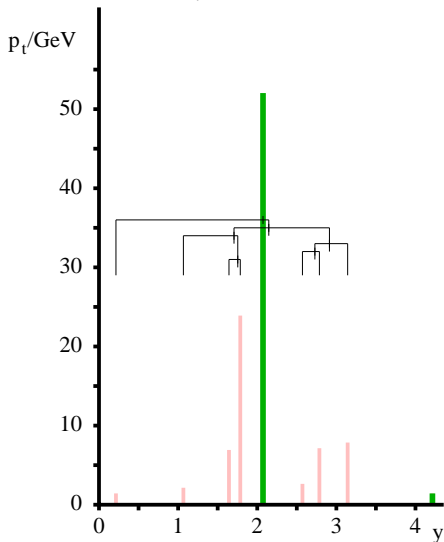
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How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

C/A identifies two hard blobs with limited soft contamination, joins them, and then adds in remaining soft junk

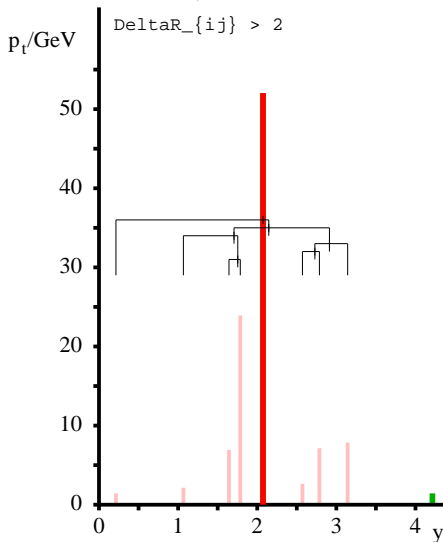
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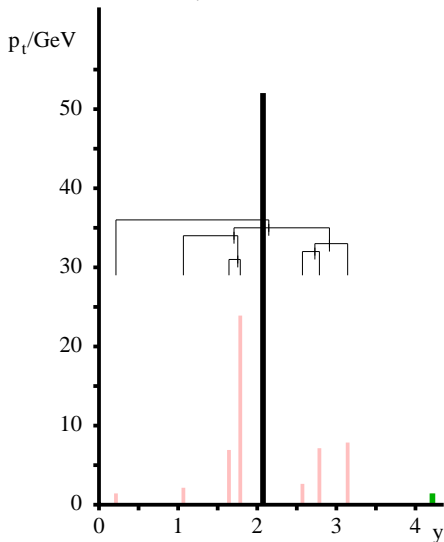
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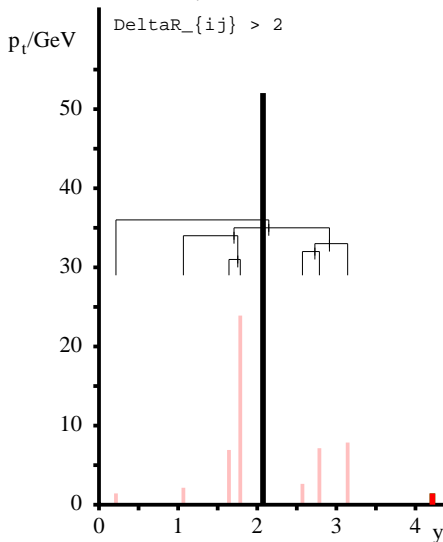
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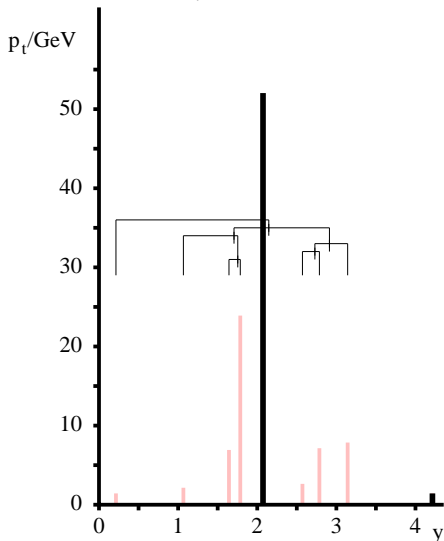
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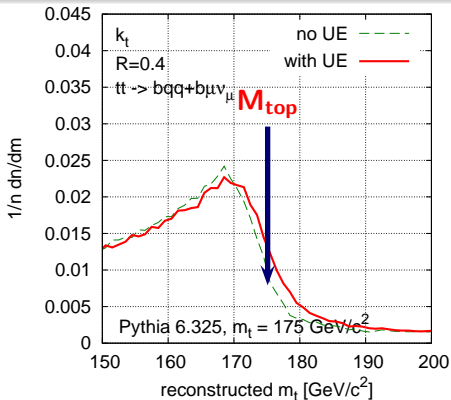
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The interesting substructure is buried inside the clustering sequence — **it's less contaminated by soft junk, but needs to be pulled out with special techniques**

