

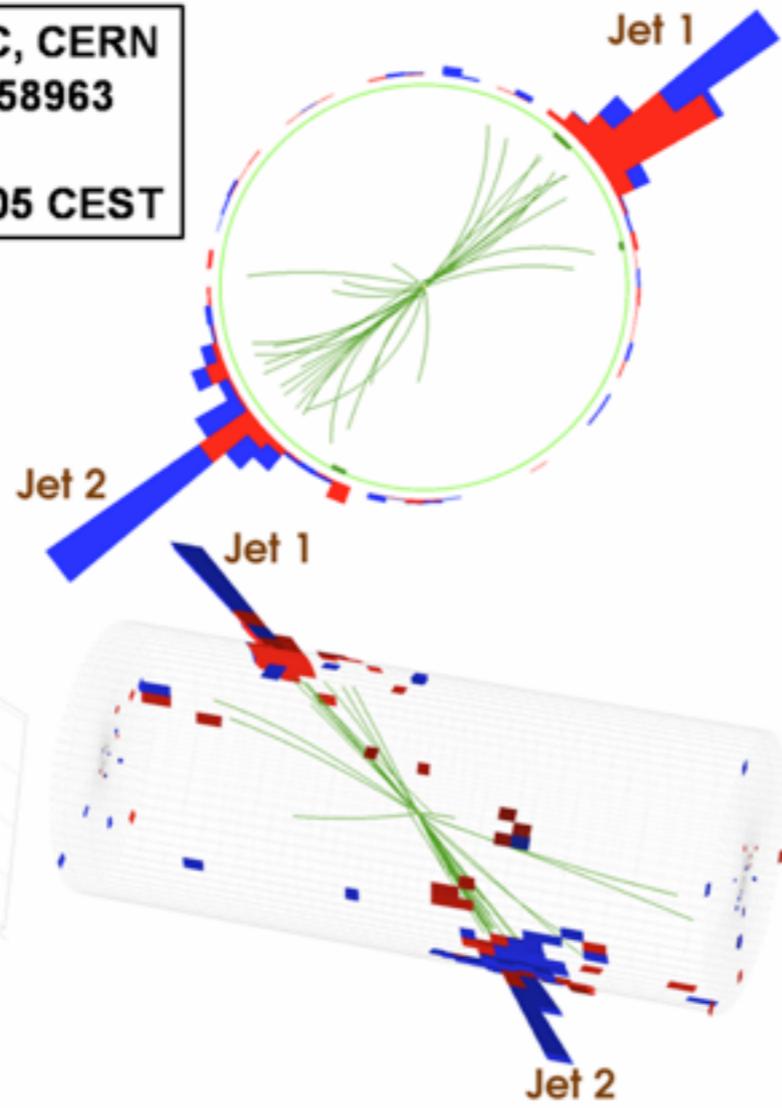
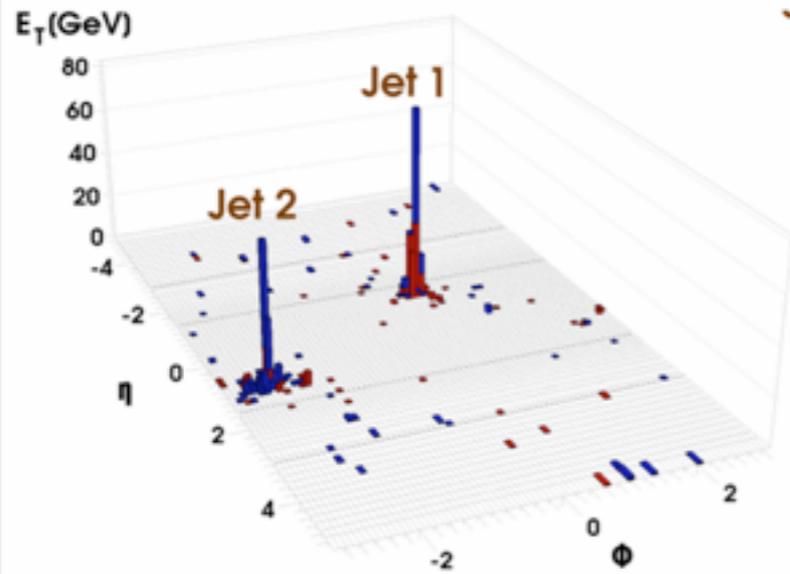
# QCD lecture 8: jets

Gavin Salam, Oxford, February 2020  
as part of Claire Gwenlan's QCD PhD course

(with extensive use of material by  
Matteo Cacciari and Gregory Soyez)

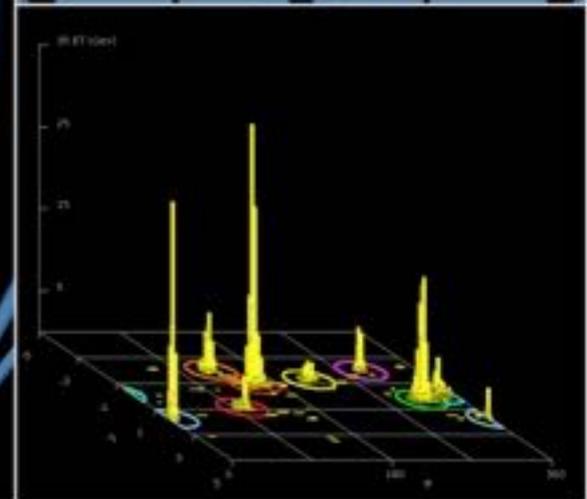
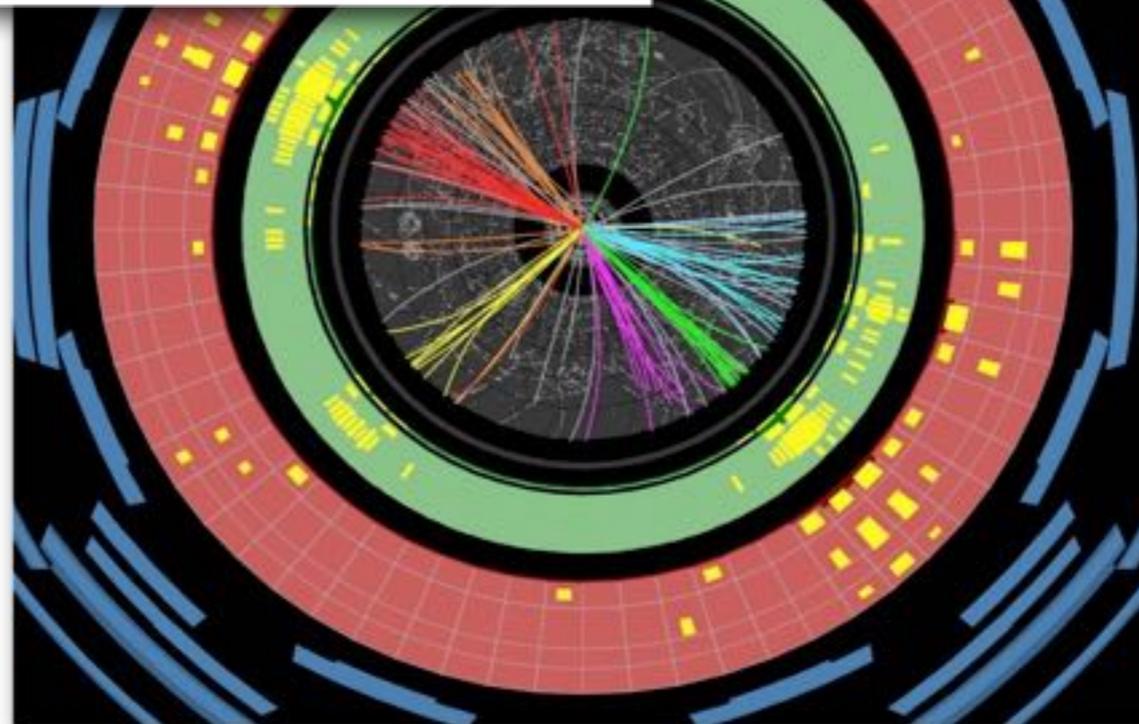
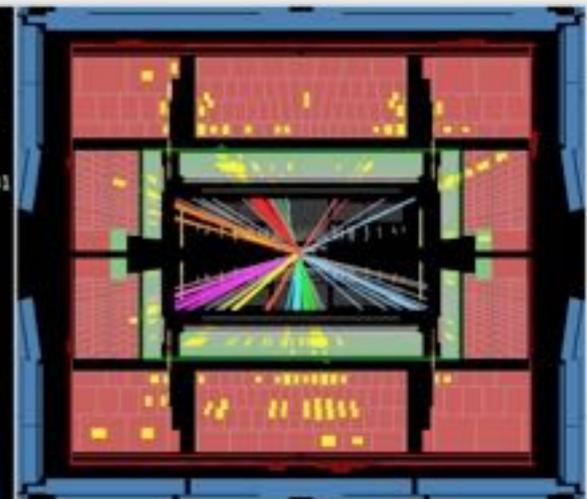


CMS Experiment at LHC, CERN  
Run 133450 Event 16358963  
Lumi section: 285  
Sat Apr 17 2010, 12:25:05 CEST



# JETS

Collimated, energetic bunches of particles



# Find all papers by ATLAS and CMS

## 2106 records found

reportnumber:CERN and (collaboration:ATLAS or collaboration:CMS)  Brief format   [Easy Search](#) [Advanced Search](#)

[find j "Phys.Rev.Lett.,105" :: more](#)

Sort by:  Display results:

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[HEP](#) 2,106 records found 1 - 25  jump to record:

### 1. Measurement of the cross section for electroweak production of a Z boson, a photon and two jets

#### TeV and constraints on anomalous quartic couplings

CMS Collaboration (Albert M Sirunyan (Yerevan Phys. Inst.) *et al.*). Feb 23, 2020.

CMS-SMP-18-007, CERN-EP-2020-007

e-Print: [arXiv:2002.09902](#) [hep-ex] | [PDF](#)

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### 2. Observation of the associated production of a top quark and a Z boson in pp collisions at $\sqrt{s} = 13$ TeV

ATLAS Collaboration (Georges Aad (Marseille, CPPM) *et al.*). Feb 18, 2020. 44 pp.

CERN-EP-2019-273

e-Print: [arXiv:2002.07546](#) [hep-ex] | [PDF](#)

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### 3. A measurement of the Higgs boson mass in the diphoton decay channel

CMS Collaboration (Albert M Sirunyan (Yerevan Phys. Inst.) *et al.*). Feb 15, 2020.

