

QCD, Higher Orders and Jets

Matteo Cacciari

LPTHE Paris and Université de Paris

Lecture 3: jet algorithms and jet substructure

Includes material from
Gavin Salam and Grégory Soyez

▶ Jet algorithms

- ▶ How are jets made

▶ Jet substructure

- ▶ What's inside them

An observable is **infrared and collinear safe** if, in the limit of a **collinear splitting**, or the **emission of an infinitely soft** particle, the observable remains **unchanged**:

$$O(X; p_1, \dots, p_n, p_{n+1} \rightarrow 0) \rightarrow O(X; p_1, \dots, p_n)$$

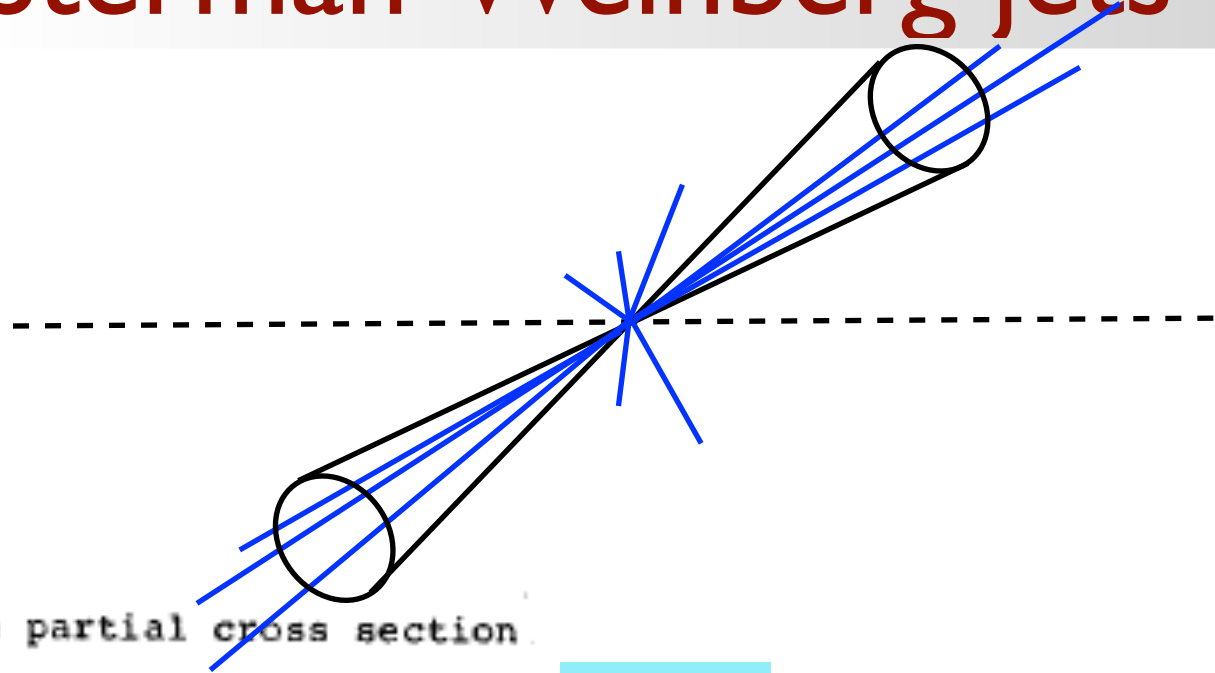
$$O(X; p_1, \dots, p_n \parallel p_{n+1}) \rightarrow O(X; p_1, \dots, p_n + p_{n+1})$$

This property ensures cancellation of **real** and **virtual** divergences in higher order calculations

If we wish to be able to calculate a jet rate in perturbative QCD the jet algorithm that we use must be IRC safe:
soft emissions and collinear splittings must not change the hard jets

Sterman-Weinberg jets

The first rigorous definition of an **infrared and collinear safe** jet in QCD is due to Sterman and Weinberg, Phys. Rev. Lett. **39**, 1436 (1977):

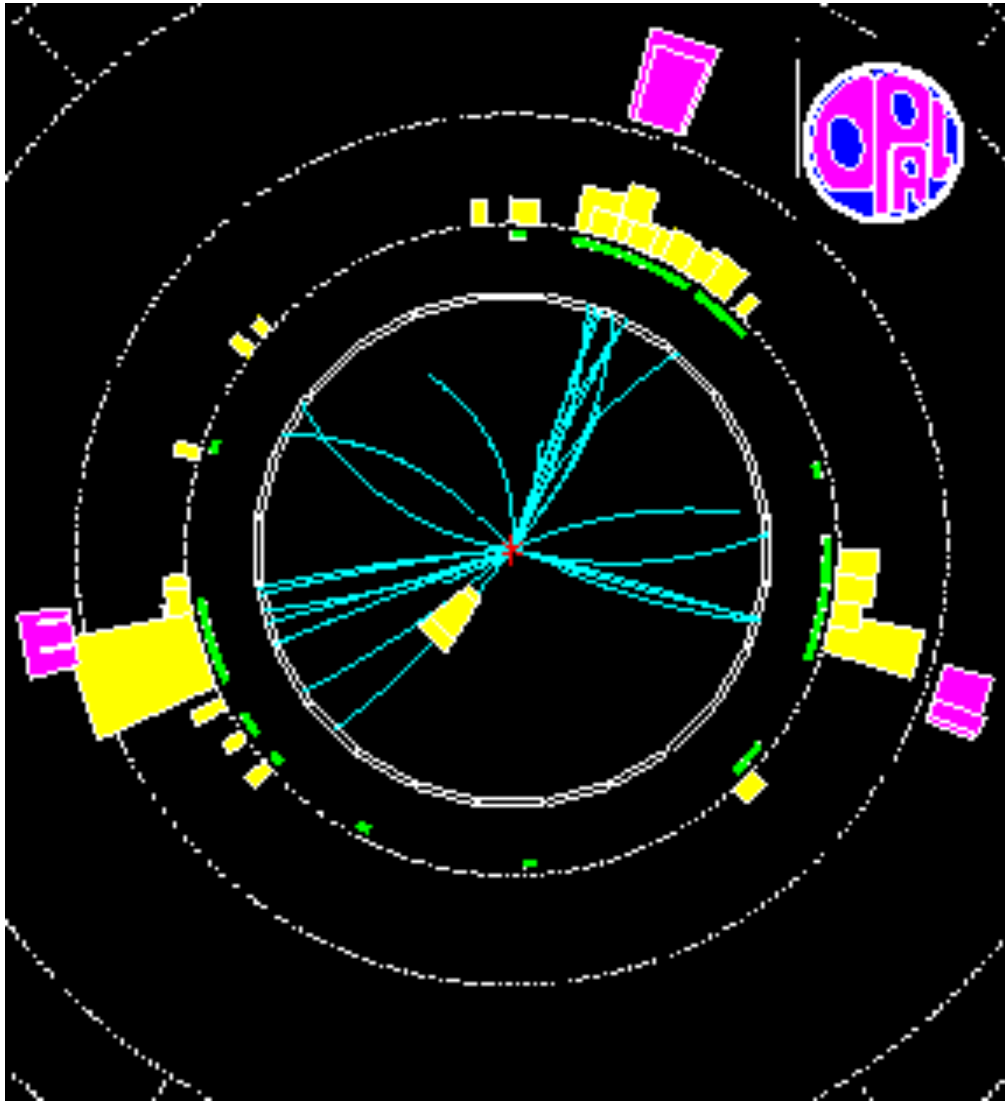


To study jets, we consider the partial cross section $\sigma(E, \theta, \Omega, \epsilon, \delta)$ for e^+e^- hadron production events, in which all but a fraction $\epsilon \ll 1$ of the total e^+e^- energy E is emitted within some pair of oppositely directed cones of half-angle $\delta \ll 1$, lying within two fixed cones of solid angle Ω (with $\pi\delta^2 \ll \Omega \ll 1$) at an angle θ to the e^+e^- beam line. We expect this to be measur-

$$\sigma(E, \theta, \Omega, \epsilon, \delta) = (d\sigma/d\Omega)_0 \Omega \left[1 - (g_E^2/3\pi^2) \left\{ 3\ln \delta + 4\ln \delta \ln 2\epsilon + \frac{\pi^3}{3} - \frac{5}{2} \right\} \right]$$

Calculable in pQCD (here is the result) but notice the soft and collinear large logs

Why jets

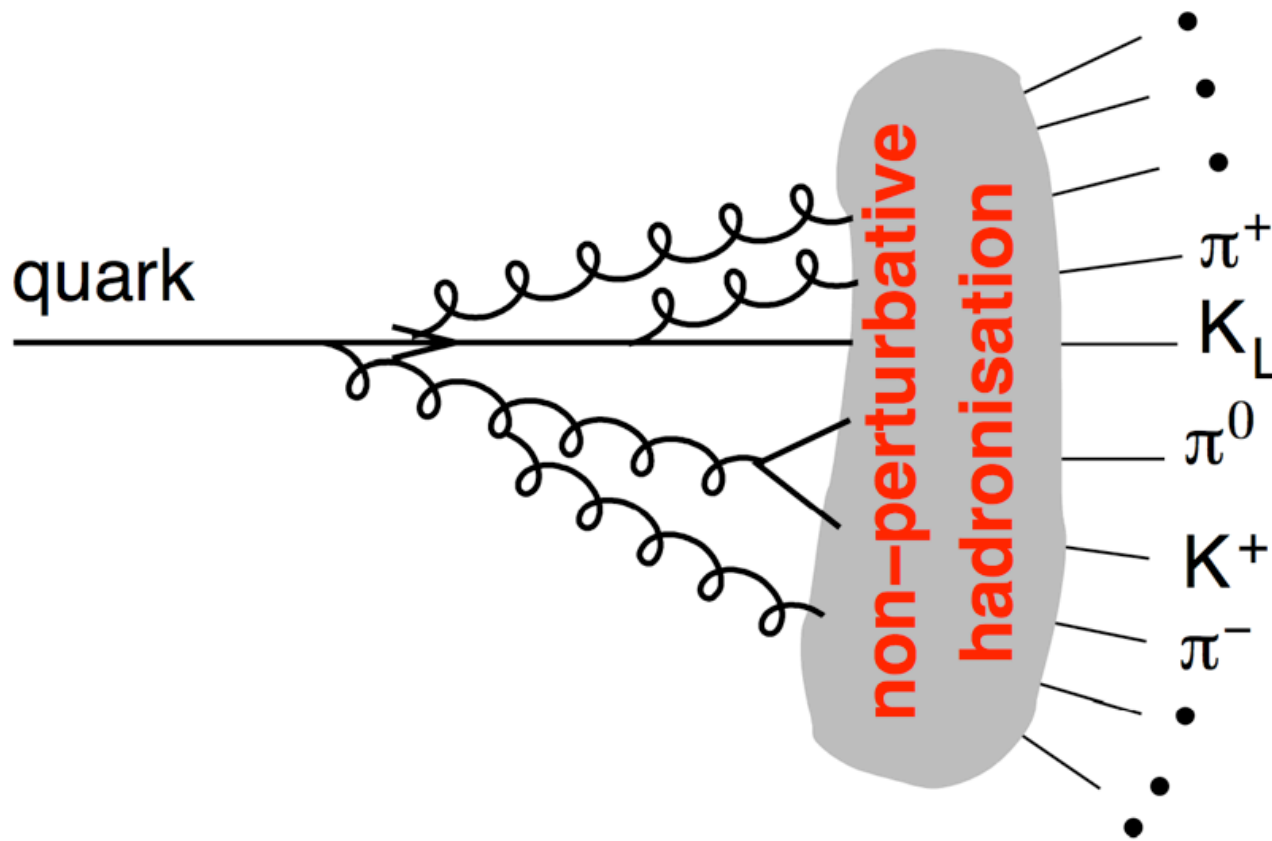


A **jet** is something that happens in high energy events: **a collimated bunch of hadrons flying roughly in the same direction**

We could eyeball the collimated bunches, but it becomes impractical with millions of events

The classification of particles into jets is best done using a **clustering algorithm**

Why do jets happen?



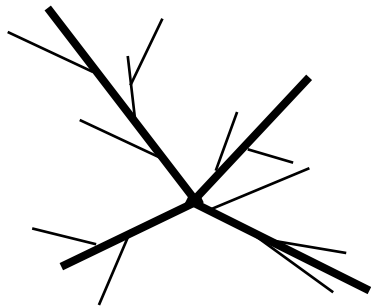
Gluon emission

$$\int \alpha_s \frac{dE}{E} \frac{d\theta}{\theta} \gg 1$$

Non-perturbative physics

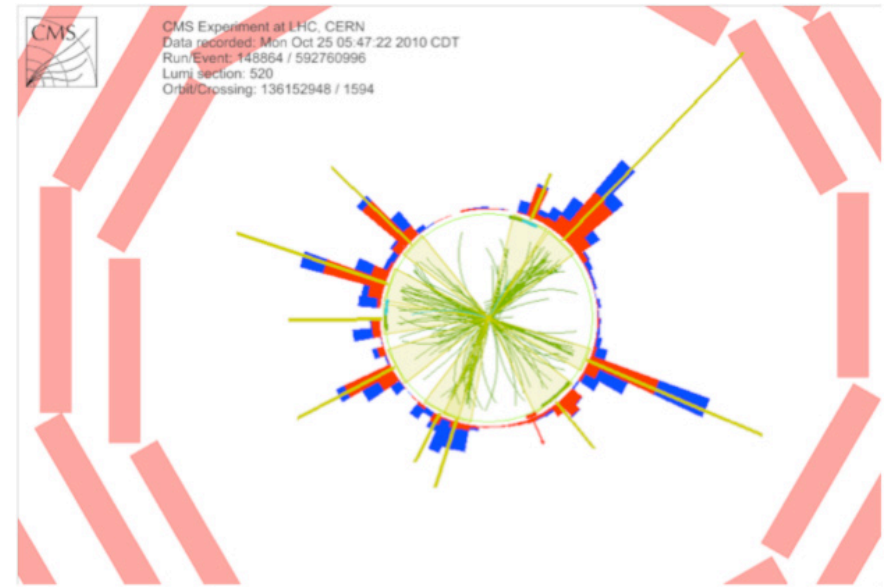
$$\alpha_s \sim 1$$

Multileg + PS



QCD predictions

??



Real data

Jets

One purpose of a 'jet clustering' algorithm is to **reduce the complexity** of the final state, simplifying many hadrons to **simpler objects** that one can hope to **calculate**

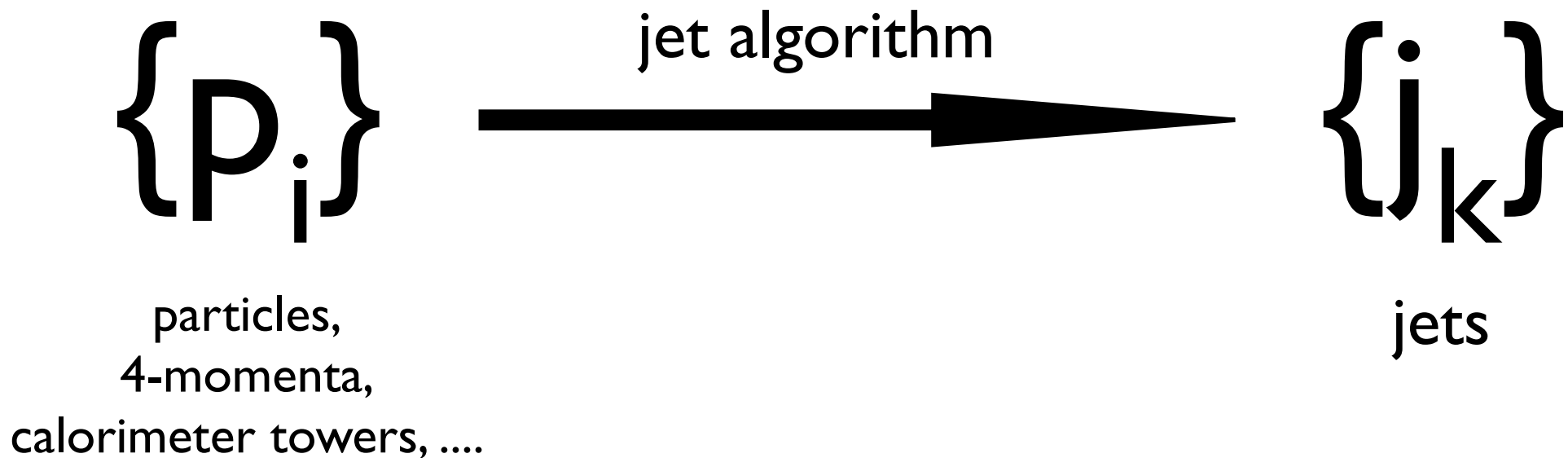
Jets can serve **two** purposes

- ▶ They can be **observables**, that one can measure and calculate
- ▶ They can be **tools**, that one can employ to extract specific properties of the final state

Different clustering algorithms have different properties and characteristics that can make them more or less appropriate for each of these tasks

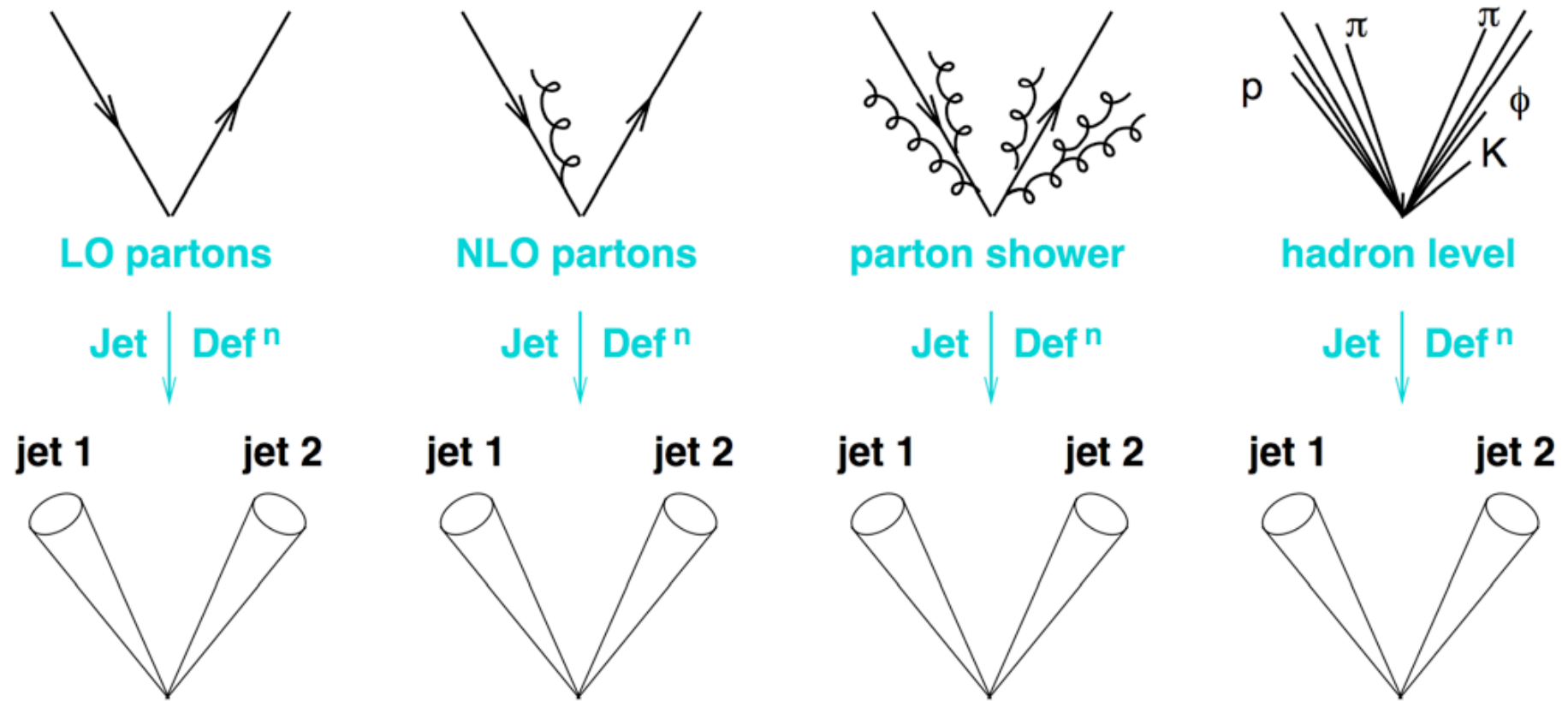
Jet clustering algorithm

A **jet algorithm** maps the momenta of the final state particles into the momenta of a certain number of jets:



Most algorithms contain a resolution parameter, **R**, which controls the extension of the jet

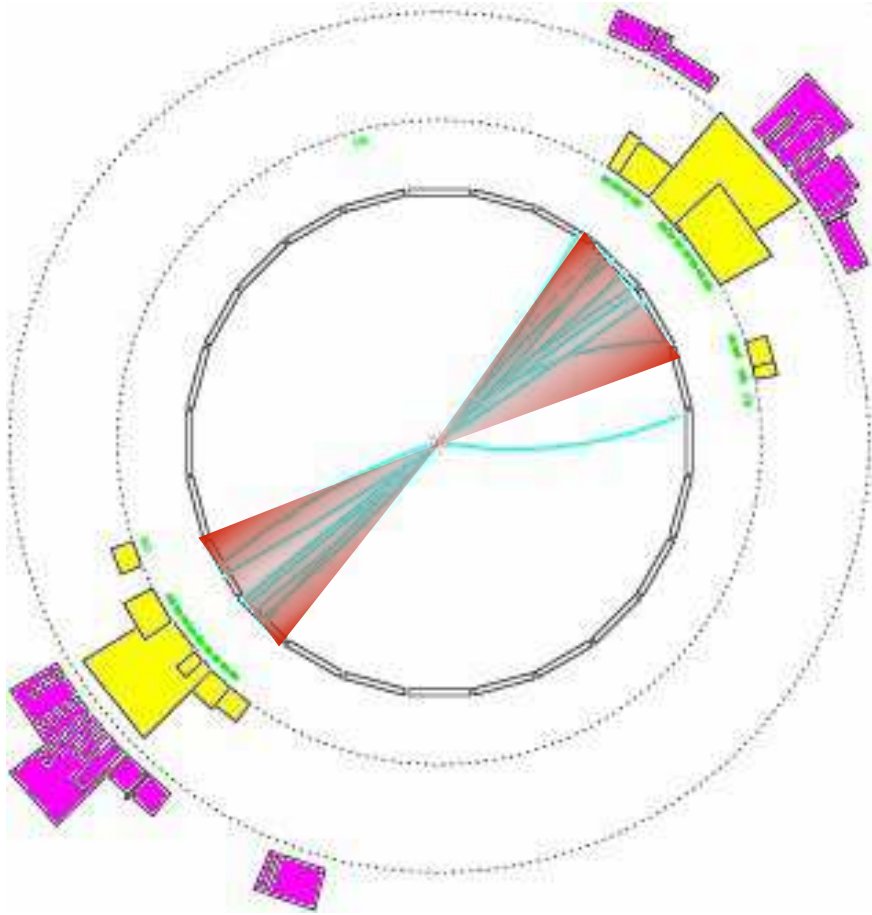
Jet definitions as projections



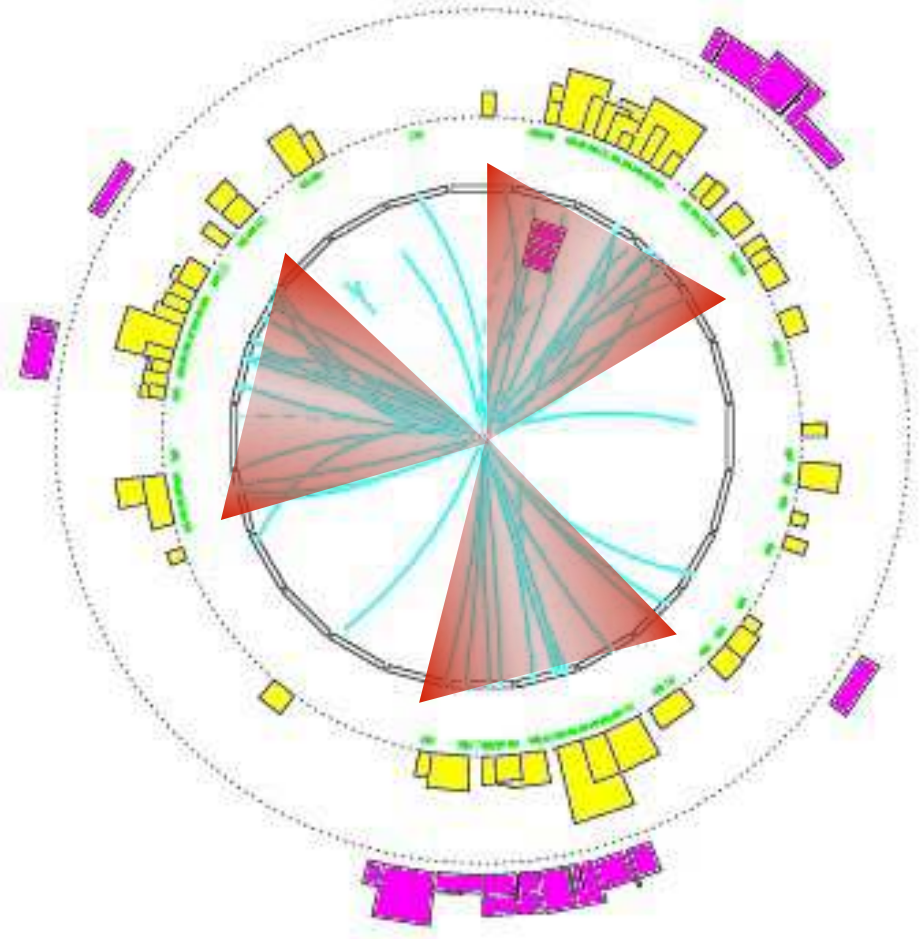
Projection to jets should be resilient to QCD effects

**NB: projections are NOT unique:
a jet is NOT EQUIVALENT to a parton**

Reconstructing jets is an ambiguous task

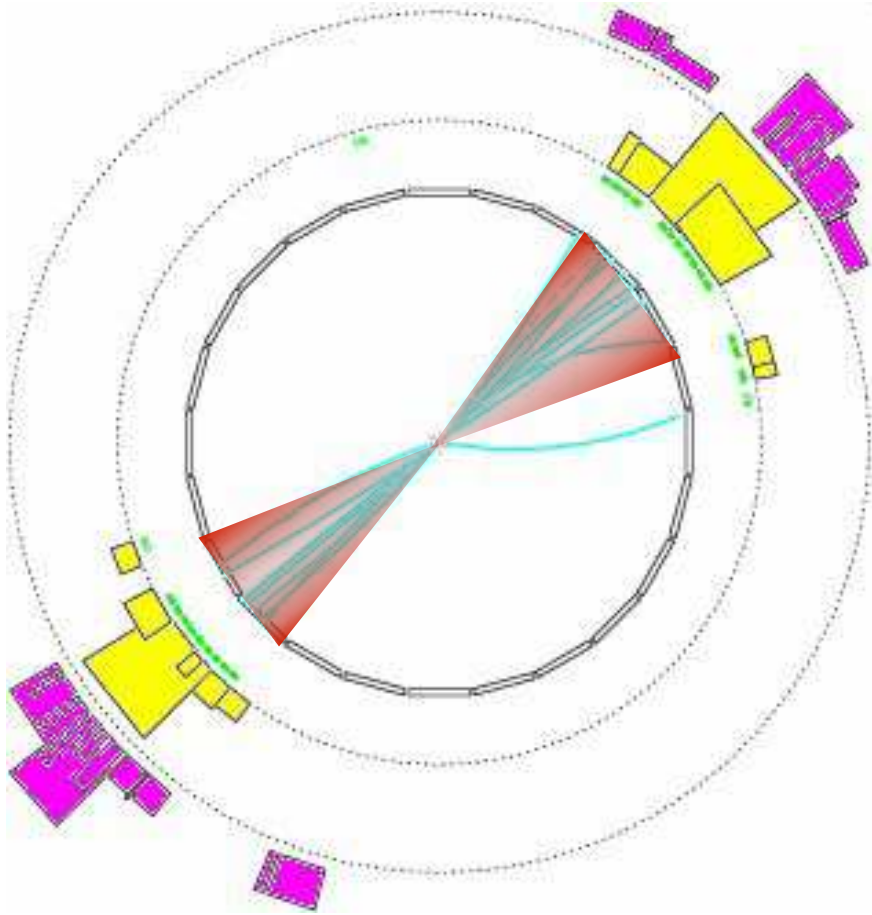


2 clear jets

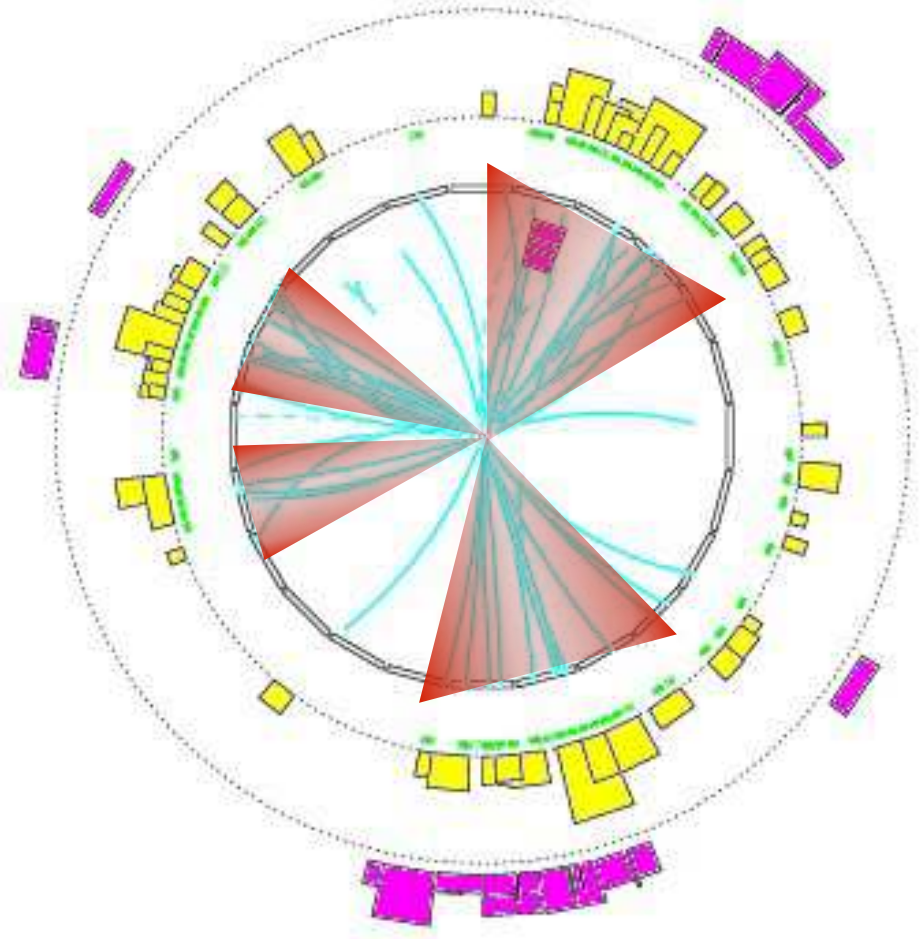


3 jets?

Reconstructing jets is an ambiguous task



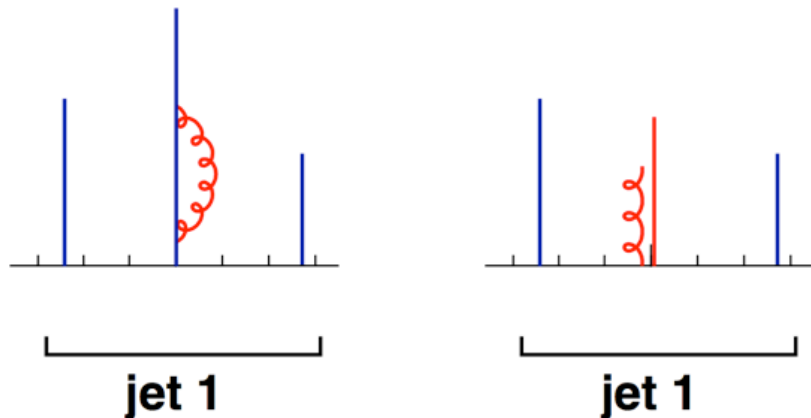
2 clear jets



3 jets?
or 4 jets?

Reconstructing jets must respect rules

Collinear Safe

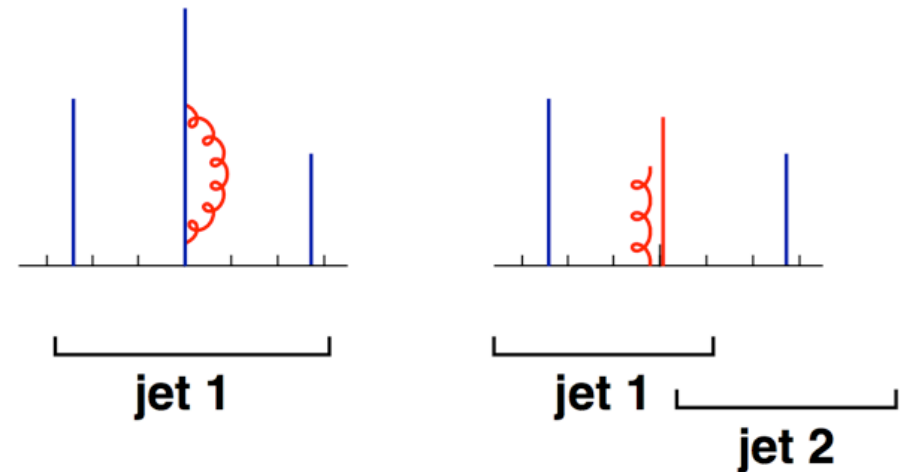


$$\alpha_S^n \times (-\infty)$$

$$\alpha_S^n \times (+\infty)$$

Infinities cancel

Collinear Unsafe



$$\alpha_S^n \times (-\infty)$$

$$\alpha_S^n \times (+\infty)$$

Infinities do not cancel

Perturbative calculations of jet observable will only be possible with collinear (and infrared) safe jet definitions

Two main classes of jet algorithms

▶ **Sequential recombination algorithms**

Bottom-up approach: combine particles starting from **closest ones**

How? Choose a **distance measure**, iterate recombination until few objects left, call them jets

Works because of mapping closeness \Leftrightarrow QCD divergence

Examples: Jade, k_t , Cambridge/Aachen, anti- k_t ,

Usually trivially made IRC safe, but their algorithmic complexity scales like N^3

▶ **Cone algorithms**

Top-down approach: find coarse regions of energy flow.

How? Find **stable cones** (i.e. their axis coincides with sum of momenta of particles in it)

Works because QCD only modifies energy flow on small scales

Examples: JetClu, MidPoint, ATLAS cone, CMS cone, SISCone.....

Can be programmed to be fairly fast, at the price of being complex and IRC unsafe

A little history

- ▶ Cone-type jets were introduced first in QCD in the 1970s (Sterman-Weinberg '77)
- ▶ In the 1980s cone-type jets were adapted for use in hadron colliders (SpS, Tevatron...) → iterative cone algorithms
- ▶ LEP was a golden era for jets: new algorithms and many relevant calculations during the 1990s
 - ▶ Introduction of the 'theory-friendly' k_t algorithm
 - ▶ sequential recombination type algorithm, IRC safe
 - ▶ it allows for all order resummation of jet rates
 - ▶ Several accurate calculations in perturbative QCD of jet properties: rates, jet mass, thrust,

$e^+e^- k_t$ (Durham) algorithm

[Catani, Dokshitzer, Olsson, Turnock, Webber '91]

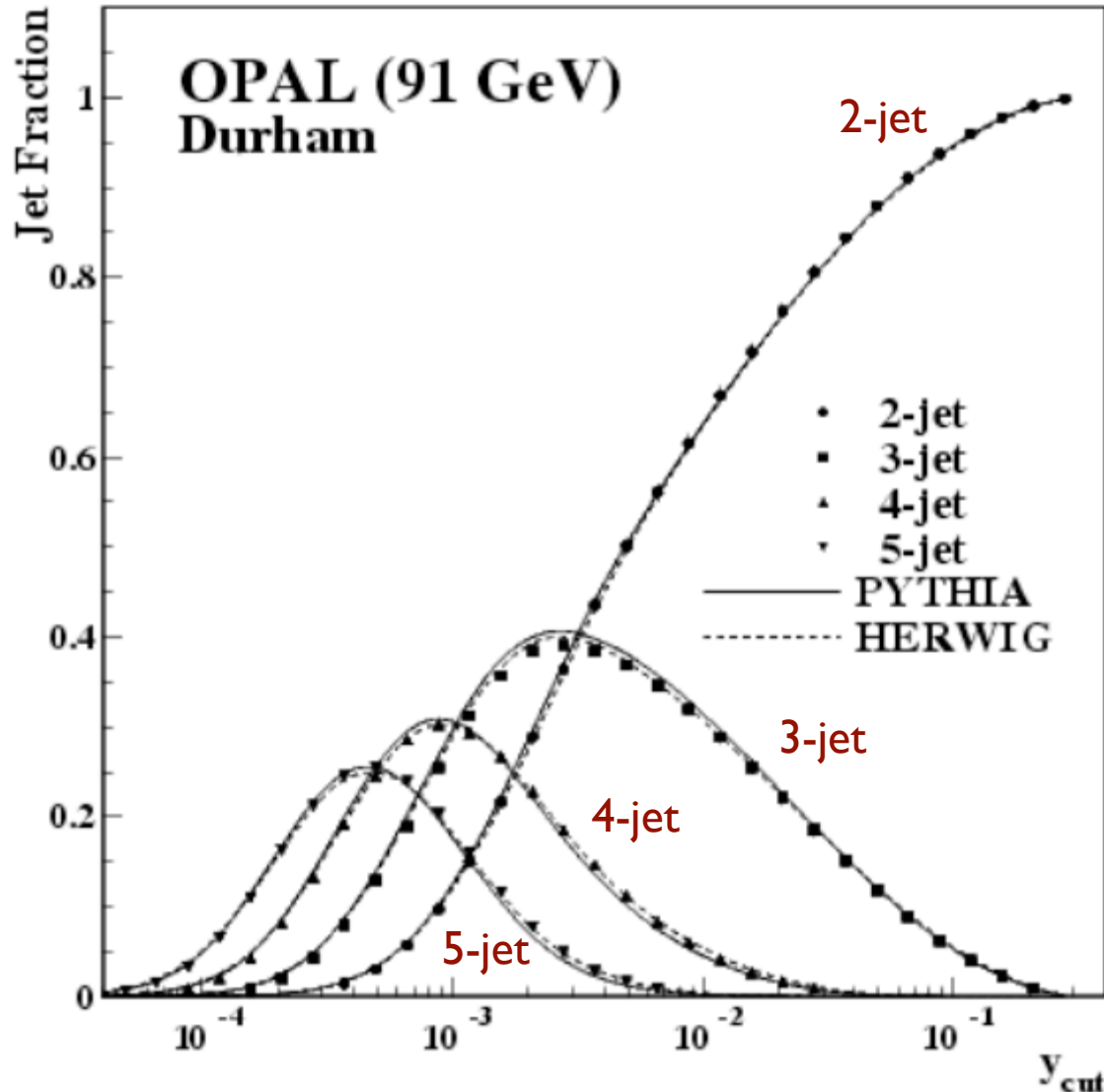
Distance:

$$y_{ij} = \frac{2 \min(E_i^2, E_j^2) (1 - \cos \theta_{ij})}{Q^2}$$

In the collinear limit, the numerator reduces to the **relative transverse momentum** (squared) of the two particles, hence the name of the algorithm

- ▶ Find the minimum y_{\min} of all y_{ij}
- ▶ If y_{\min} is below some jet resolution threshold y_{cut} , recombine i and j into a single new particle ('pseudojet'), and repeat
- ▶ If no $y_{\min} < y_{\text{cut}}$ are left, all remaining particles are jets

$e^+e^- k_t$ (Durham) algorithm in action



Characterise events
in terms of number of jets
(as a function of y_{cut})

Resummed calculations for distributions of y_{cut} doable with the k_t algorithm

$e^+e^- k_t$ (Durham) algorithm v. QCD

k_t is a sequential recombination type algorithm

One key feature of the k_t algorithm is its relation to the structure of QCD divergences:

$$\frac{dP_{k \rightarrow ij}}{dE_i d\theta_{ij}} \sim \frac{\alpha_s}{\min(E_i, E_j)\theta_{ij}}$$

The y_{ij} distance is the inverse of the emission probability

- ▶ The k_t algorithm roughly inverts the QCD branching sequence (the pair which is recombined first is the one with the largest probability to have branched)
- ▶ The history of successive clusterings has **physical meaning**

Two parameters, R and $p_{t,min}$

(These are the two parameters in essentially every widely used hadron-collider jet algorithm)

$$d_{ij} = \min(p_{ti}^2, p_{tj}^2) \frac{\Delta R_{ij}^2}{R^2}, \quad \Delta R_{ij}^2 = (y_i - y_j)^2 + (\phi_i - \phi_j)^2$$

Sequential recombination algorithm

1. Find smallest of d_{ij} , d_{iB}
2. If ij , recombine them
3. If iB , call i a jet and remove from list of particles
4. repeat from step 1 until no particles left

Only use jets with $p_t > p_{t,min}$

Inclusive k_t algorithm

S.D. Ellis & Soper, 1993

Catani, Dokshitzer, Seymour & Webber, 1993

The k_t algorithm and its siblings

$$d_{ij} = \min(p_{ti}^{2p}, p_{tj}^{2p}) \frac{\Delta y^2 + \Delta \phi^2}{R^2} \quad d_{iB} = p_{ti}^{2p}$$

$p = 1$ k_t algorithm

S. Catani, Y. Dokshitzer, M. Seymour and B. Webber, Nucl. Phys. B406 (1993) 187
S.D. Ellis and D.E. Soper, Phys. Rev. D48 (1993) 3160

$p = 0$ Cambridge/Aachen algorithm

Y. Dokshitzer, G. Leder, S. Moretti and B. Webber, JHEP 08 (1997) 001
M. Wobisch and T. Wengler, hep-ph/9907280

$p = -1$ anti- k_t algorithm

MC, G. Salam and G. Soyez, arXiv:0802.1189

NB: in anti- k_t pairs with a **hard** particle will cluster first: if no other hard particles are close by, the algorithm will give **perfect cones**

Quite ironically, a sequential recombination algorithm is the 'perfect' cone algorithm

IRC safety of generalised- k_t algorithms

$$d_{ij} = \min(p_{ti}^{2p}, p_{tj}^{2p}) \frac{\Delta y^2 + \Delta \phi^2}{R^2} \quad d_{iB} = p_{ti}^{2p}$$

$p > 0$

New **soft** particle ($p_t \rightarrow 0$) means that $d \rightarrow 0 \Rightarrow$ clustered first, no effect on jets

New **collinear** particle ($\Delta y^2 + \Delta \phi^2 \rightarrow 0$) means that $d \rightarrow 0 \Rightarrow$ clustered first, no effect on jets

$p = 0$

New **soft** particle ($p_t \rightarrow 0$) can be new jet of zero momentum \Rightarrow no effect on hard jets

New **collinear** particle ($\Delta y^2 + \Delta \phi^2 \rightarrow 0$) means that $d \rightarrow 0 \Rightarrow$ clustered first, no effect on jets

$p < 0$

New **soft** particle ($p_t \rightarrow 0$) means $d \rightarrow \infty \Rightarrow$ clustered last or new zero-jet, no effect on hard jets

New **collinear** particle ($\Delta y^2 + \Delta \phi^2 \rightarrow 0$) means that $d \rightarrow 0 \Rightarrow$ clustered first, no effect on jets

IRC safe algorithms

k_t	<p>SR</p> $d_{ij} = \min(p_{ti}^2, p_{tj}^2) \Delta R_{ij}^2 / R^2$ <p>hierarchical in rel p_t</p>	<p>Catani et al '91 Ellis, Soper '93</p>	$N \ln N$
Cambridge/ Aachen	<p>SR</p> $d_{ij} = \Delta R_{ij}^2 / R^2$ <p>hierarchical in angle</p>	<p>Dokshitzer et al '97 Wengler, Wobish '98</p>	$N \ln N$
anti- k_t	<p>SR</p> $d_{ij} = \min(p_{ti}^{-2}, p_{tj}^{-2}) \Delta R_{ij}^2 / R^2$ <p>gives perfectly conical hard jets</p>	<p>MC, Salam, Soyez '08 (Delsart, Loch)</p>	$N^{3/2}$
SISCone	<p>Seedless iterative cone with split-merge gives 'economical' jets</p>	<p>Salam, Soyez '07</p>	$N^2 \ln N$