Butterworth, Davison, Rubin & GPS '08
Kaplan, Schwartz, Reherman & Tweedie '08
Butterworth, Ellis, Rubin & GPS '09
Ellis, Vermilion & Walsh '09

A more complex example: top reconstruction



Game: measure top mass to 1 GeV

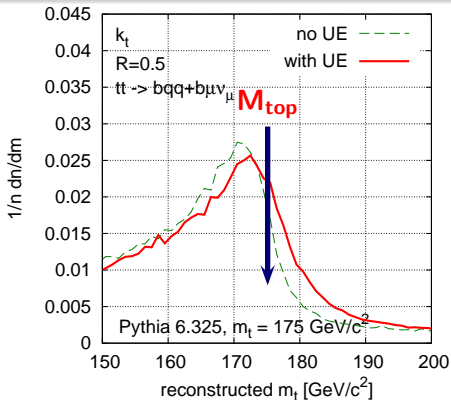
example for Tevatron
 $m_t = 175 \text{ GeV}$

▶ Small R : lose 6 GeV to PT radiation and hadronisation, UE and pileup irrelevant

▶ Large R : hadronisation and PT radiation leave mass at $\sim 175 \text{ GeV}$, UE adds 2 – 4 GeV

Is the final top mass (after W jet-energy-scale and Monte Carlo unfolding) independent of R used to measure jets?

Flexibility in jet finding gives powerful cross-check of systematic effects
 cf. Seymour & Tevlin '06