CMS-HIG-19-004, CERN-EP-2020-004

e-Print: [arXiv:2002.06398](#) [hep-ex] | [PDF](#)

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Pull out those that refer to one widely used jet-alg  
1289 records found

reportnumber:CERN and (collaboration:ATLAS or collaboration:CMS) and refersto:recid:779060   [Easy Search](#)  
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Sort by:    Display results:

> 60% of papers use jets!

**HEP** 1,289 records found 1 - 25

1. **Measurement of the cross section for electroweak production of a Z boson, a photon and two jets in pp collisions at  $\sqrt{s} = 13$  TeV and constraints on anomalous quartic couplings**

CMS Collaboration (Albert M Sirunyan (Yerevan Phys. Inst.) *et al.*). Feb 23, 2020.

CMS-SMP-18-007, CERN-EP-2020-007

e-Print: [arXiv:2002.09902 \[hep-ex\]](#) | [PDF](#)

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CERN-EP-2019-273

e-Print: [arXiv:2002.07546 \[hep-ex\]](#) | [PDF](#)

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[CERN Document Server](#); [ADS Abstract Service](#); [Link to ATLAS News article](#)

[Detailed record](#)

3. **Search for physics beyond the standard model in events with jets and two same-sign or at least three opposite-sign dileptons in pp collisions at  $\sqrt{s} = 13$  TeV**

CMS Collaboration (Albert M Sirunyan (Yerevan Phys. Inst.) *et al.*). Jan 27, 2020. 47 pp.

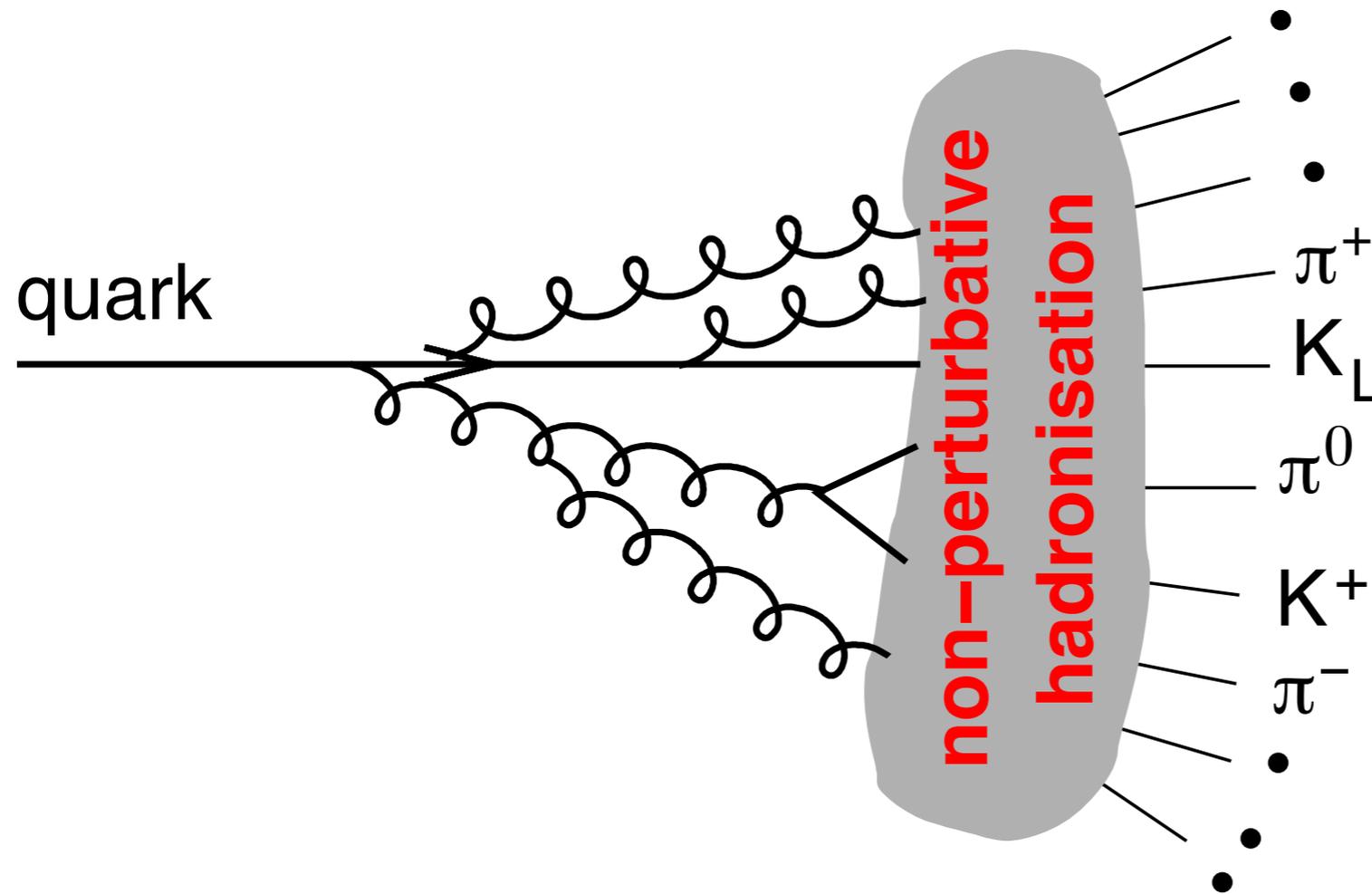
CMS-SUS-19-008, CERN-EP-2020-001

e-Print: [arXiv:2001.10086 \[hep-ex\]](#) | [PDF](#)

[References](#) | [BibTeX](#) | [LaTeX\(US\)](#) | [LaTeX\(EU\)](#) | [Harvmac](#) | [EndNote](#)

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# Why do we see jets?