'second-generation' algorithms

All are available in FastJet, <http://fastjet.fr>

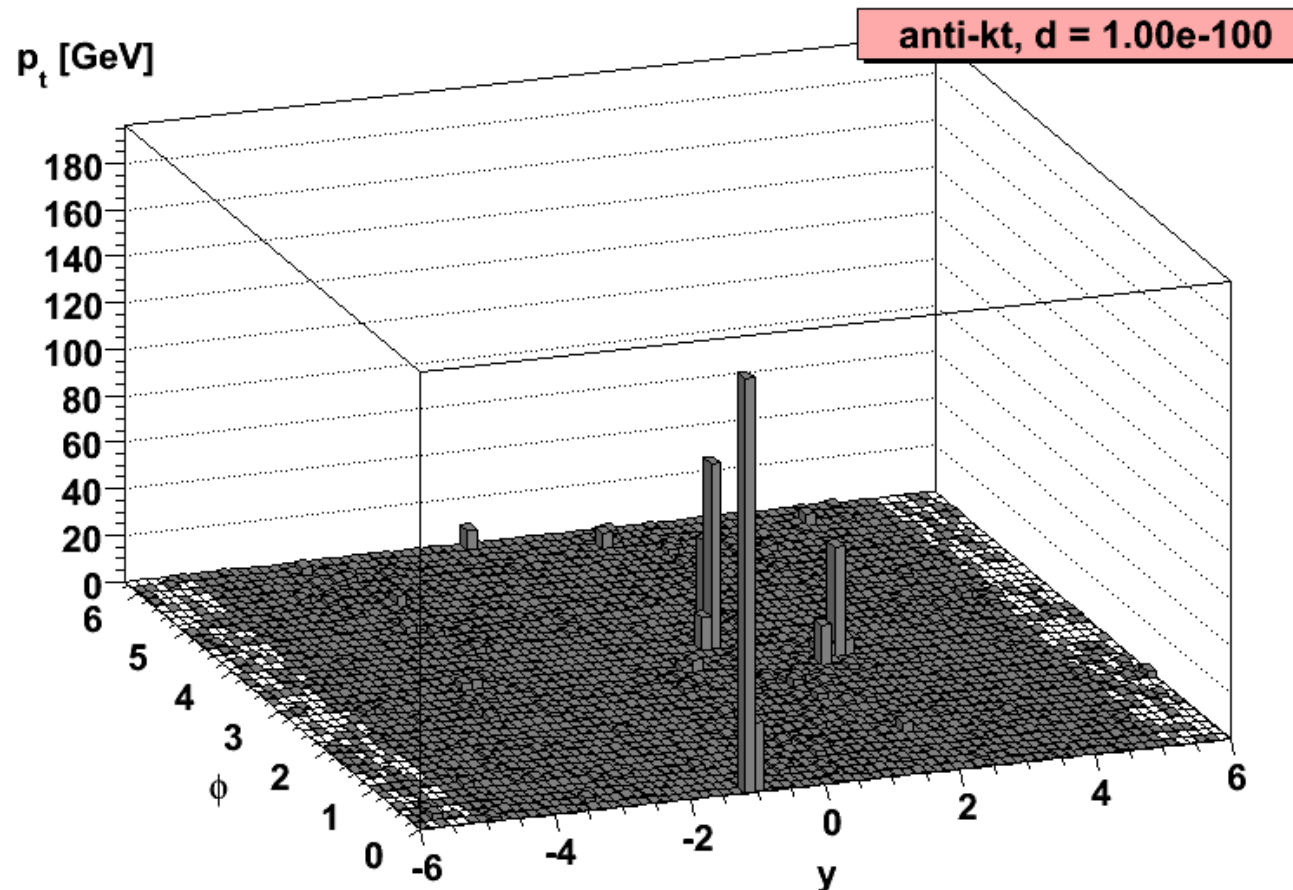
(As well as many IRC unsafe ones)

Clustering grows
around hard cores

$$d_{ij} = \frac{1}{\max(p_{ti}^2, p_{tj}^2)} \frac{\Delta R_{ij}^2}{R^2}, \quad d_{iB} = \frac{1}{p_{ti}^2}$$

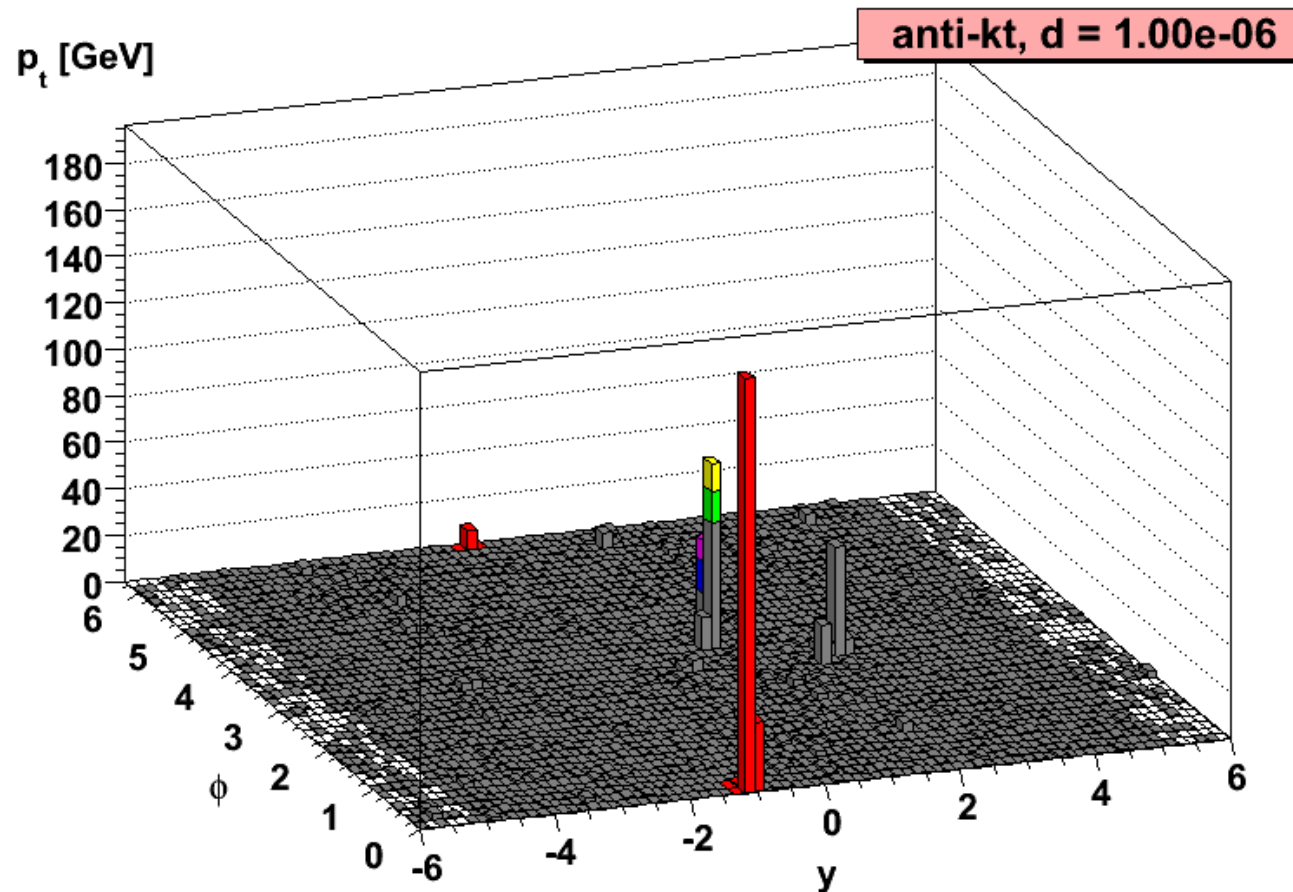
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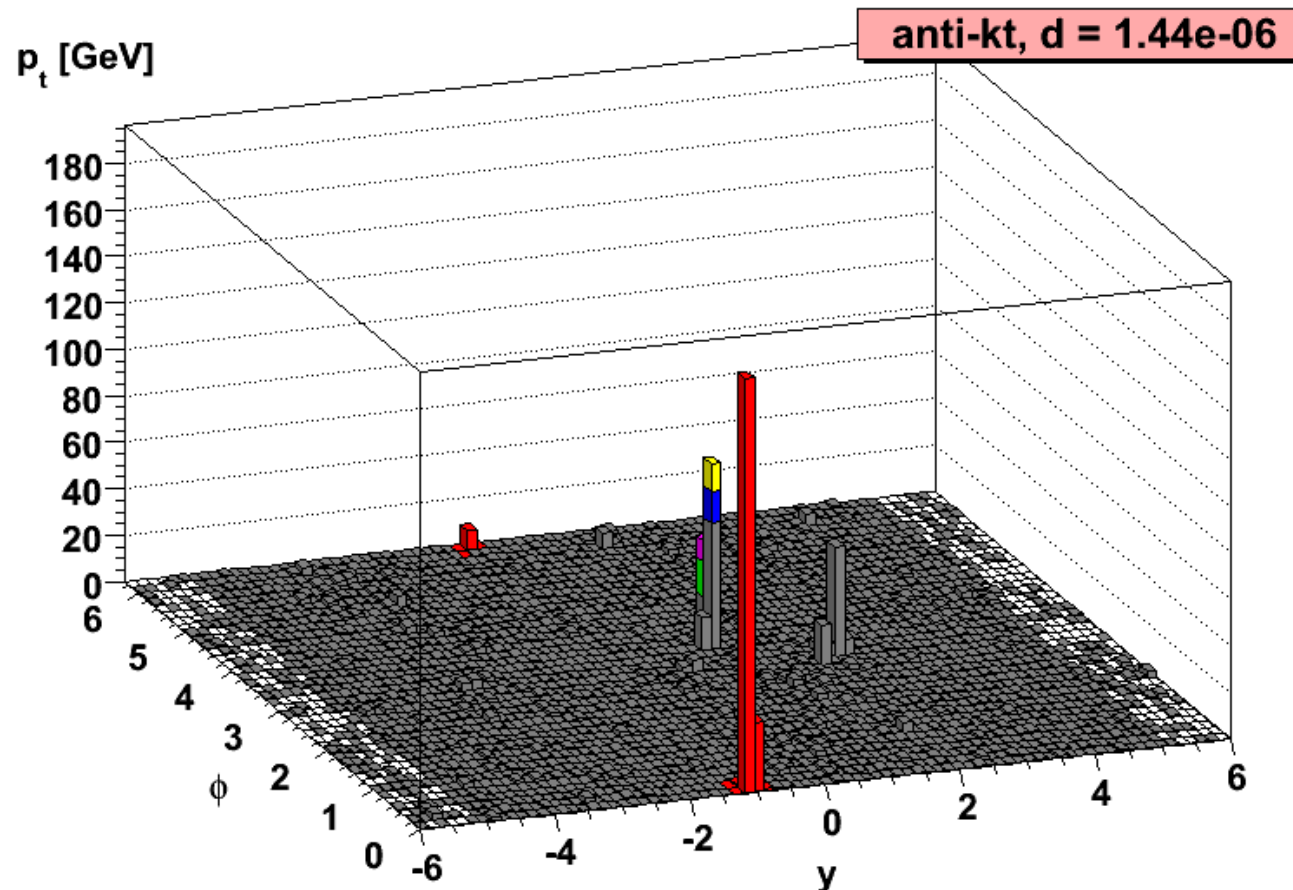
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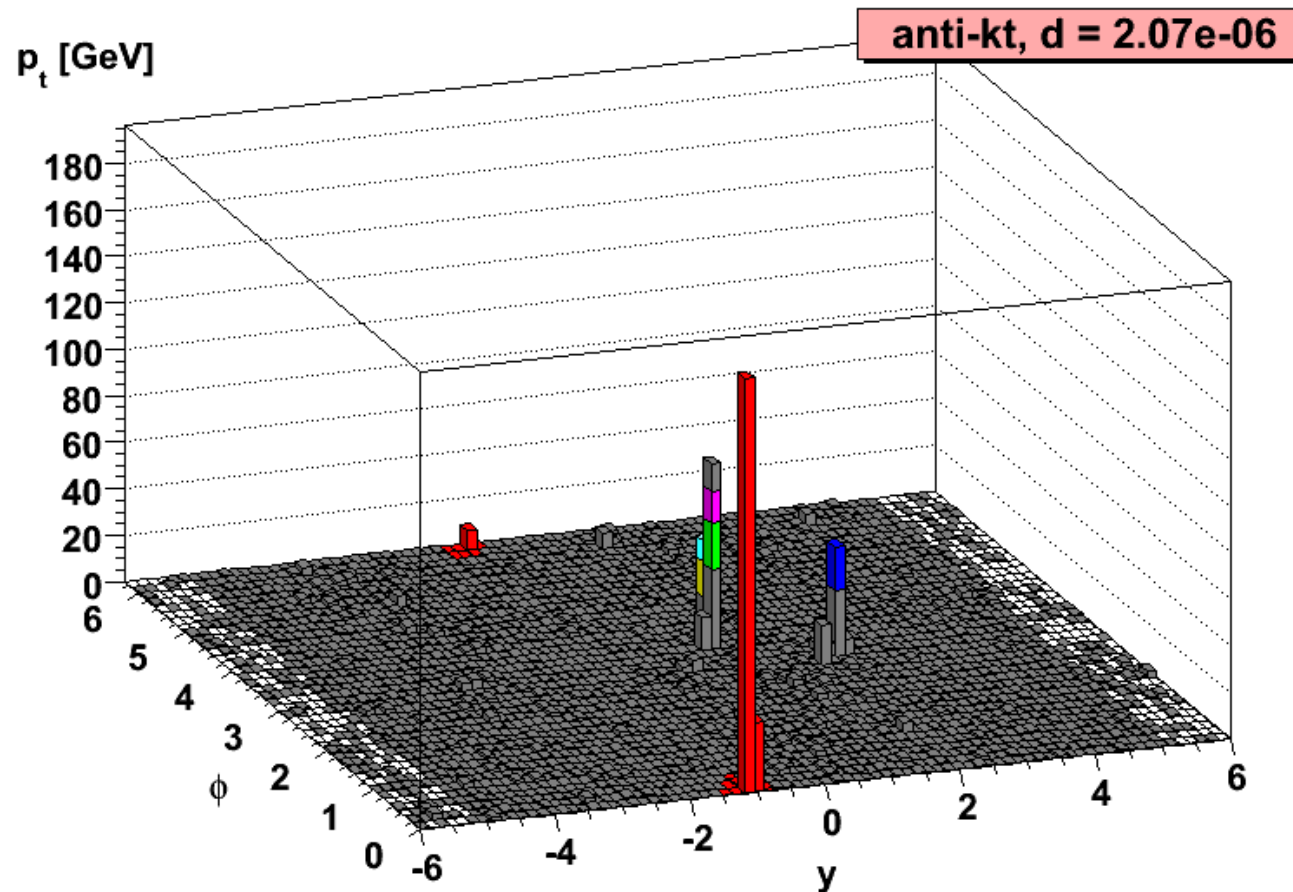
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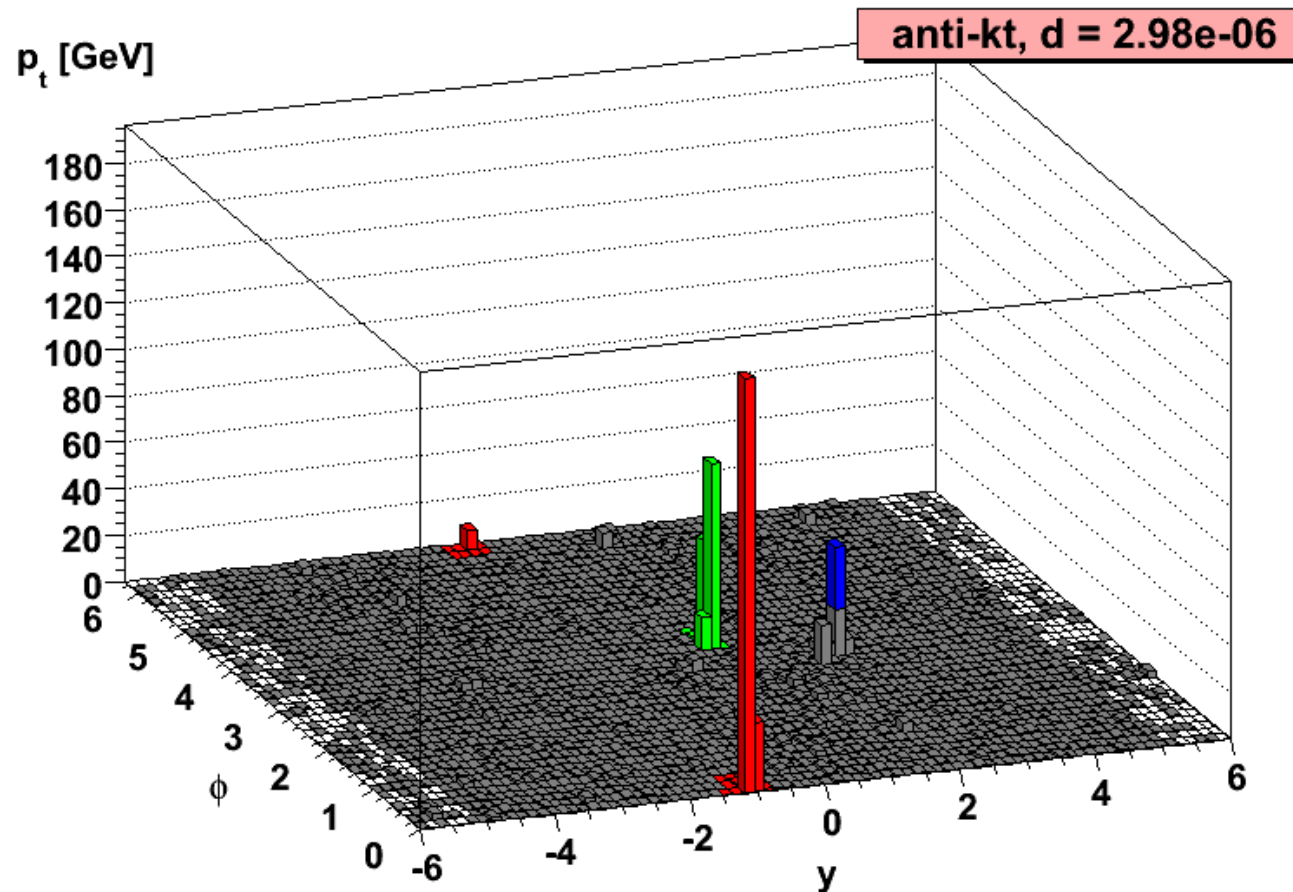
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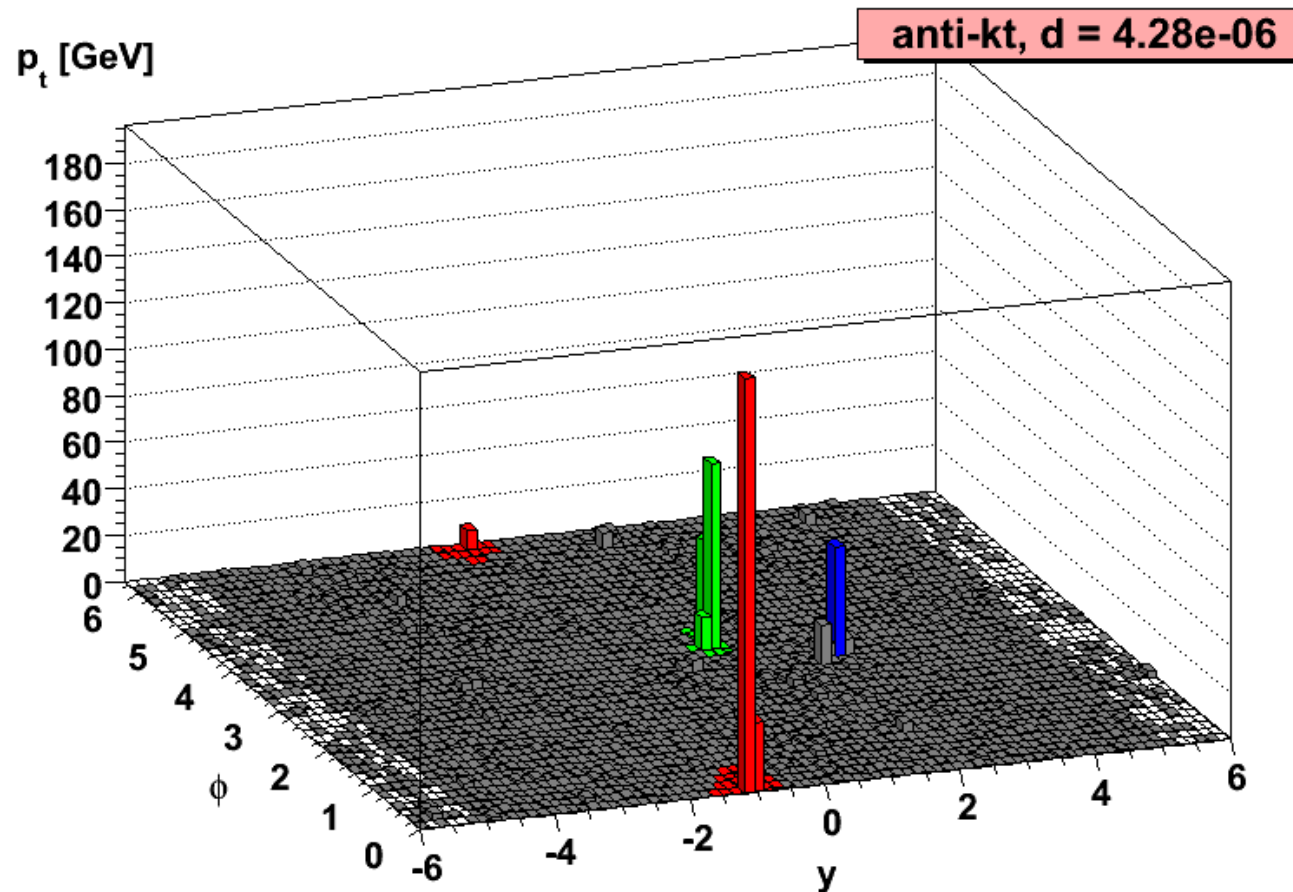
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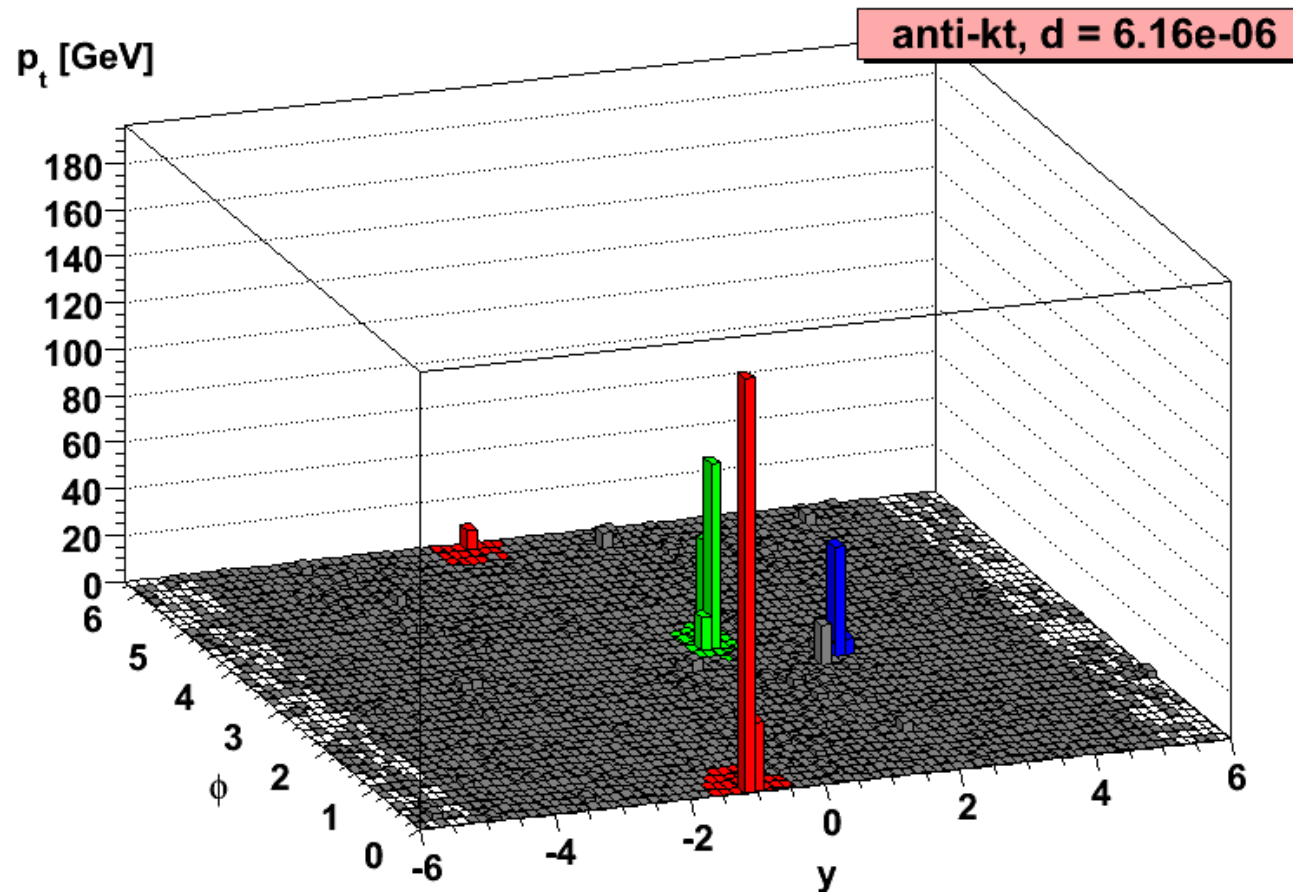
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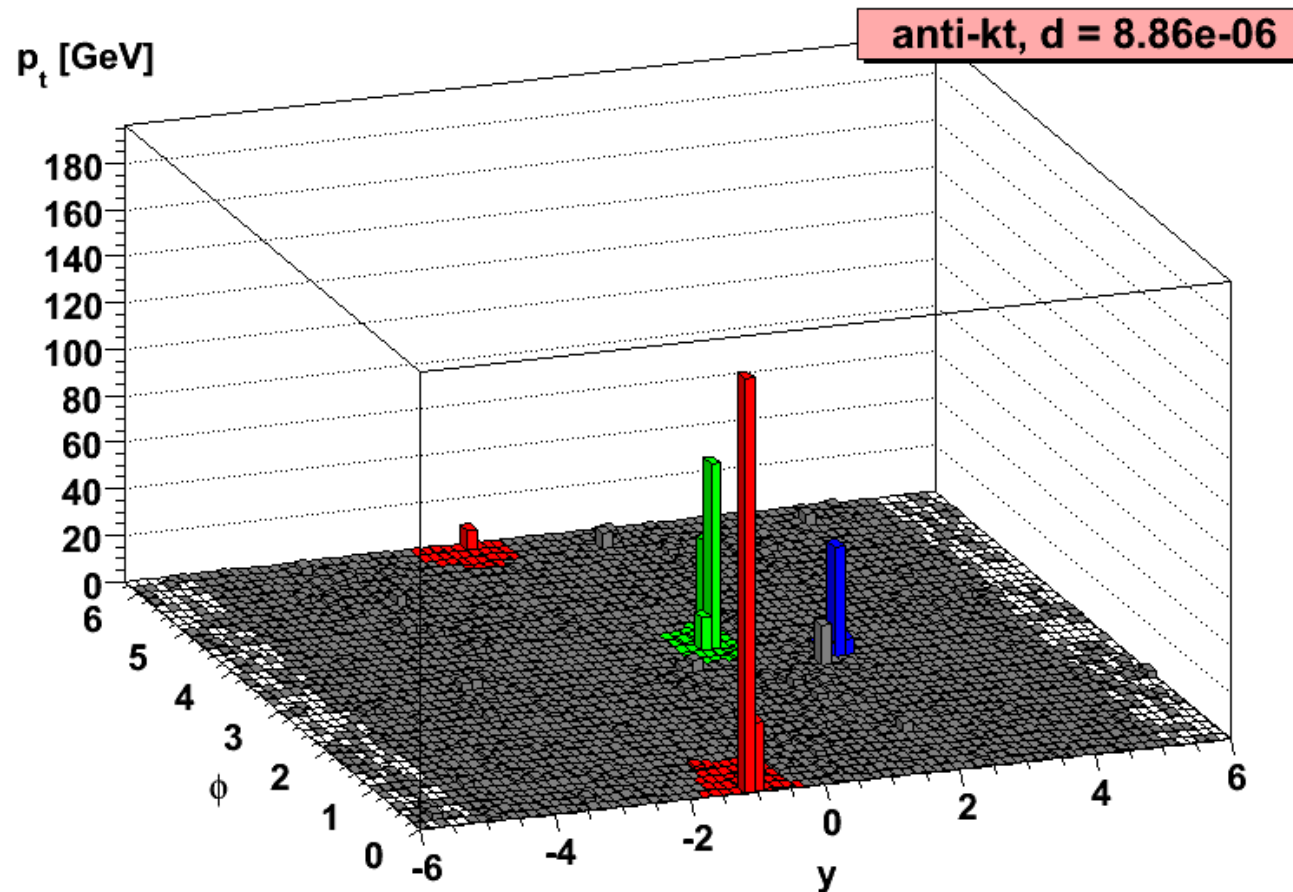
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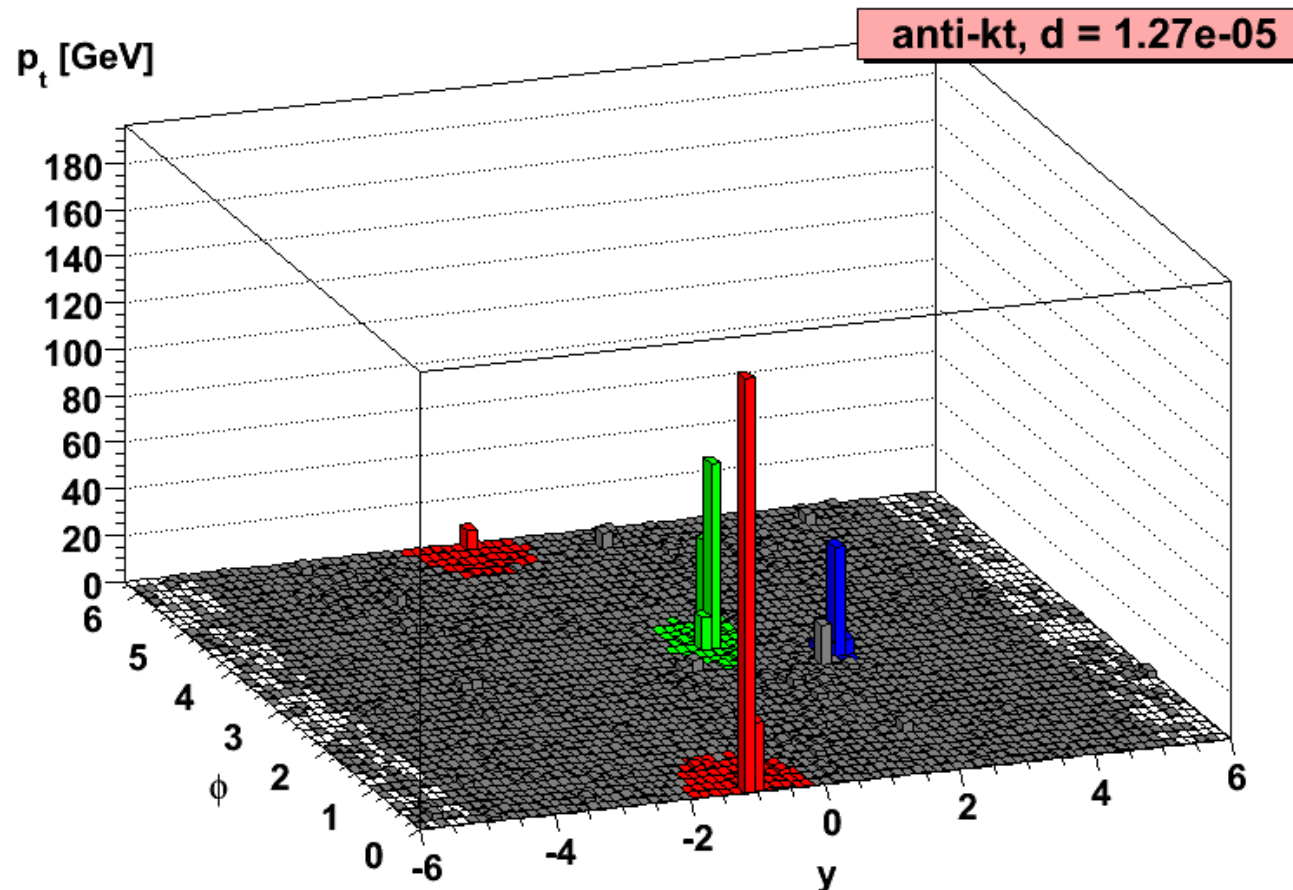
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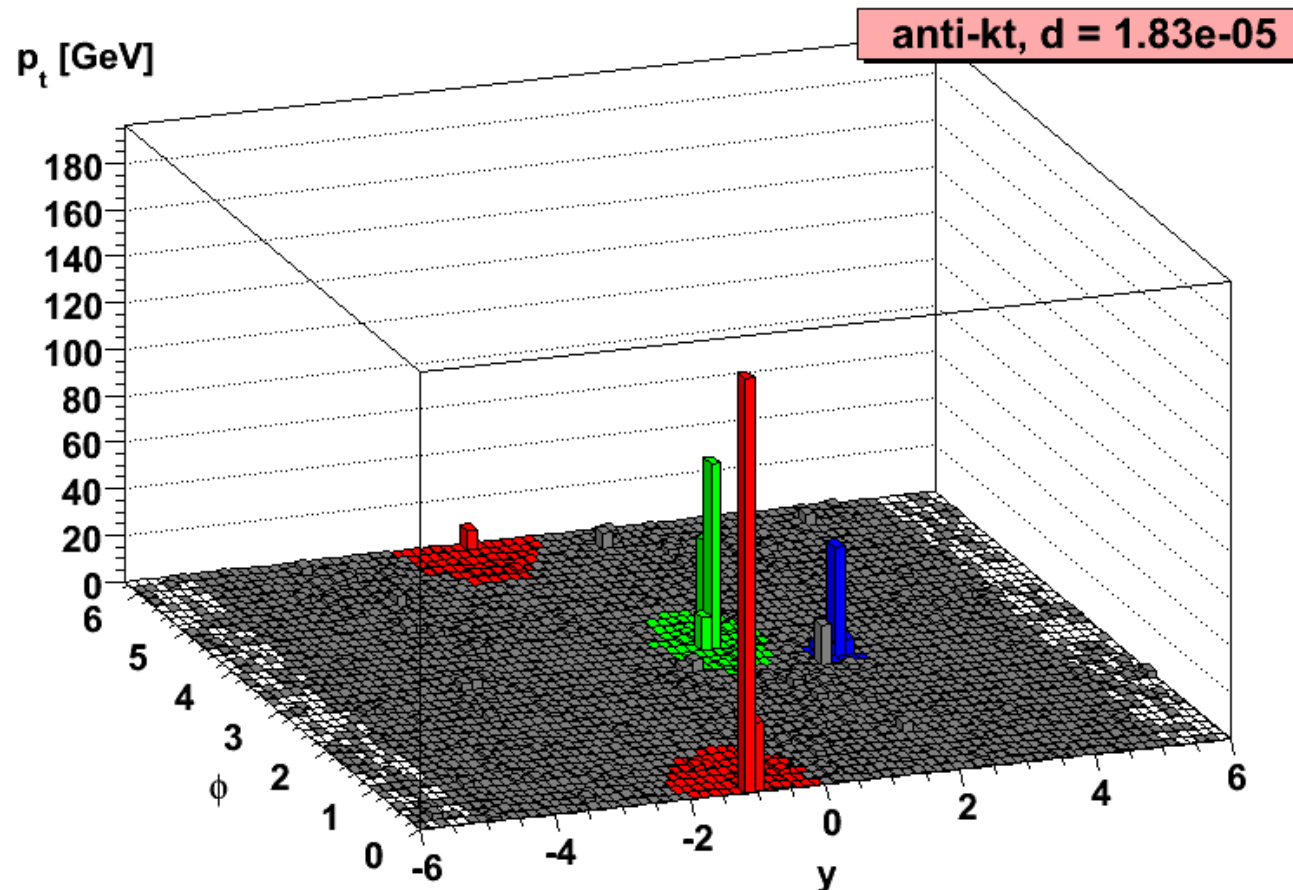
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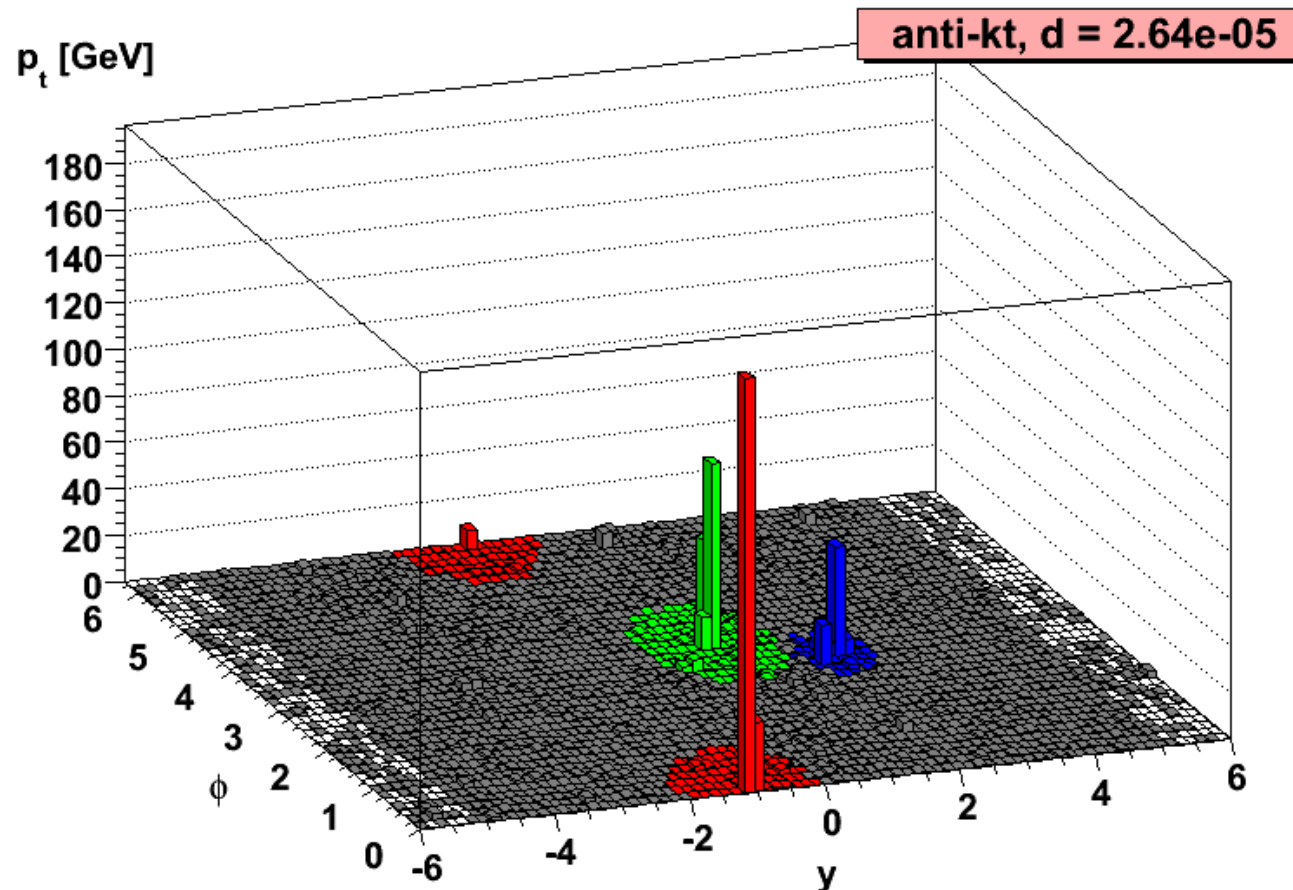
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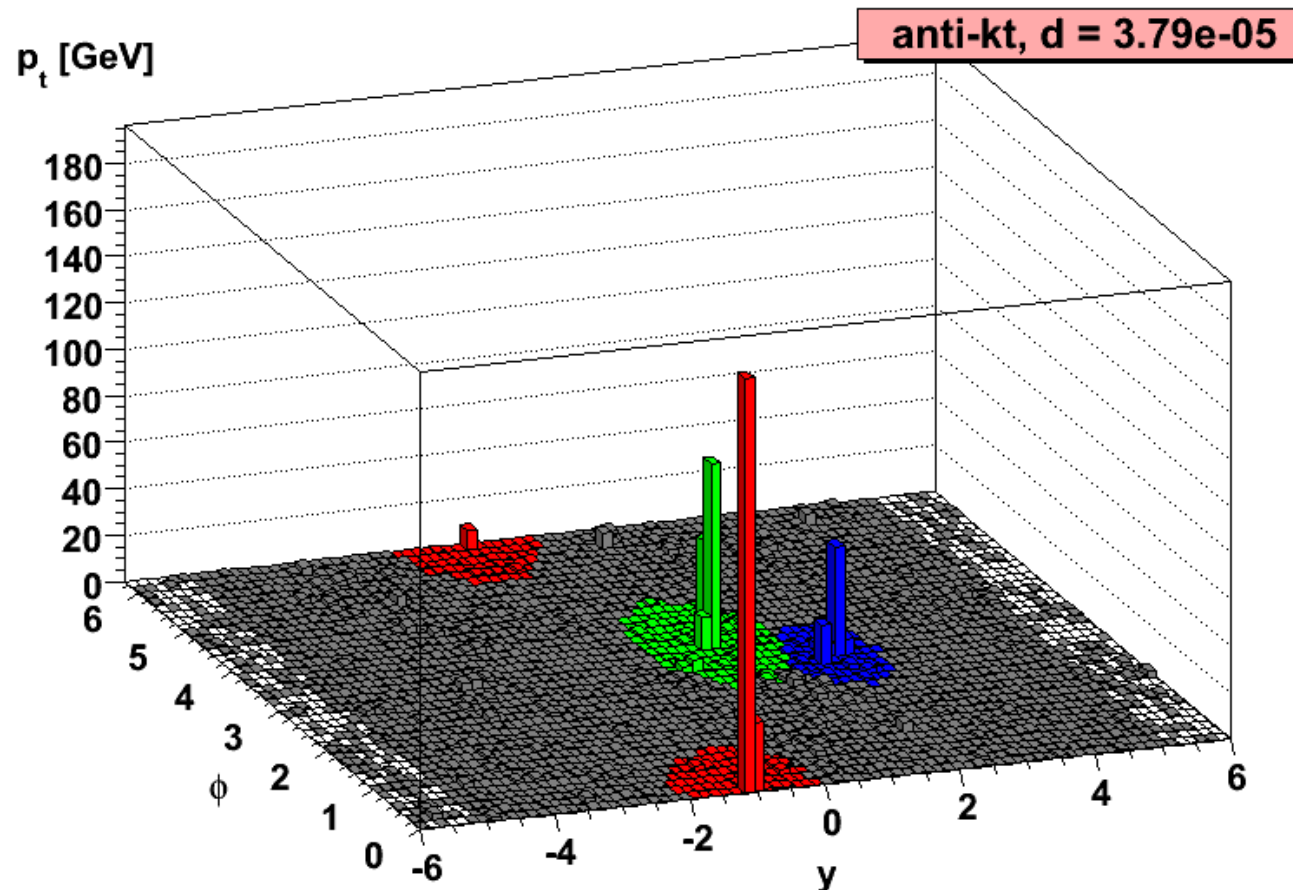
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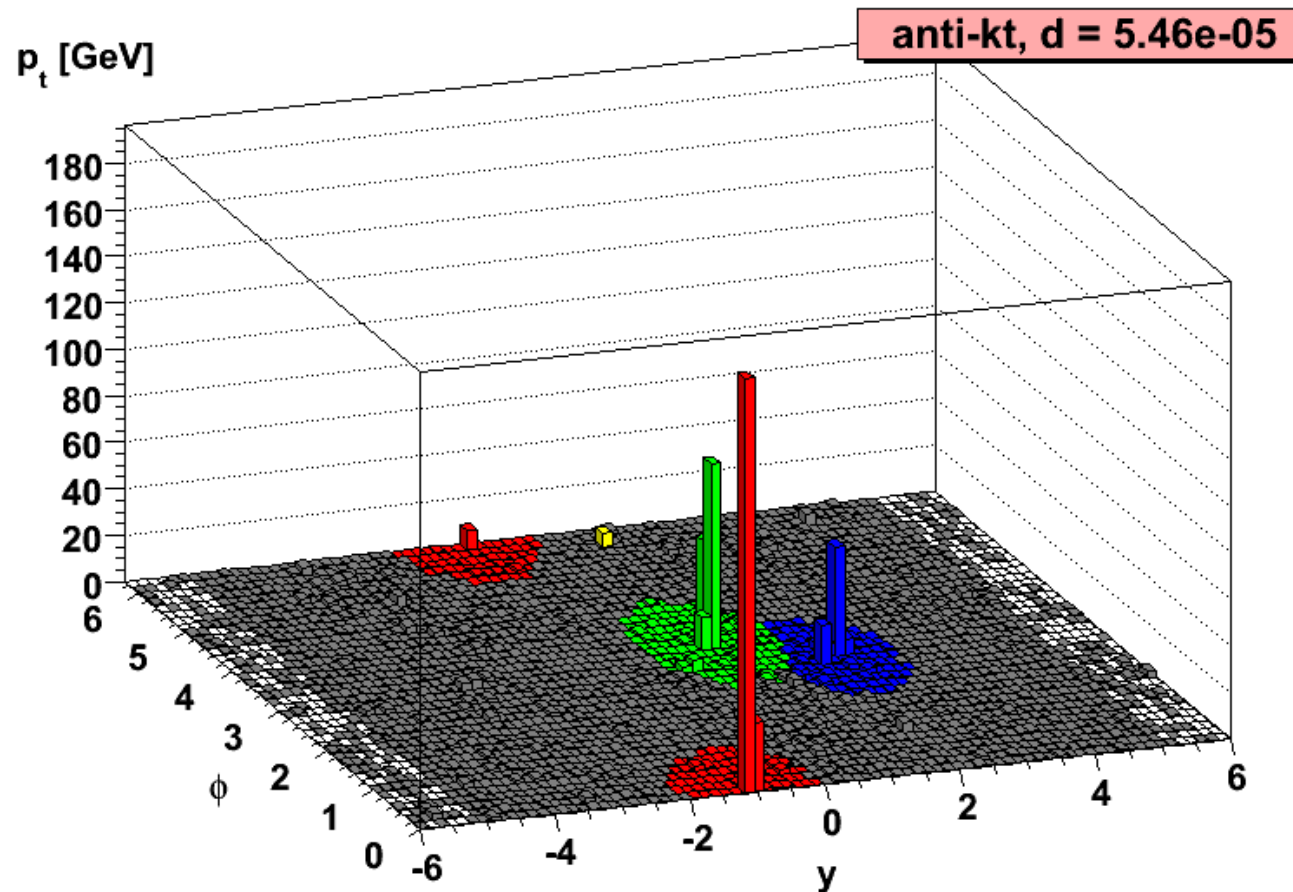
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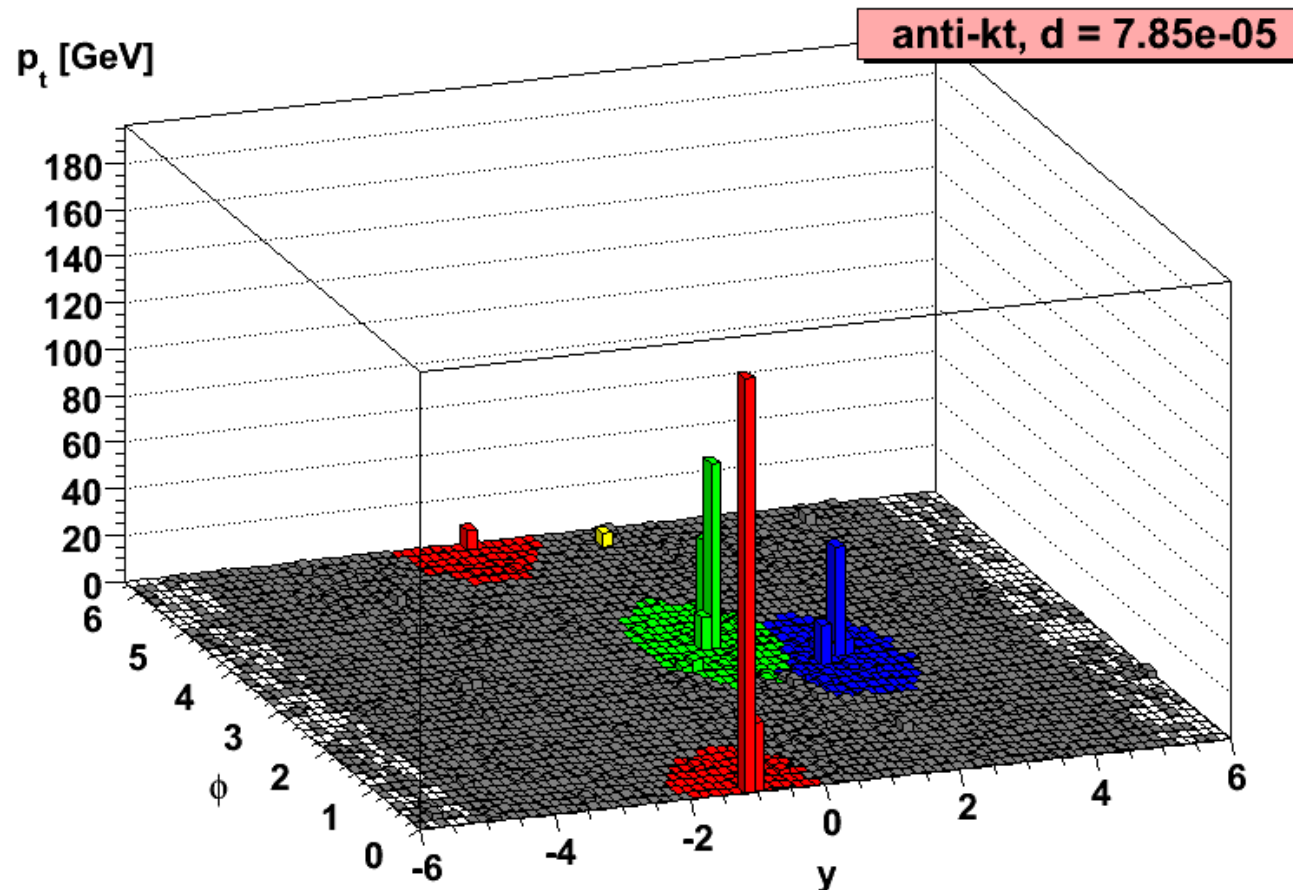
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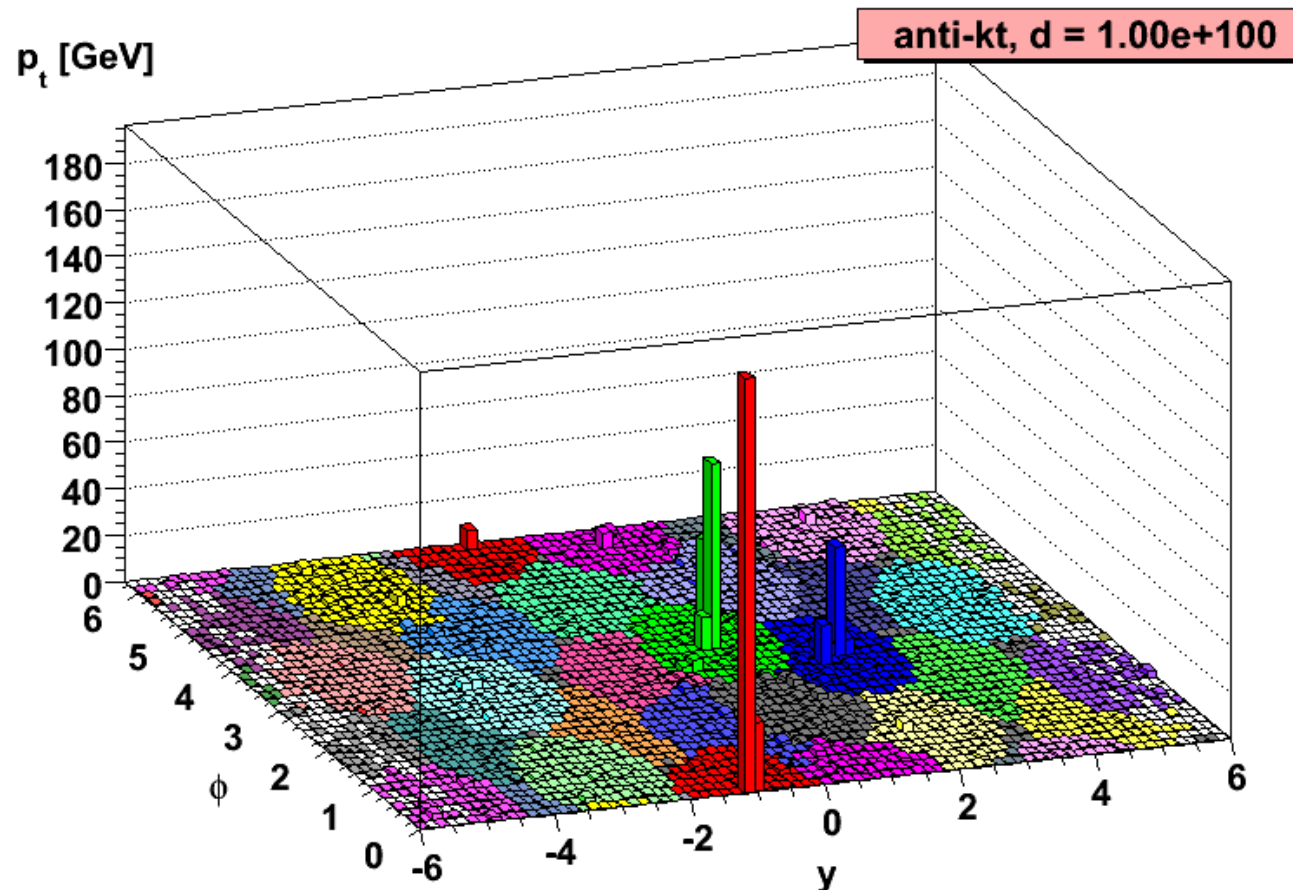
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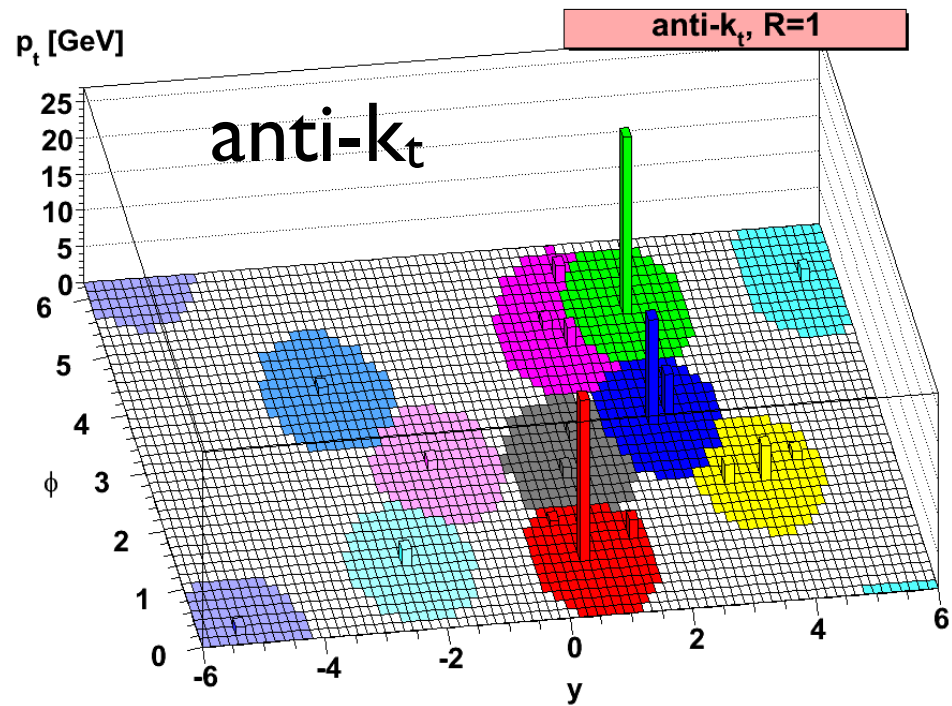
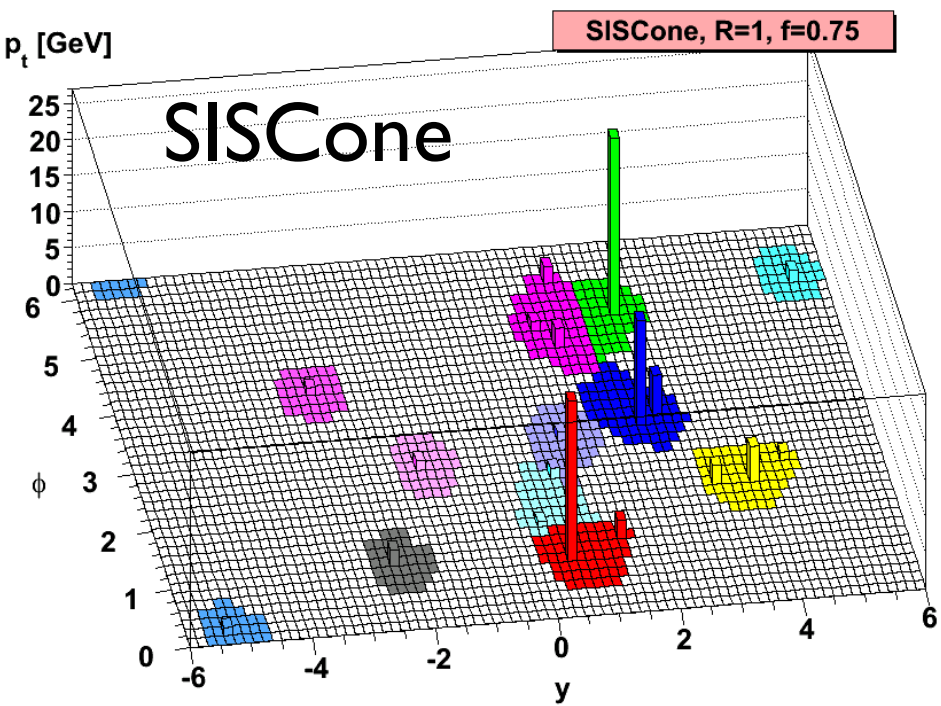
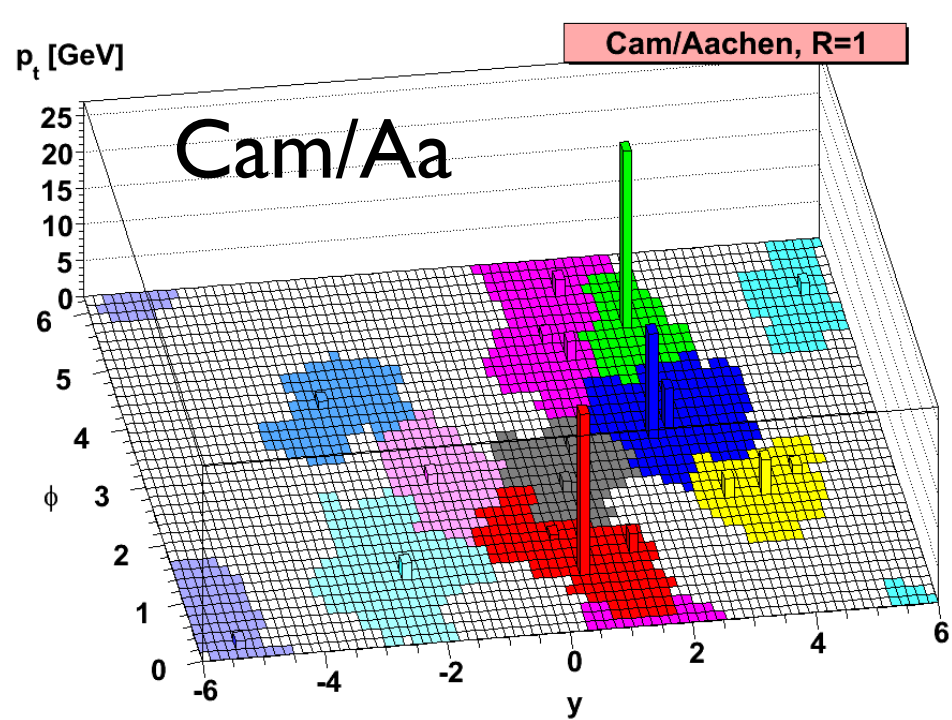
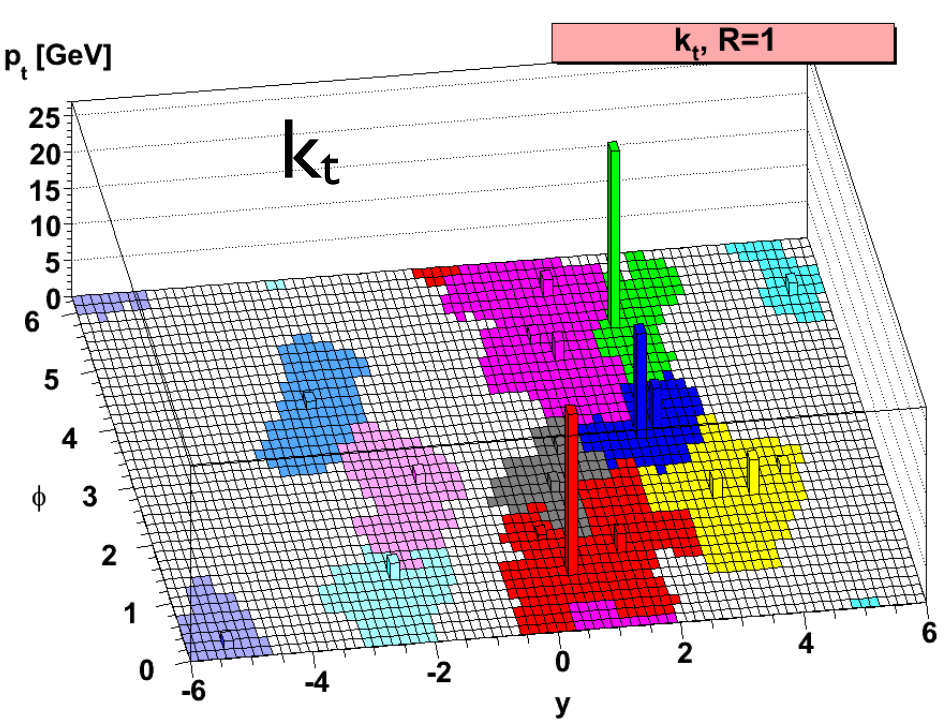