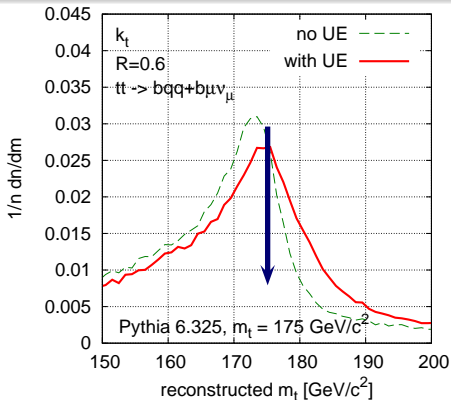
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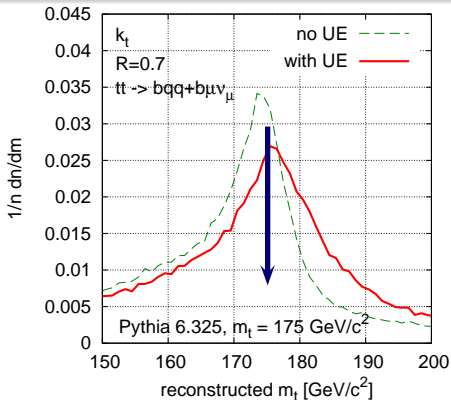
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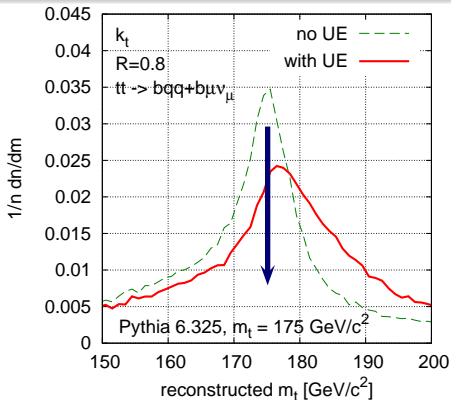
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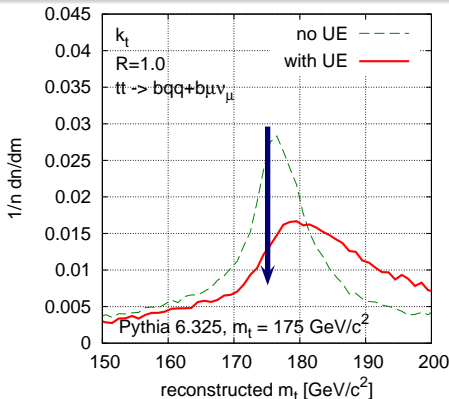
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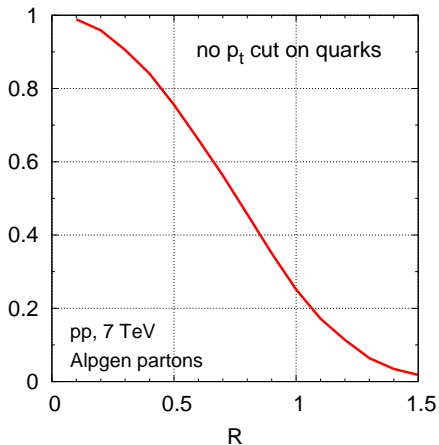
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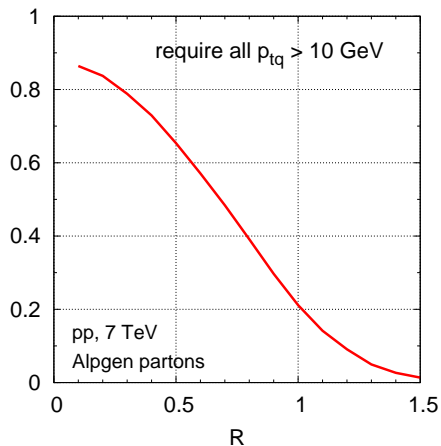
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fraction of $pp \rightarrow t\bar{t} \rightarrow 6q$ events with all $R_{qq} > R$



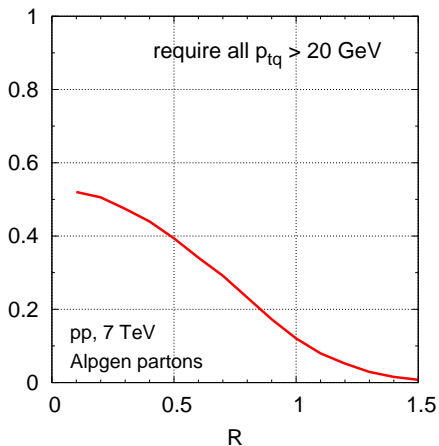
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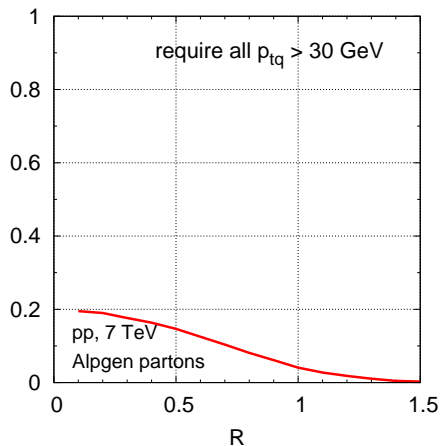
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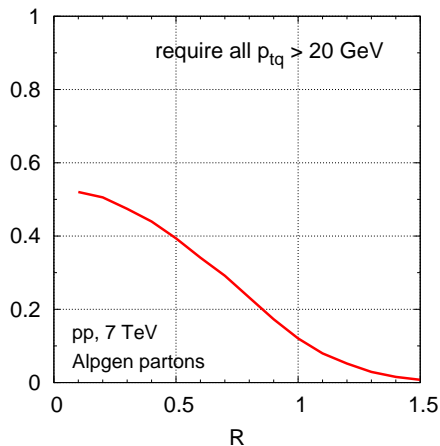
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Herwig $pp \rightarrow t\bar{t} \rightarrow \text{hadrons}$

Distribution of number of jets