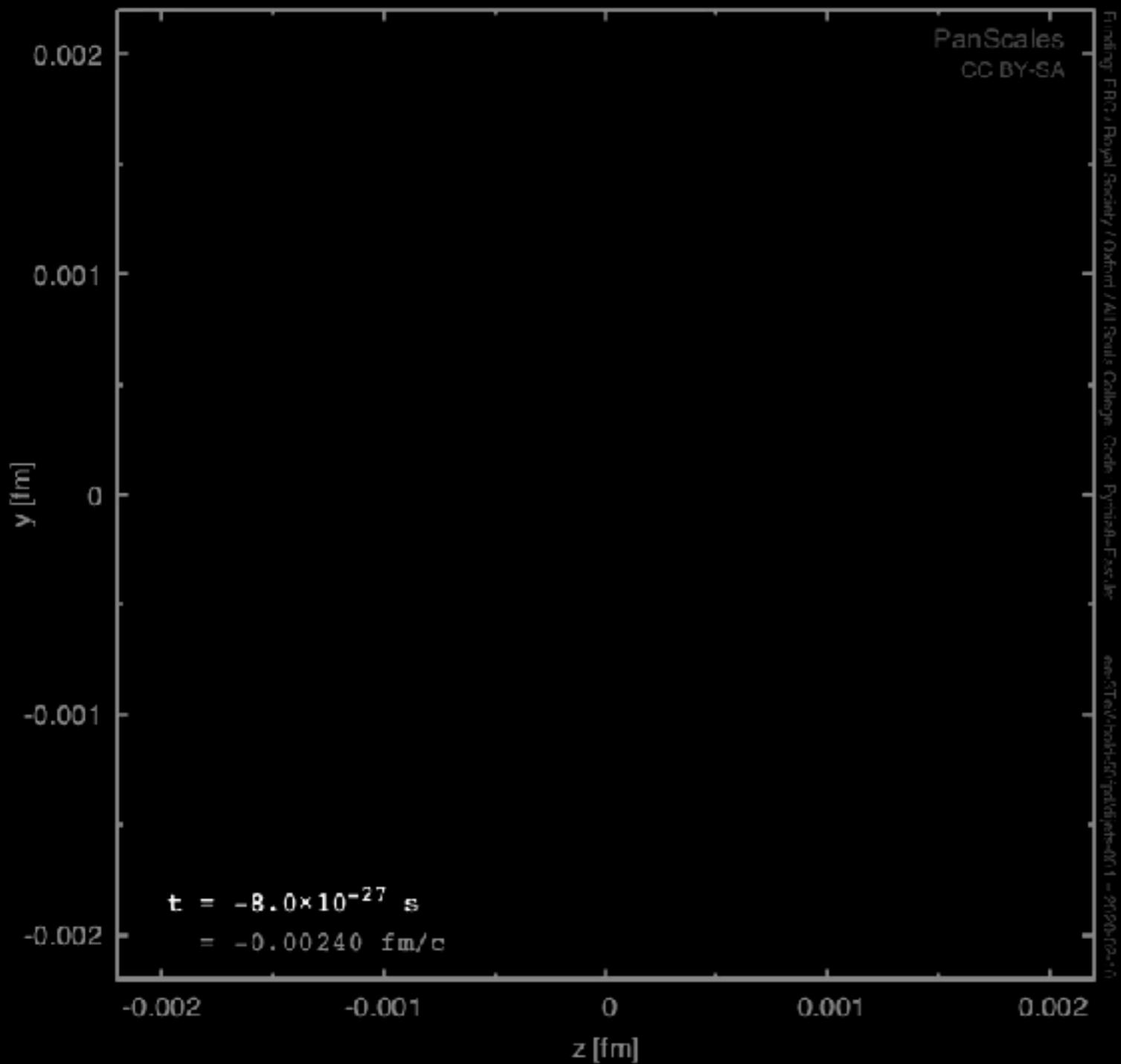


Gluon emission

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

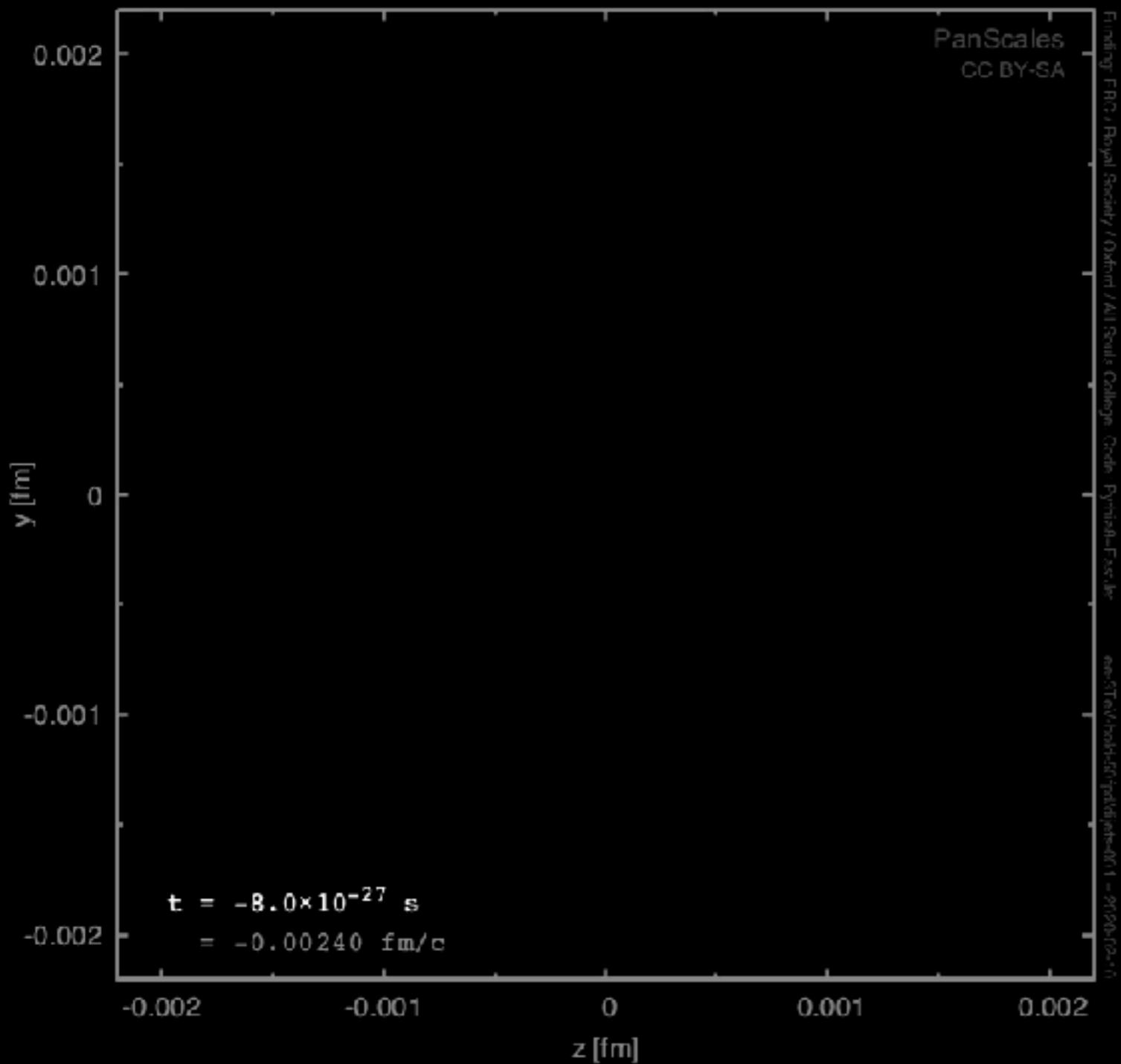
Non-perturbative physics

$$\alpha_s \sim 1$$



- incoming beam particle
- intermediate particle
- final particle

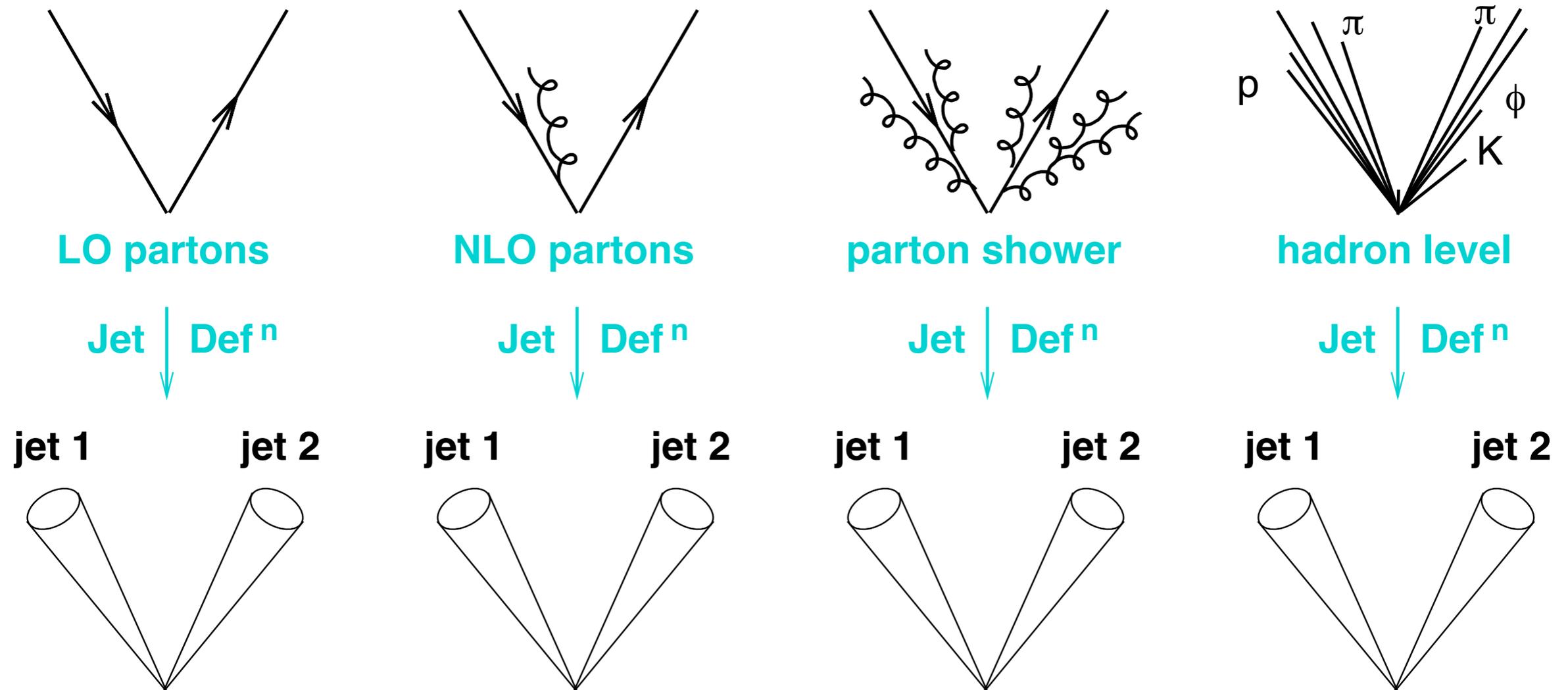
Event evolution spans 7 orders of magnitude in space-time