Clustering grows around hard cores

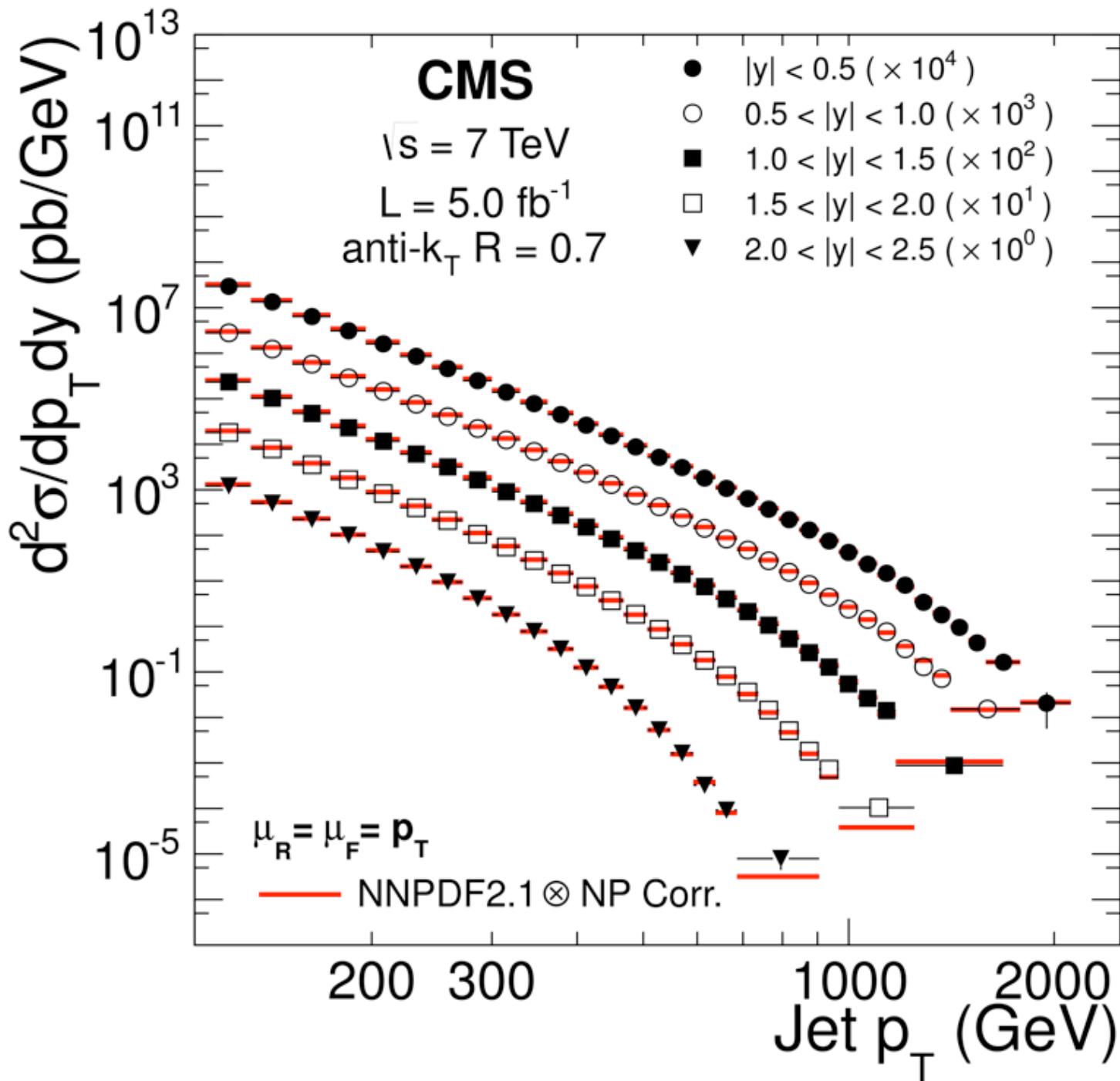
$$d_{ij} = \frac{1}{\max(p_{ti}^2, p_{tj}^2)} \frac{\Delta R_{ij}^2}{R^2}, \quad d_{iB} = \frac{1}{p_{ti}^2}$$



Anti- k_t gives circular jets (“cone-like”) in a way that’s infrared safe



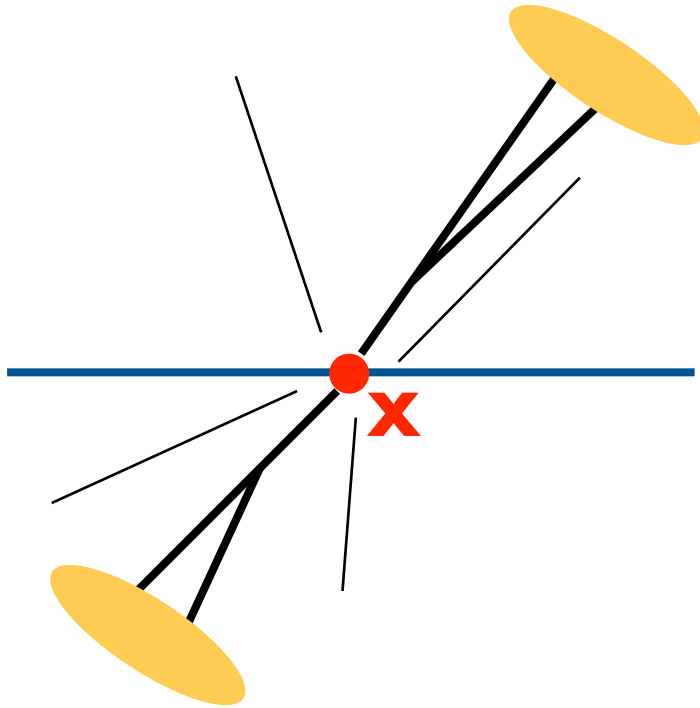
Example of jet observable



Inclusive
jet cross
section

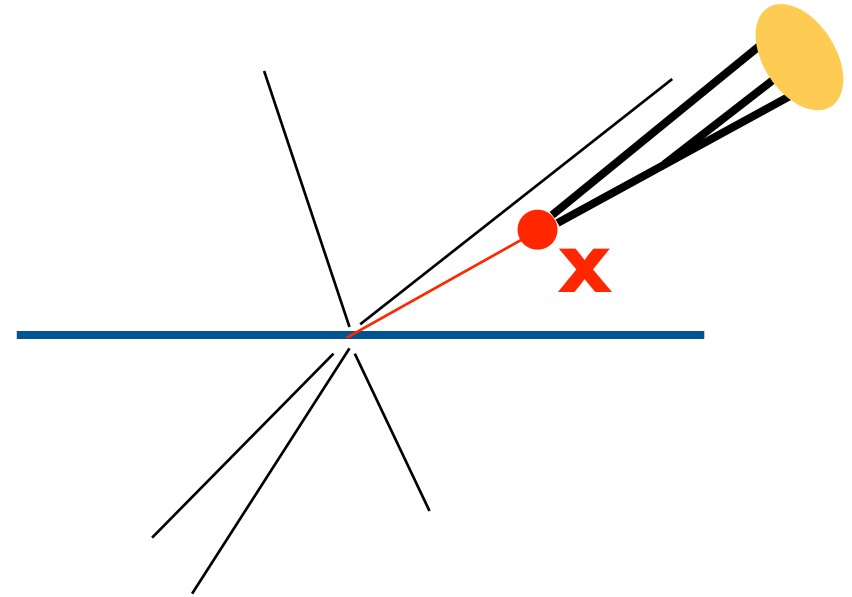
Excellent
theory-data
agreement over
many orders of
magnitude

Why boosted objects



Heavy particle X at **rest**

Easy to resolve jets and calculate invariant mass, but signal very likely swamped by background (eg $H \rightarrow bb$ v. $tt \rightarrow WbWb$)

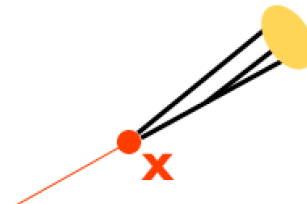
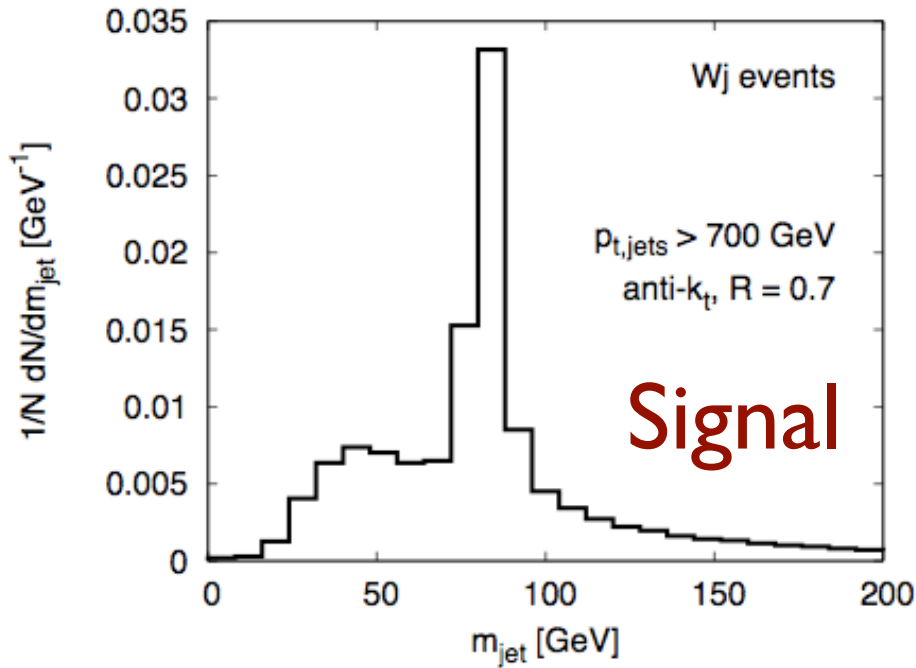


Boosted heavy particle X

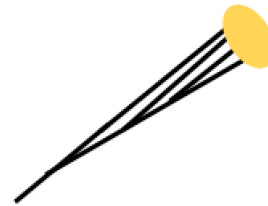
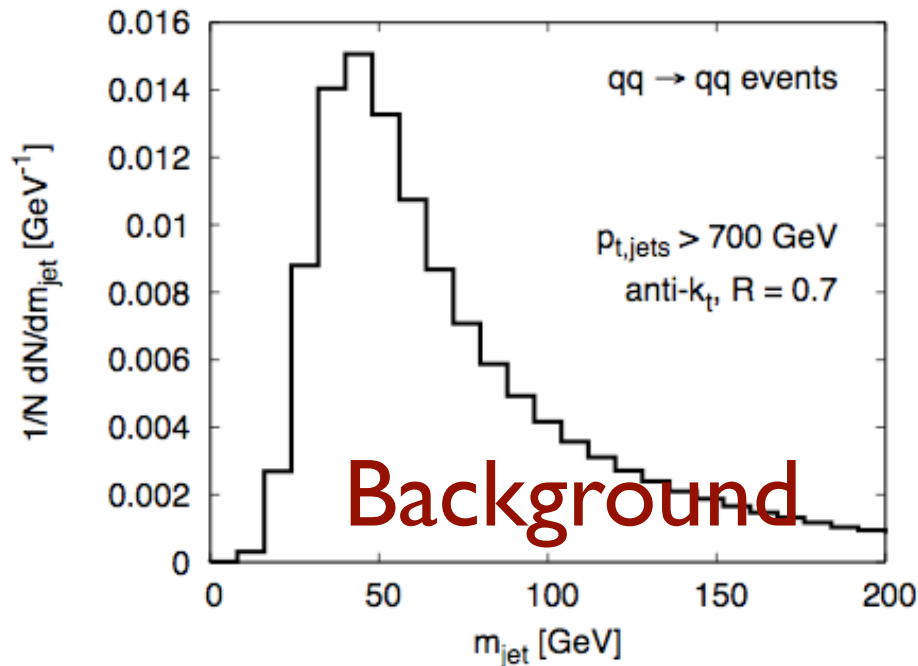
Cross section very much reduced, but acceptance better and some backgrounds smaller/reducible

Mass of a single jet

G. Salam



A heavy object decaying into a single jet naturally gives it a mass...

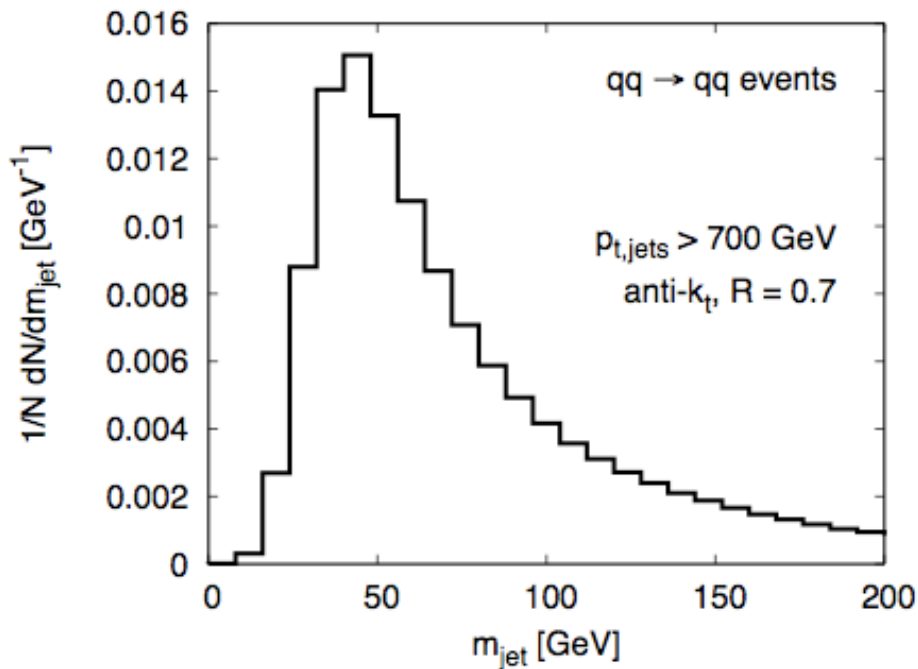


... but pure QCD jets can be massive too:

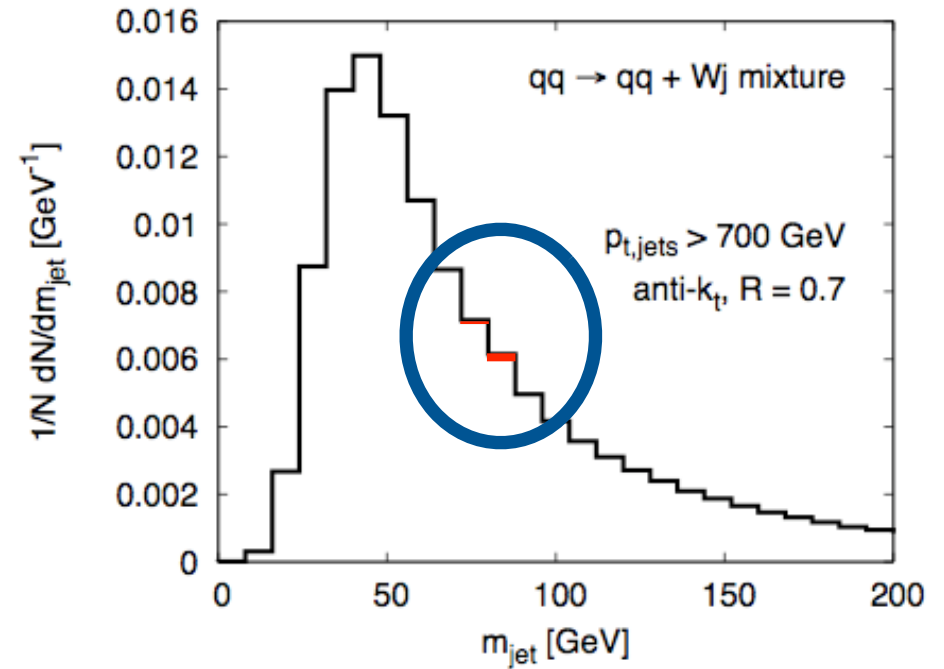
$$\frac{dN}{d \ln m} \sim \alpha_s \ln \frac{p_t R}{m} \times \text{Sudakov}$$

Mass of a single jet

Summing 'signal' and 'background' (with appropriate cross sections) shows how much the background dominates



Background only

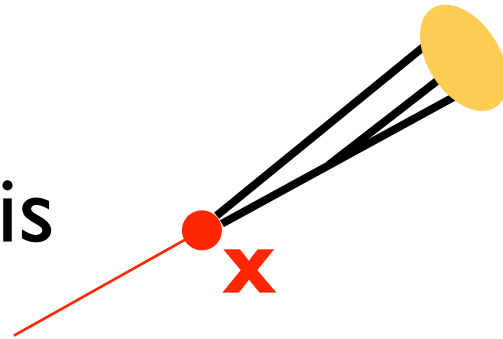


Signal + background

Practically identical

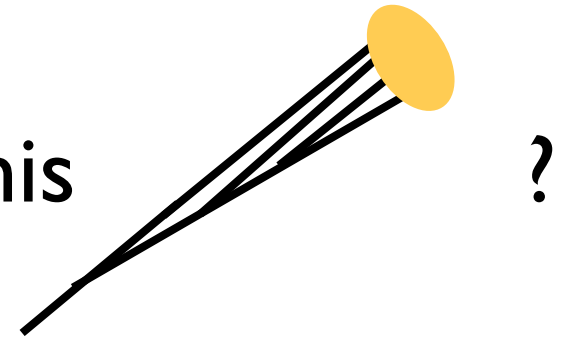
This means that one can't rely on the invariant mass only.
An appropriate strategy must be found to reduce the background and enhance the signal

How to tell this



Decay of a heavy
(boosted) object

from this



Light parton
fragmentation

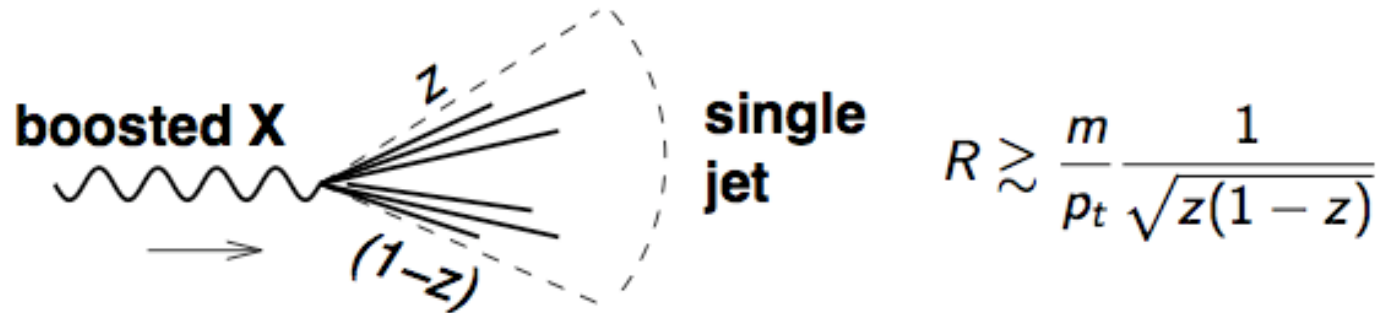
Tagging and Grooming

- ▶ The substructure of a jet can be exploited to
 - ▶ **tag** a particular structure inside the jet, i.e. a massive particle
 - ▶ First examples: Higgs (2-prong decay), top (3-prong decay)
 - ▶ remove background contamination from the jet or its components, while keeping the bulk of the perturbative radiation (often generically denoted as **grooming**)
 - ▶ First examples: filtering, trimming, pruning

Why substructure

Scales: $m \sim 100 \text{ GeV}$, $p_t \sim 500 \text{ GeV}$

(e.g. electroweak particle from decay of $\sim 1 \text{ TeV}$ BSM particle)



- ▶ need **small R** ($< 2m/p_t \sim 0.4$) to resolve **two** prongs
- ▶ need **large R** ($> \sim 3m/p_t \sim 0.6$) to cluster into a **single** jet

Possible strategies

- ▶ Use large R, get a single jet : **background large**
- ▶ Use small R, resolve the jets : **what is the right scale?**
 - ▶ Also: small jets lead to huge combinatorial issues

Let an algorithm find the 'right' substructure

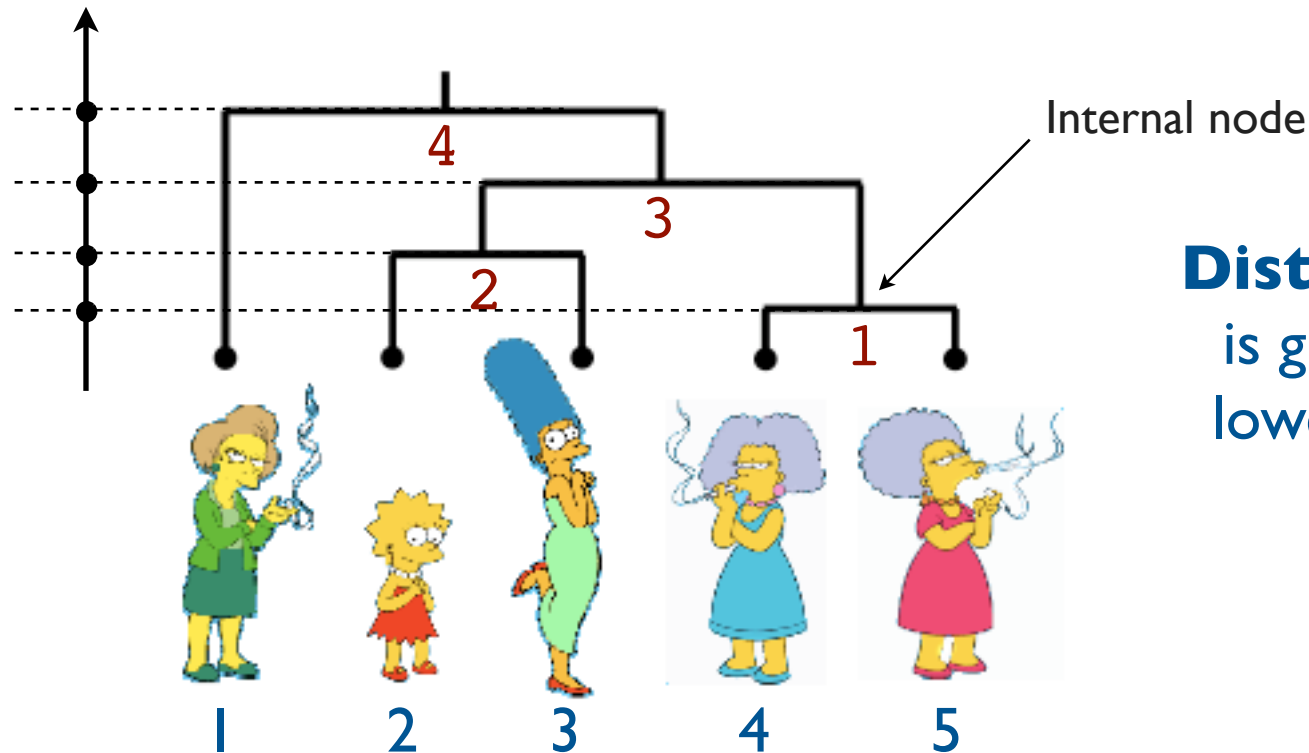
What jets to use for substructure?

Different jet algorithms will give different ‘pictures’
of what’s inside a jet

Dendrogram

Used to represent graphically the sequence of clustering steps in a sequential recombination algorithm

Distance



Distance between two objects is given by the **height** of the lowest internal node that they share.