- incoming beam particle
- intermediate particle
- final particle

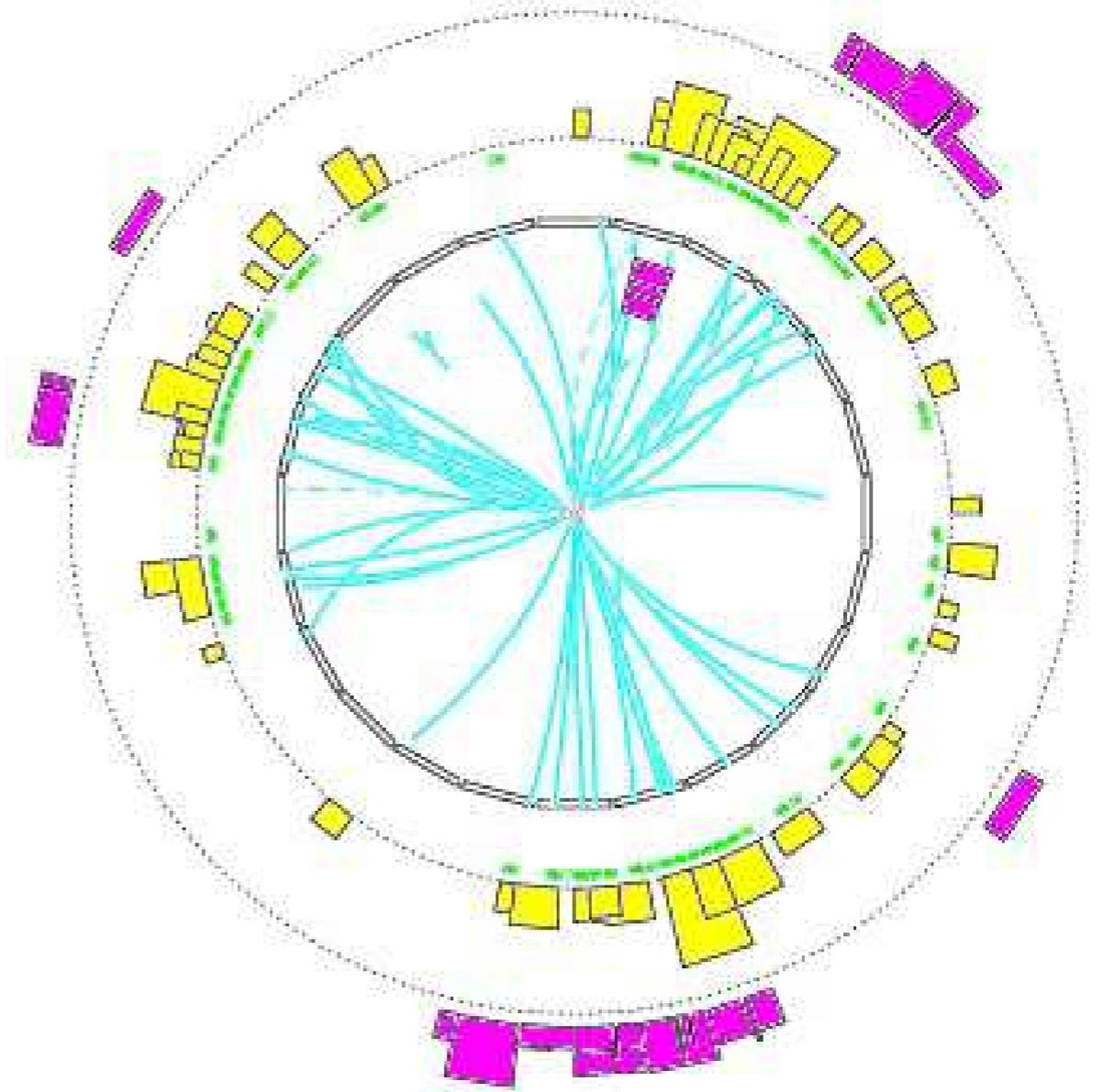
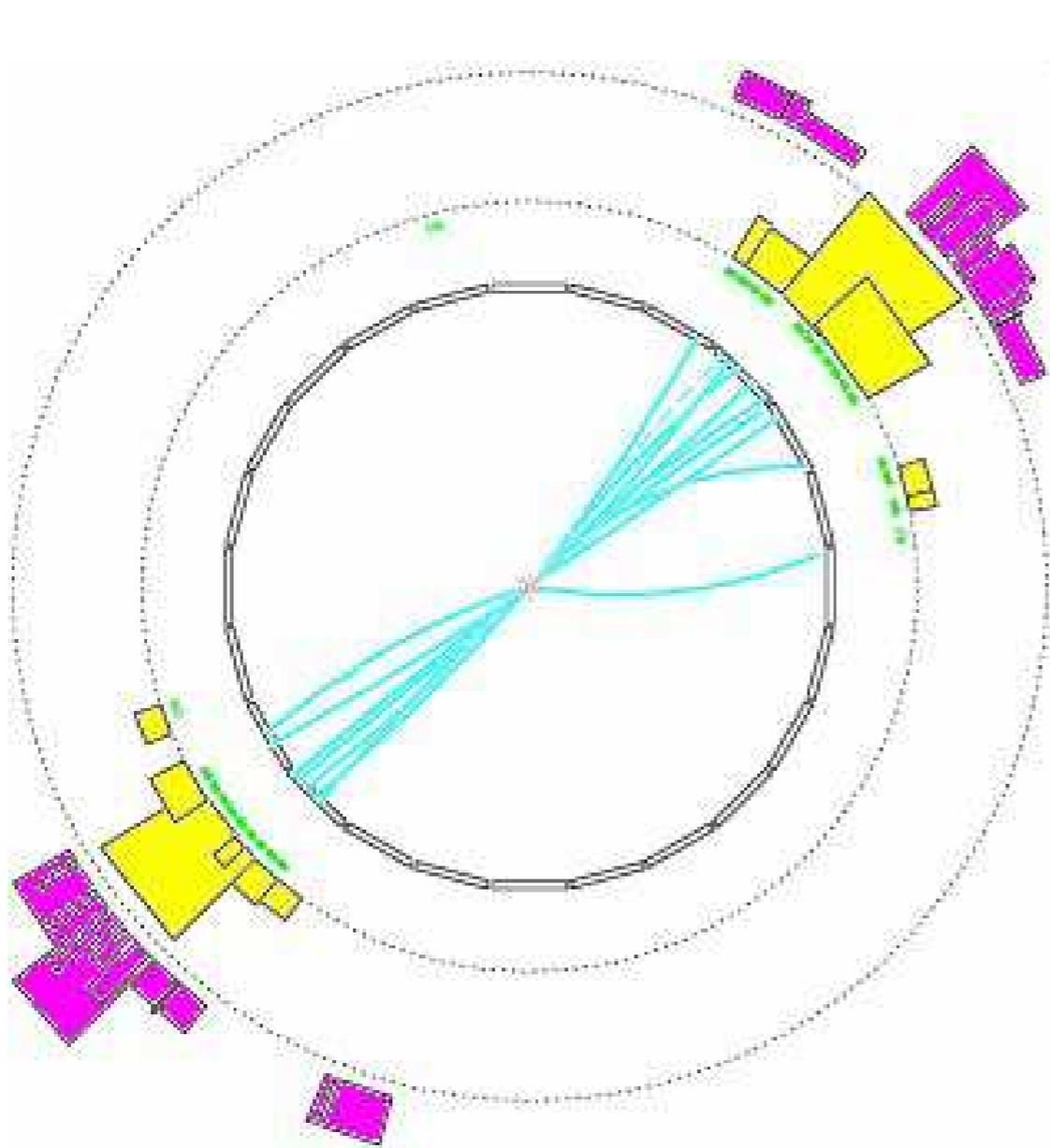
Event evolution spans 7 orders of magnitude in space-time

# Jet finding as a form of projection

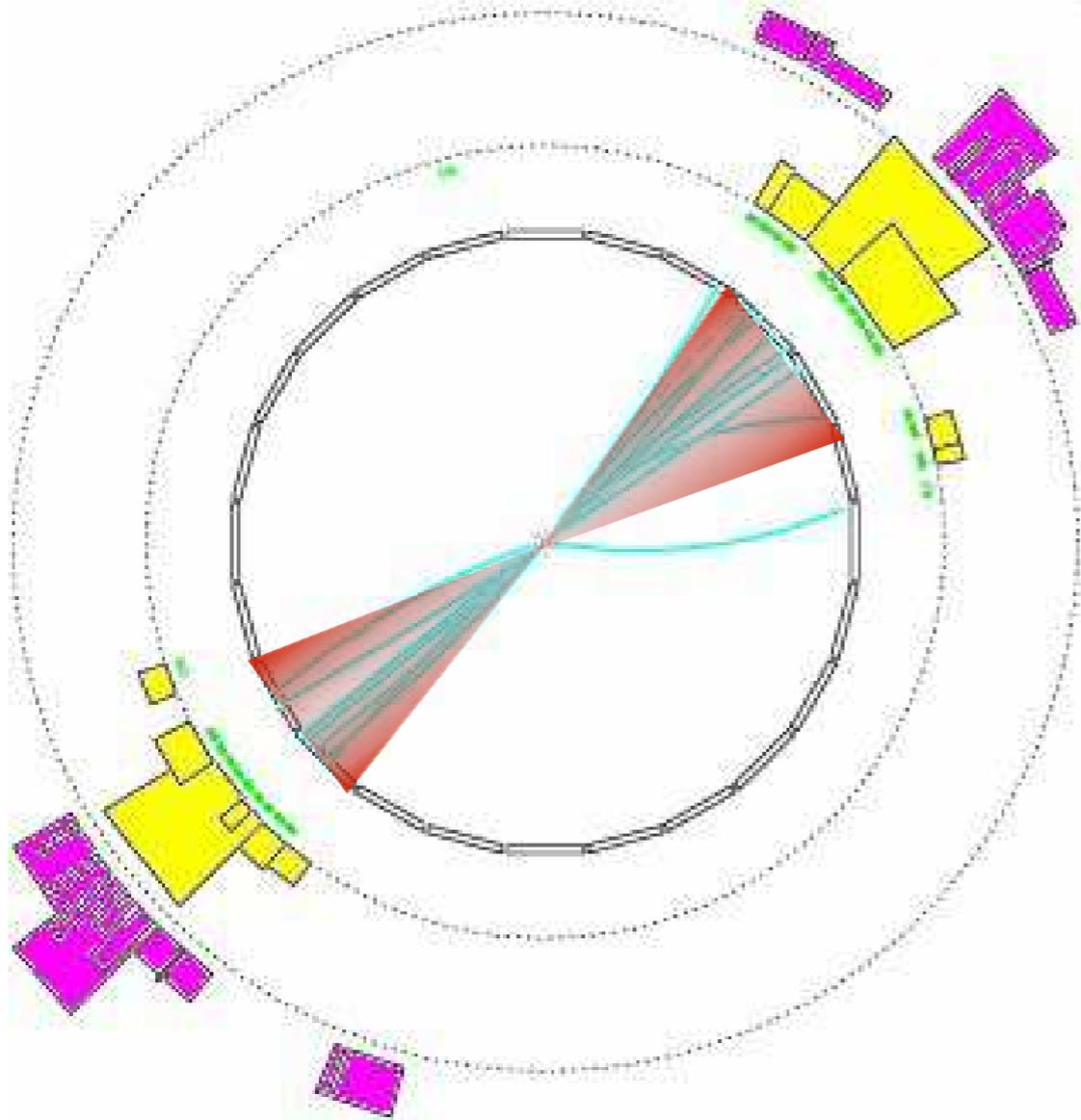


Projection to jets should be resilient to QCD effects

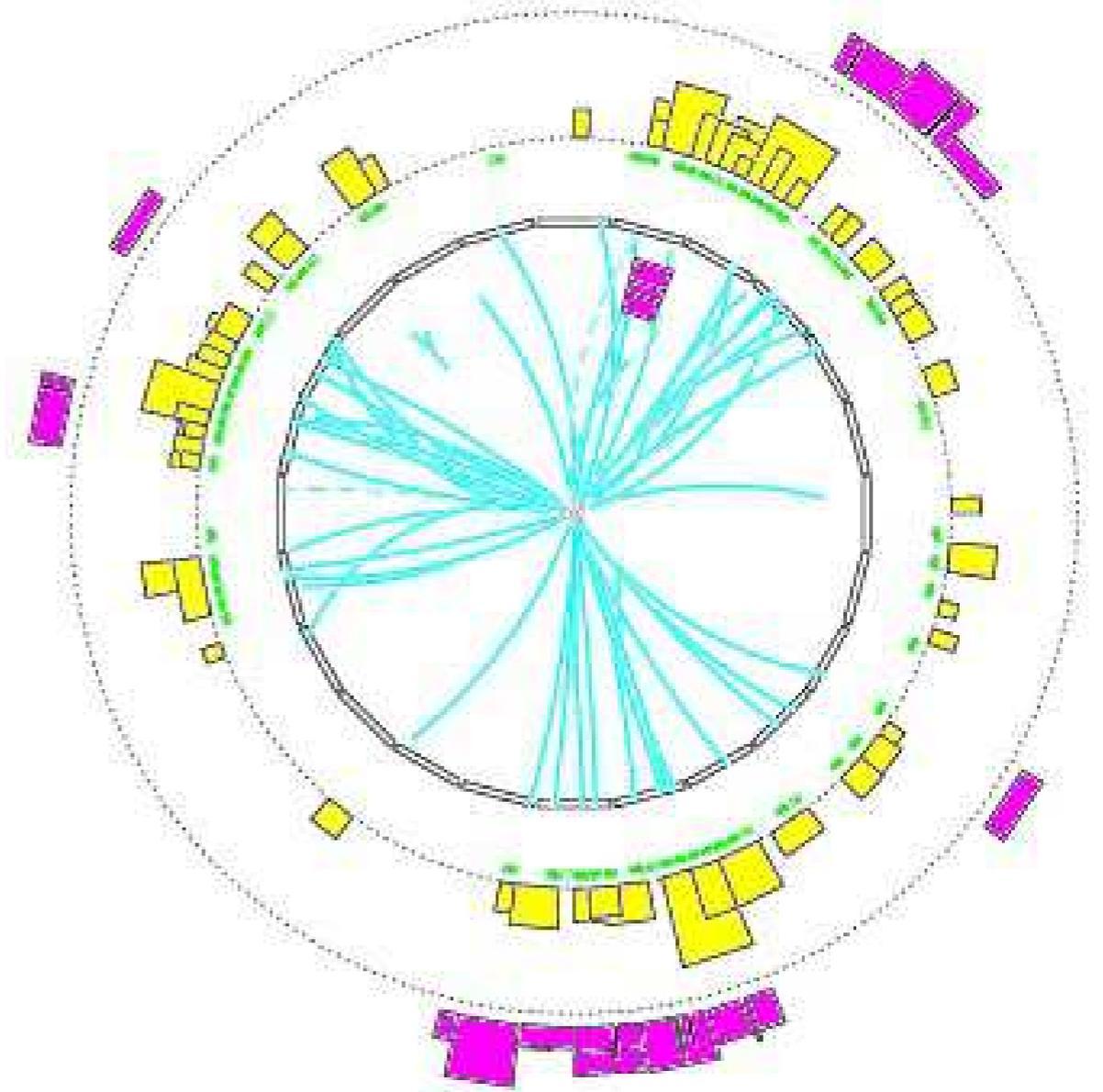
# Reconstructing jets is an ambiguous task



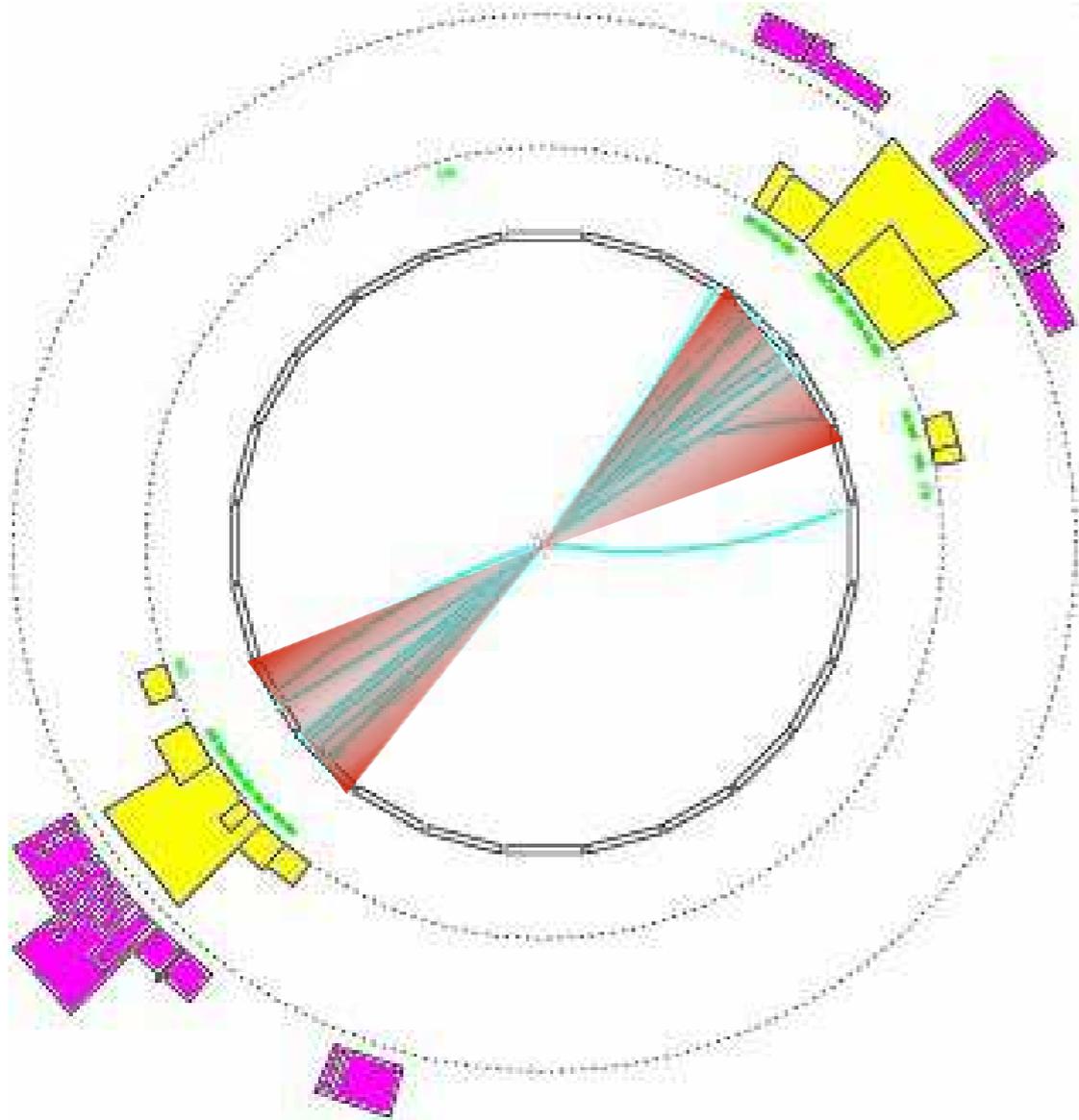
# Reconstructing jets is an ambiguous task