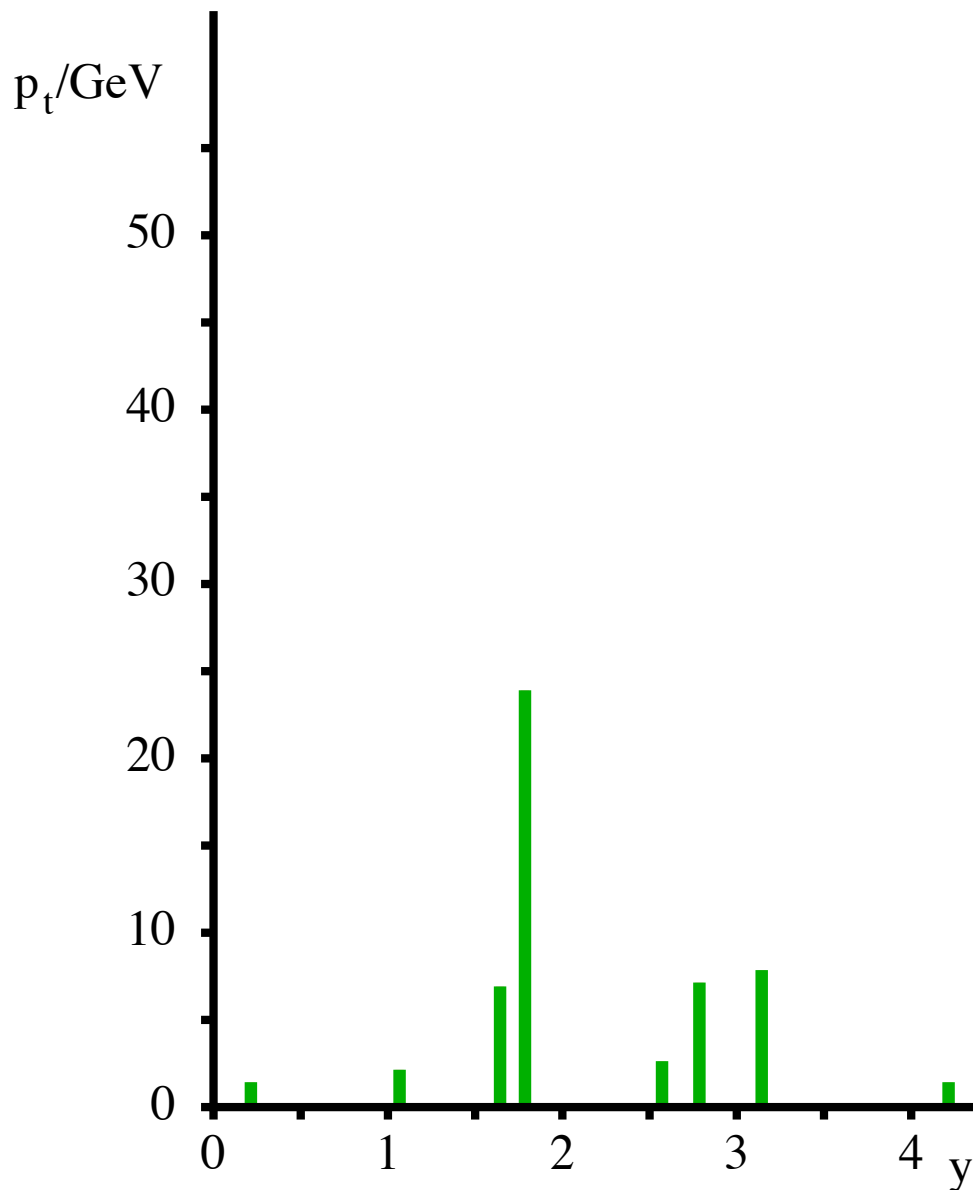
Order of clustering here is 1,2,3,4

The **clustering sequence** is 4-5 (1), 2-3 (2), 23-45 (3), 1-2345 (4)

anti-kt

Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



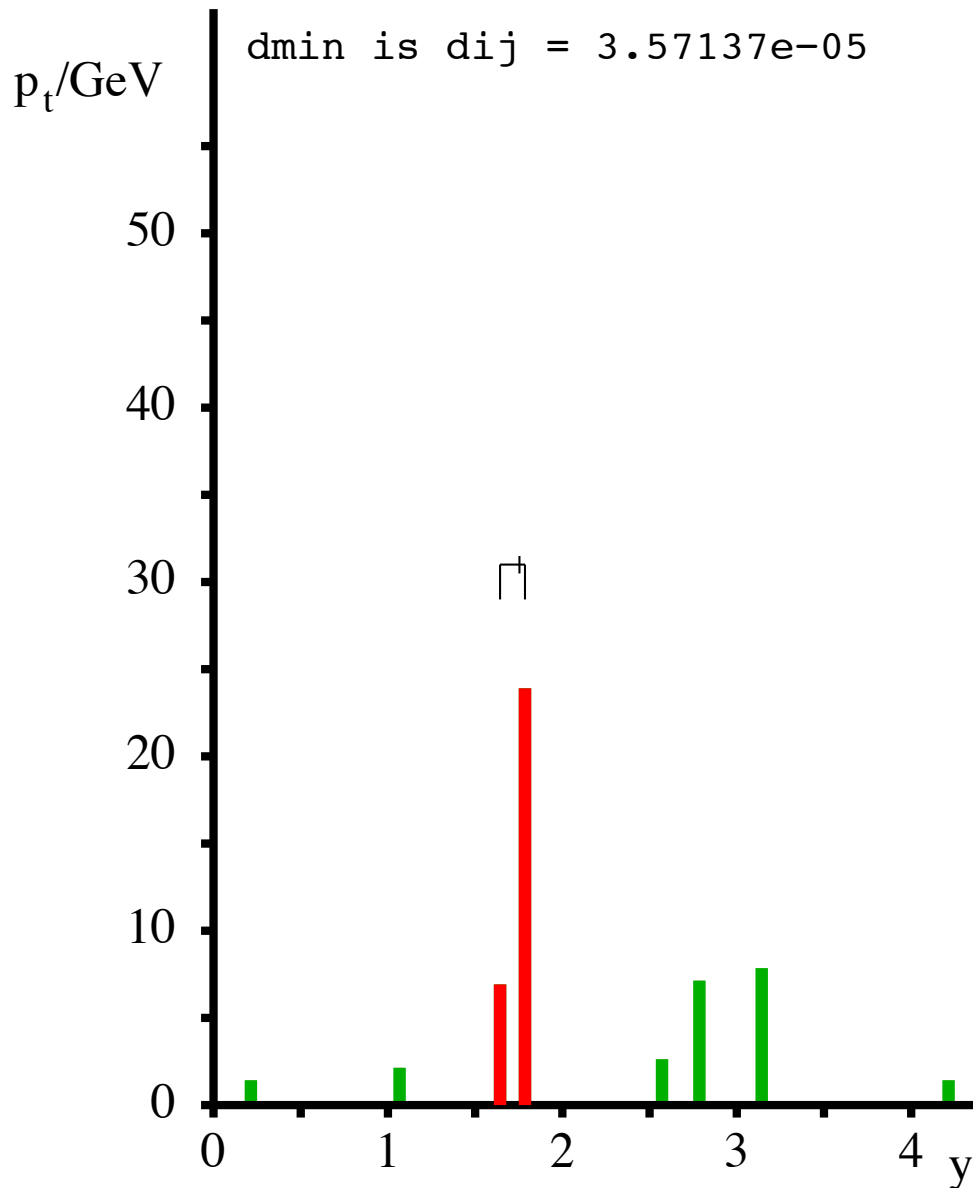
How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

Identifying jet substructure: try out anti- k_t

anti- k_t algorithm

d_{\min} is $d_{ij} = 3.57137e-05$

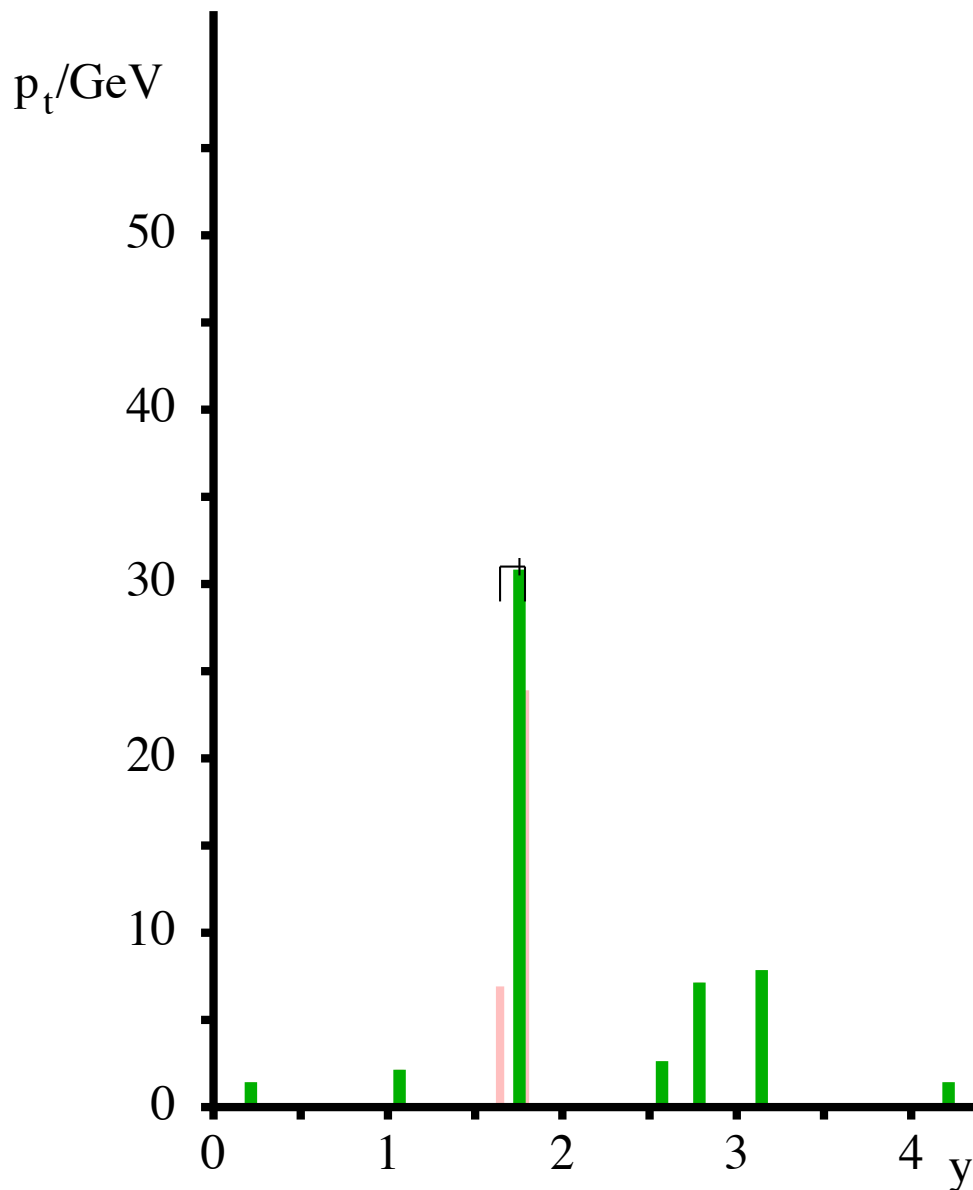


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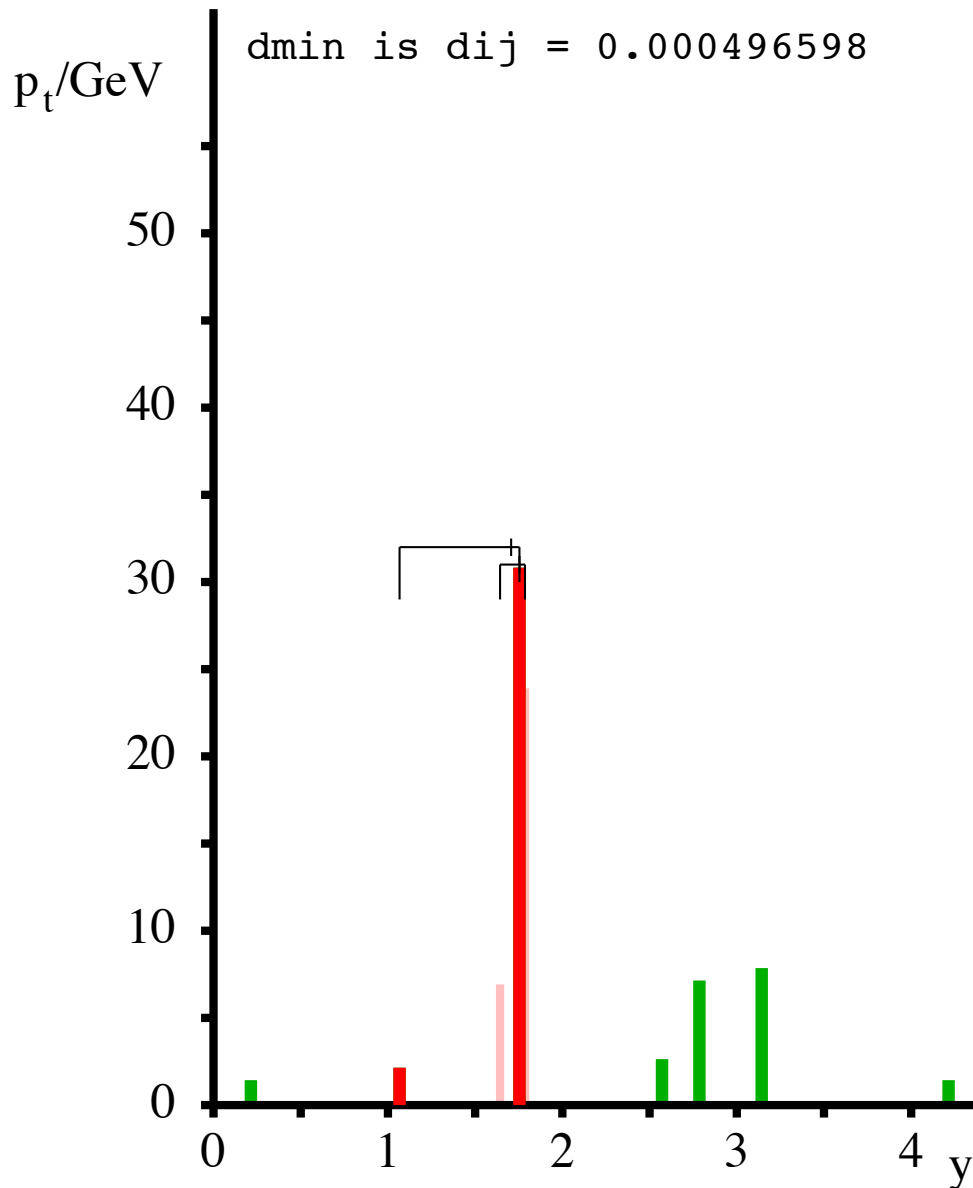
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d_{\min} is $d_{ij} = 0.000496598$

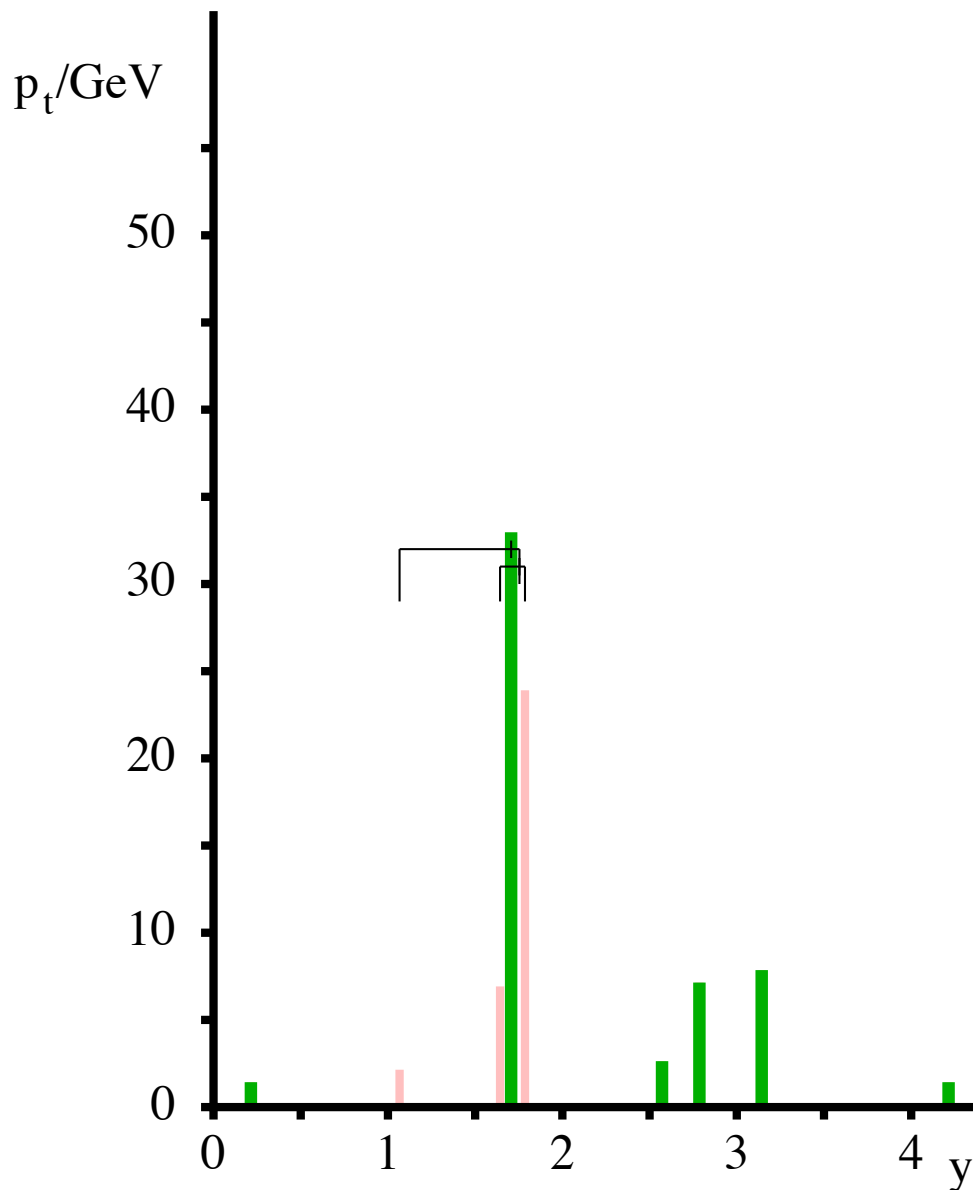


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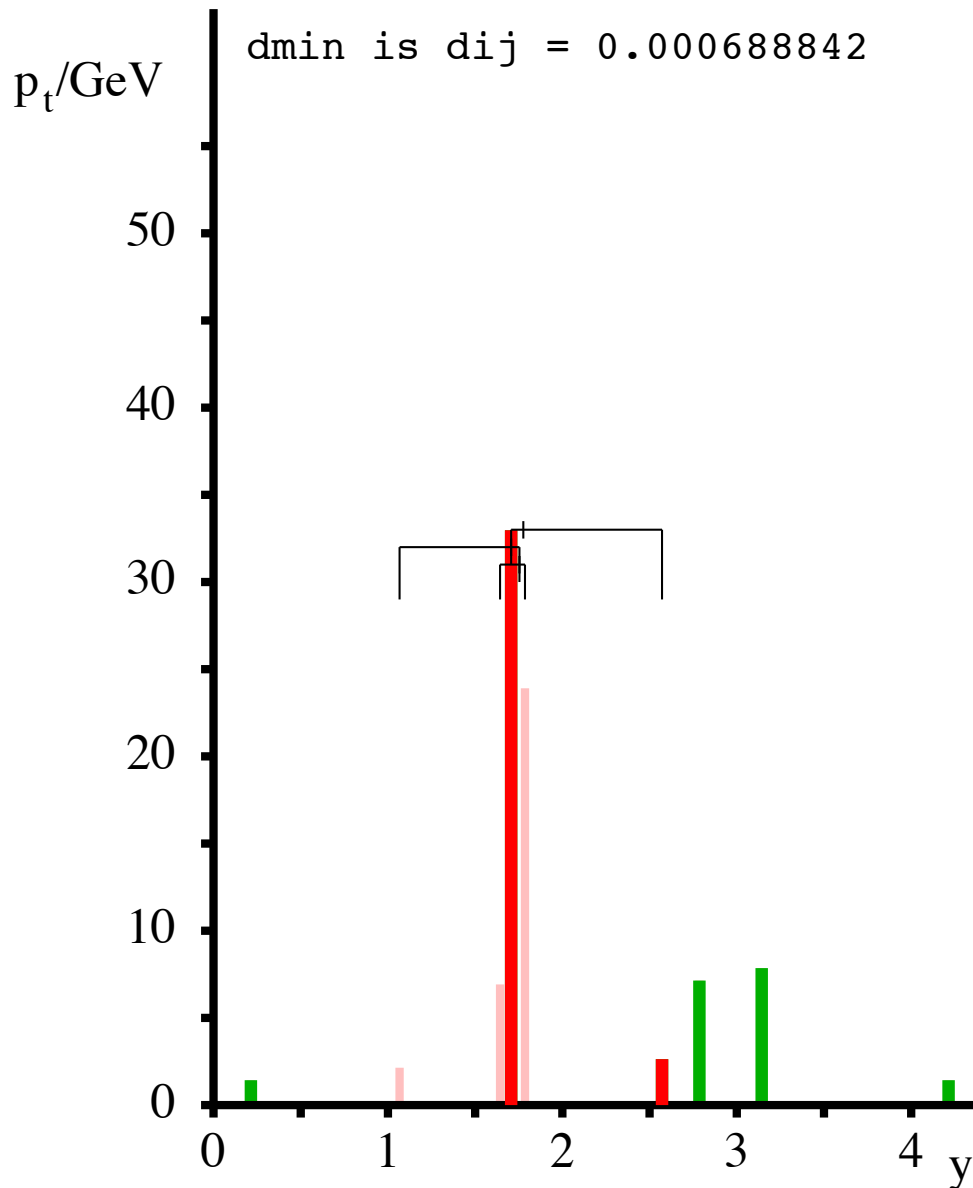
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anti- k_t algorithm

d_{\min} is $d_{ij} = 0.000688842$

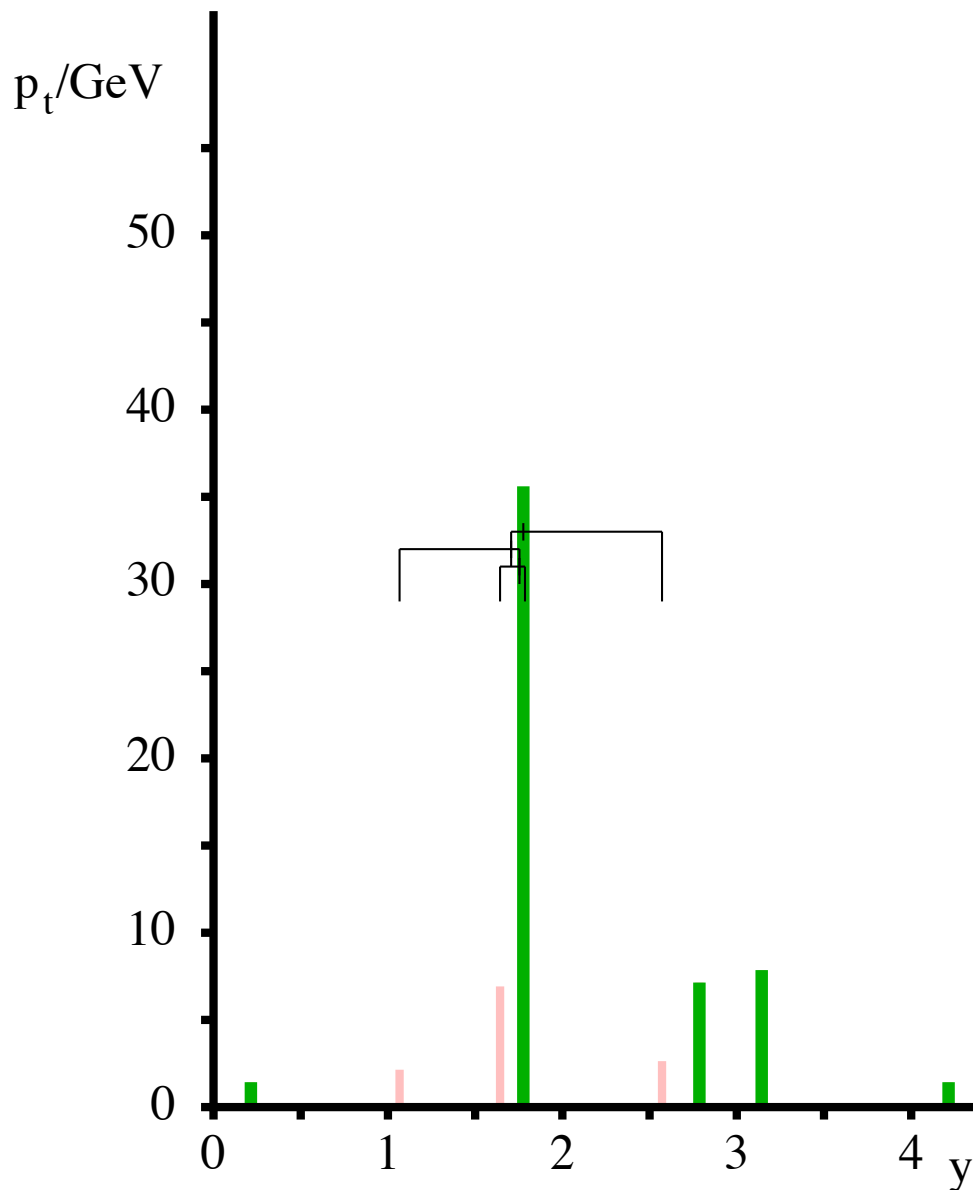


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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



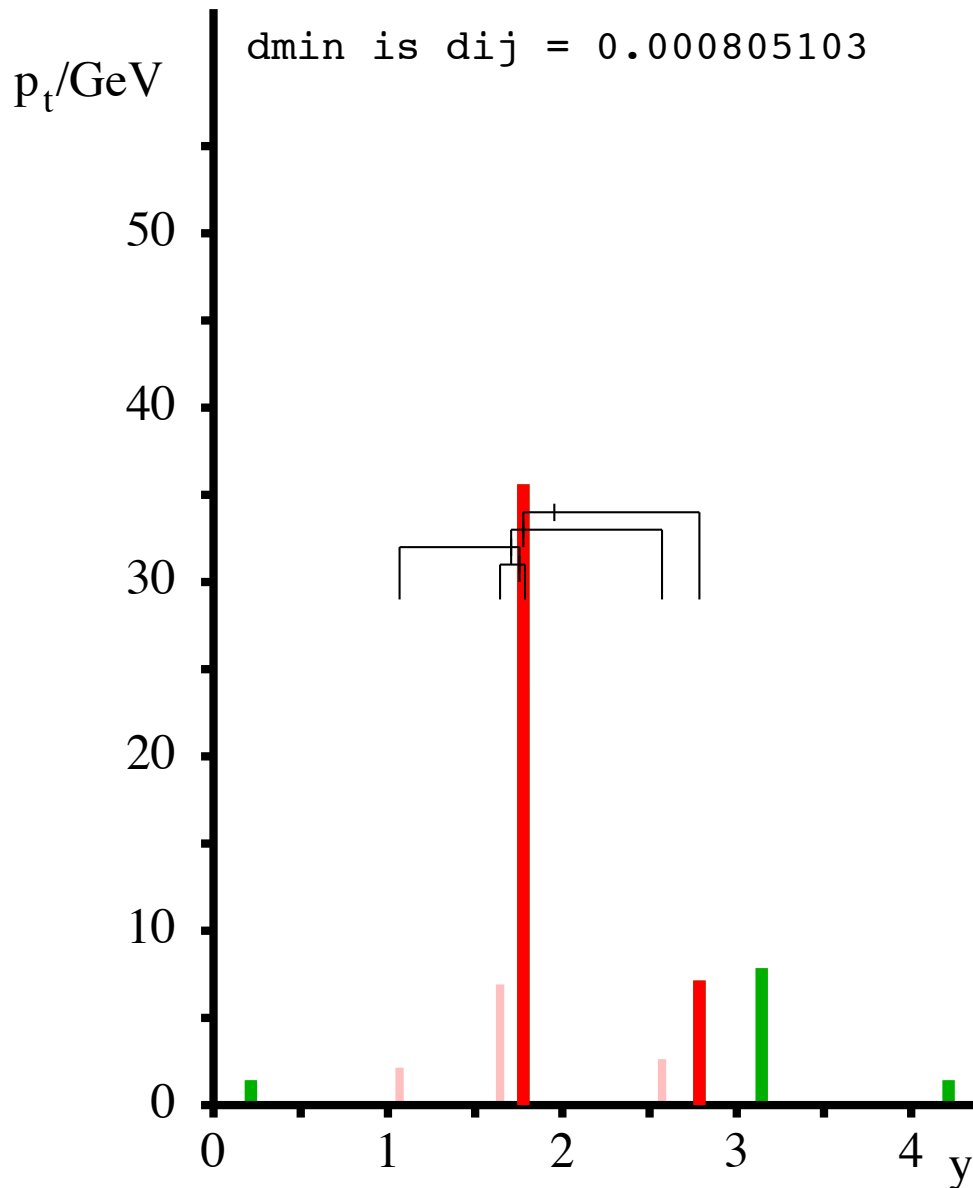
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d_{\min} is $d_{ij} = 0.000805103$



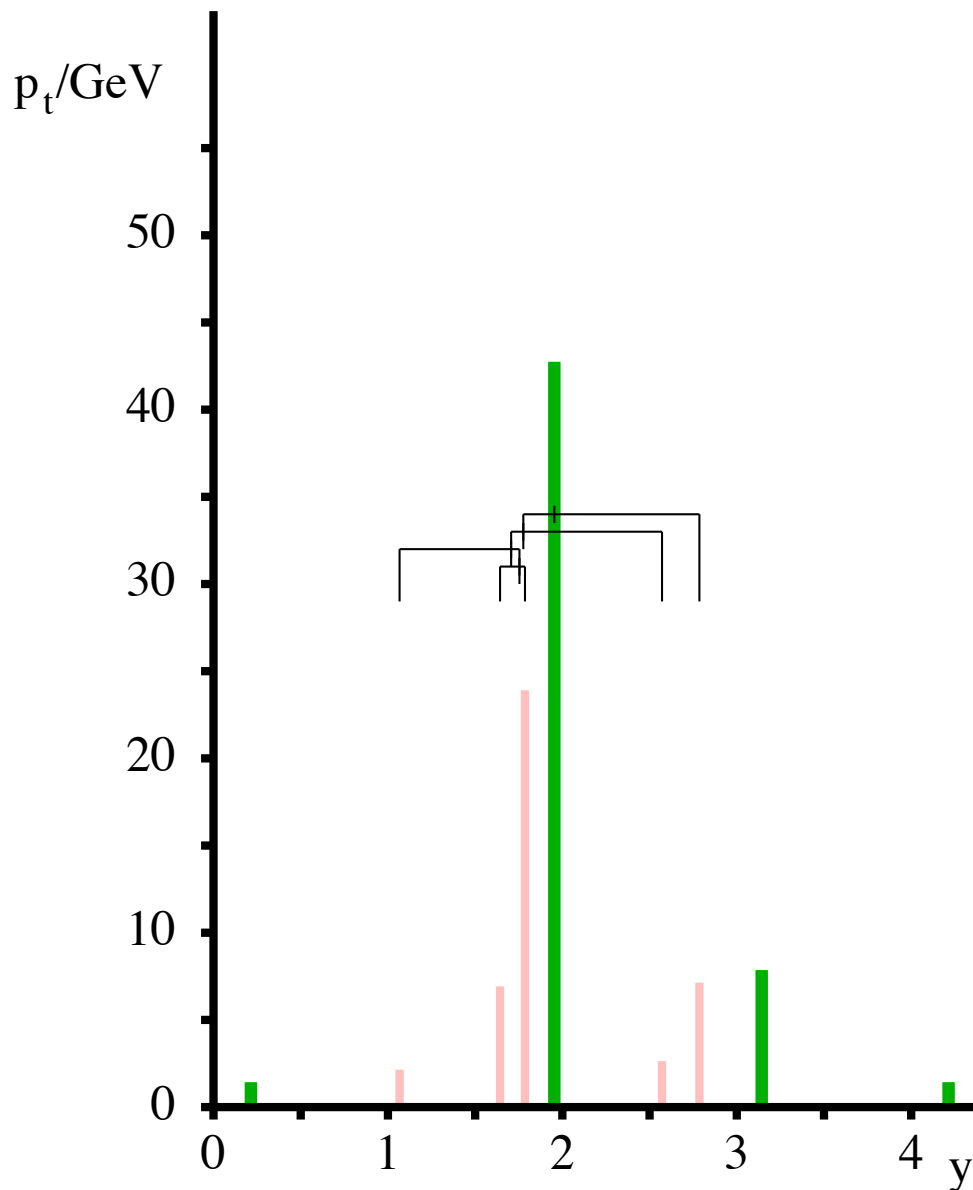
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Anti- k_t gradually makes its way through the secondary blob \rightarrow no clear identification of substructure associated with 2nd parton.

Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



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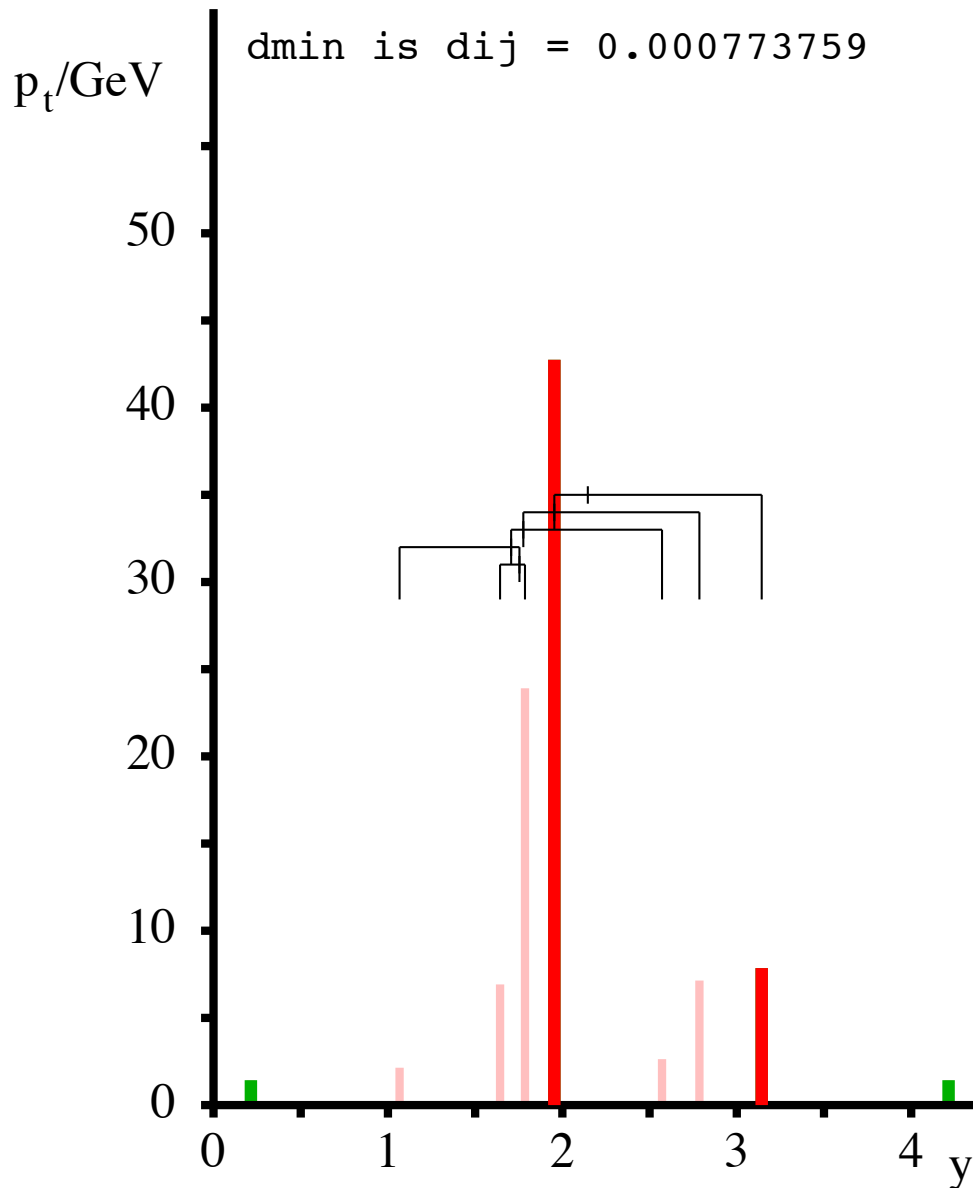
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm

d_{\min} is $d_{ij} = 0.000773759$



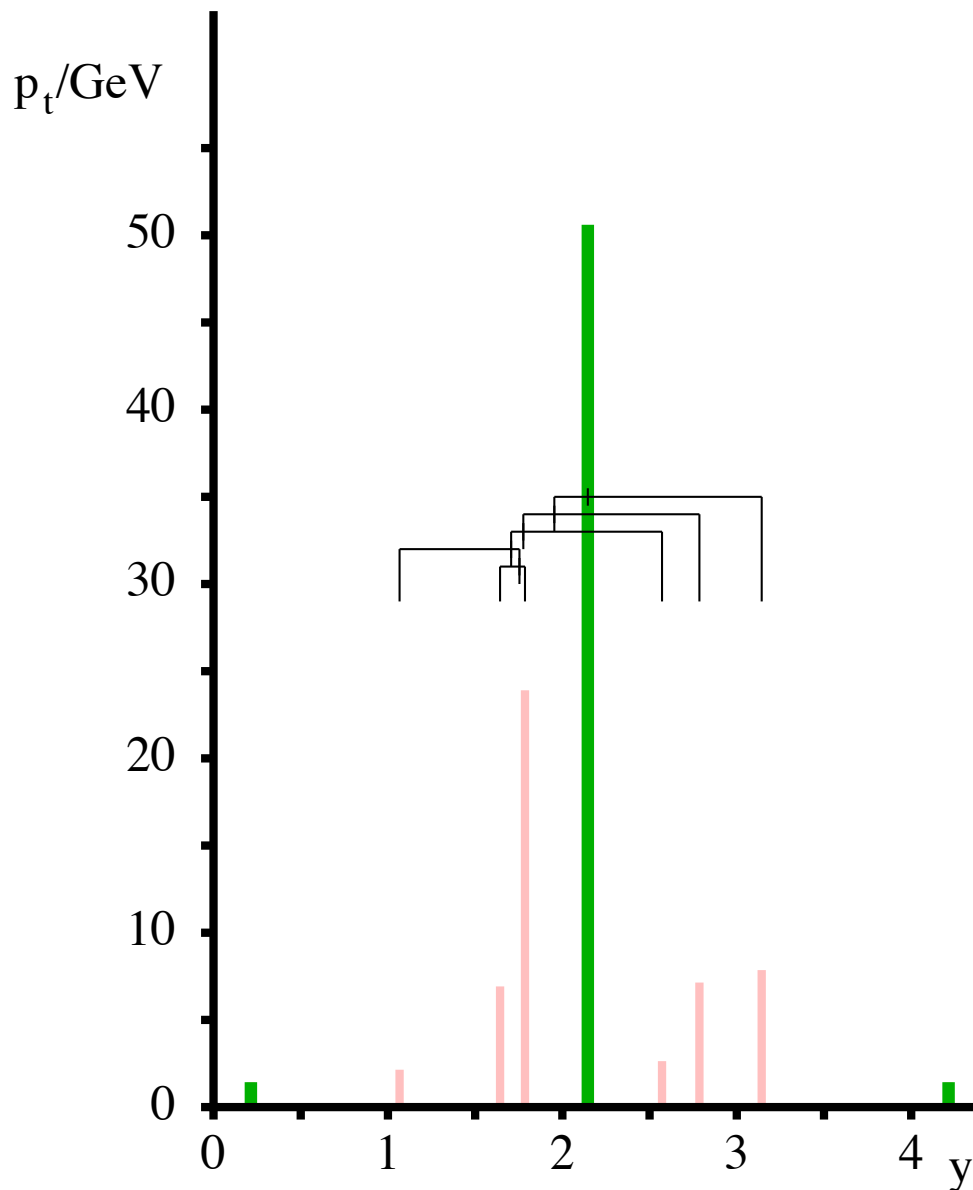
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



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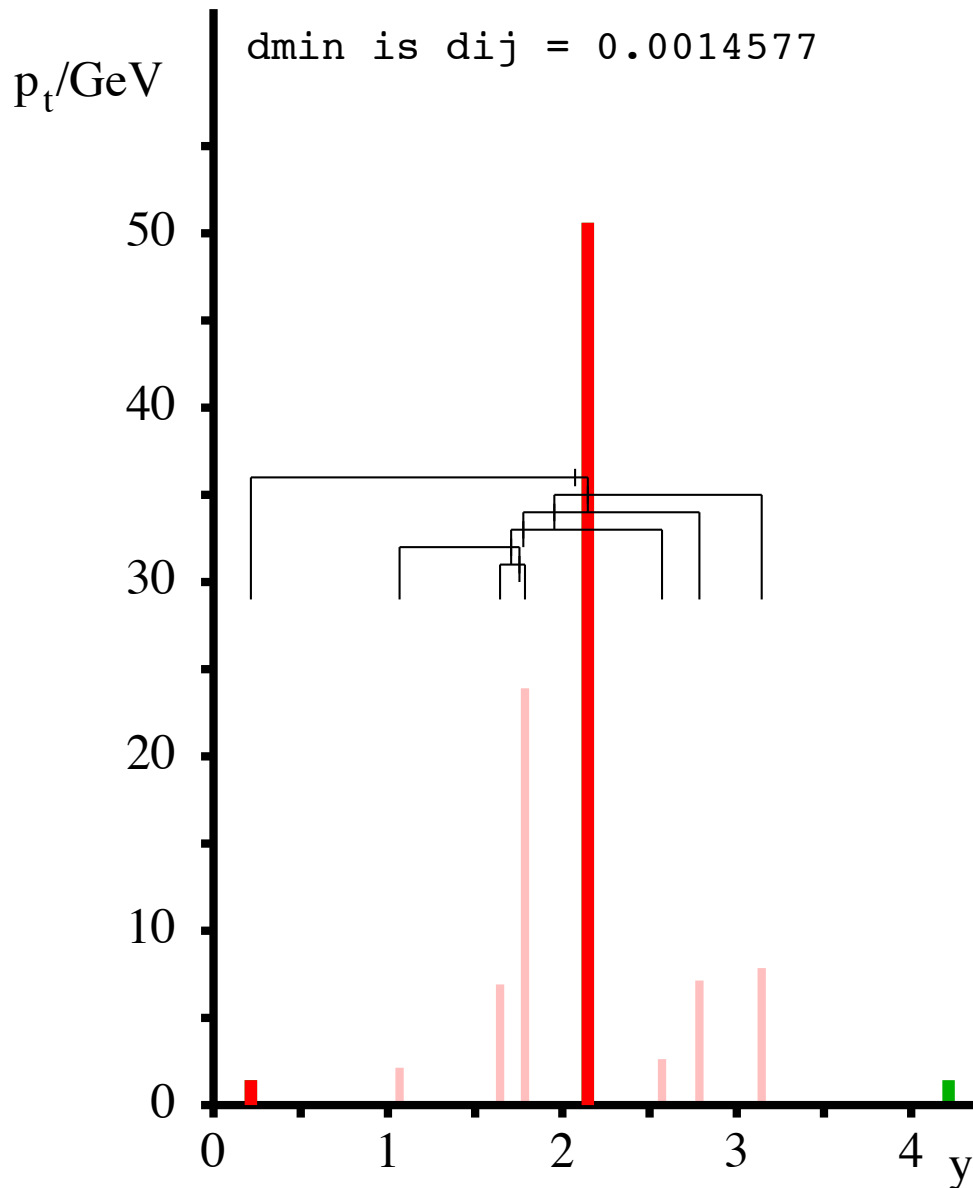
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm

d_{\min} is $d_{ij} = 0.0014577$



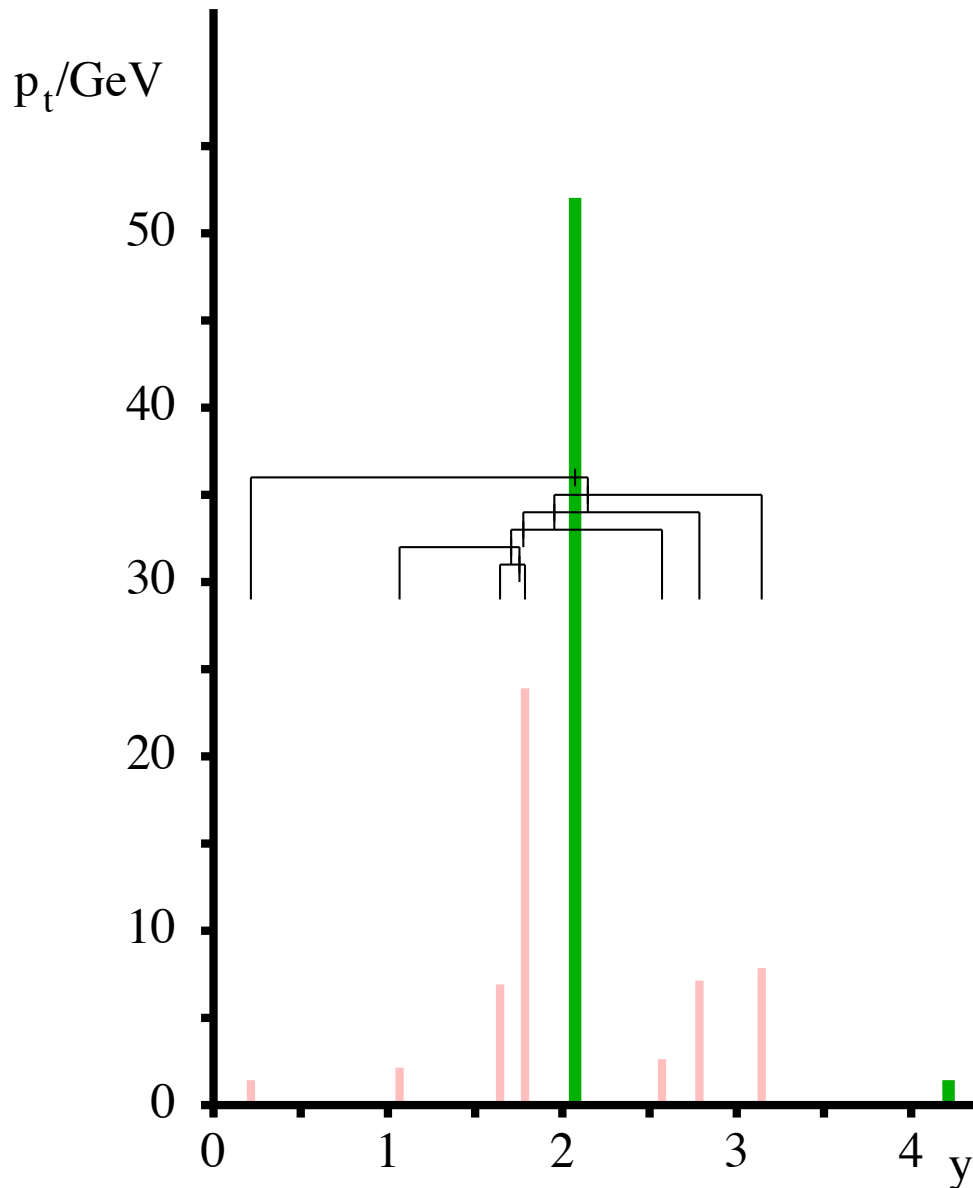
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anti- k_t algorithm



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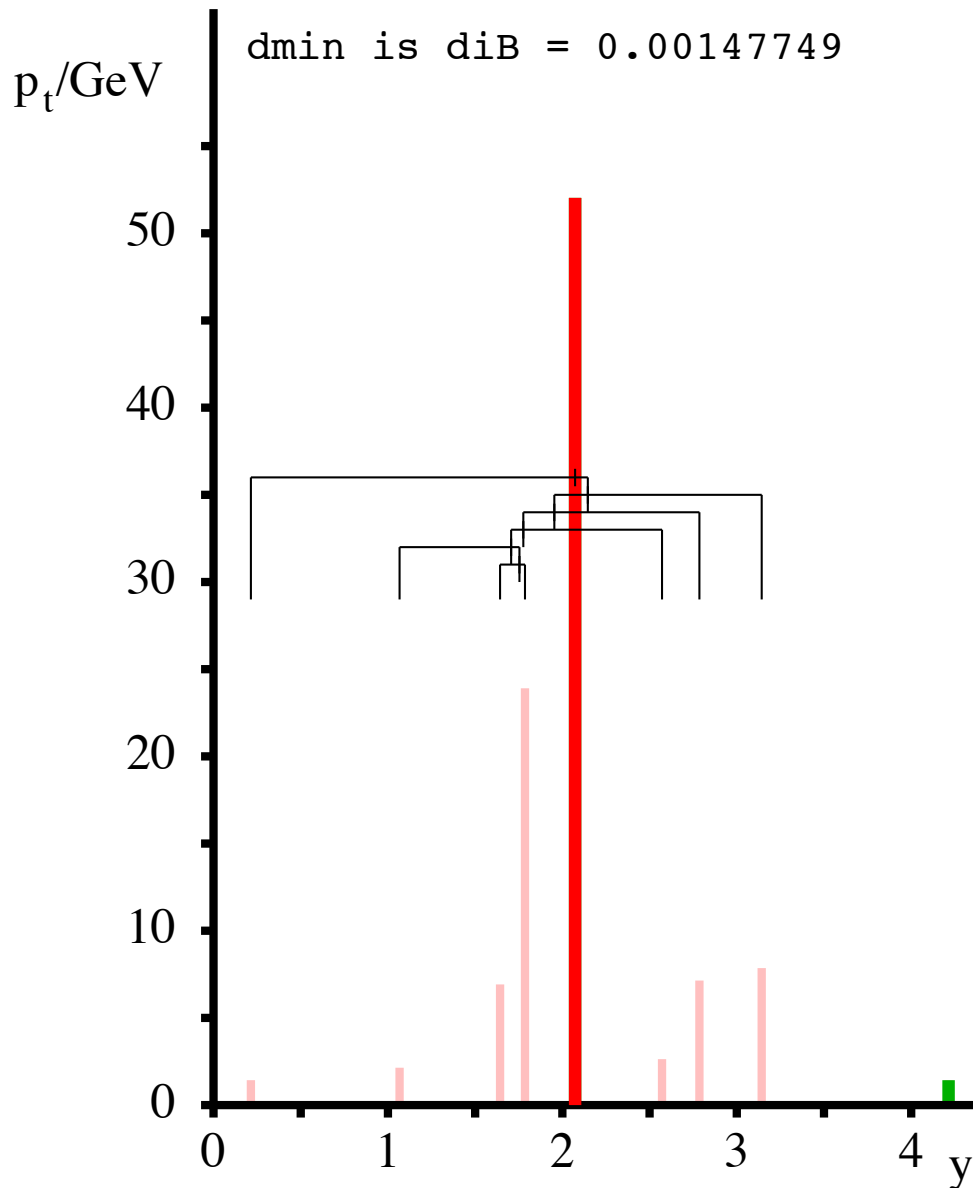
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm

d_{\min} is $d_{iB} = 0.00147749$



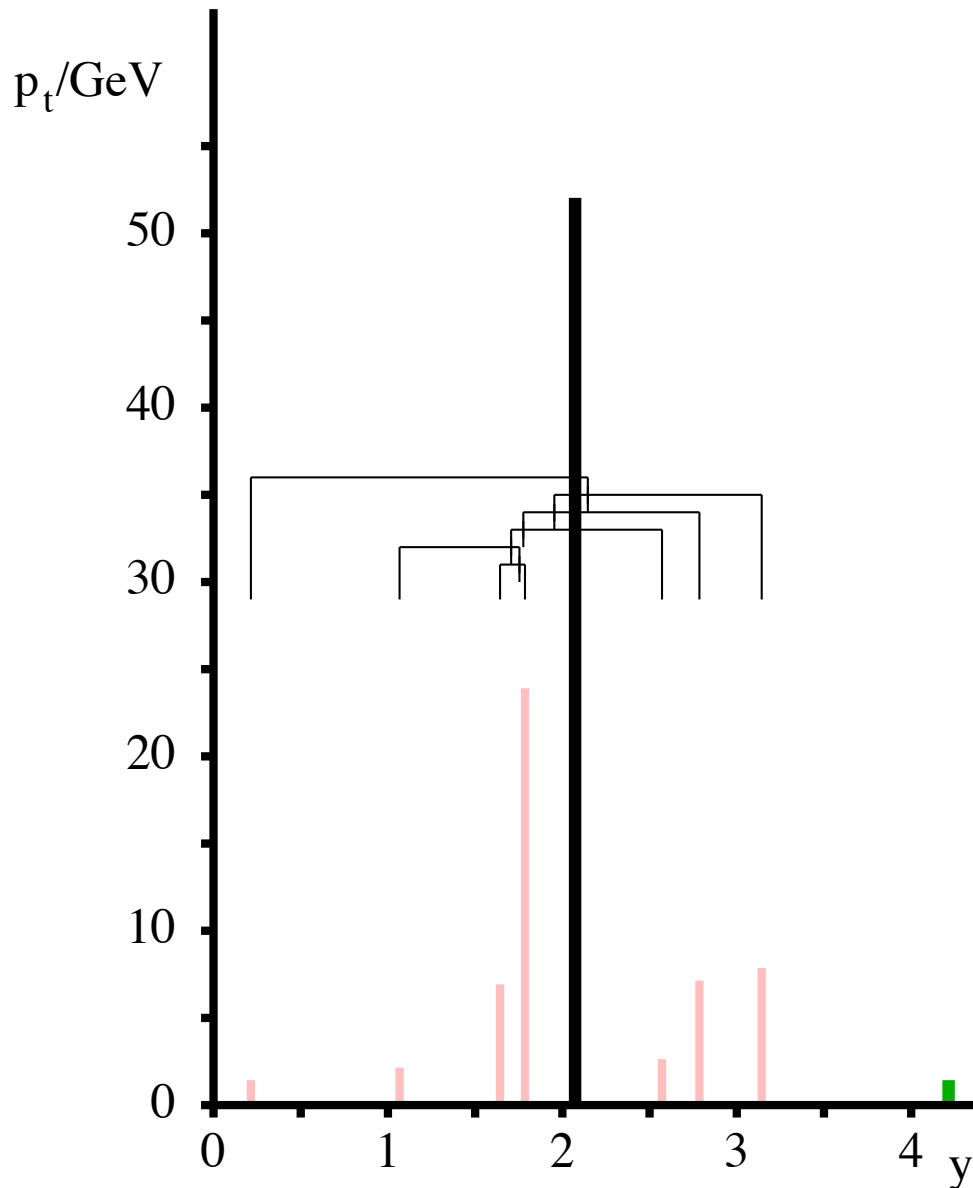
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



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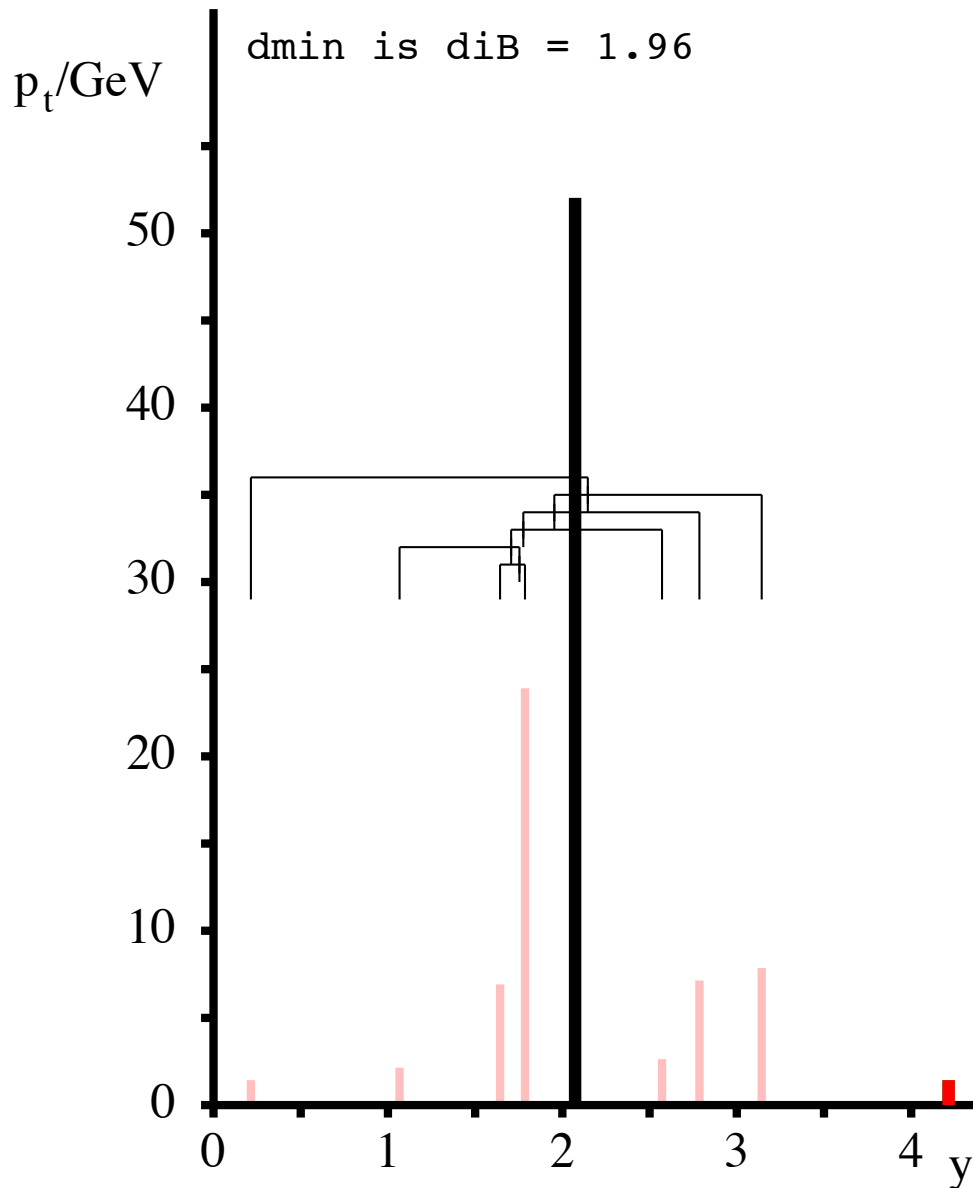
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm

d_{\min} is $d_{iB} = 1.96$



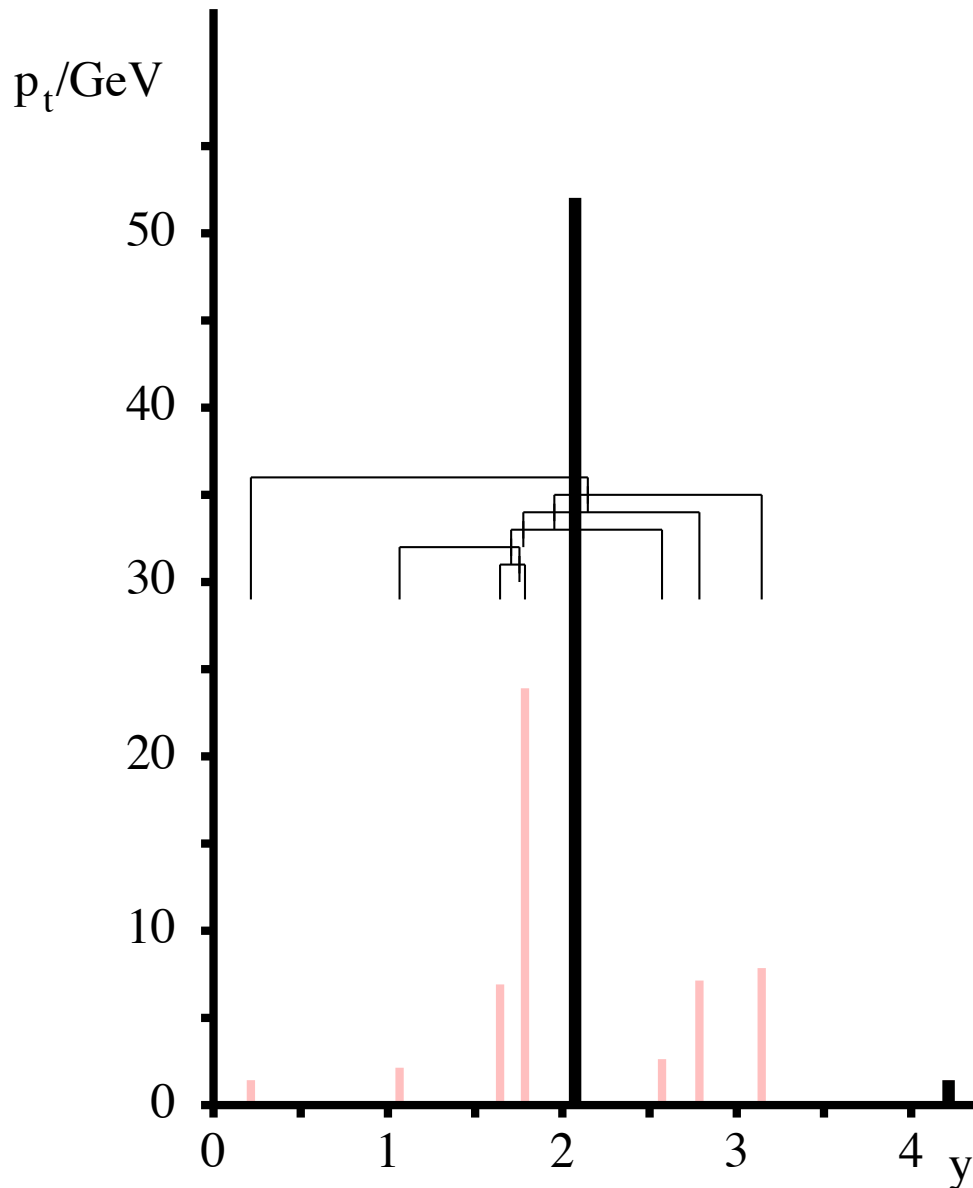
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Identifying jet substructure: try out anti- k_t

anti- k_t algorithm



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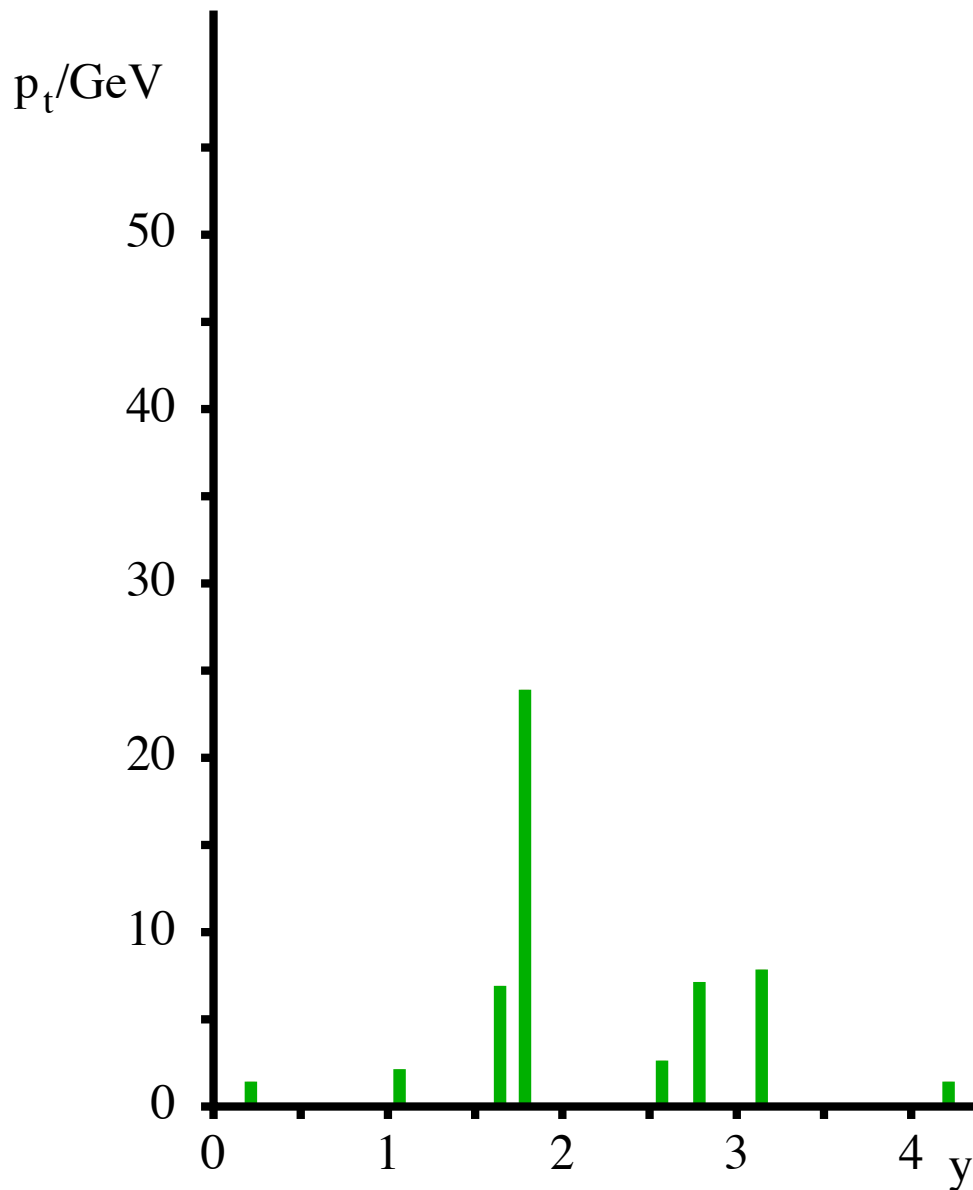
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kt

Identifying jet substructure: try out k_t

k_t algorithm



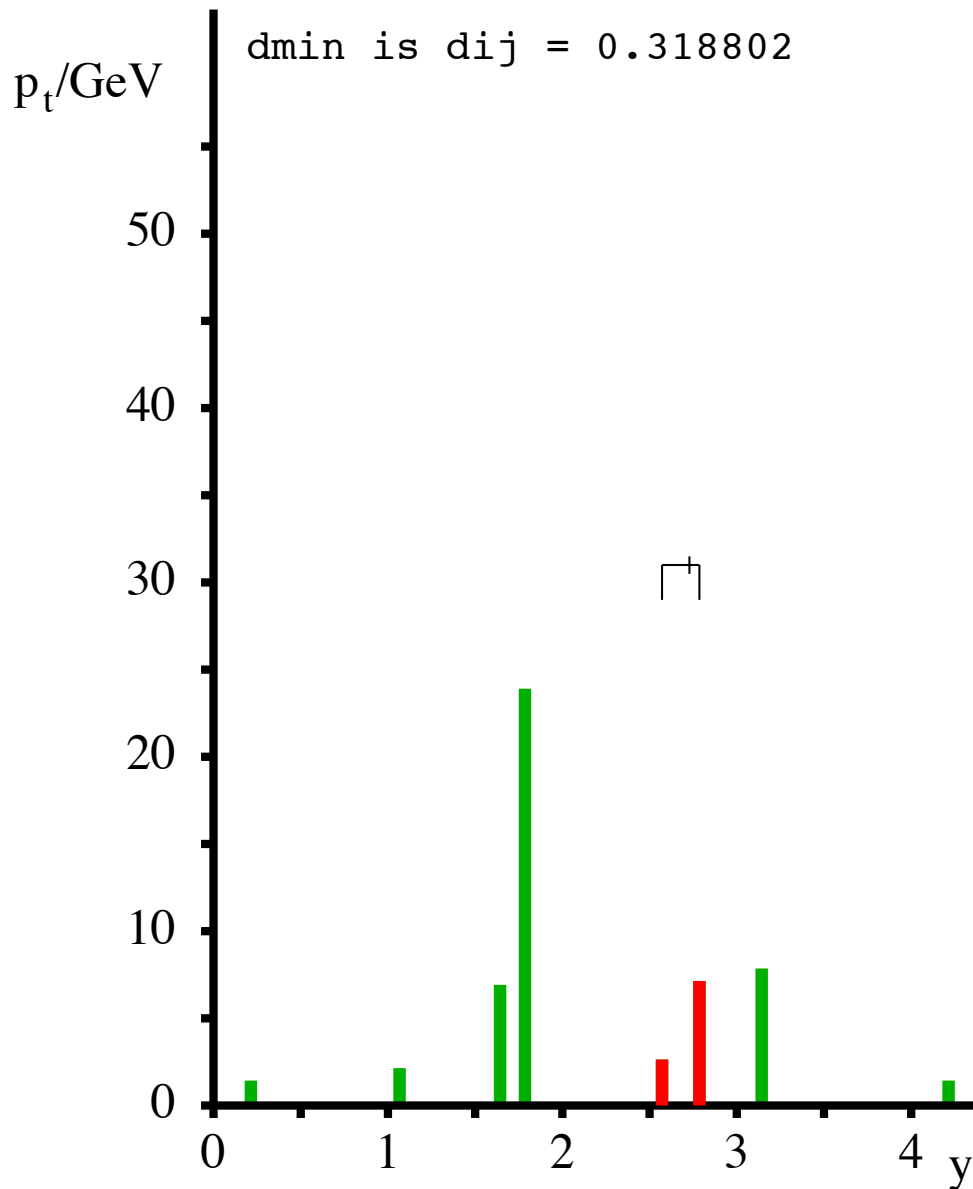
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Identifying jet substructure: try out k_t

k_t algorithm

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

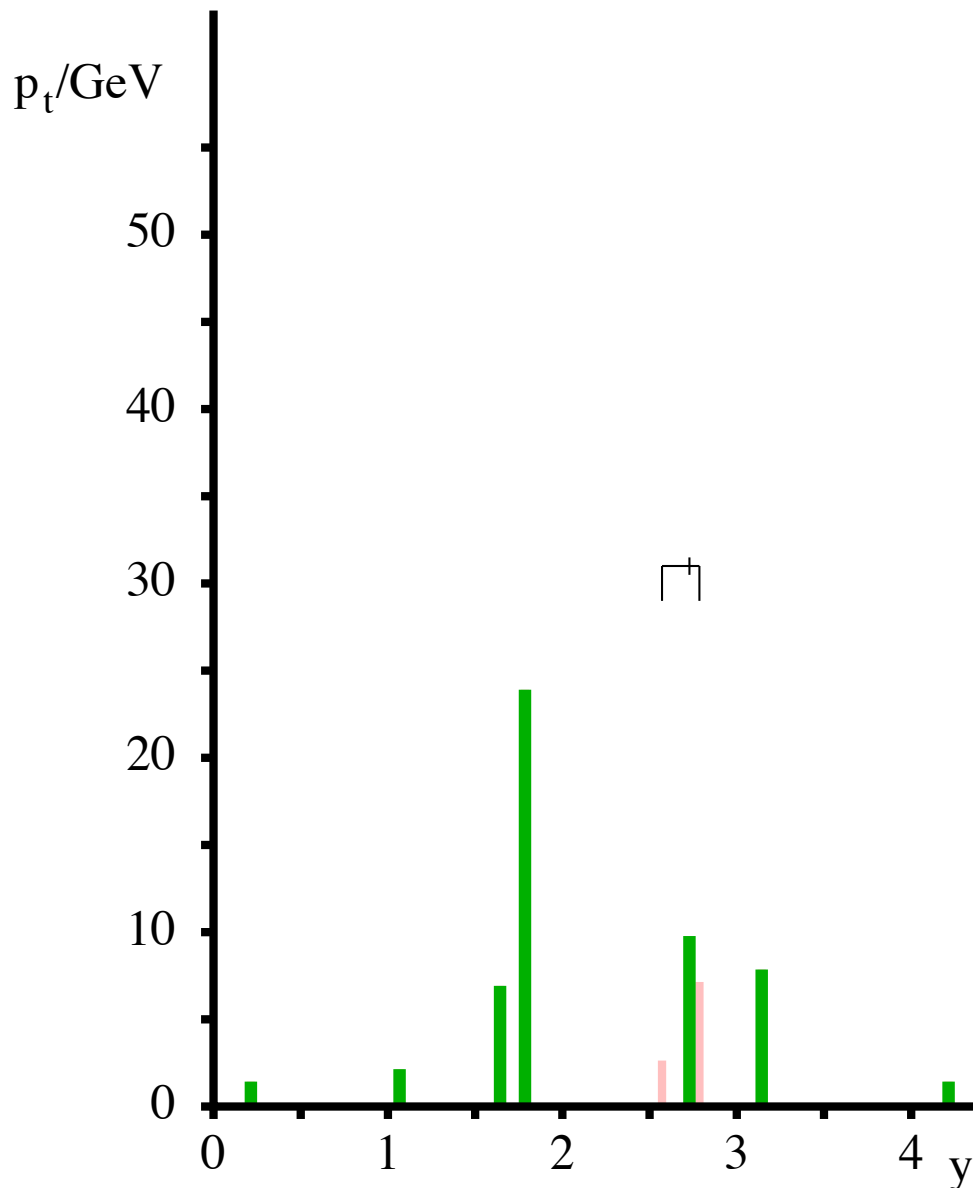


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Identifying jet substructure: try out k_t

k_t algorithm



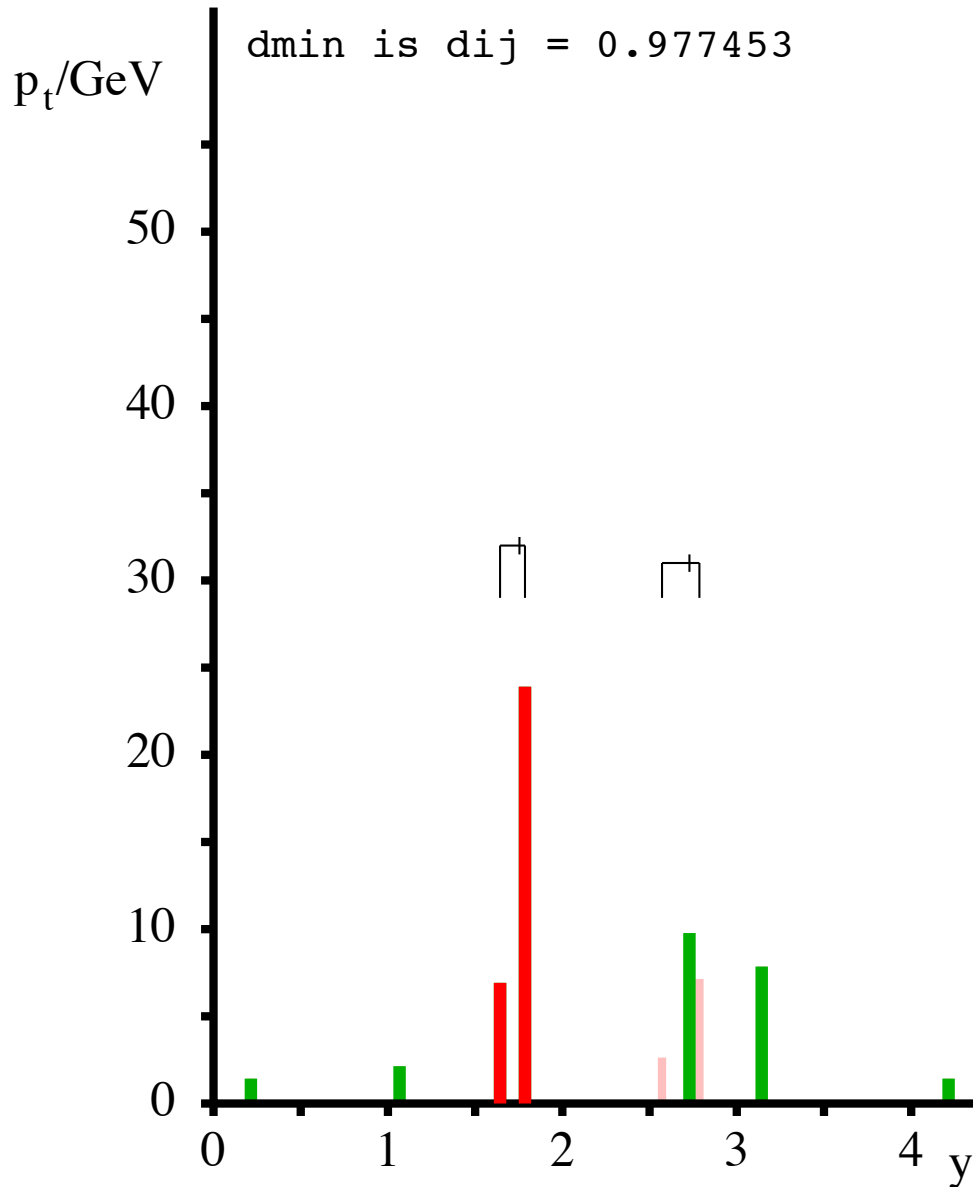
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Identifying jet substructure: try out k_t

k_t algorithm

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

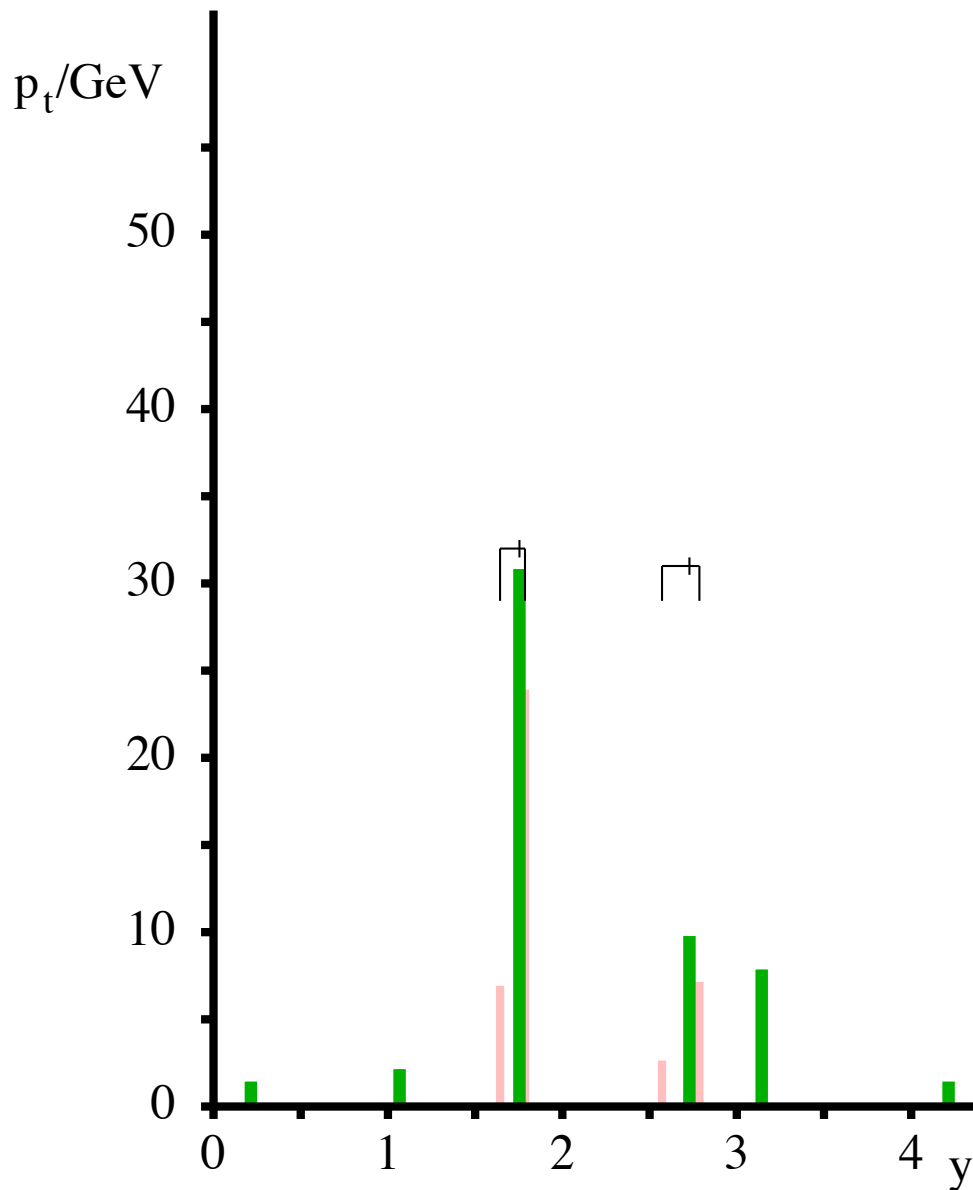


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Identifying jet substructure: try out k_t

k_t algorithm



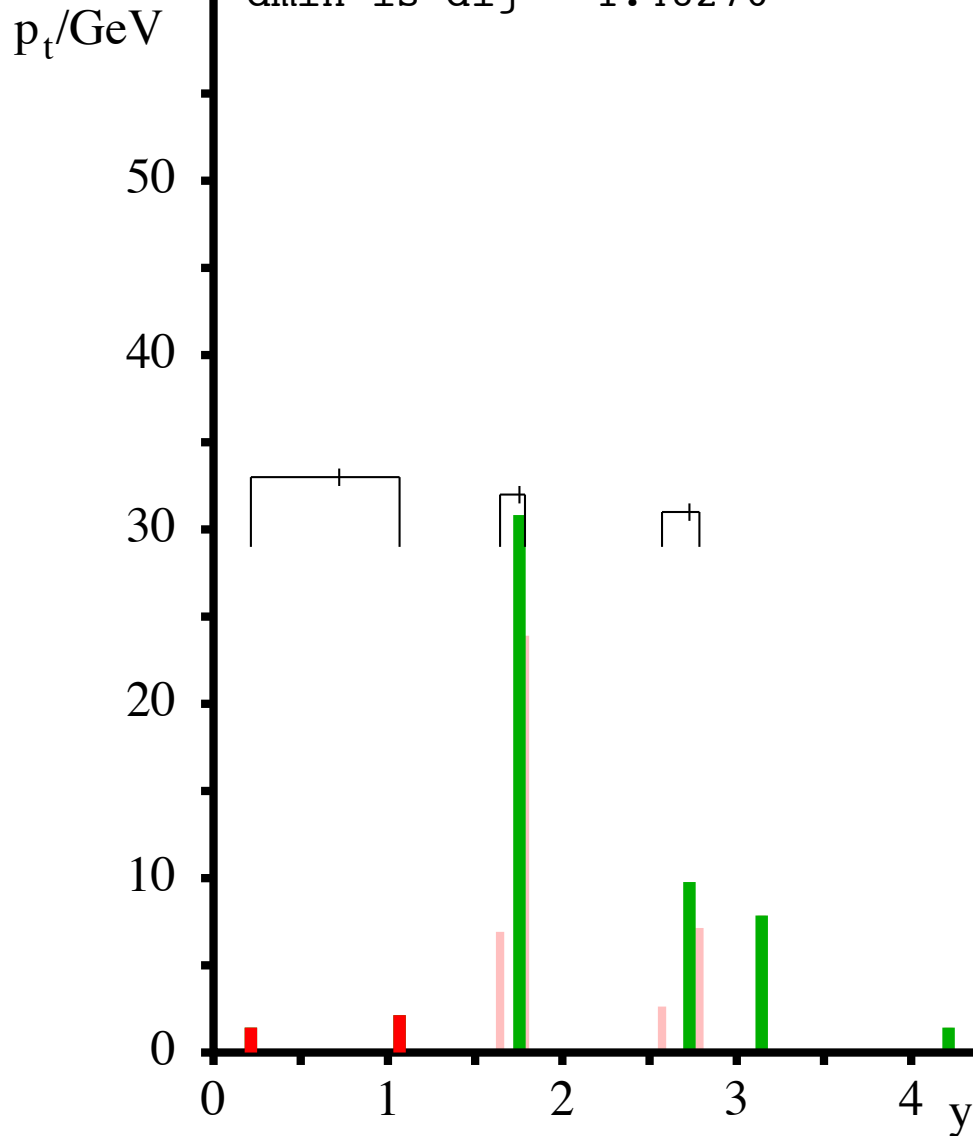
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Identifying jet substructure: try out k_t

k_t algorithm

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



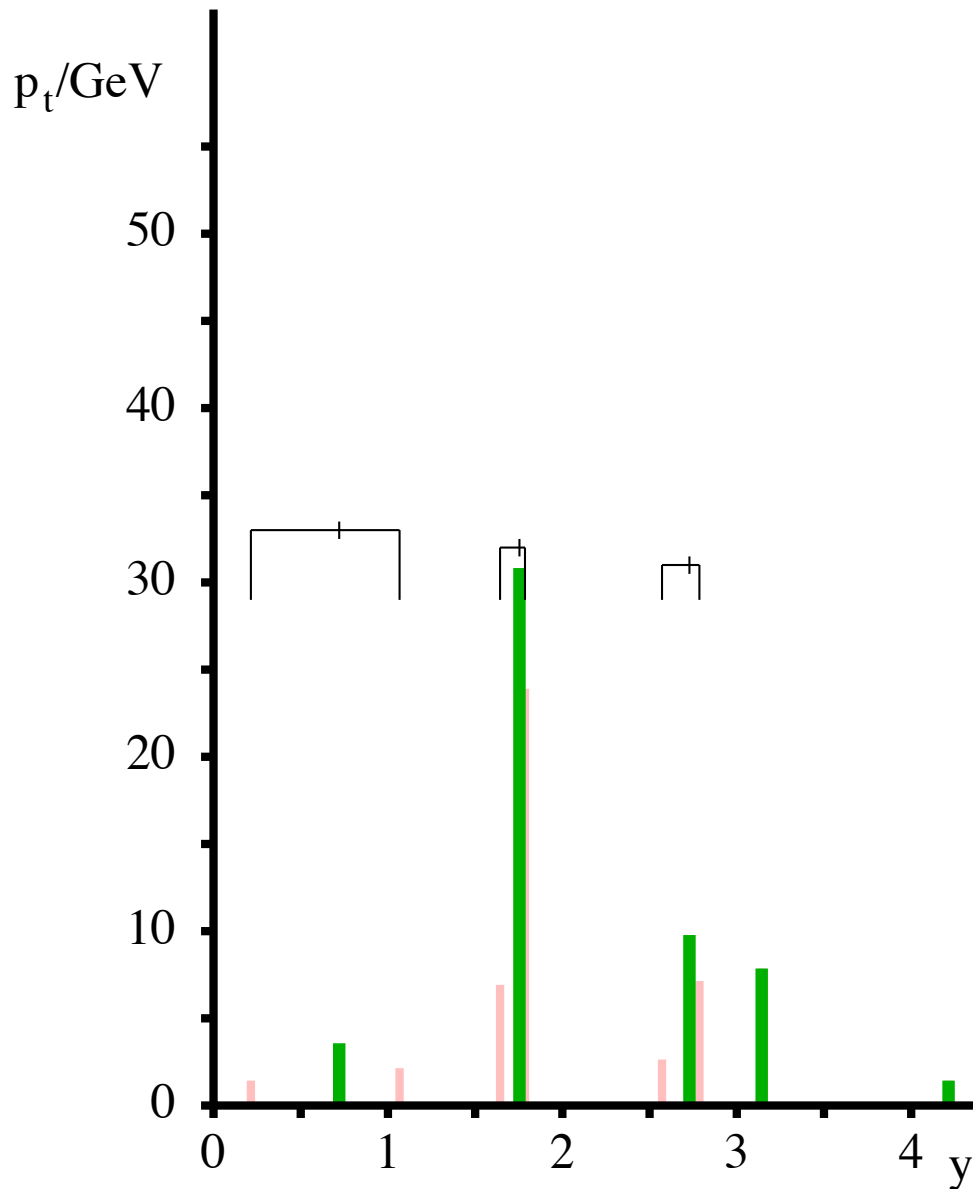
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k_t clusters soft “junk” early on in the clustering