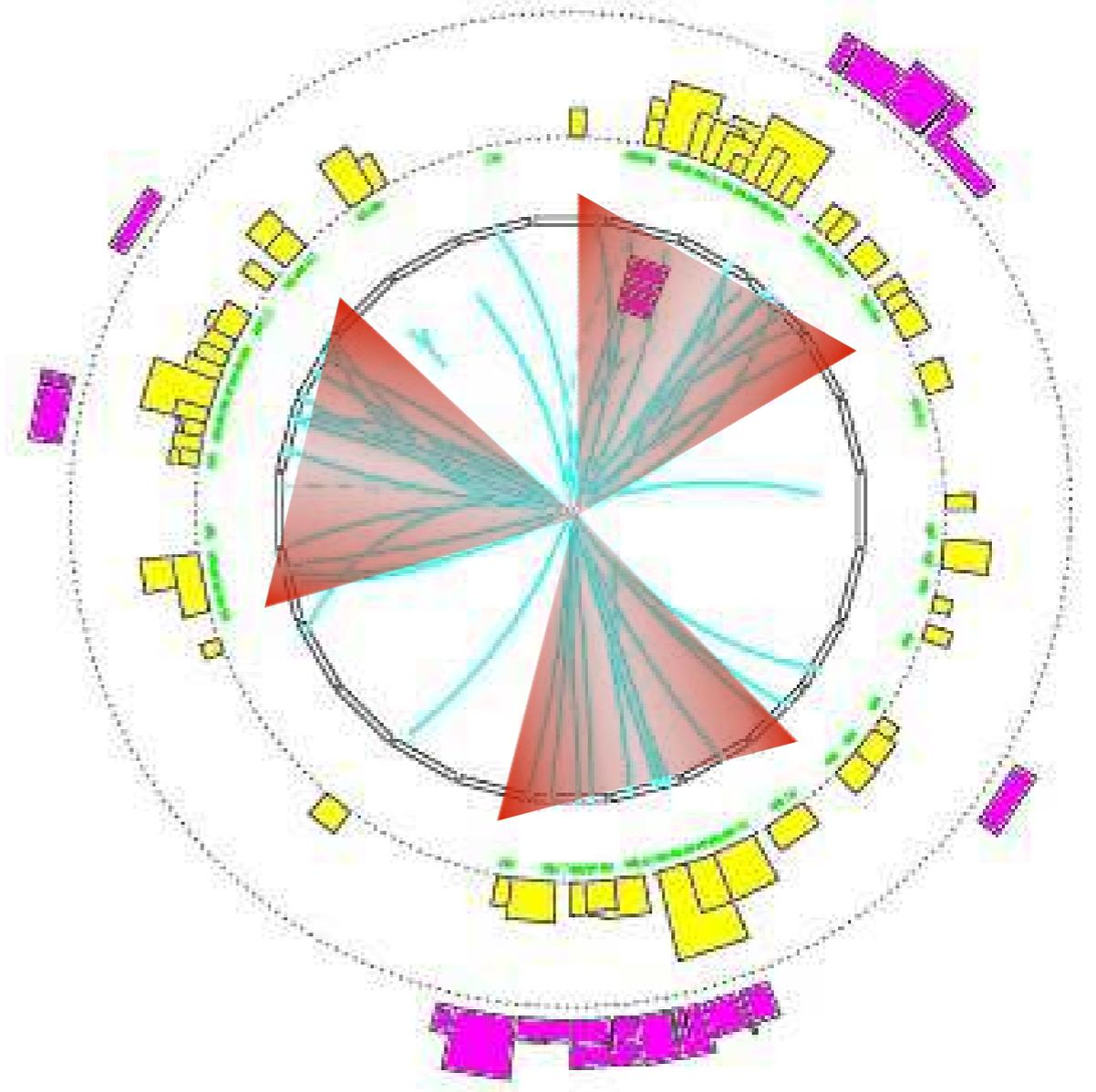
2 clear jets



# Reconstructing jets is an ambiguous task

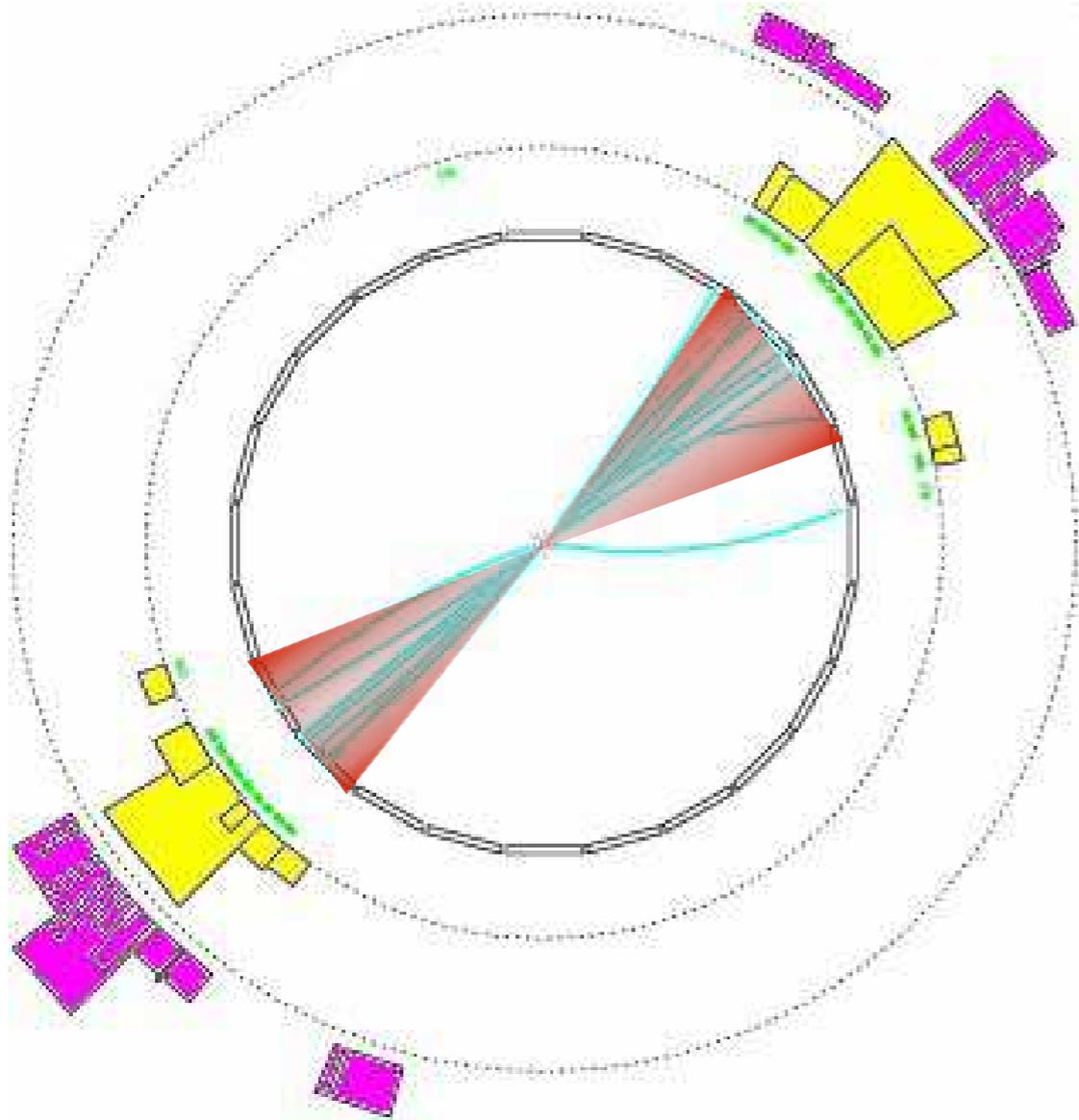


2 clear jets

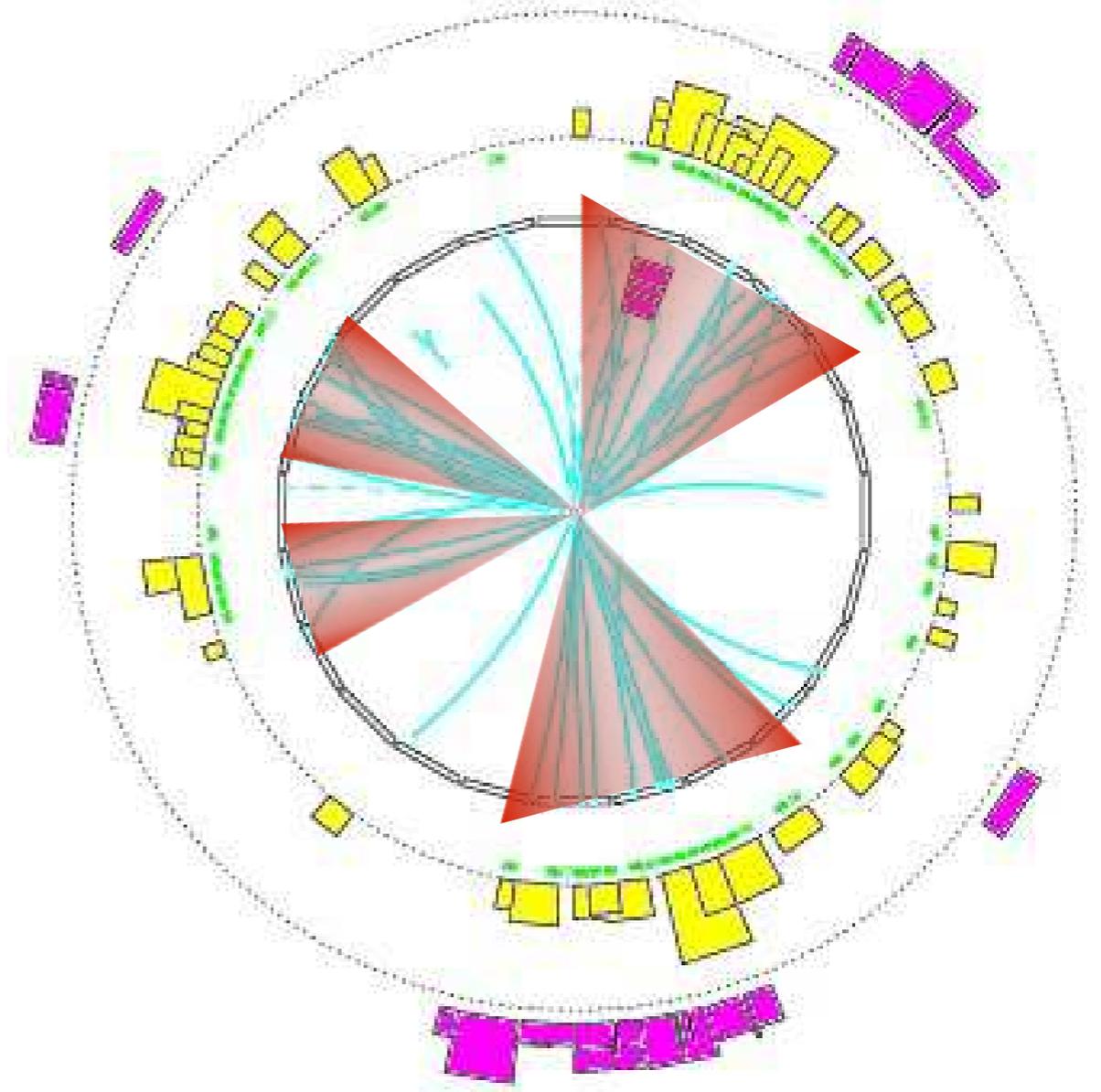


3 jets?

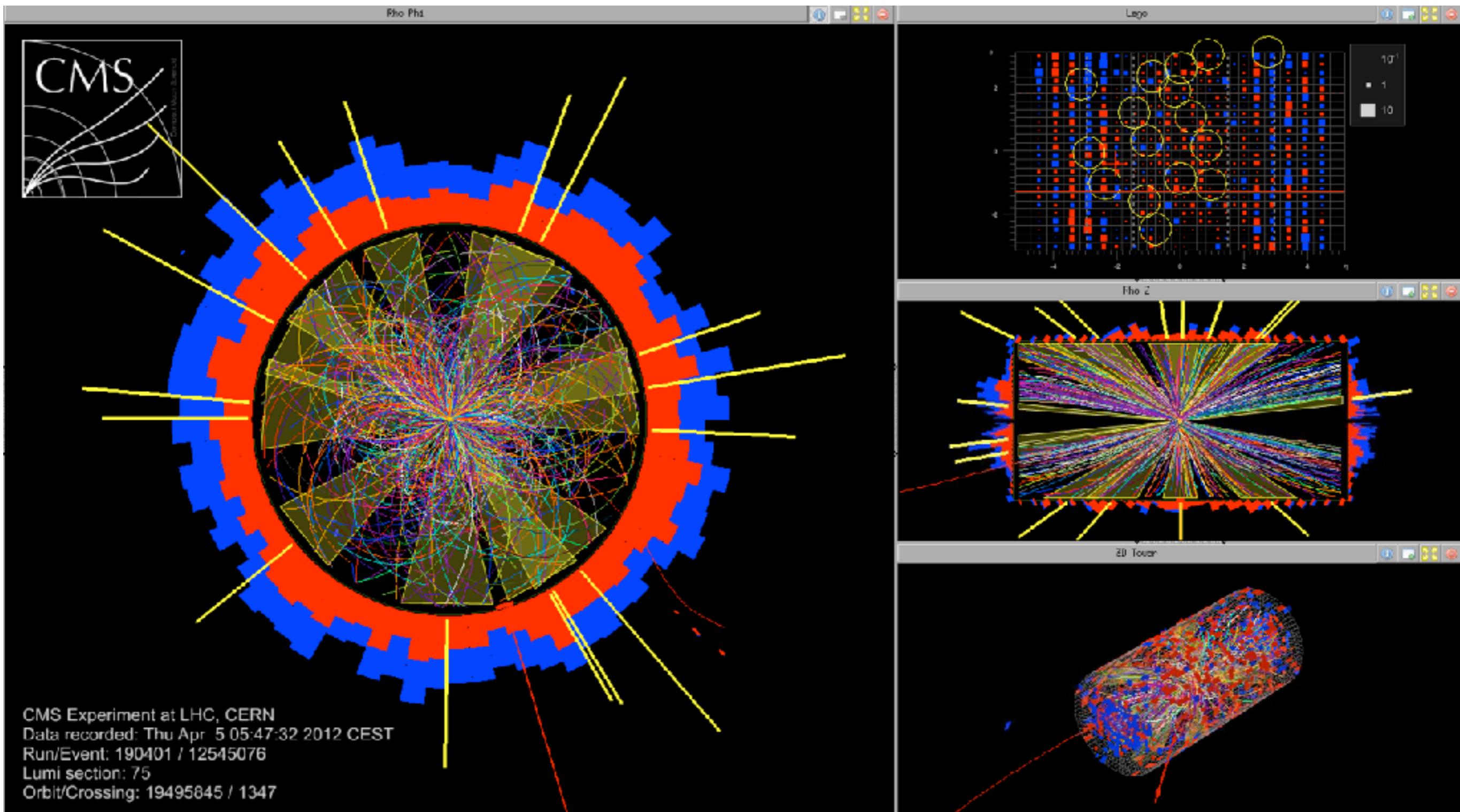
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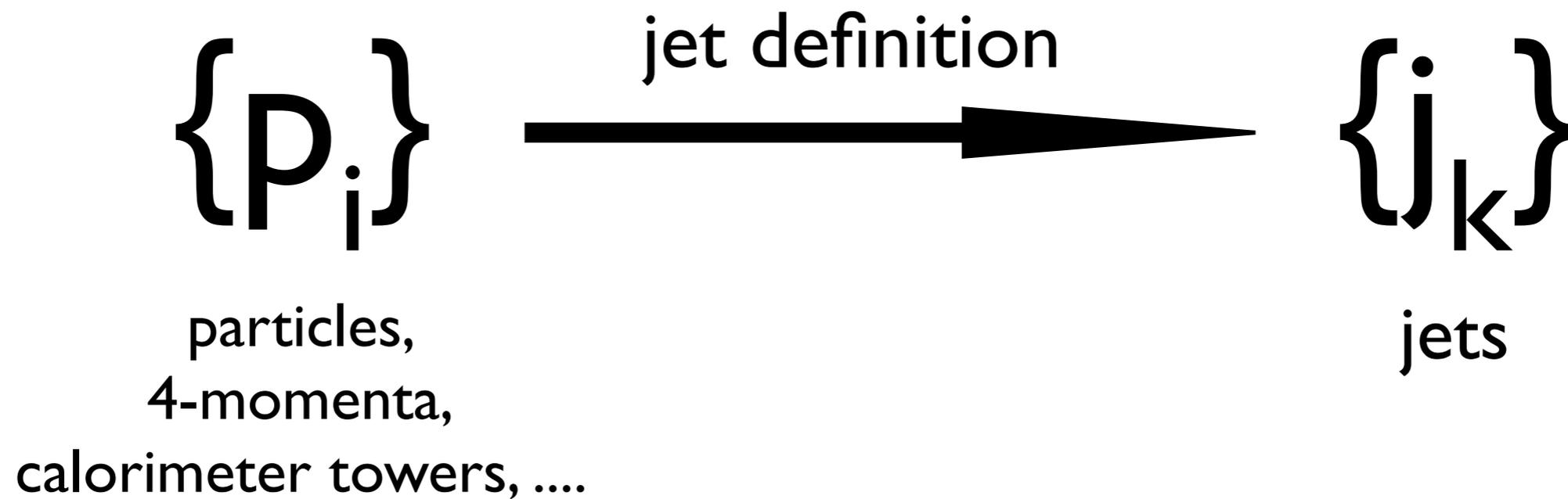
2 clear jets