Identifying jet substructure: try out k_t

k_t algorithm



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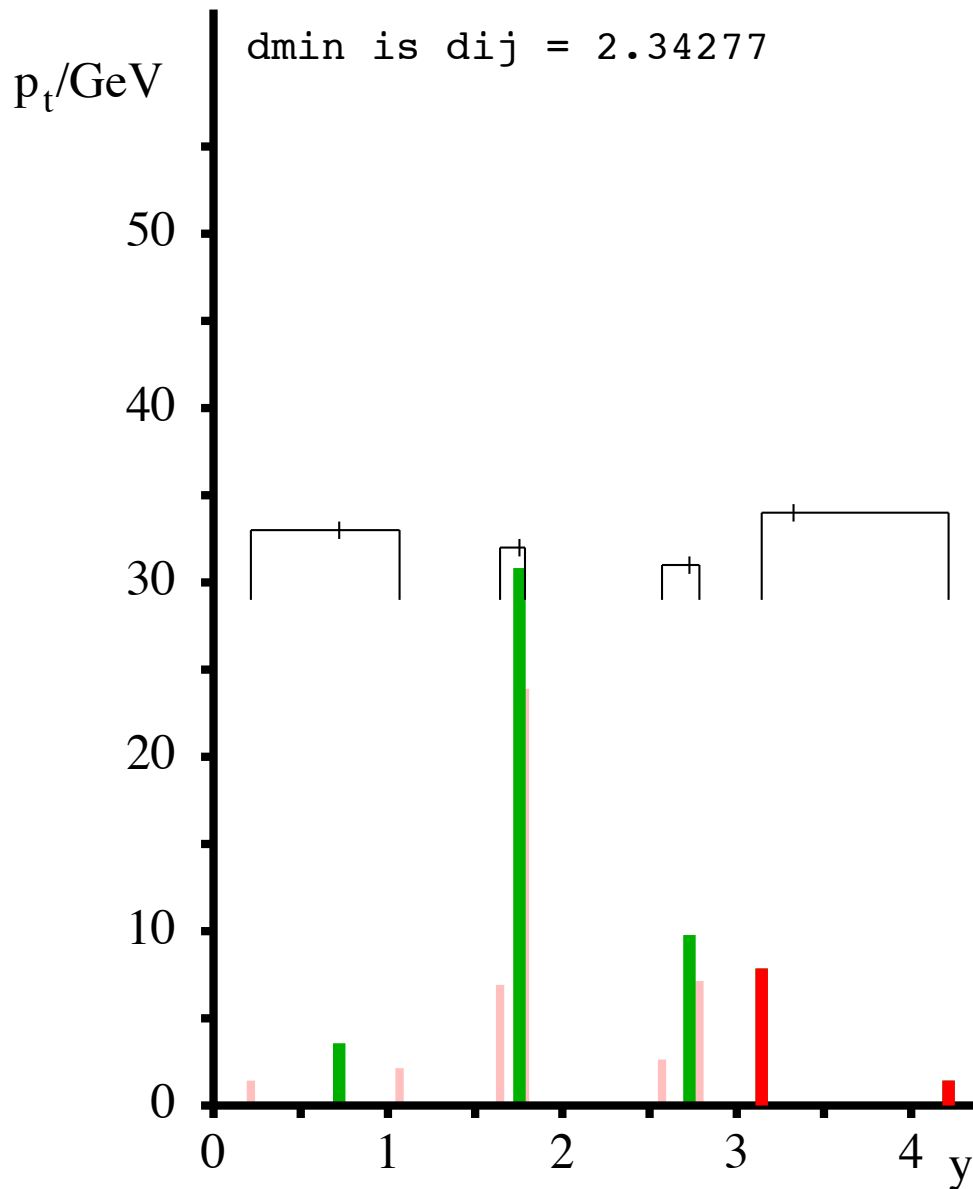
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Identifying jet substructure: try out k_t

k_t algorithm

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



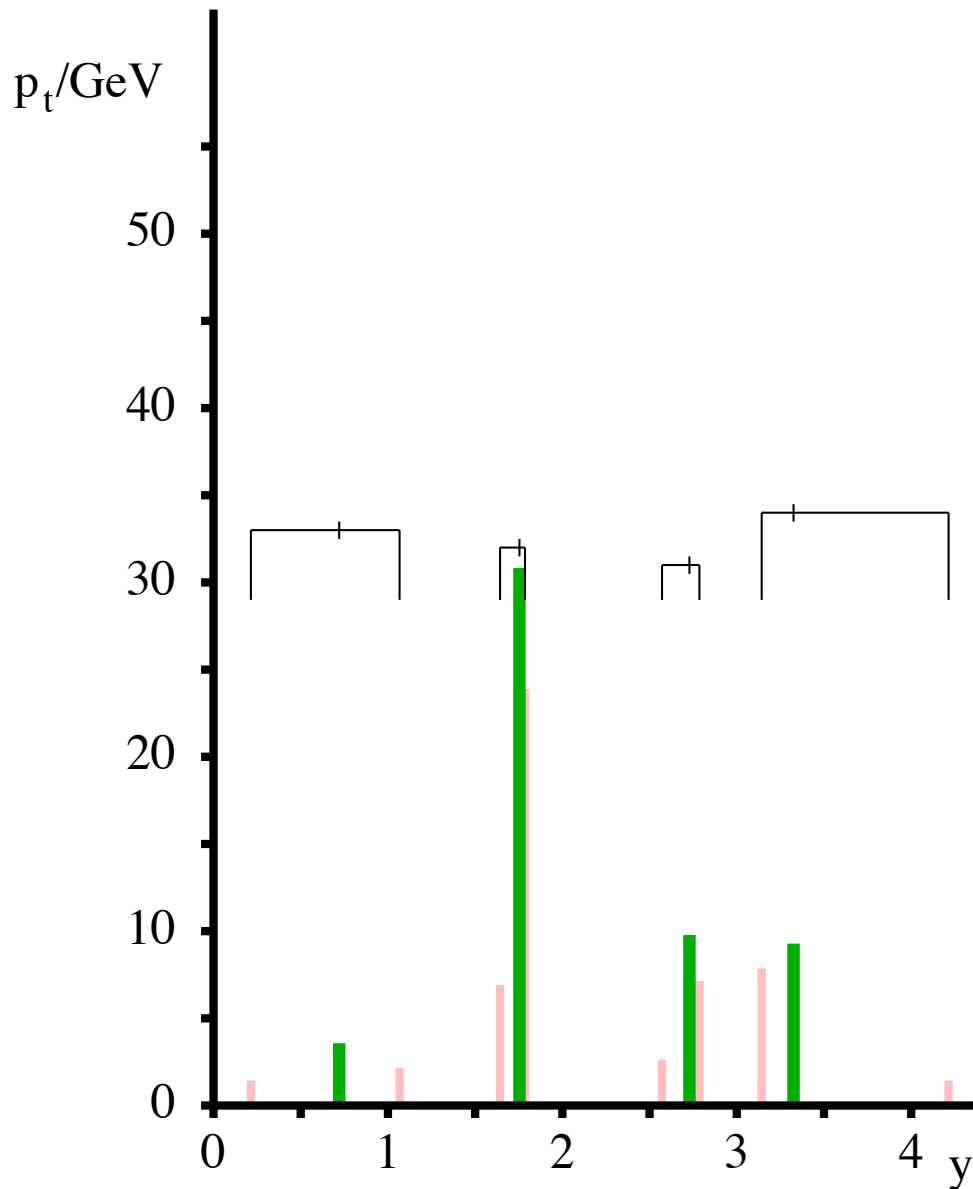
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Identifying jet substructure: try out k_t

k_t algorithm



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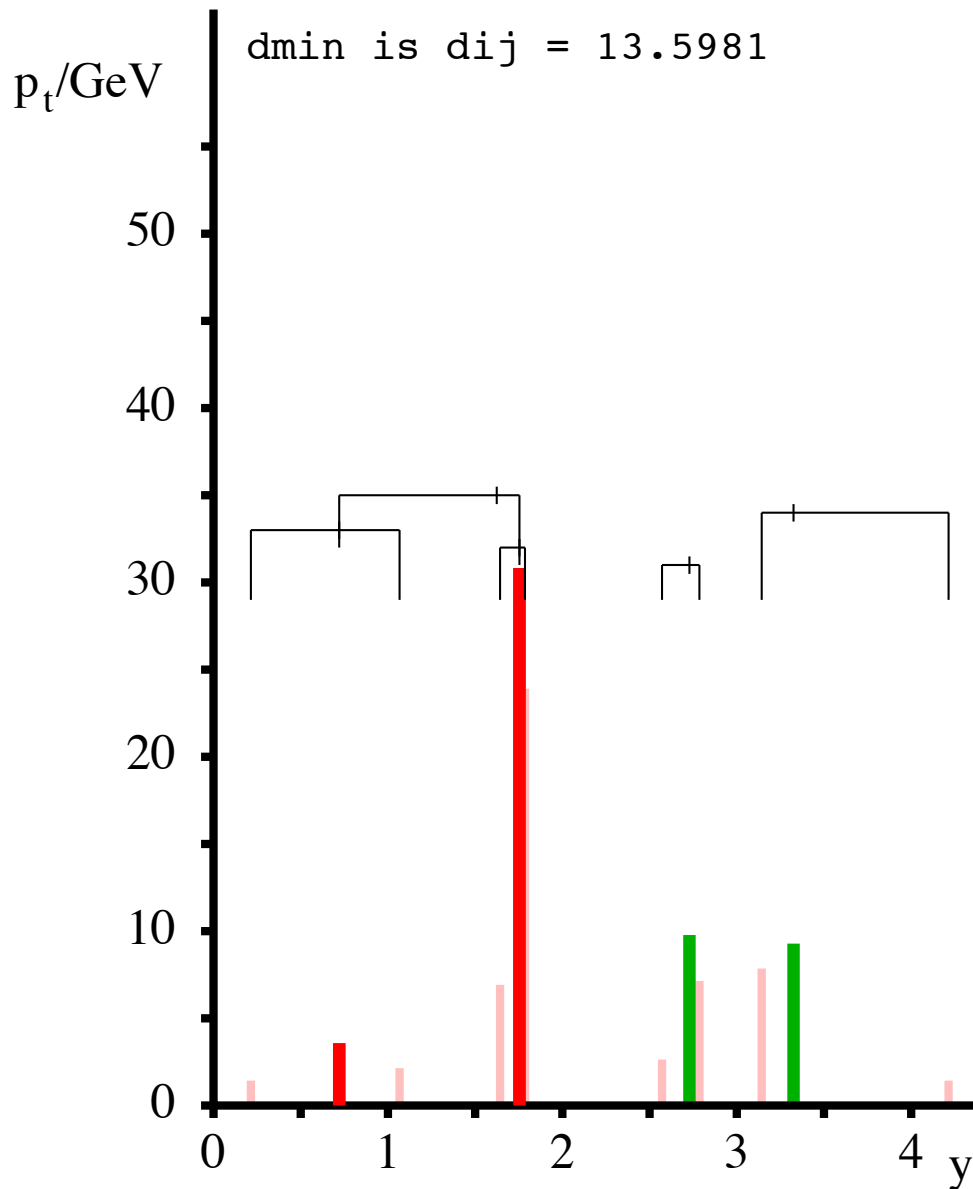
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Identifying jet substructure: try out k_t

k_t algorithm

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



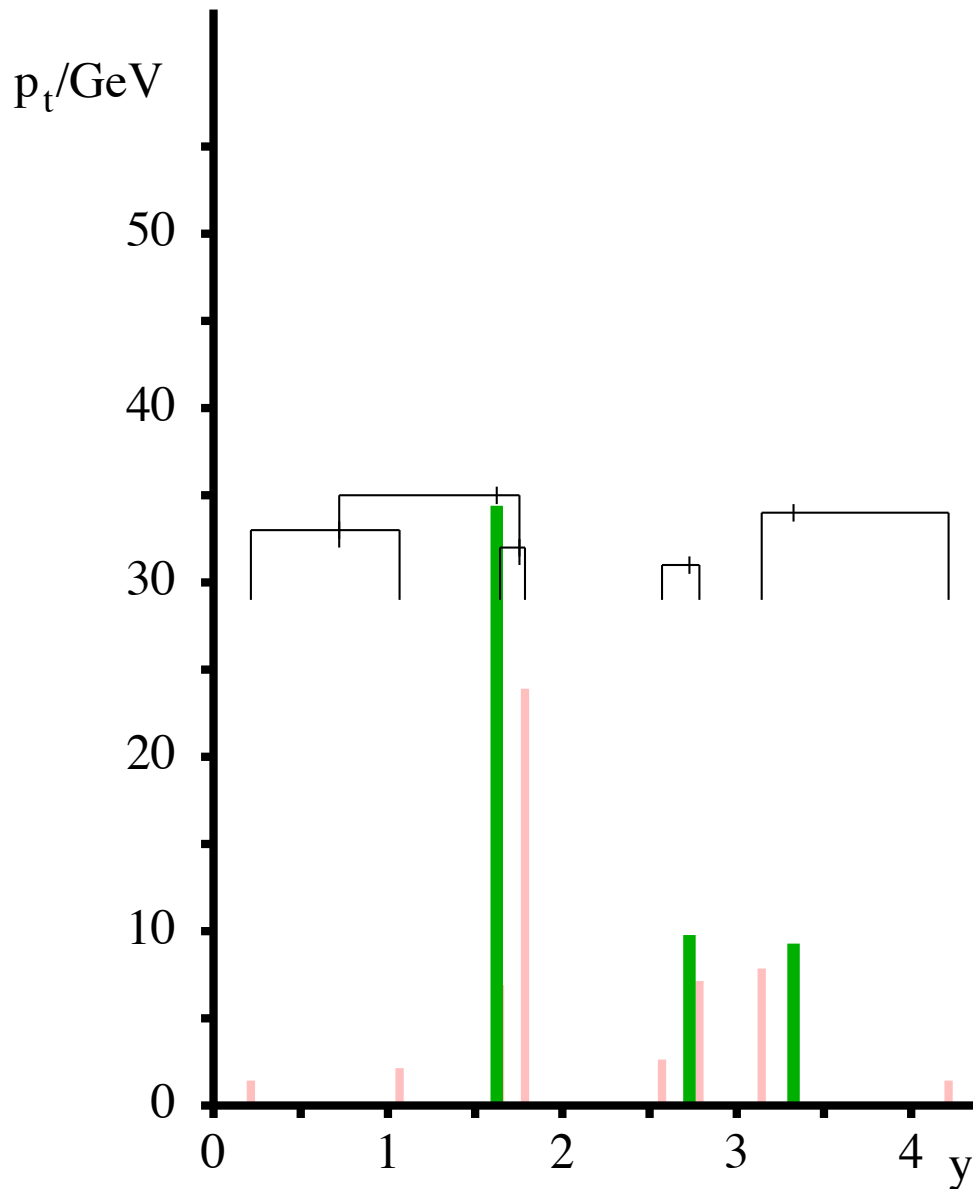
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k_t algorithm



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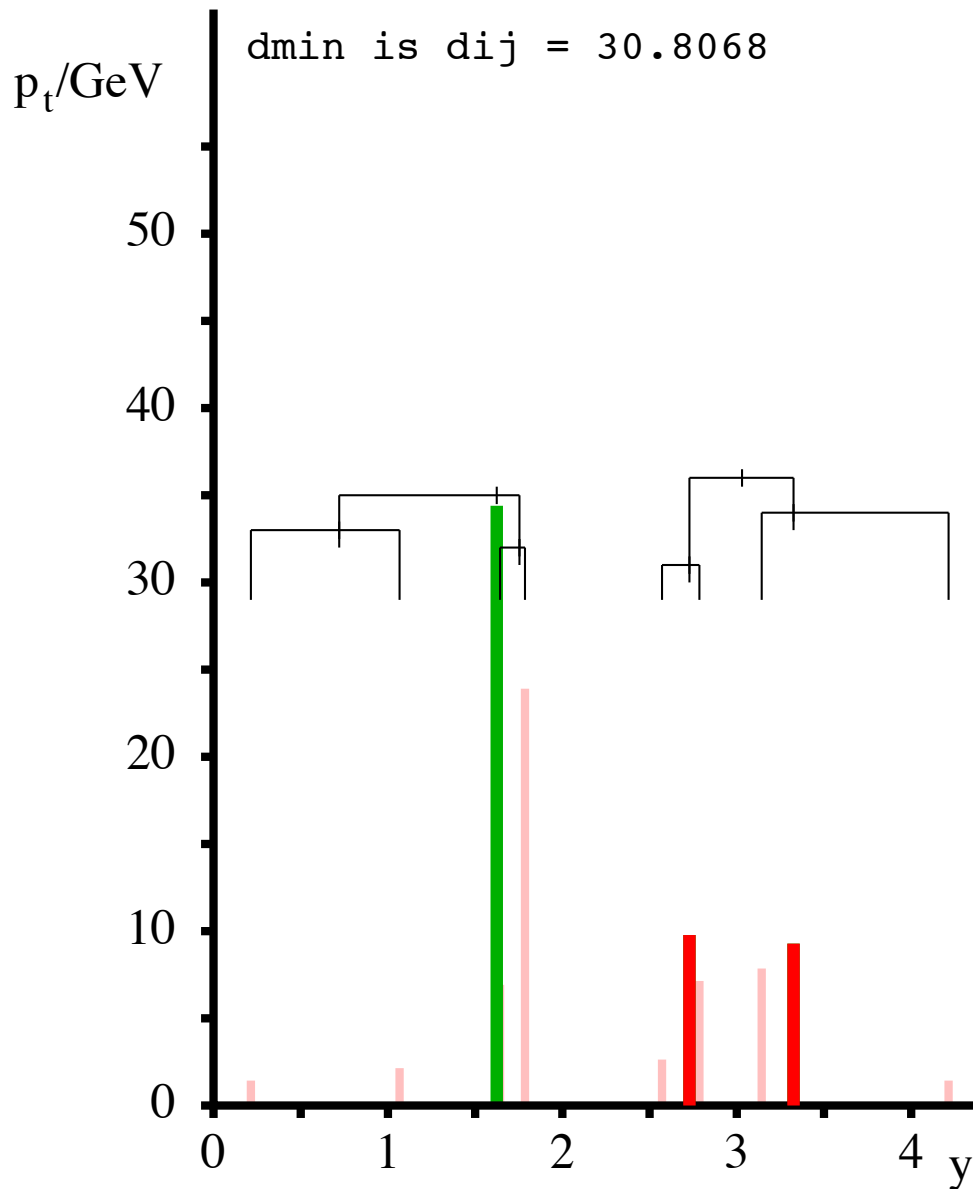
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Identifying jet substructure: try out k_t

k_t algorithm

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



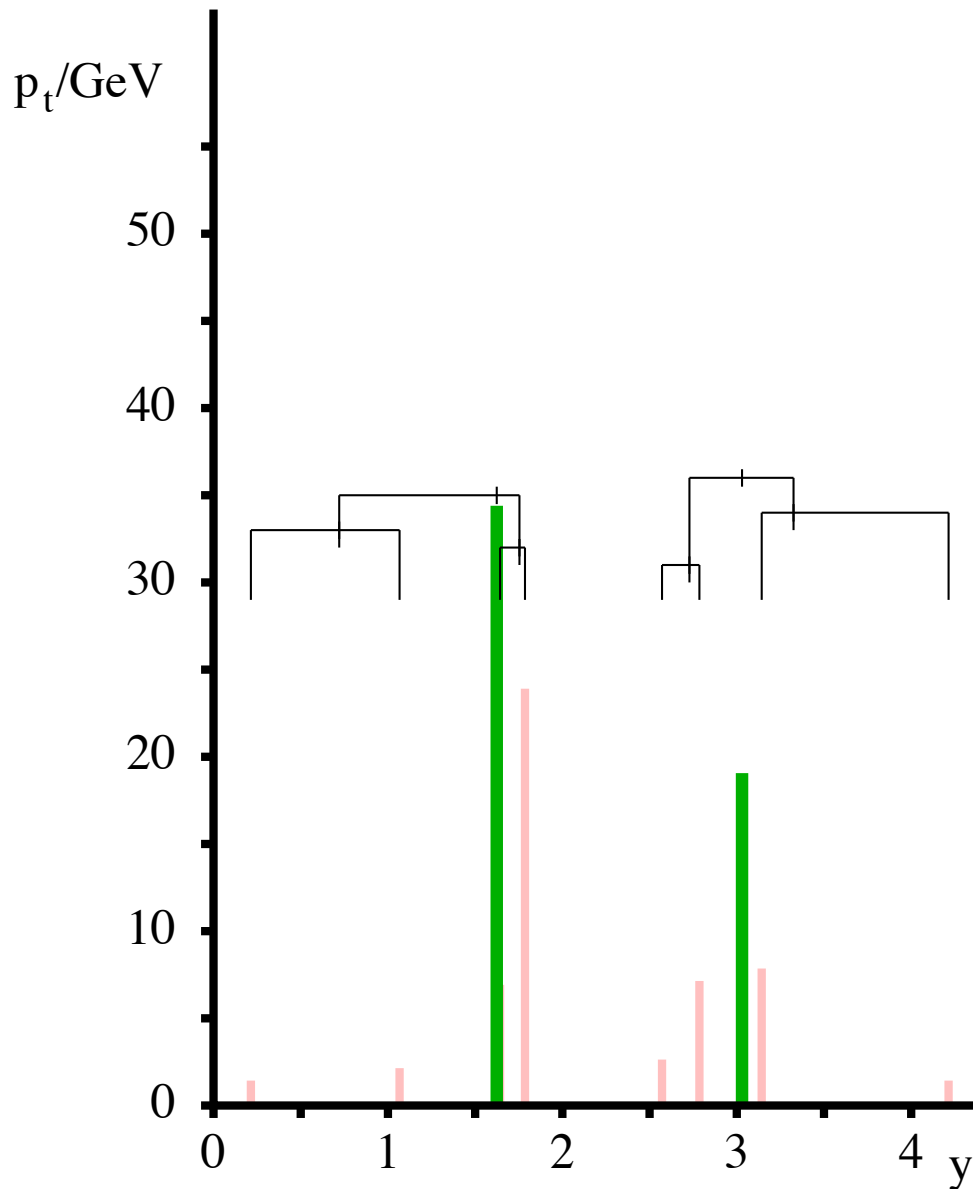
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Identifying jet substructure: try out k_t

k_t algorithm



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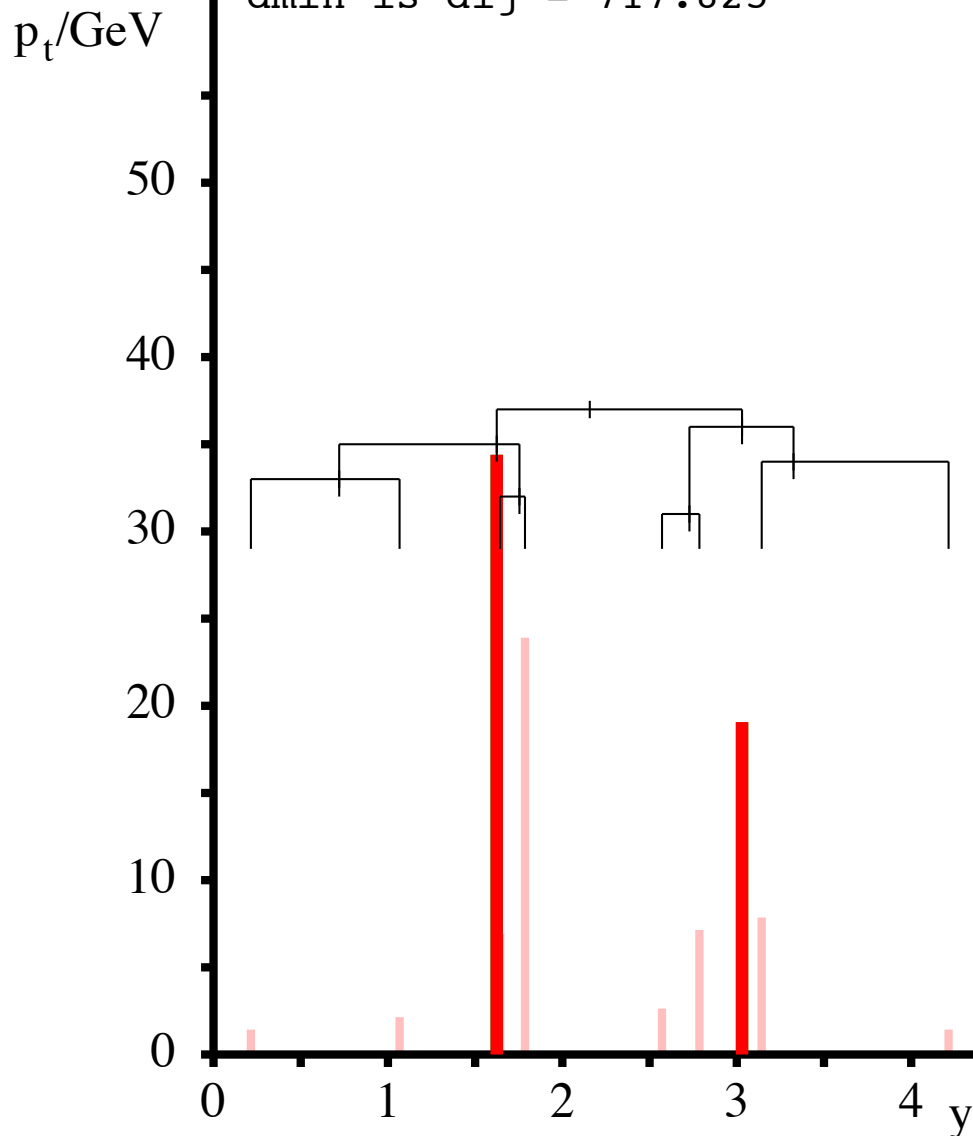
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Identifying jet substructure: try out k_t

k_t algorithm

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



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

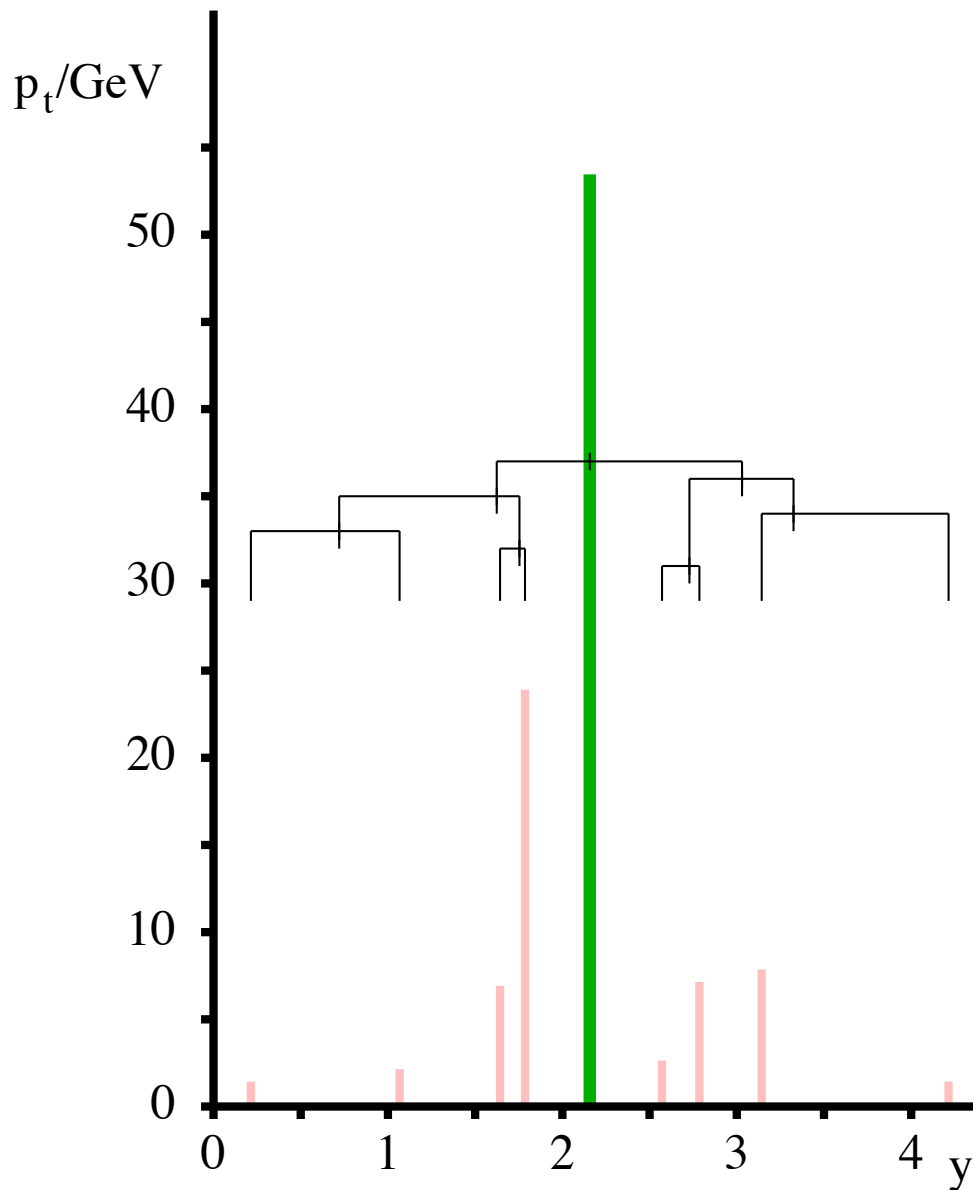
This is crucial for identifying the kinematic variables of the partons in the jet (e.g. z).

k_t clusters soft “junk” early on in the clustering

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

Identifying jet substructure: try out k_t

k_t algorithm



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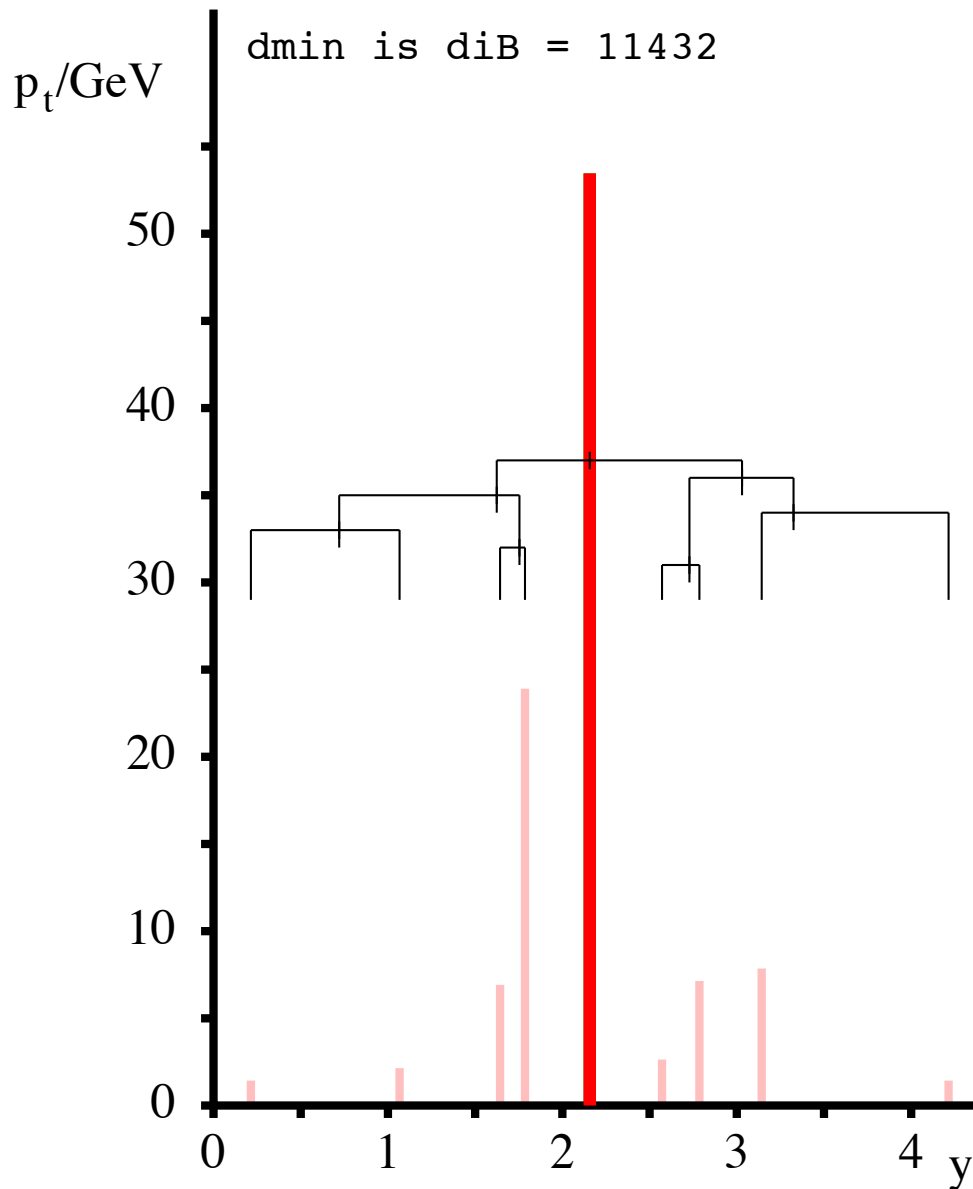
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Identifying jet substructure: try out k_t

k_t algorithm

d_{\min} is $d_{iB} = 11432$



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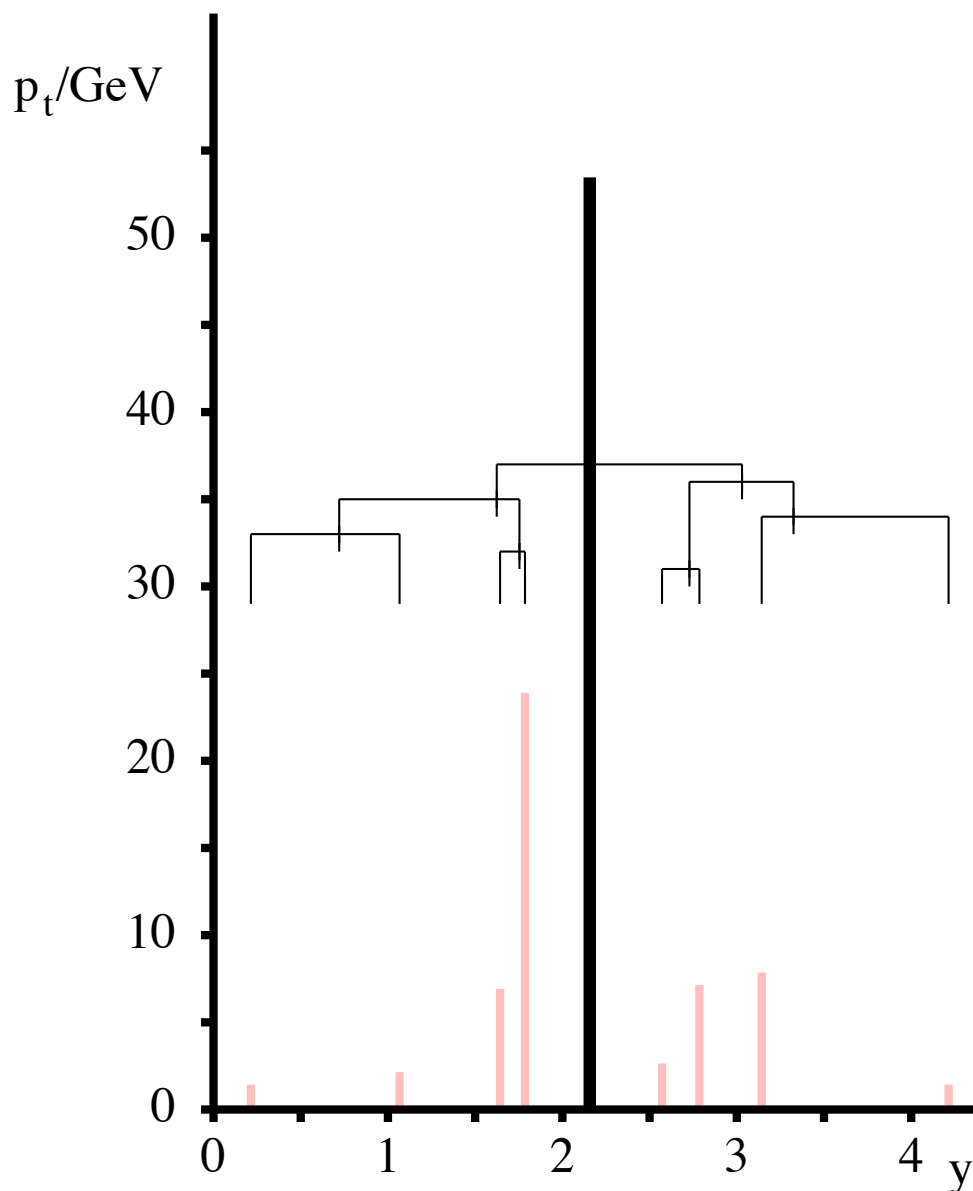
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k_t algorithm



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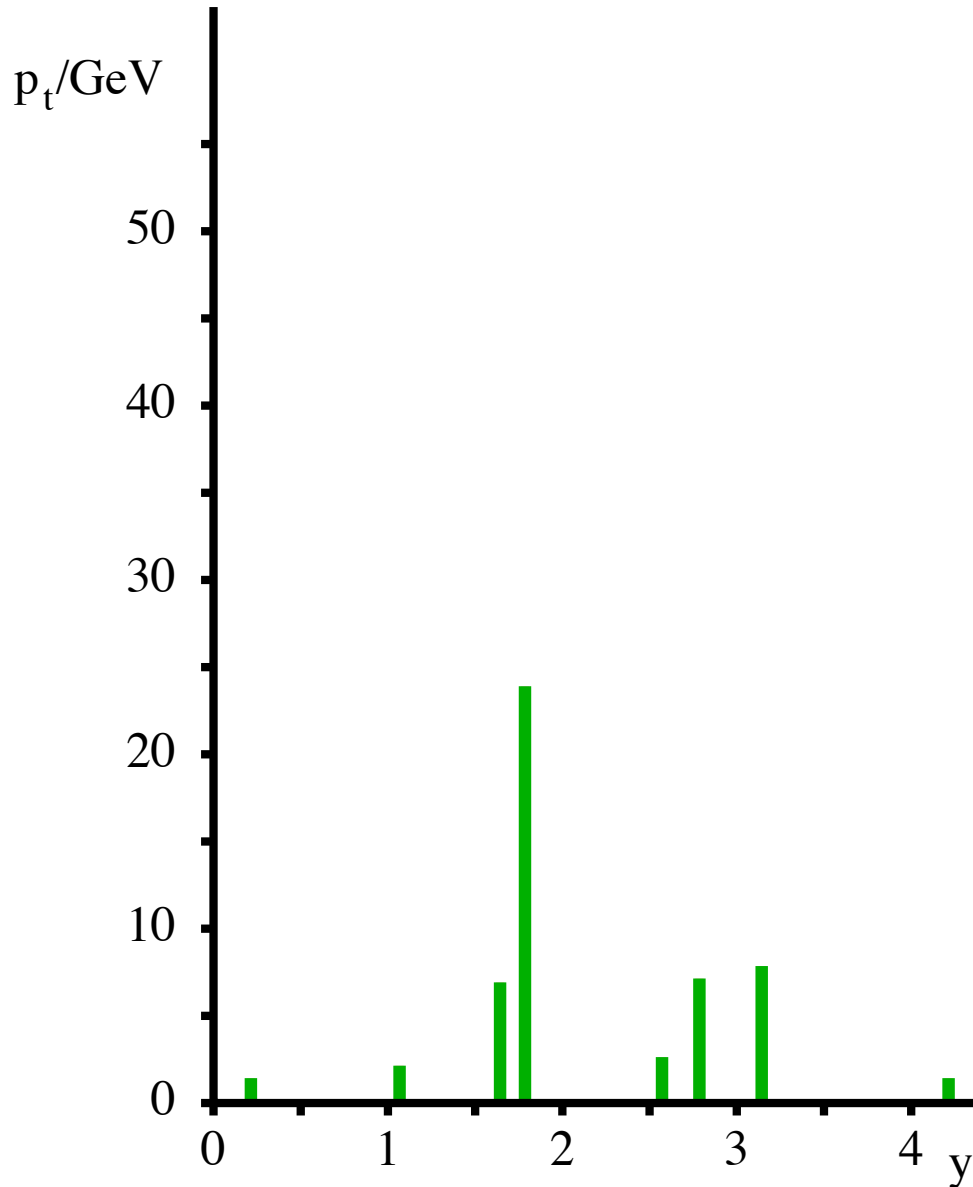
This meant it was the first algorithm to be used for jet substructure.