3 jets?  
**or 4 jets?**



# Make a choice: specify a jet definition



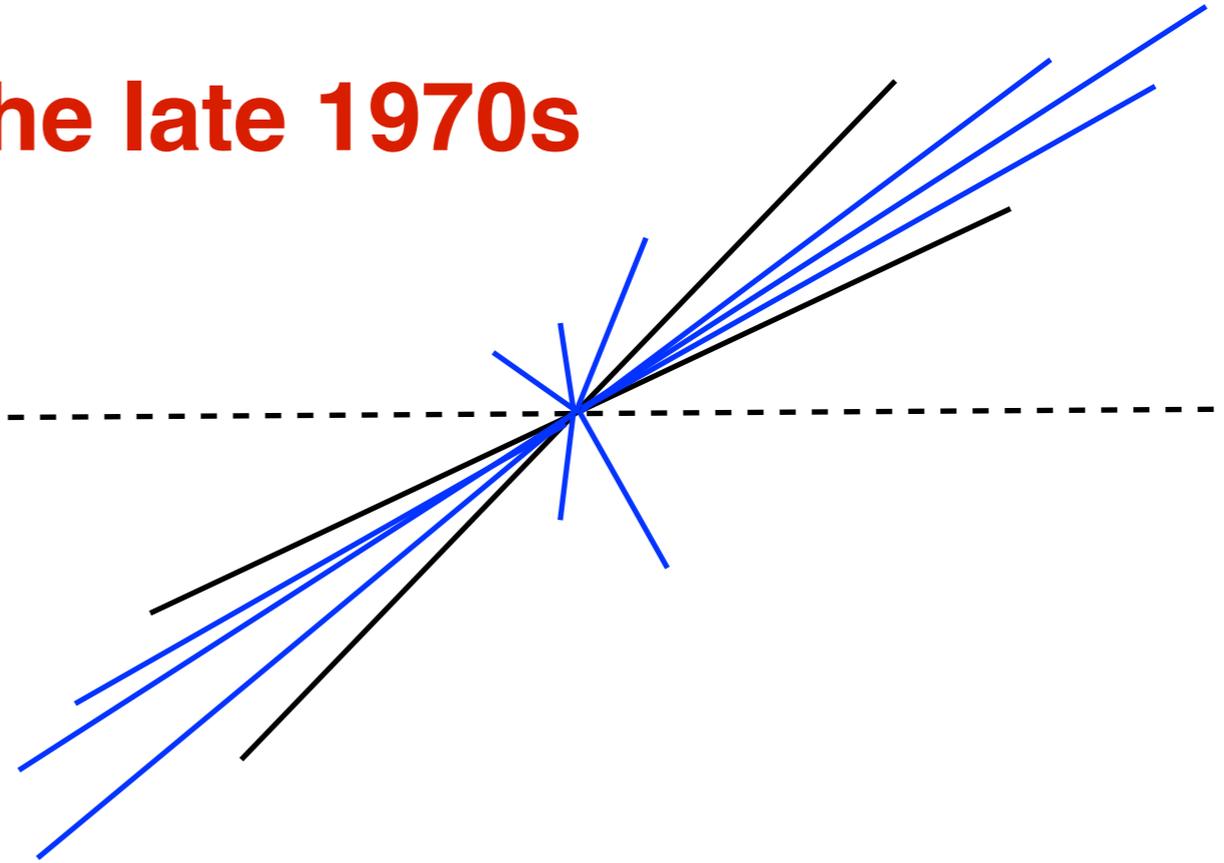
- Which particles do you put together into a same jet?
- How do you recombine their momenta (4-momentum sum is the obvious choice, right?)

*“Jet [definitions] are legal contracts between theorists and experimentalists”*  
-- MJ Tannenbaum

They're also a way of organising the information in an event  
1000's of particles per events, up to 20.000,000 events per second

# Jet definitions date back to the late 1970s

Sterman and Weinberg,  
Phys. Rev. Lett. 39, 1436 (1977):

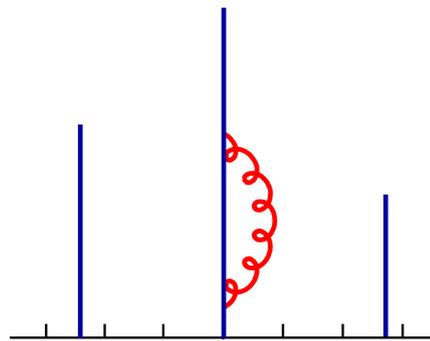


To study jets, we consider the partial cross section  $\sigma(E, \theta, \Omega, \epsilon, \delta)$  for  $e^+e^-$  hadron production events, in which all but a fraction  $\epsilon \ll 1$  of the total  $e^+e^-$  energy  $E$  is emitted within some pair of oppositely directed cones of half-angle  $\delta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

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

# Key requirement: infrared and collinear safety

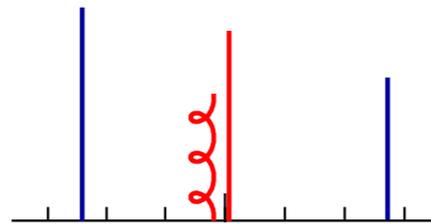
## Collinear Safe



jet 1

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

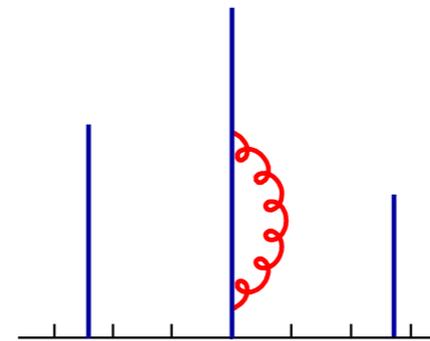
**Infinities cancel**



jet 1

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

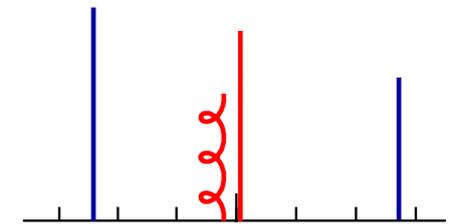
## Collinear Unsafe



jet 1

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

**Infinities do not cancel**



jet 1

jet 2

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

**Invalidates perturbation theory**

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

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

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

## 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

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

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

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

## 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**

If a particle  $i$  has no neighbours  $j$  within a distance  $\Delta R_{ij} \leq R$ , then  $d_{iB} < \text{all } d_{ij}$ , and  $i$  becomes a jet.