Seymour '93

Butterworth, Cox & Forshaw '02

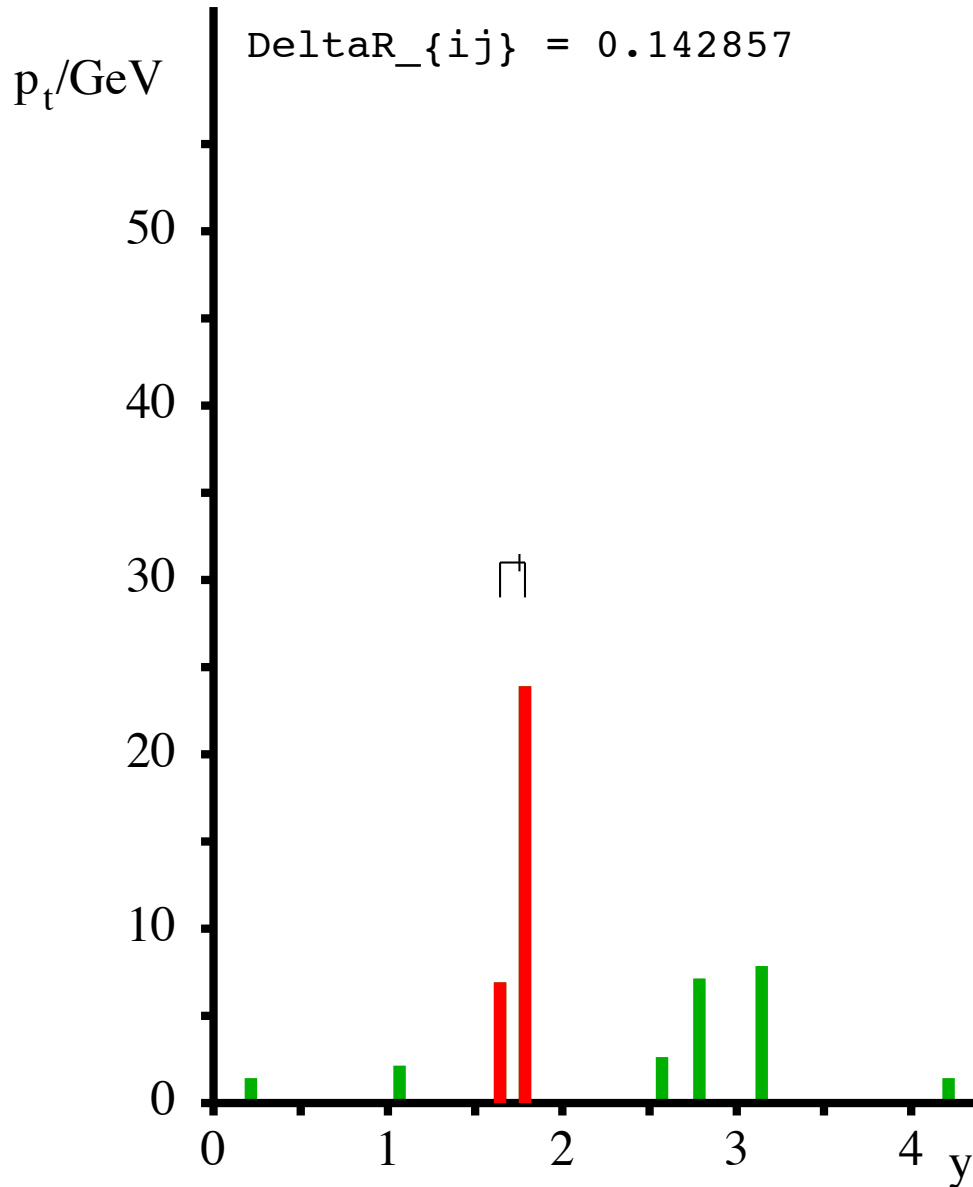
Cambridge/Aachen

Cambridge/Aachen algorithm



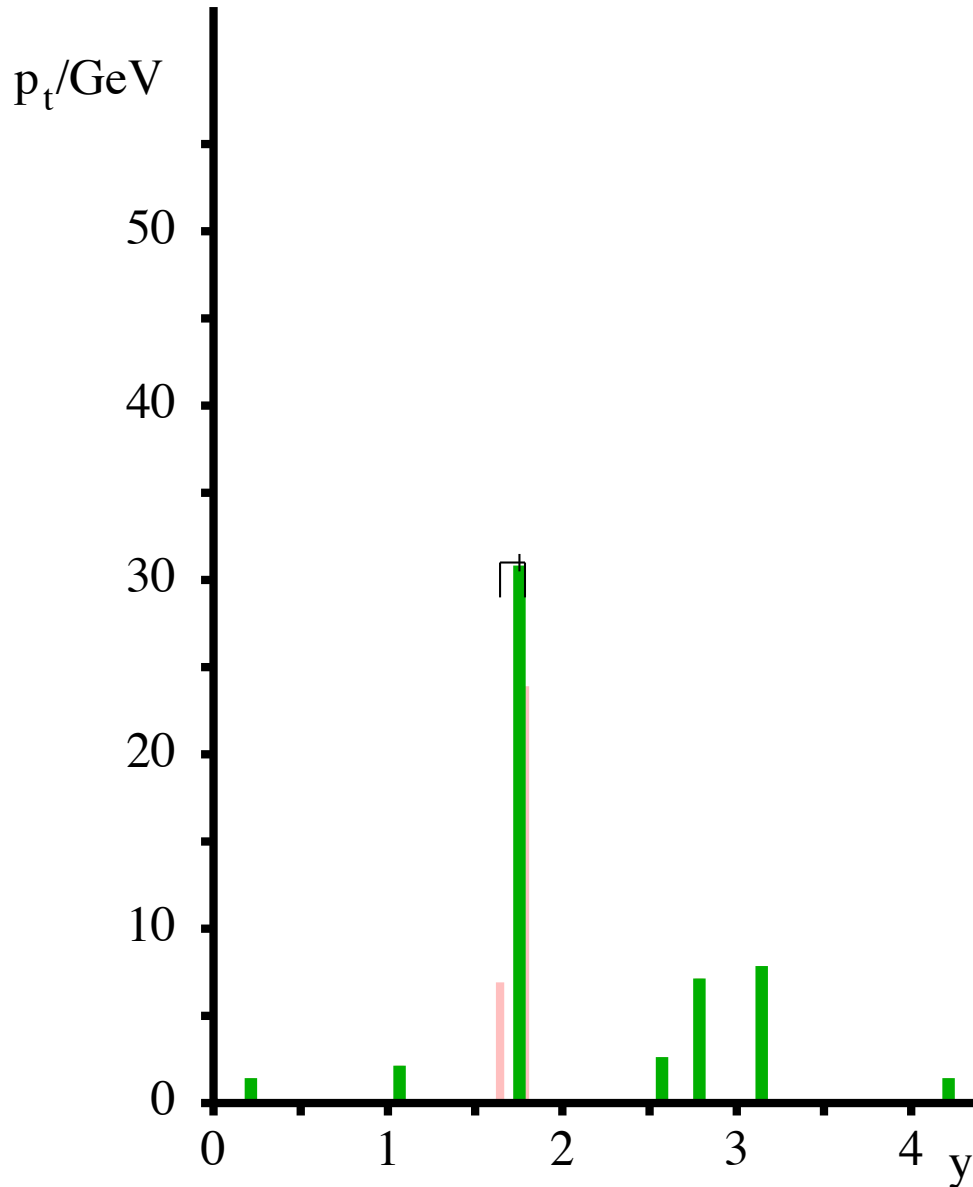
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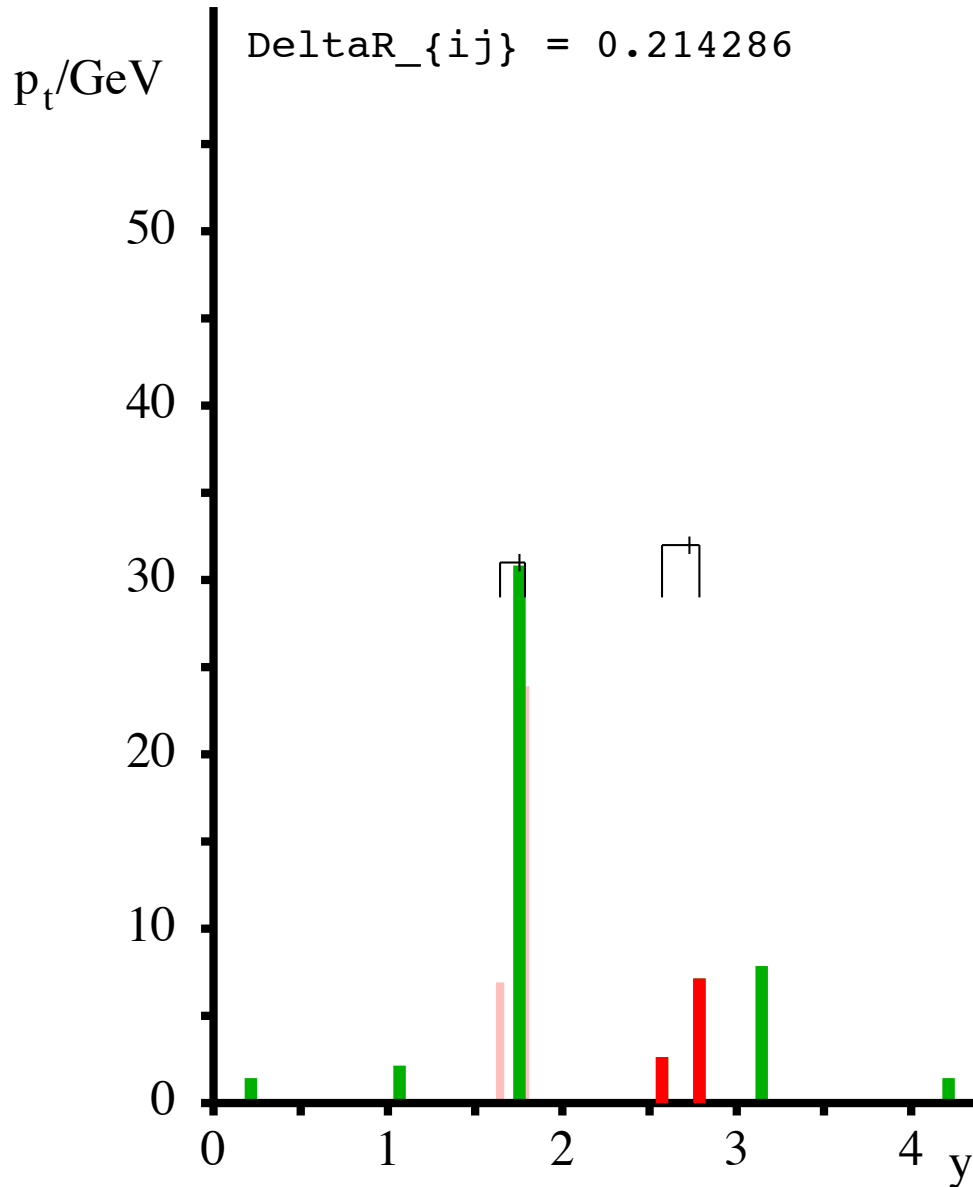
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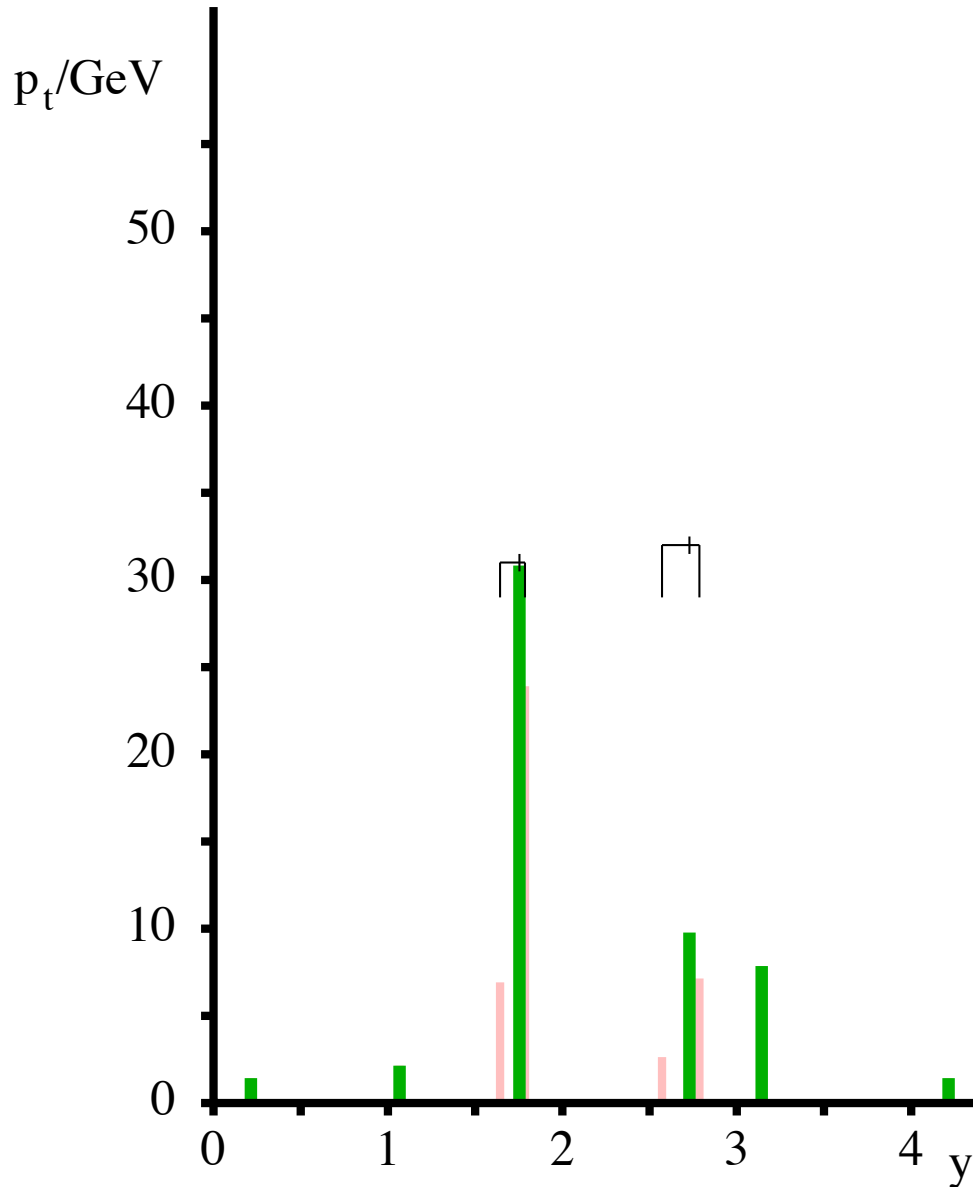
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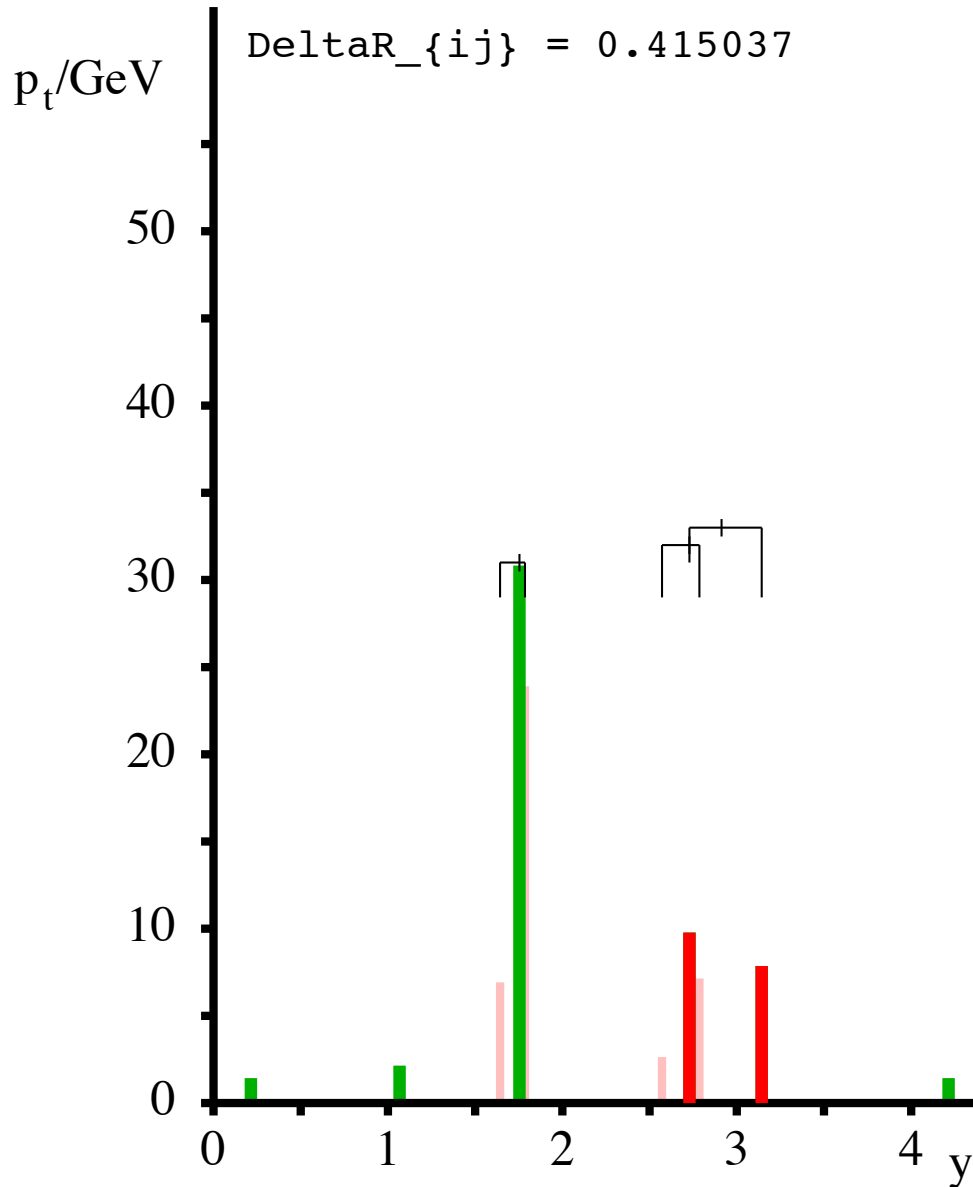
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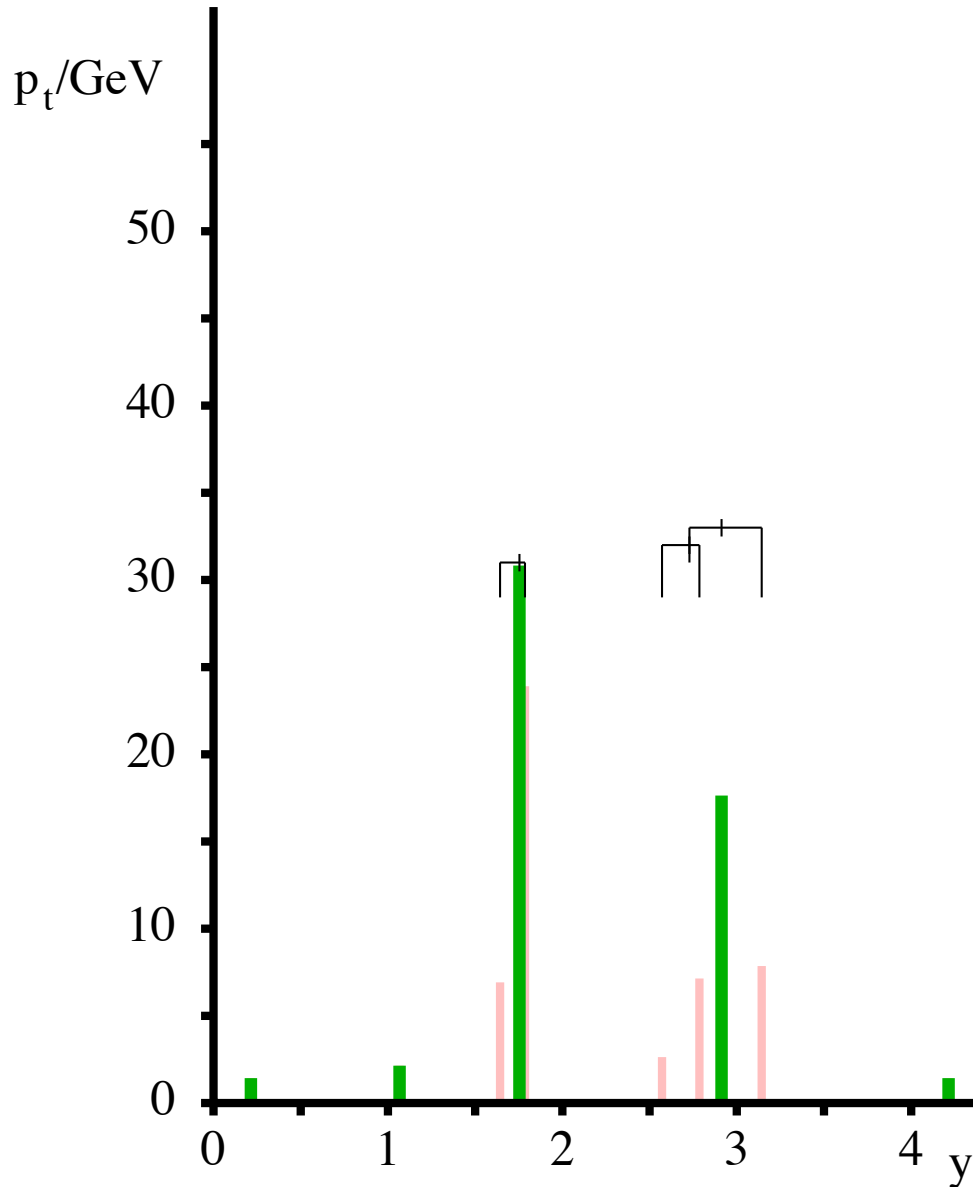
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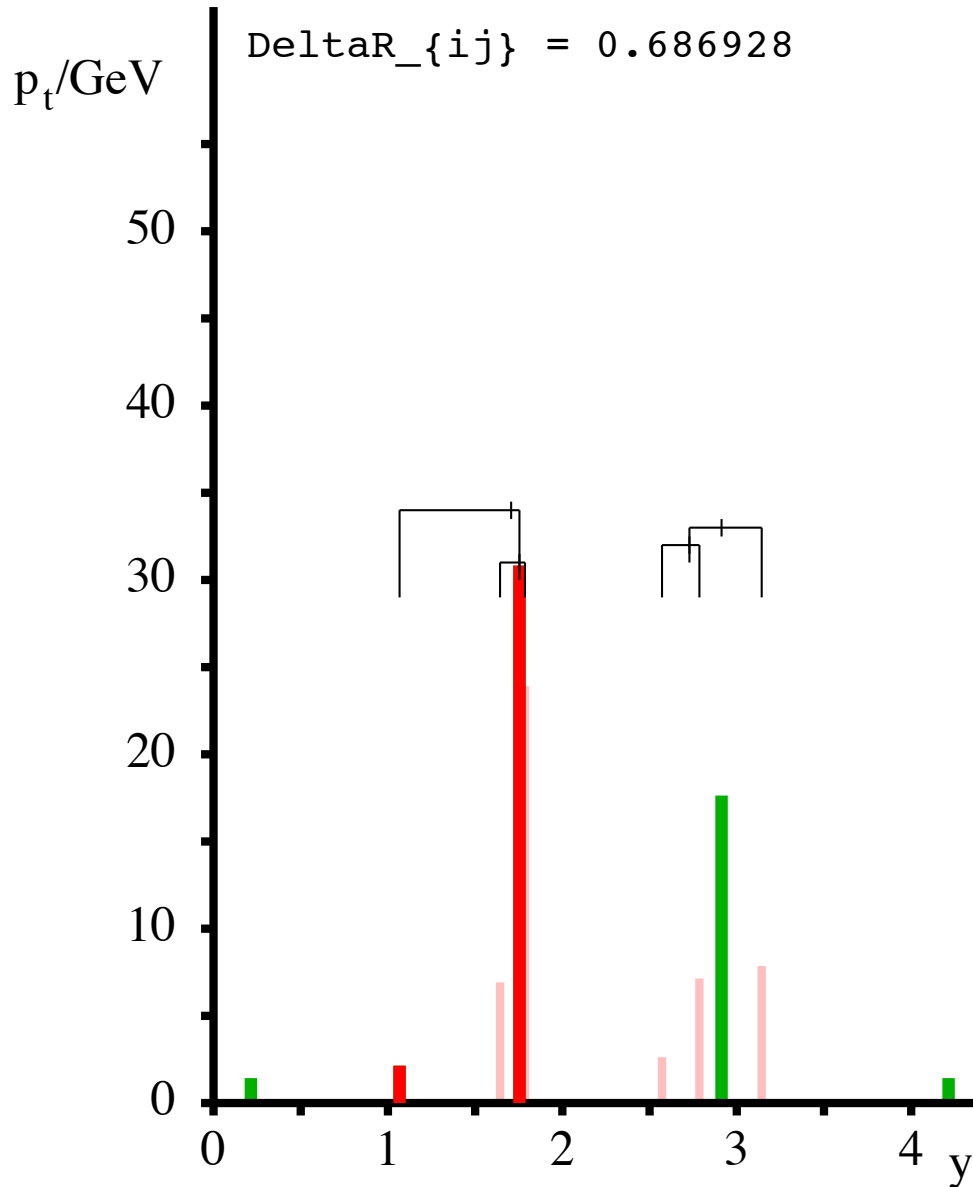
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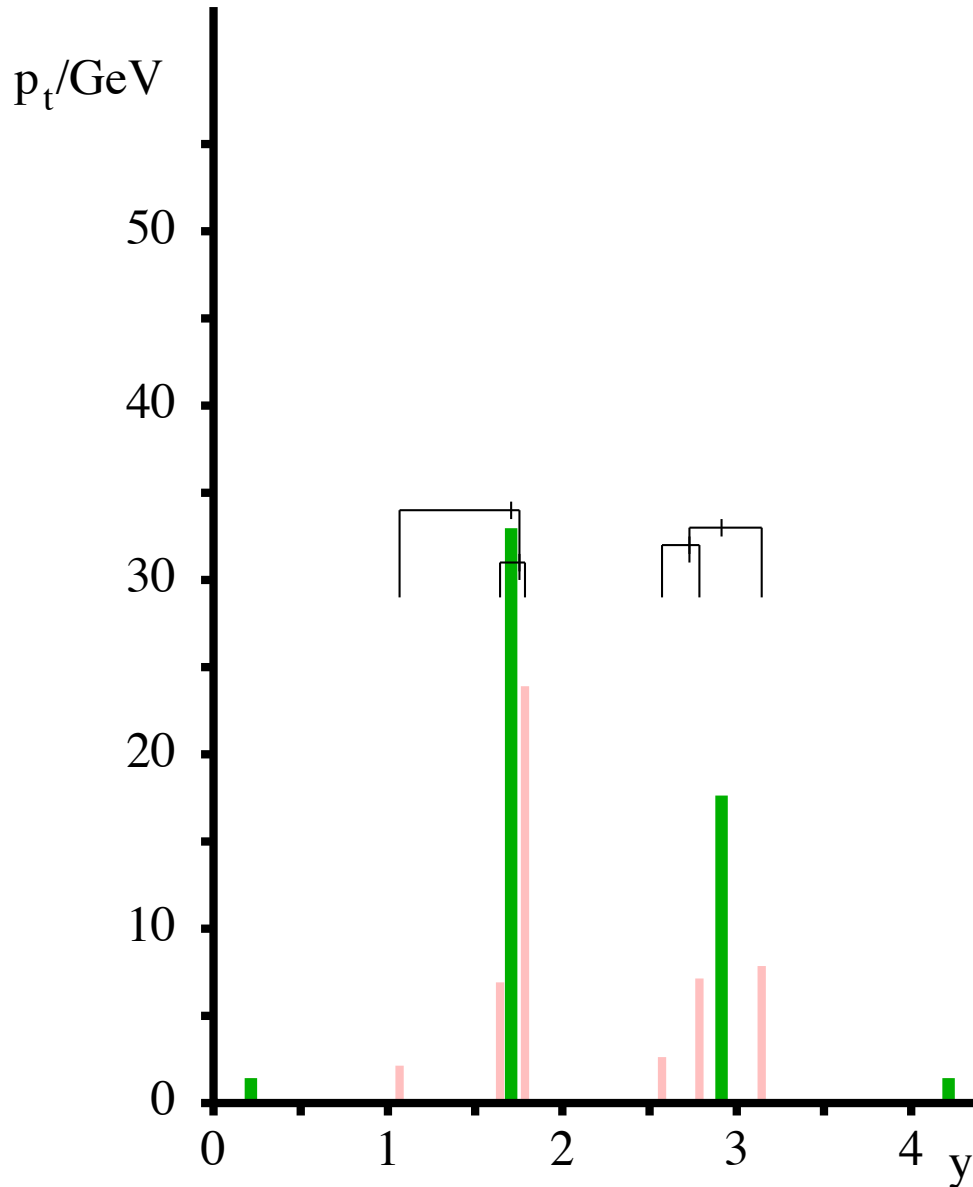
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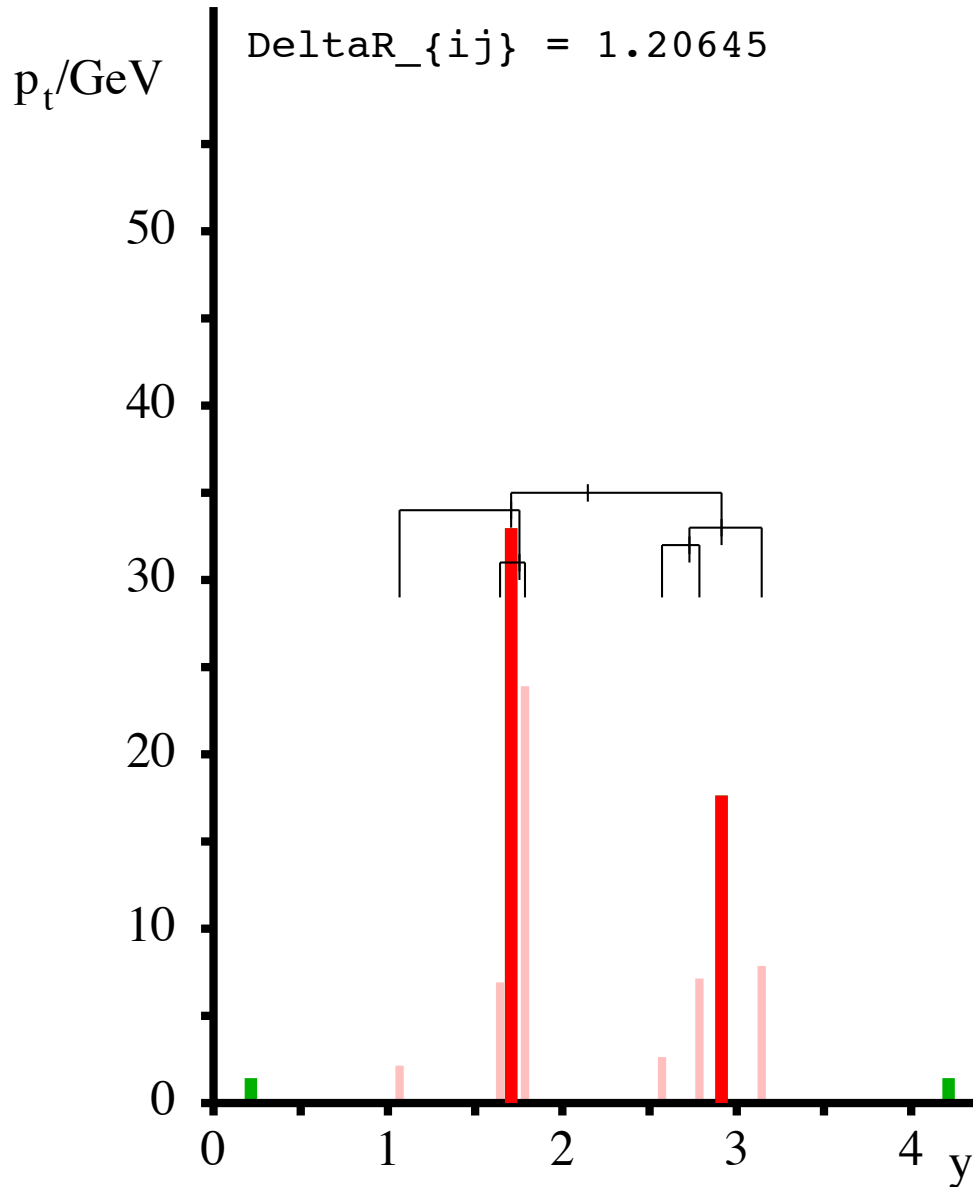
Cambridge/Aachen algorithm



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

Cambridge/Aachen algorithm

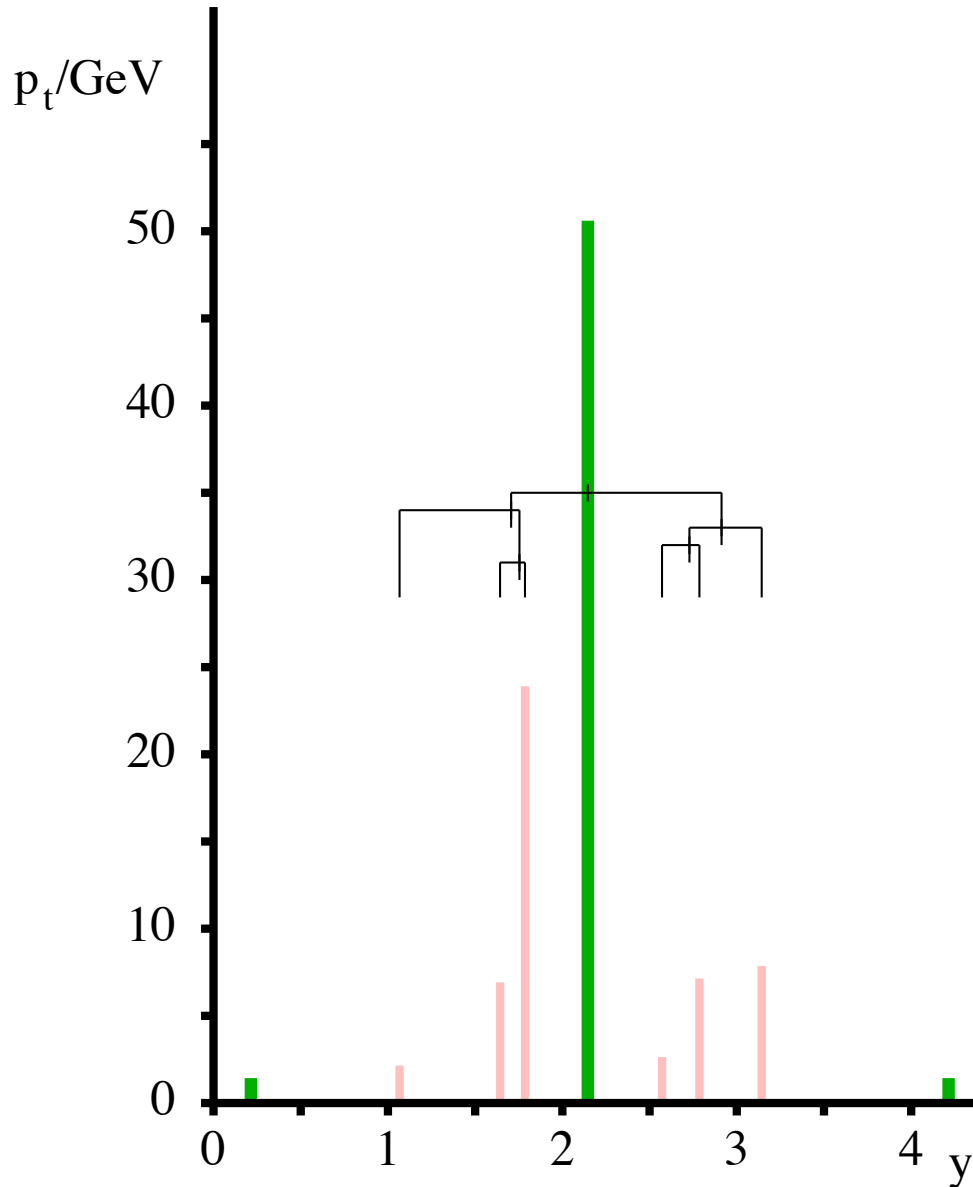


How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

Identifying jet substructure: Cam/Aachen

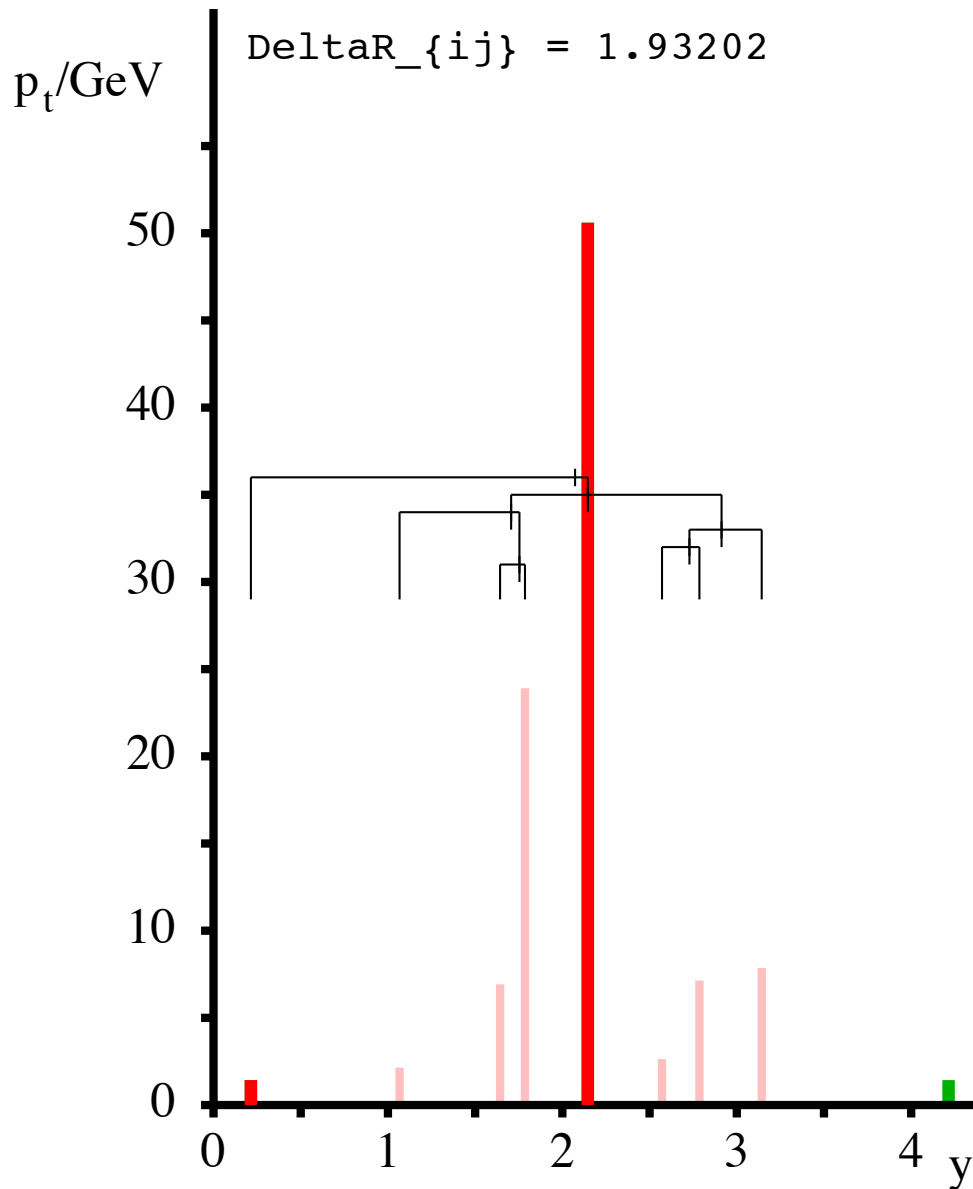
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How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

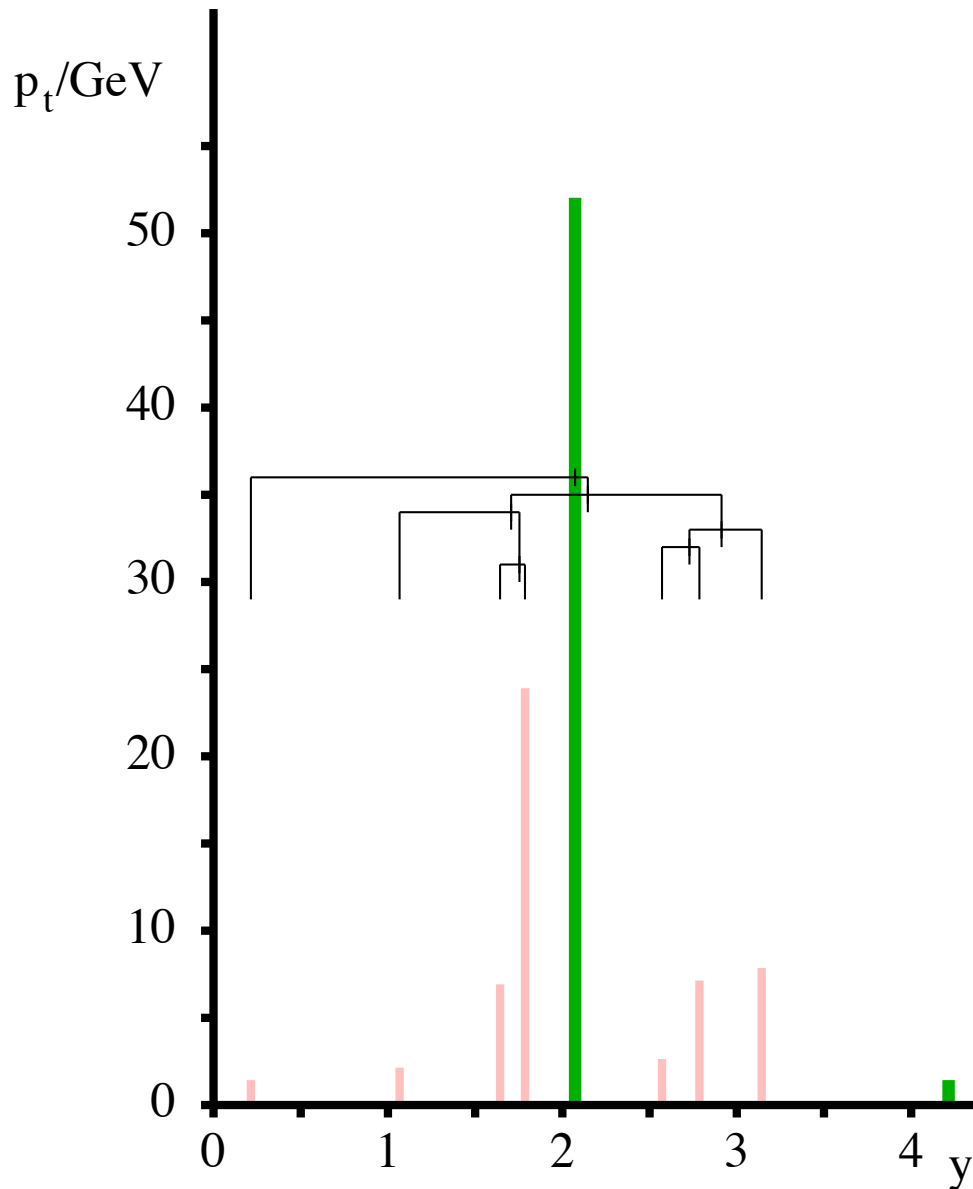
Cambridge/Aachen algorithm



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

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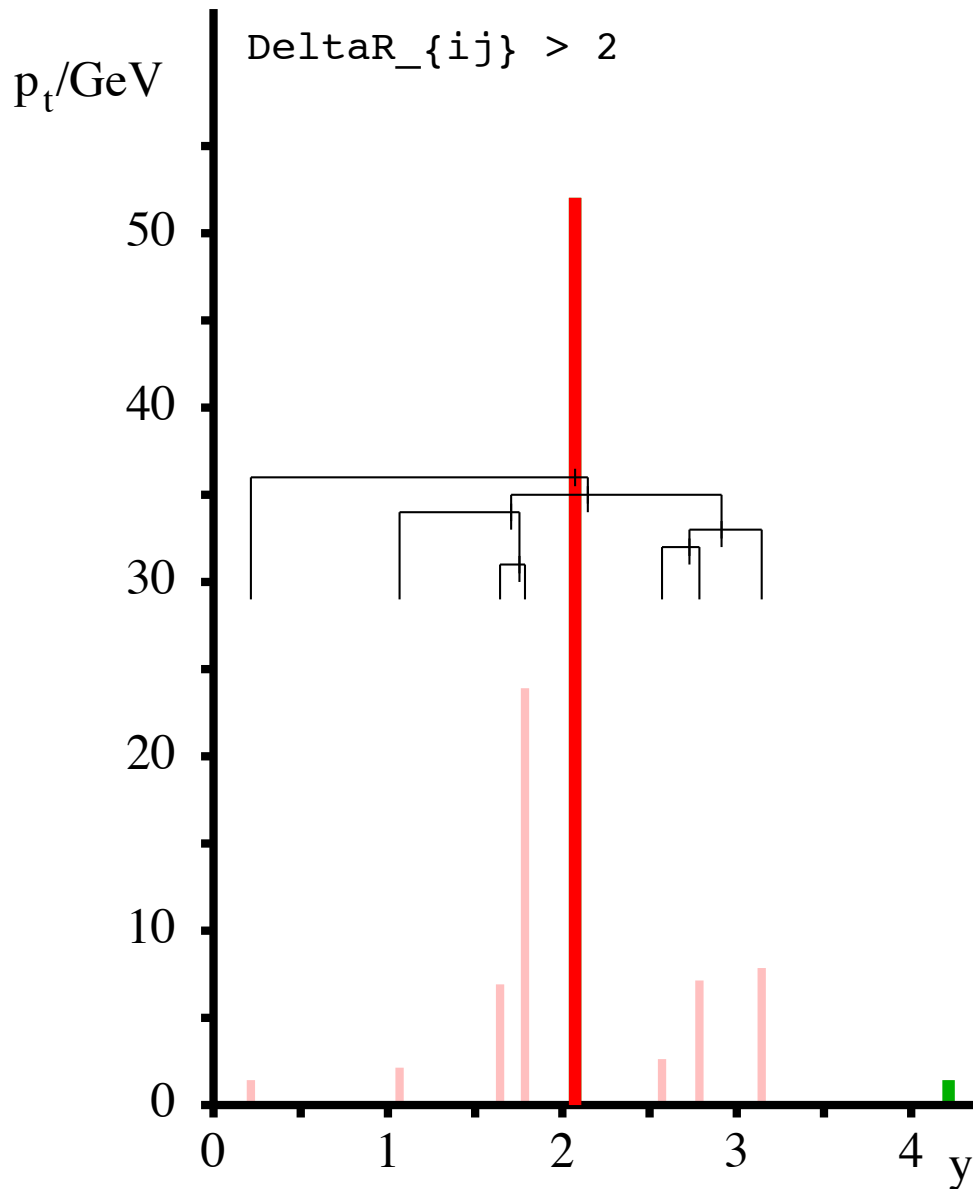
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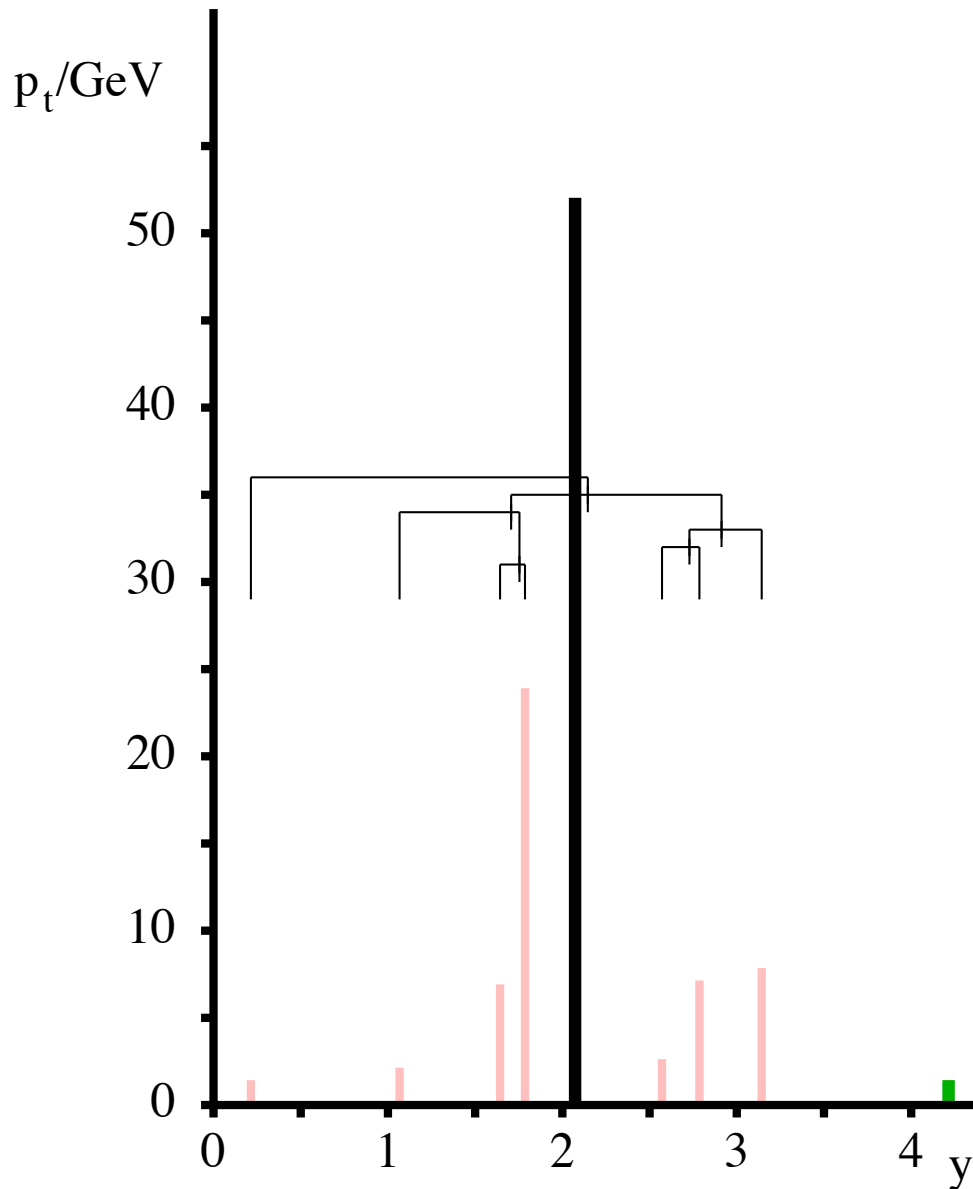
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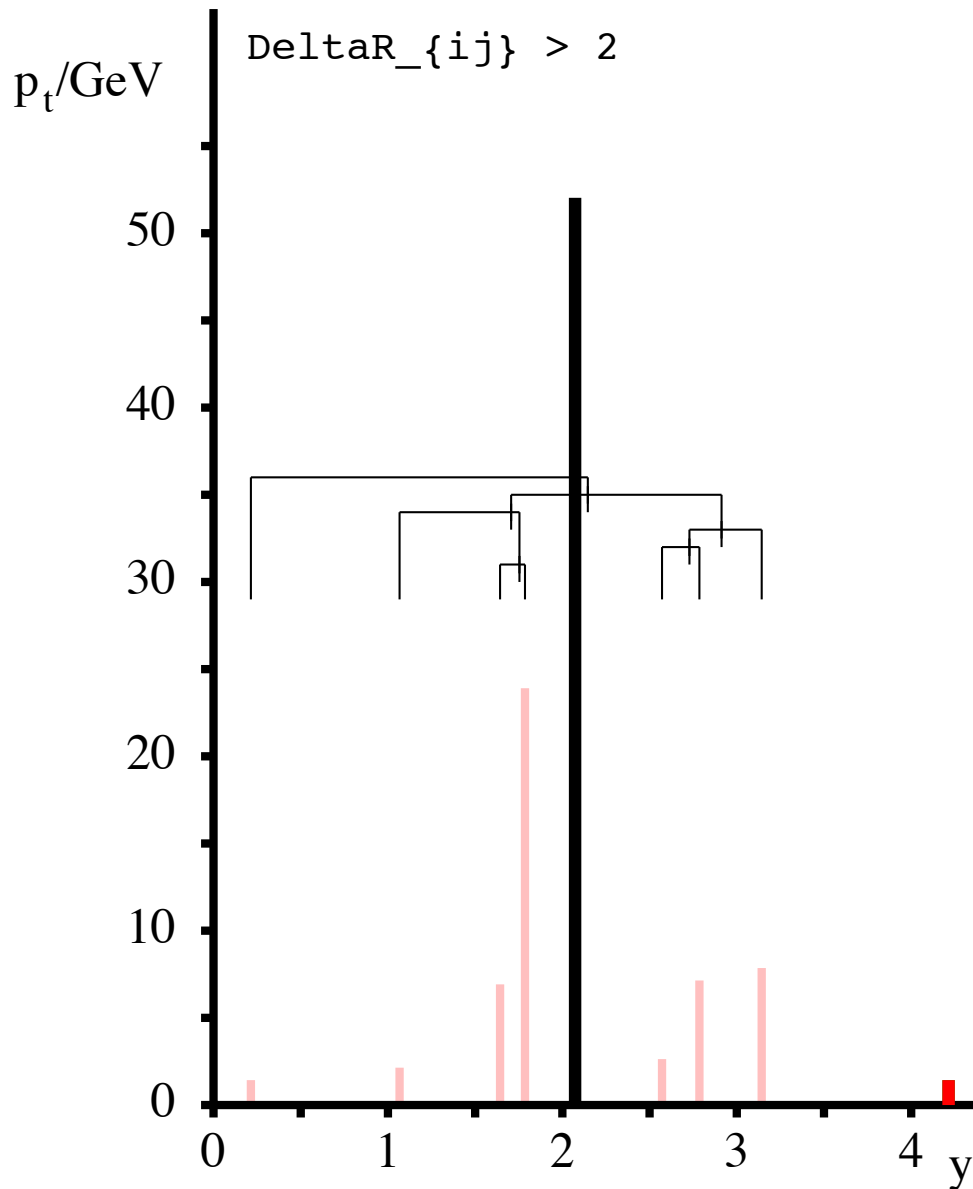
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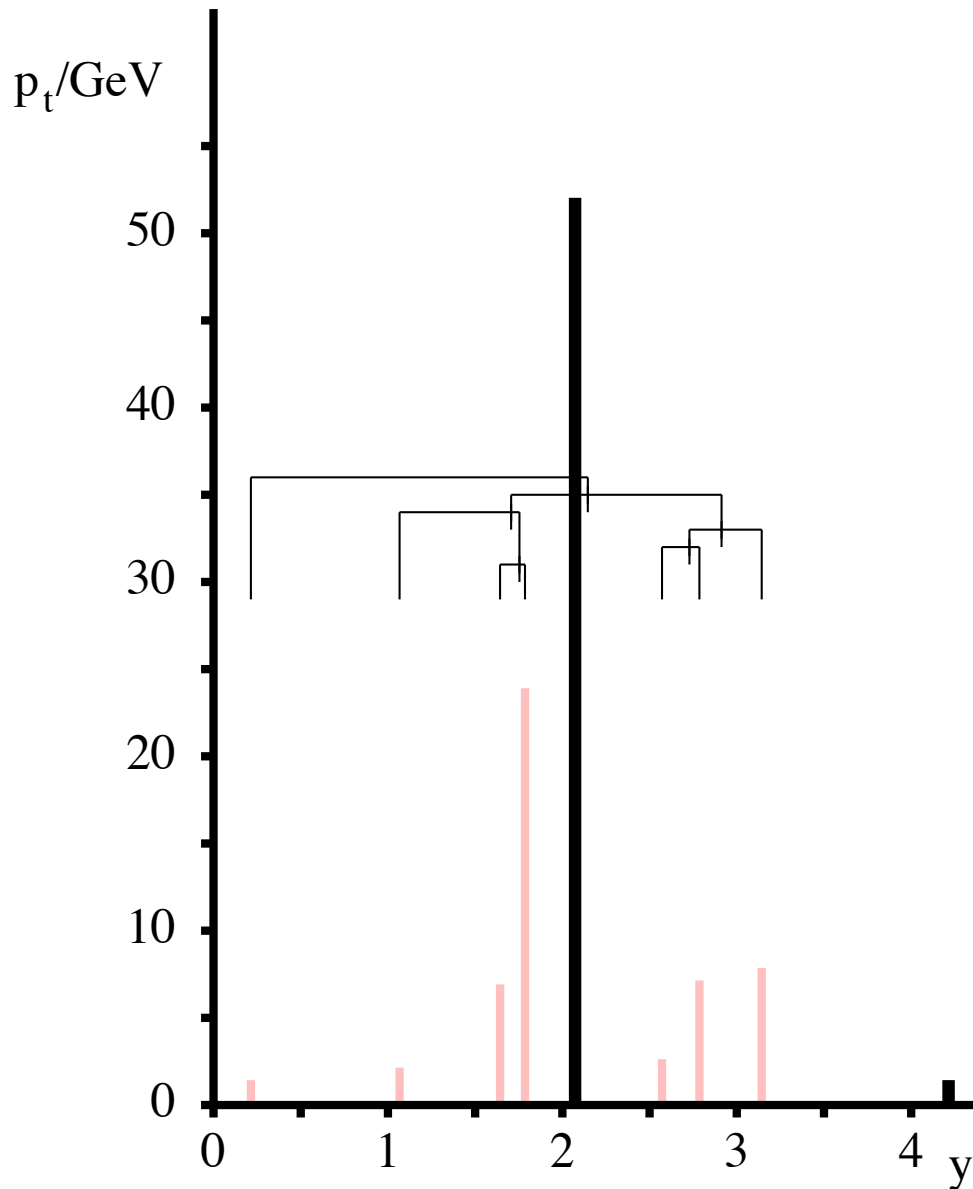
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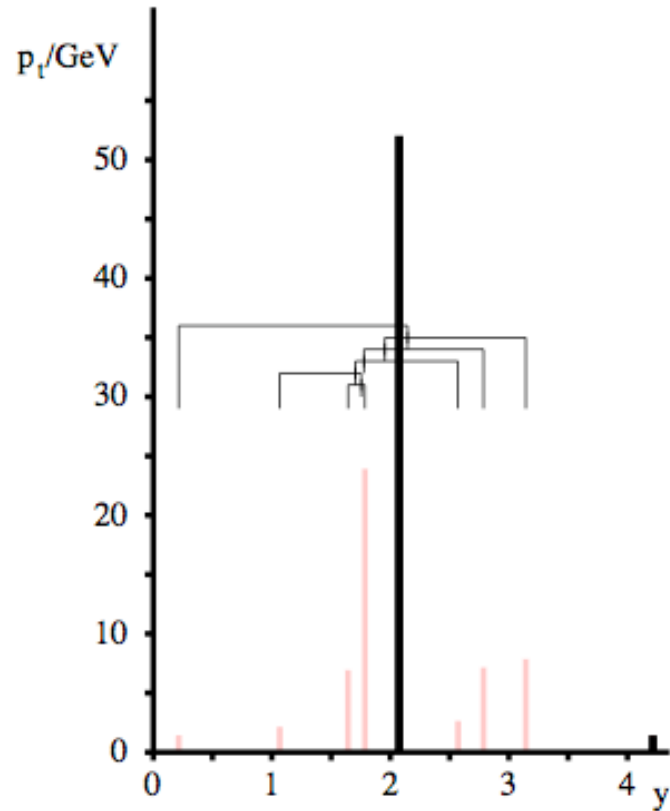
C/A identifies two hard blobs with limited soft contamination, joins them, and then adds in remaining soft junk

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**

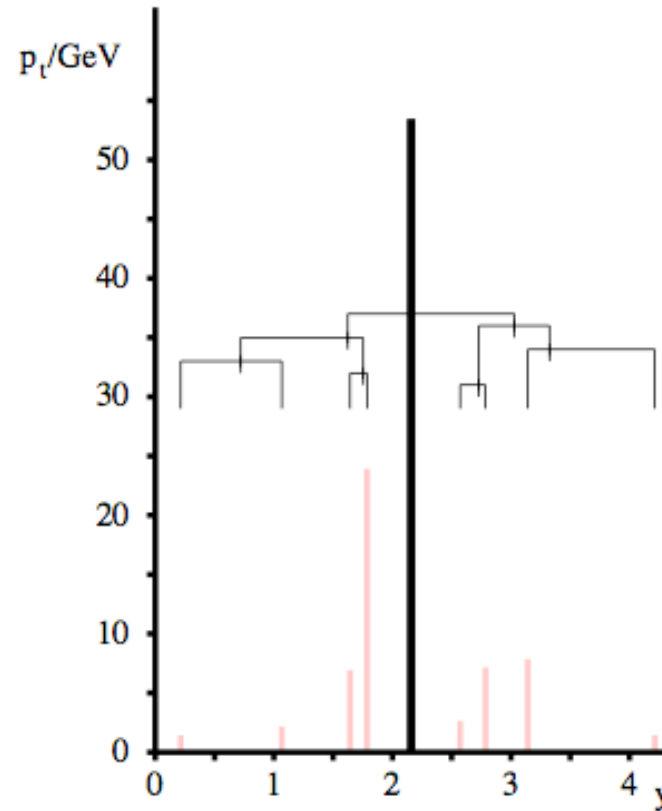
Butterworth, Davison, Rubin & GPS '08
Kaplan, Schwartz, Reherman & Tweedie '08
Butterworth, Ellis, Rubin & GPS '09
Ellis, Vermilion & Walsh '09

Hierarchical substructure

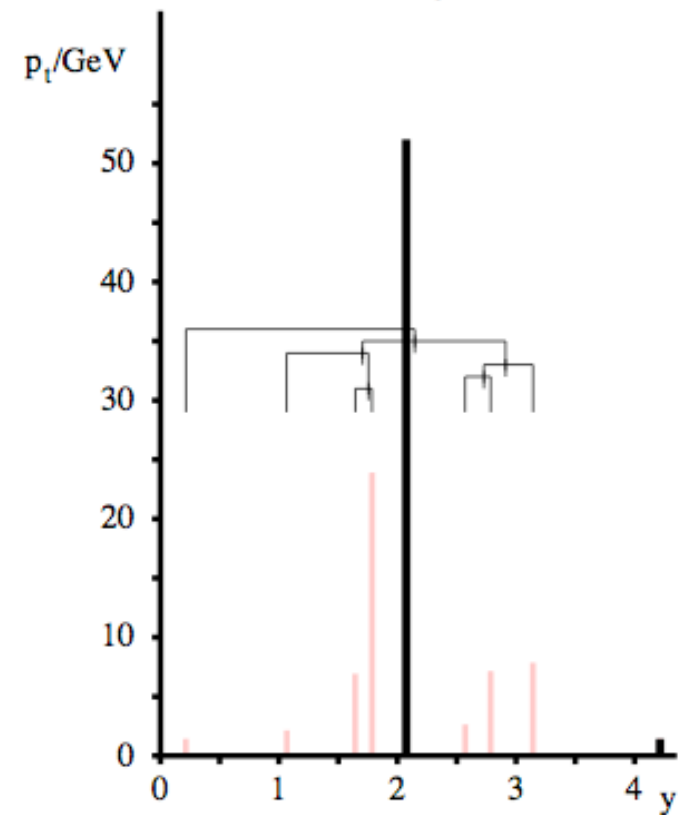
anti- k_t algorithm



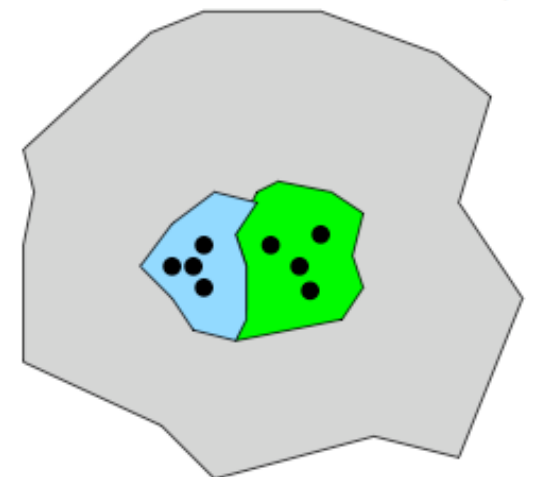
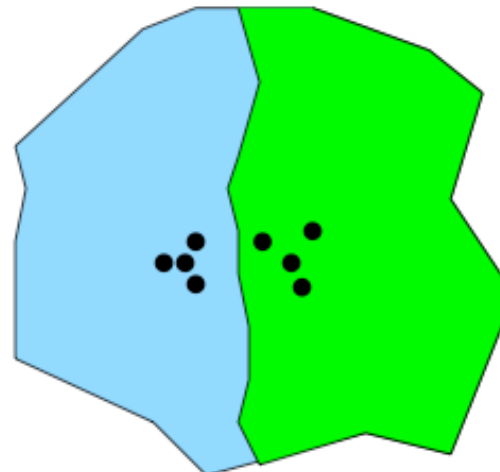
k_t algorithm



Cambridge/Aachen



Undo the last
clustering step(s)



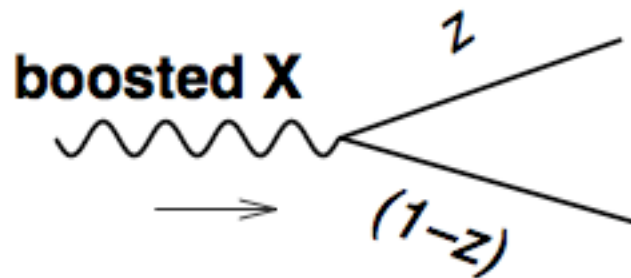
The IRC safe algorithms

	Speed	Regularity	UE contamination	Backreaction	Hierarchical substructure
k_t	☺☺☺	☂	☂☂	☁☁	☺☺
Cambridge /Aachen	☺☺☺	☂	☂	☁☁	☺☺☺
anti- k_t	☺☺☺	☺☺	☁/☺	☺☺	✗
SISCone	☺	☁	☺☺	☁	✗

Array of tools with different characteristics.
Pick the right one for the job

QCD v. heavy decay

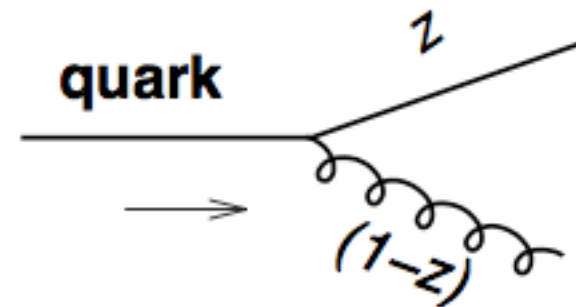
A possible approach for reducing the QCD background is to identify the two prongs of the heavy particle decay, and put a cut on their momentum fraction



Signal:

$$P(z) \sim 1$$

Will split mainly **symmetrically**



Background:

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

Will split mainly **asymmetrically**

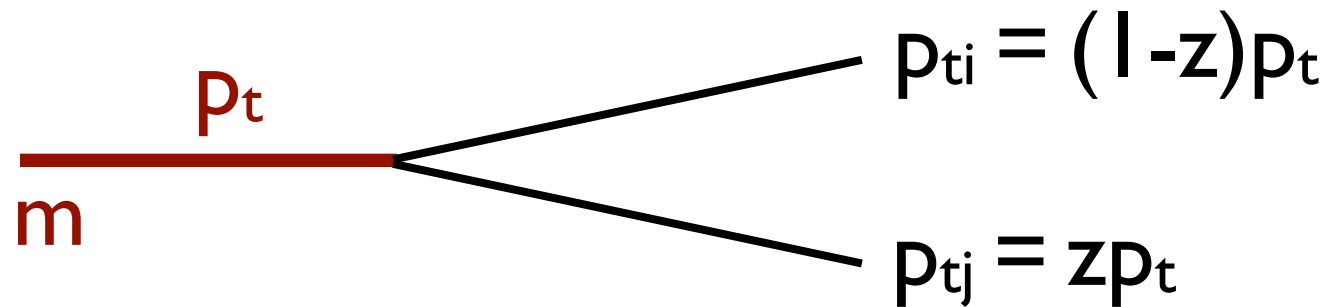
Potential tagger: asymmetric splitting

Possibly implemented via a cut on

$$y = \min(p_{ti}^2, p_{tj}^2) \frac{\Delta R_{ij}^2}{m^2} \simeq \frac{\min(p_{ti}, p_{tj})}{\max(p_{ti}, p_{tj})}$$

Splittings and distances

Quasi-collinear
splitting ($p_{tj} < p_{ti}$)



Invariant mass:
$$m^2 \simeq p_{ti}p_{tj}\Delta R_{ij}^2 = (1-z)zp_t^2\Delta R_{ij}^2$$

k_t distance:
$$d_{ij}^{(p_{tj} < p_{ti})} \stackrel{z^2 p_t^2 \Delta R_{ij}^2}{=} \simeq \frac{z}{1-z} m^2$$

For a given mass, the **background** will have **smaller distance** d_{ij} than the signal, i.e. it will tend to **cluster earlier** in the k_t algorithm

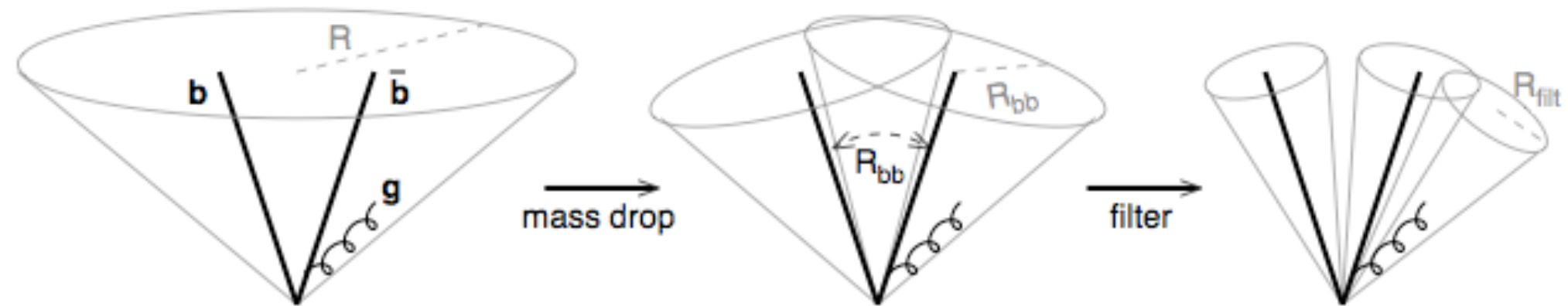
Potential tagger: last clustering in k_t algorithm

This is where the hierarchy of the k_t algorithm becomes relevant. QCD radiation is clustered first, and only at the end the symmetric, large-angle splittings due to decays are reclustered

$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$$

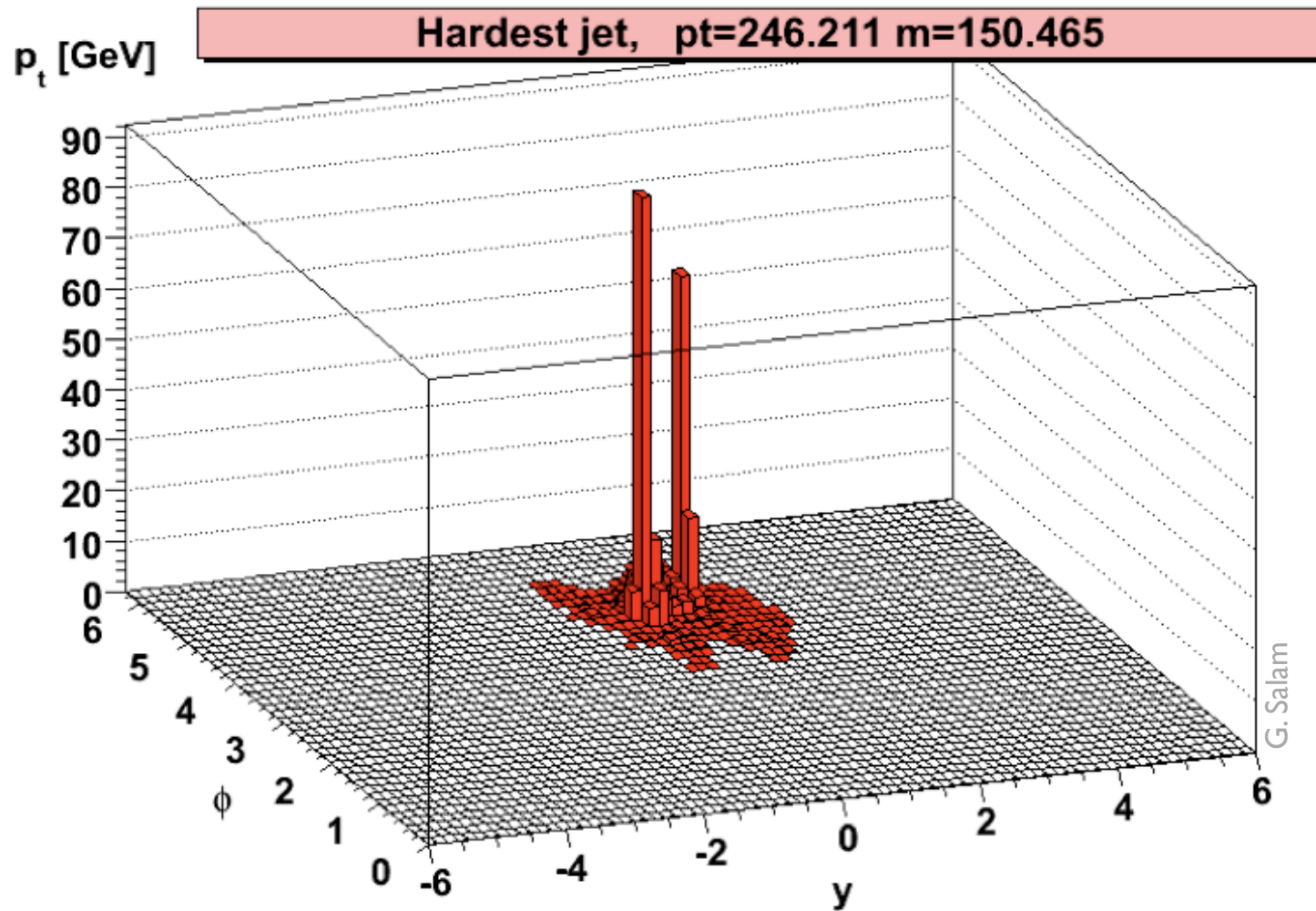
The BDRS tagger/groomer

Butterworth, Davison, Rubin, Salam, 2008



- ▶ A two-prong tagger/groomer for boosted Higgs, which
 - ▶ Uses the **Cambridge/Aachen** algorithm (because it's 'physical')
 - ▶ Employs a **Mass-Drop** condition, as well as an **asymmetry cut** to find the **relevant splitting** (i.e. '**tag**' the heavy particle)
 - ▶ Includes a post-processing step, using '**filtering**' (introduced in the same paper) to clean as much as possible the resulting jets of UE contamination ('**grooming**')

$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$$

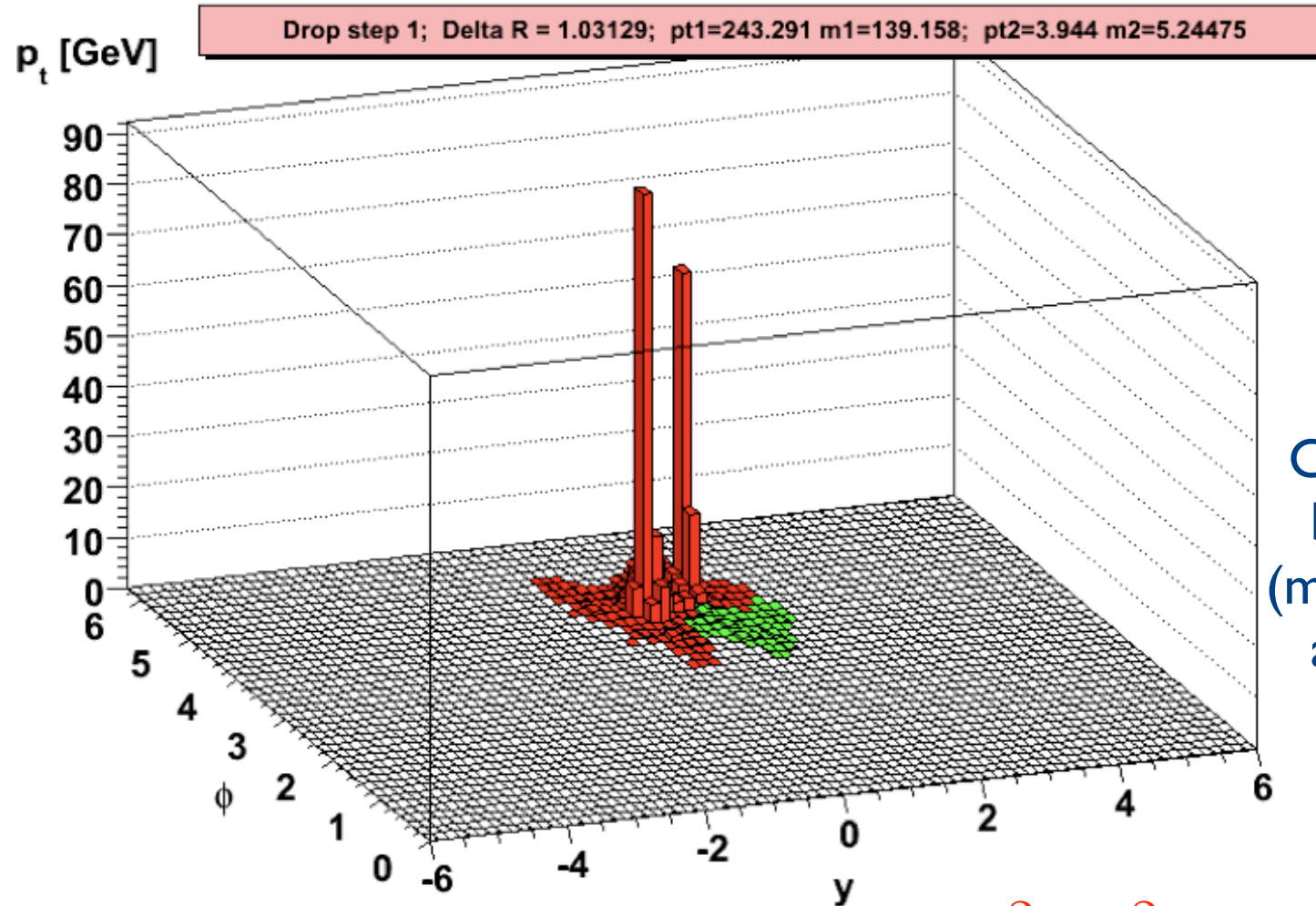


Start with the
hardest jet

Use C/A with
large $R=1.2$

$m_j = 150$ GeV

$pp \rightarrow ZH \rightarrow \nu\nu b\bar{b}$

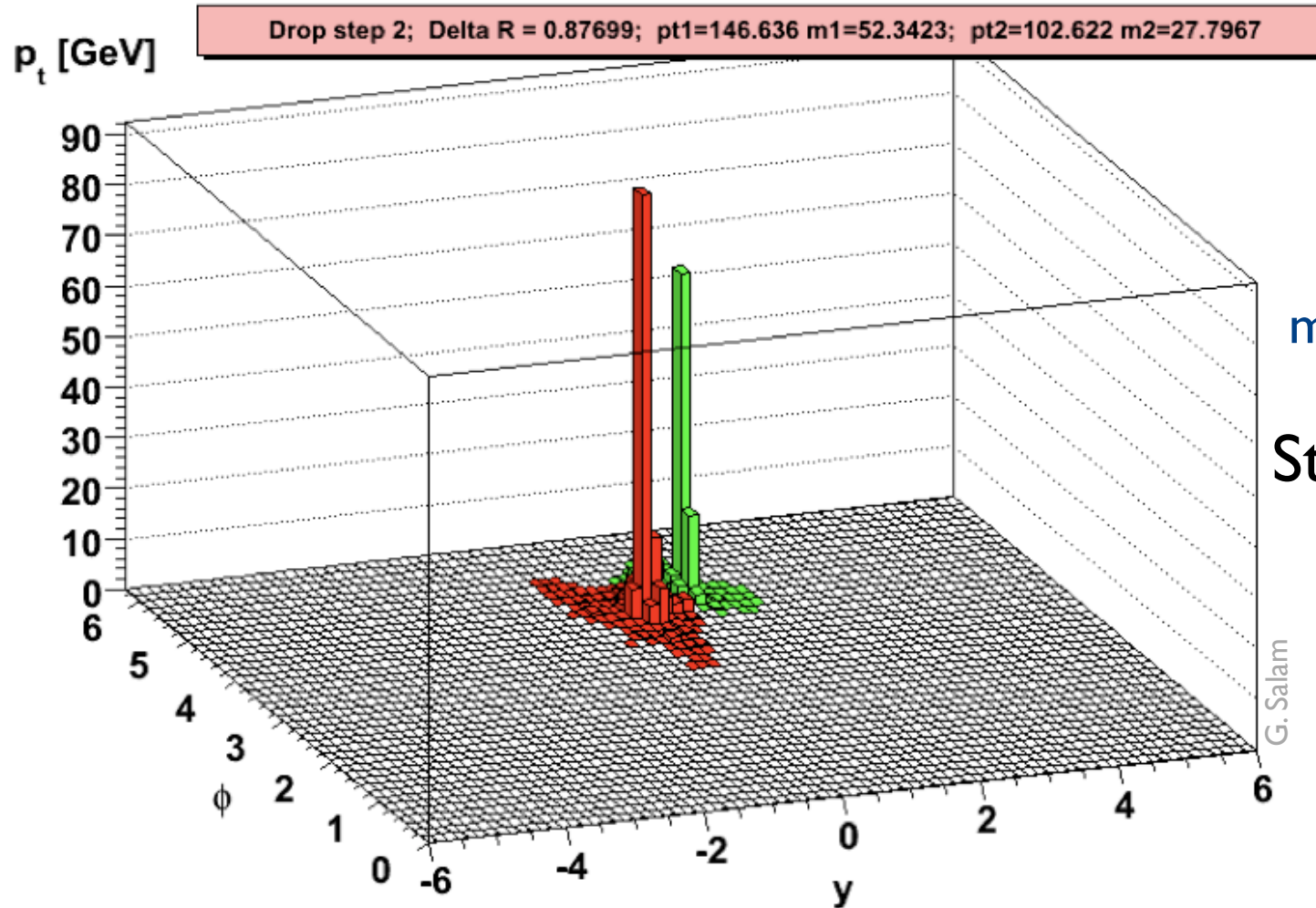


Undo last step of clustering

Check how the mass splits between the two subjects ($m_1 = 139$ GeV, $m_2 = 5$ GeV) and how asymmetric the splitting is

If $\frac{\max(m_1, m_2)}{m_j} > \mu$ or $\frac{\min(p_{t1}^2, p_{t2}^2)}{m_j^2} \Delta R_{12}^2 < y_{cut}$ repeat

$pp \rightarrow ZH \rightarrow \nu\nu bb$

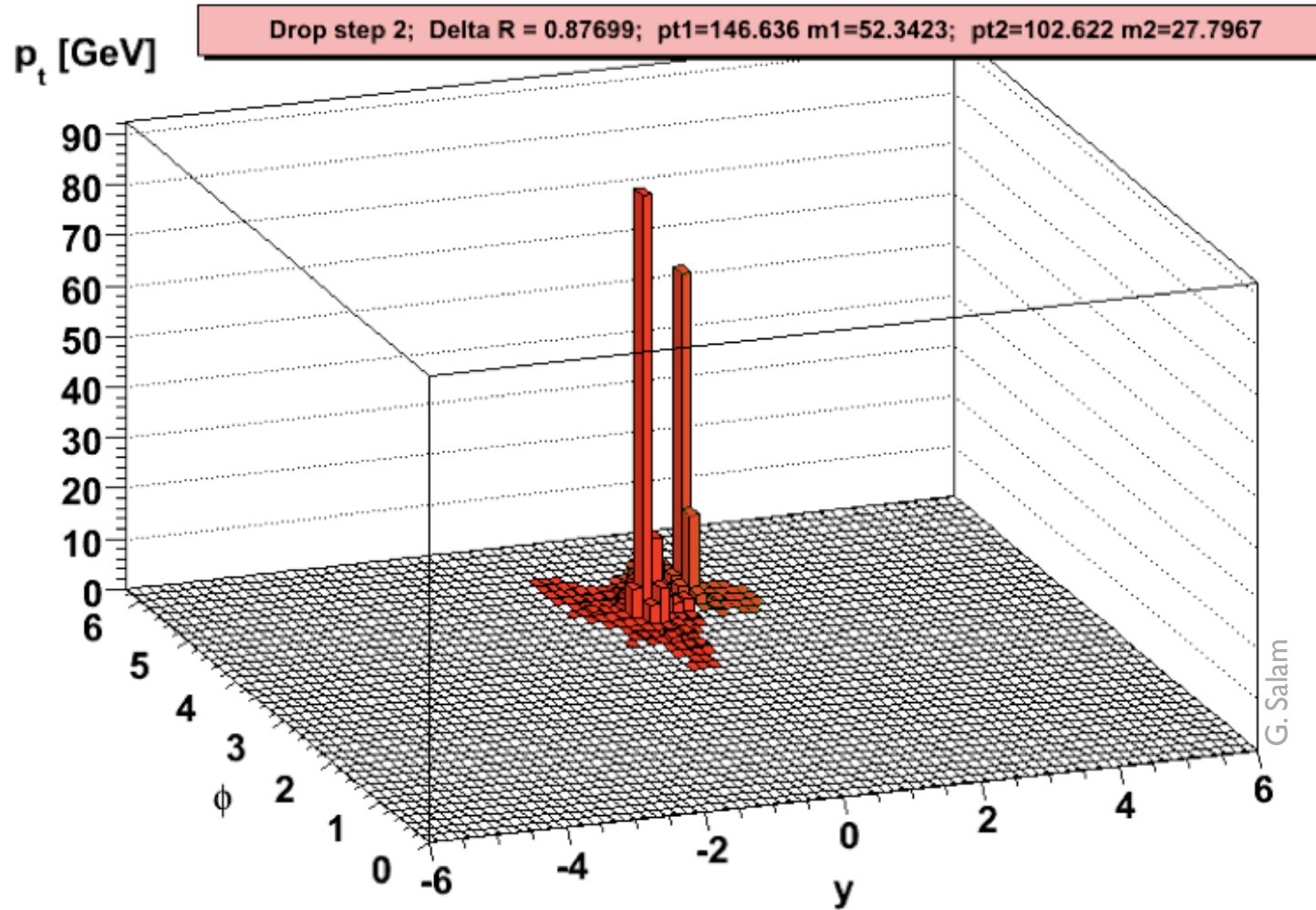


$m_1 = 52 \text{ GeV}, m_2 = 28 \text{ GeV}$

Stop when a **large mass drop** is observed
(and **recombine** these two jets)

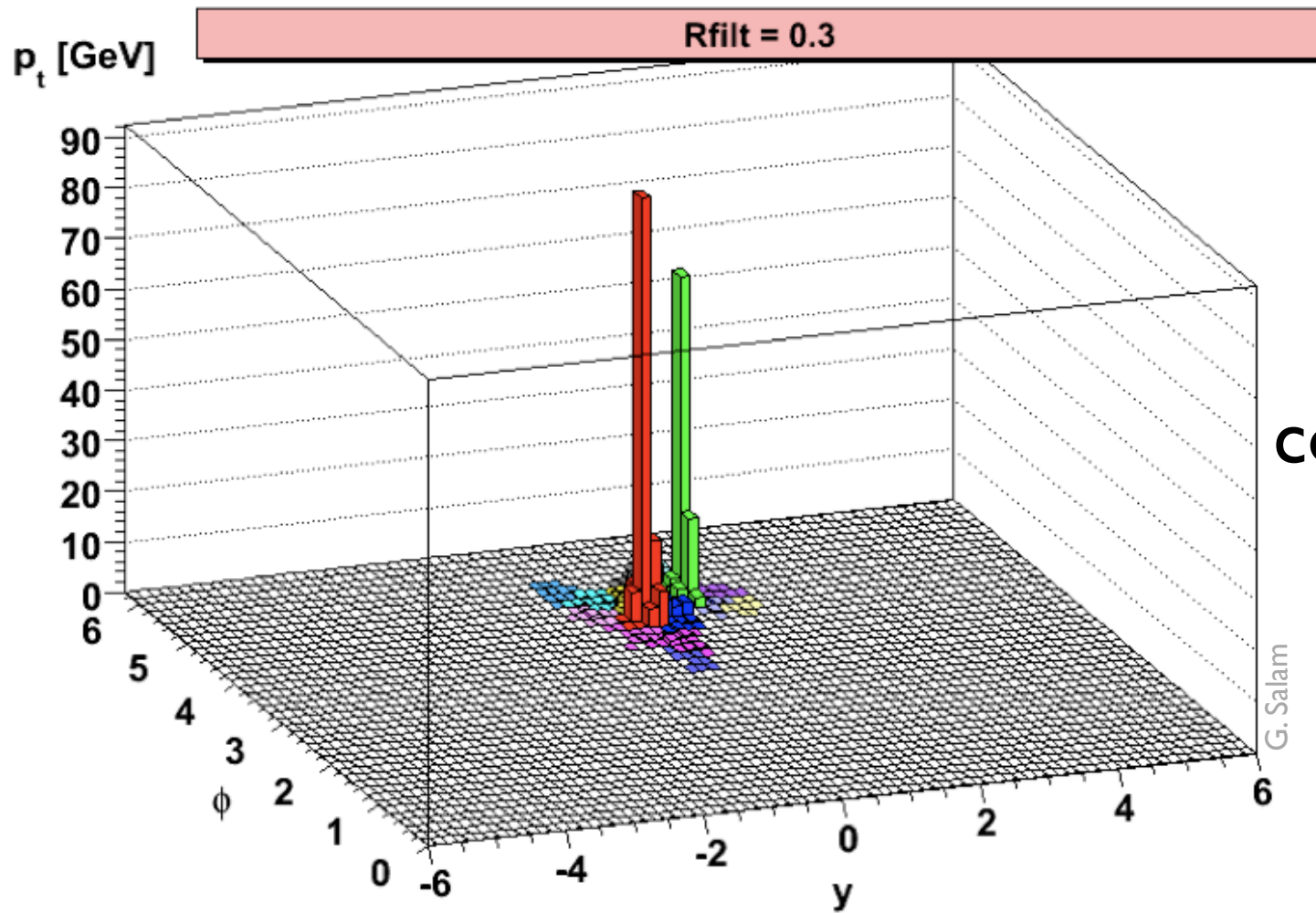
[NB. Parameters used $\mu = 0.67$ and $y_{\text{cut}} = 0.09$]

$pp \rightarrow ZH \rightarrow \nu\nu b\bar{b}$



Start with the recombined jet

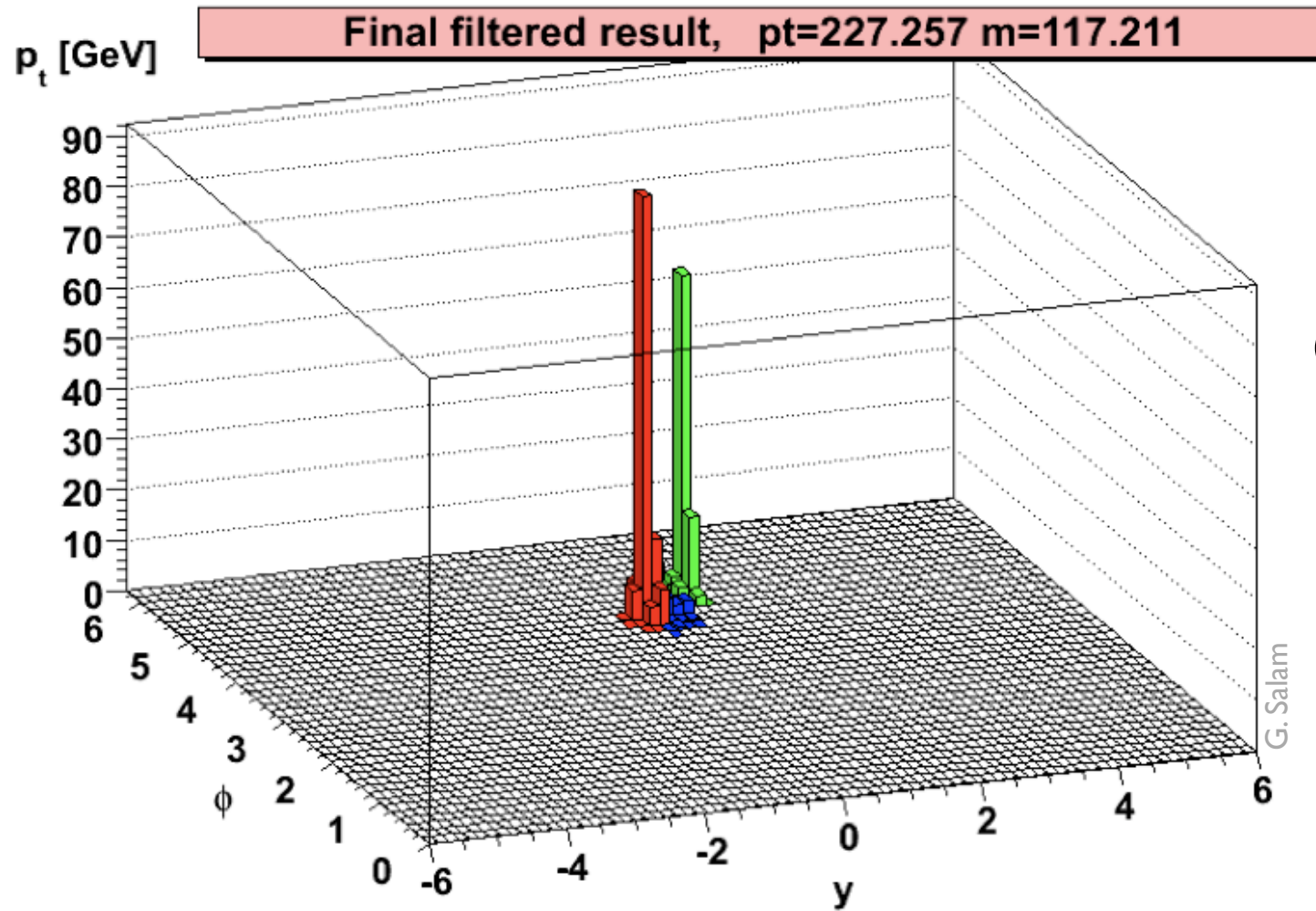
$pp \rightarrow ZH \rightarrow \nu\nu b\bar{b}$



Recluster the
constituents with R_{filt}

G. Salam

$pp \rightarrow ZH \rightarrow \nu\nu b\bar{b}$



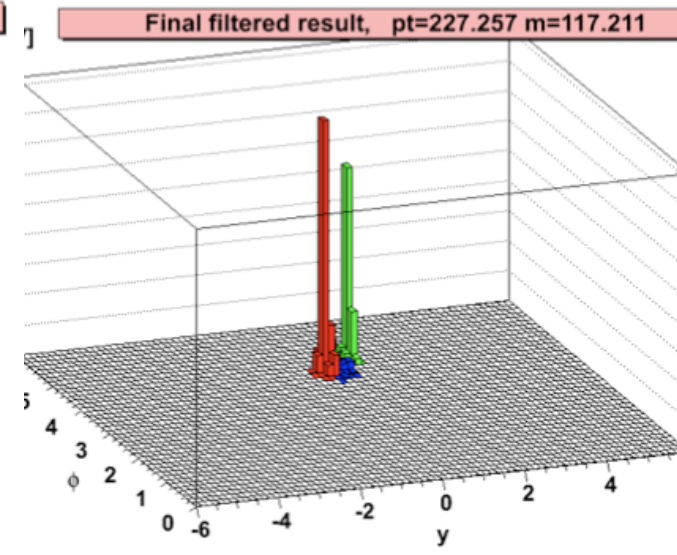
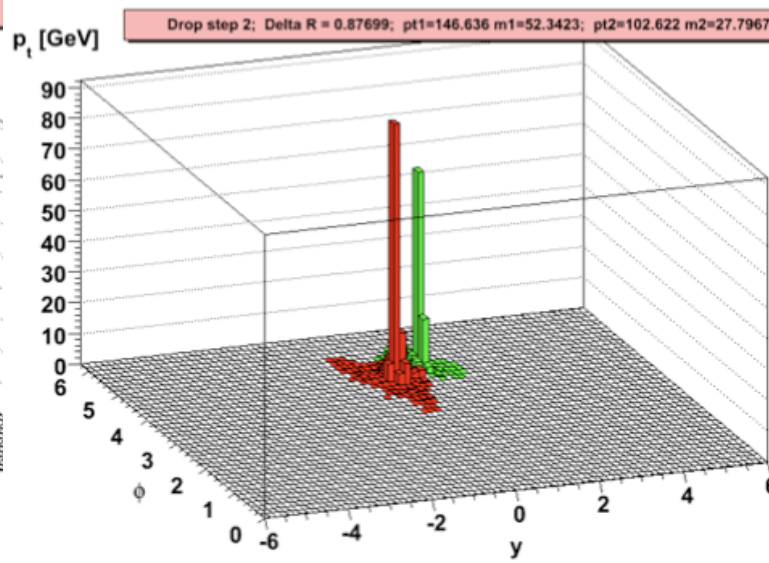
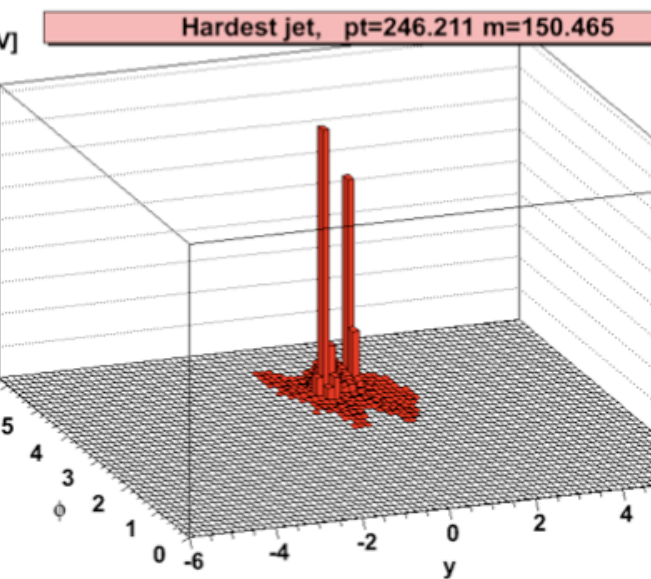
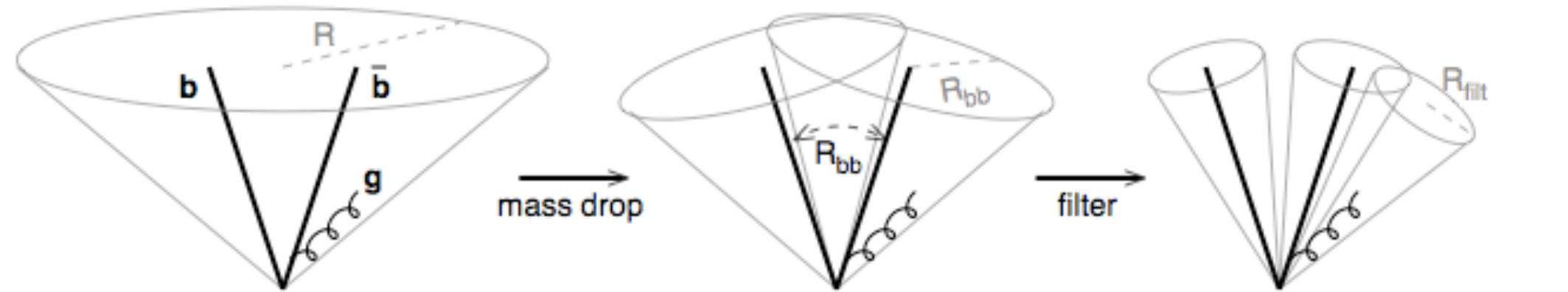
Only keep the n_{filt} hardest jets

The low-momentum stuff surrounding the hard particles has been removed

$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$$

Visualisation of BDRS

Butterworth, Davison, Rubin, Salam, 2008



Cluster with a large R

Undo the clustering into subjects, until a large asymmetry/mass drop is observed: tagging step

Re-cluster with smaller R, and keep only 3 hardest jets: grooming step

Take-home messages

- ▶ A number of different IRC-safe jet algorithms exist
 - ▶ They all try to be good proxies for hard partons, but they have different characteristics, especially with respect to soft particles
- ▶ Jets from all algorithms inevitably suffer from pileup contamination
 - ▶ Techniques exist (though not covered today) to subtract it, either at jet-level, or at particle-level
- ▶ The past ~10 years have seen the emergence of jet-substructure analysis techniques, aimed at helping classifying a jet according to its origin (quark or gluon, heavy particle, etc)
 - ▶ These techniques have been successfully validated in Standard Model analyses and are being employed in BSM searches and even in heavy ion analyses.