**$k_t$  alg.:** Find smallest of

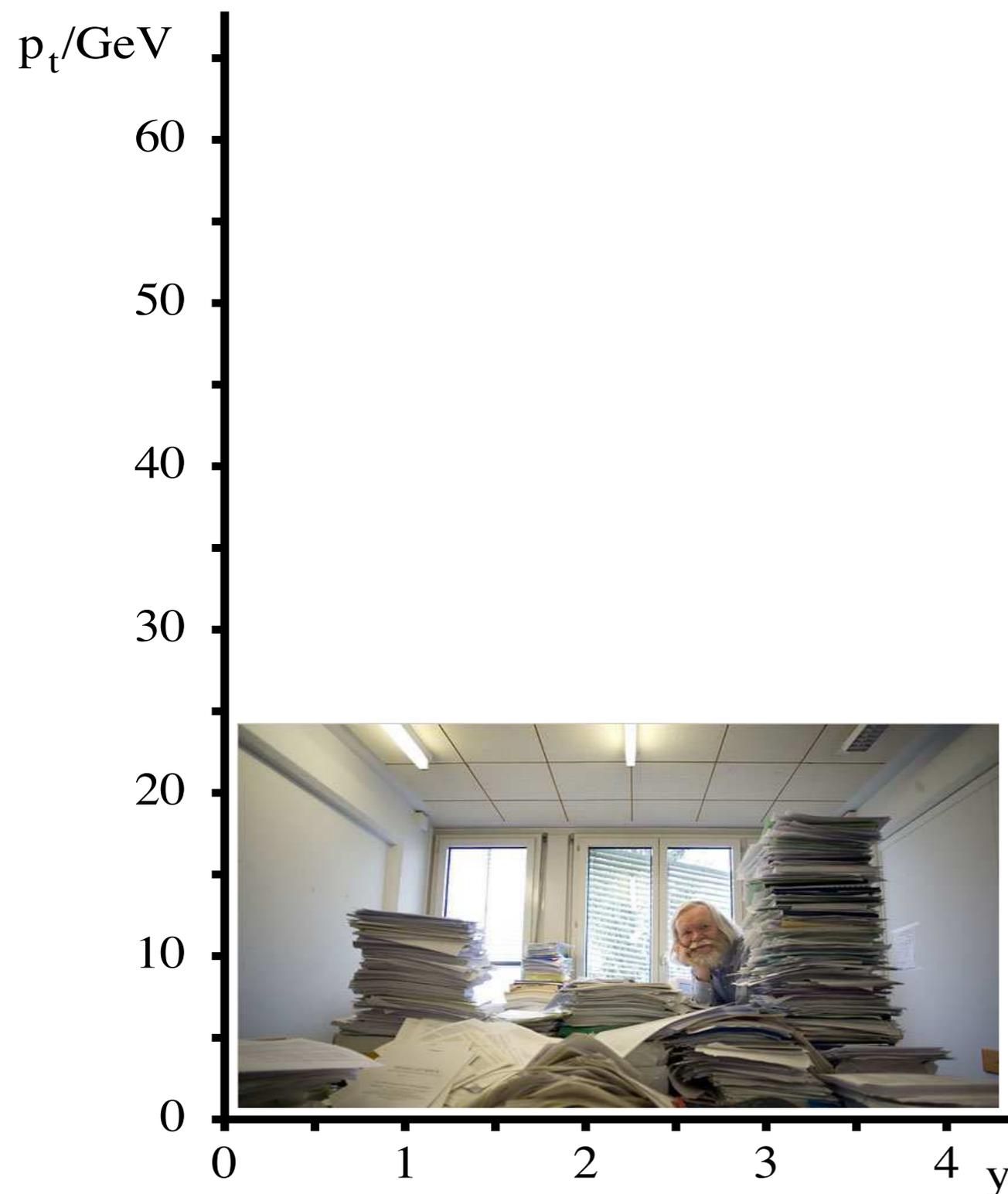
$$d_{ij} = \min(k_{ti}^2, k_{tj}^2) \frac{\Delta R_{ij}^2}{R^2}, \quad d_{iB} = k_{ti}^2$$

- ▶ If  $d_{ij}$  recombine
- ▶ if  $d_{iB}$ ,  $i$  is a jet

Example clustering with  $k_t$  algorithm,  $R = 1.0$

$\phi$  assumed 0 for all towers





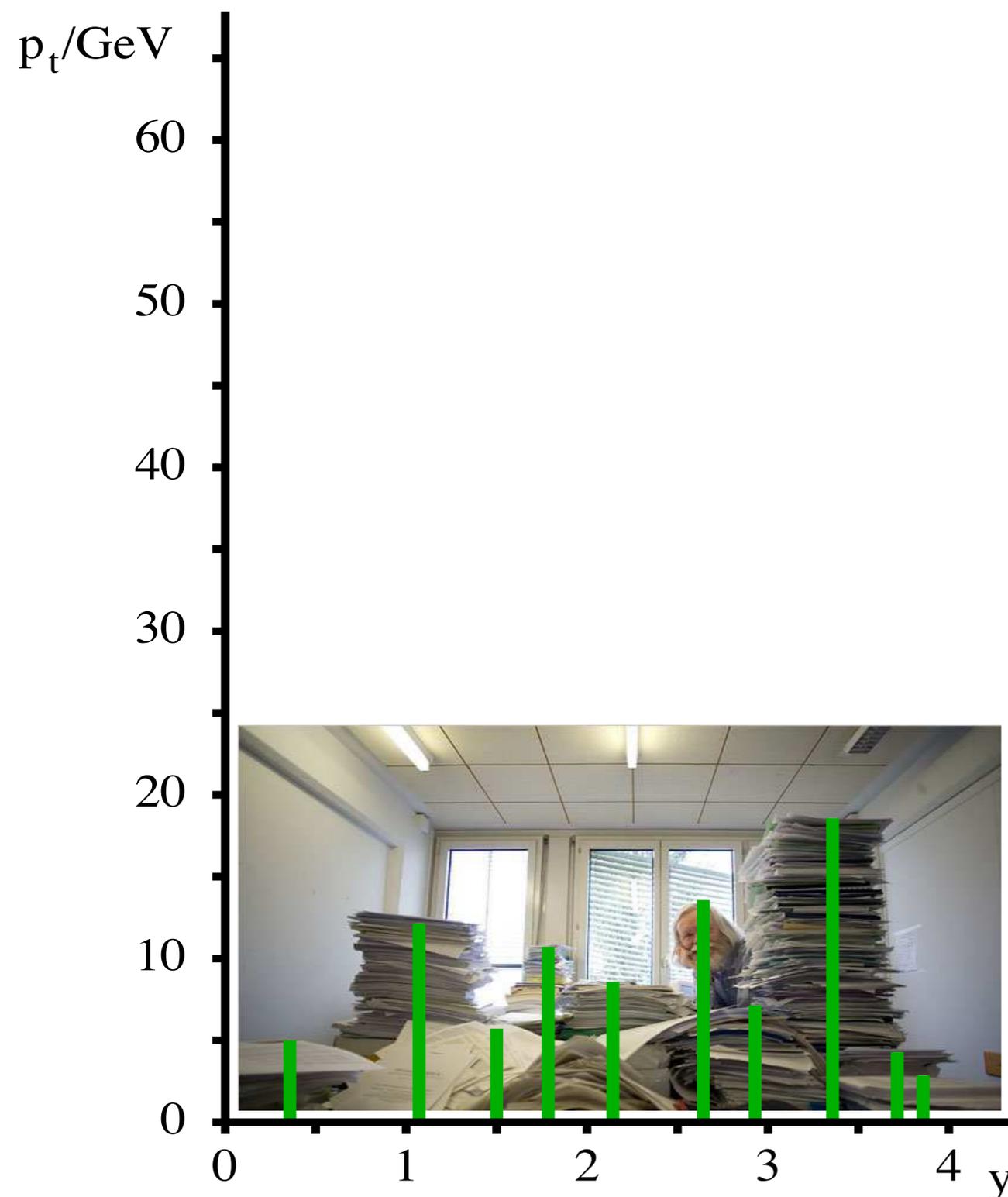
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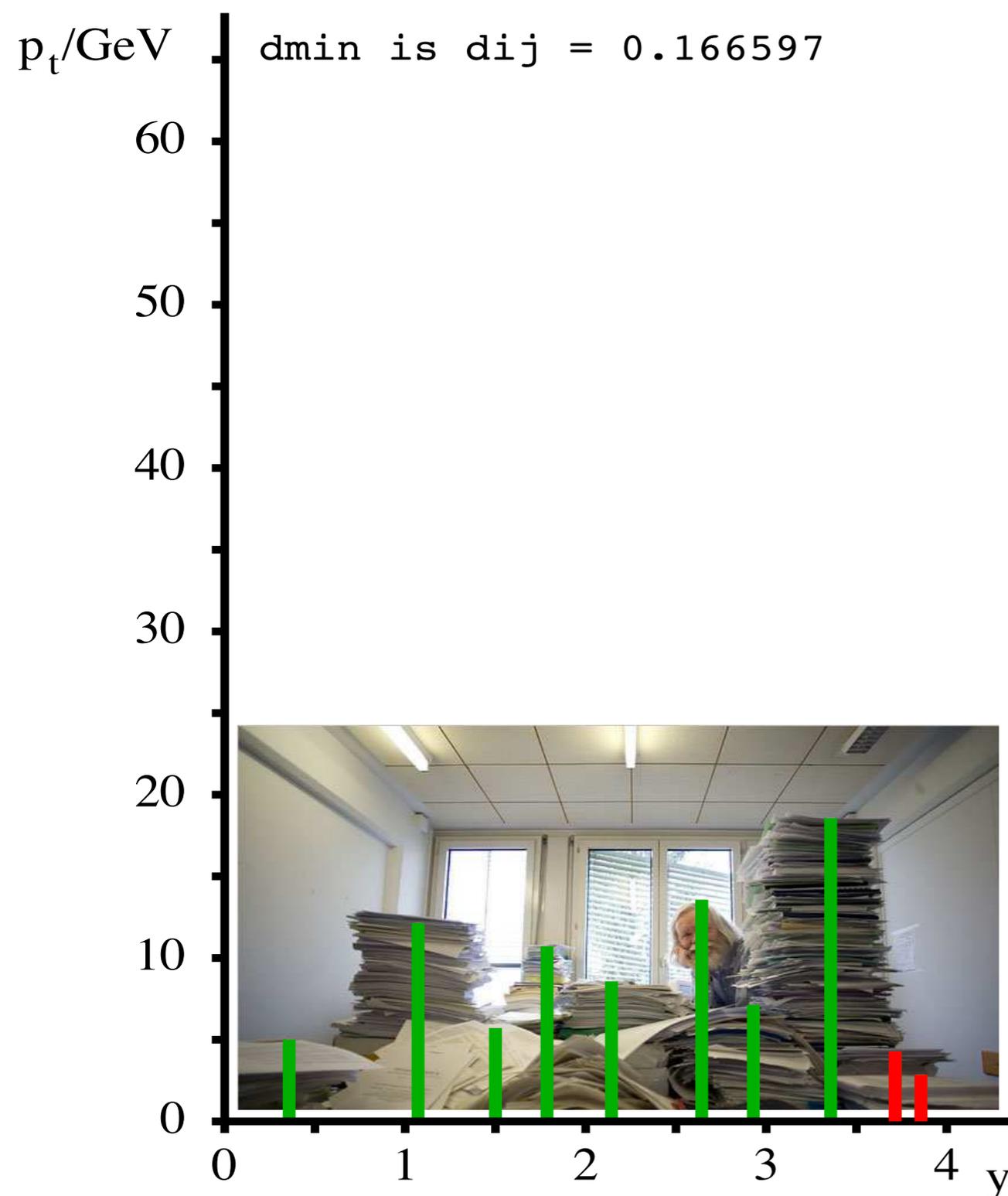
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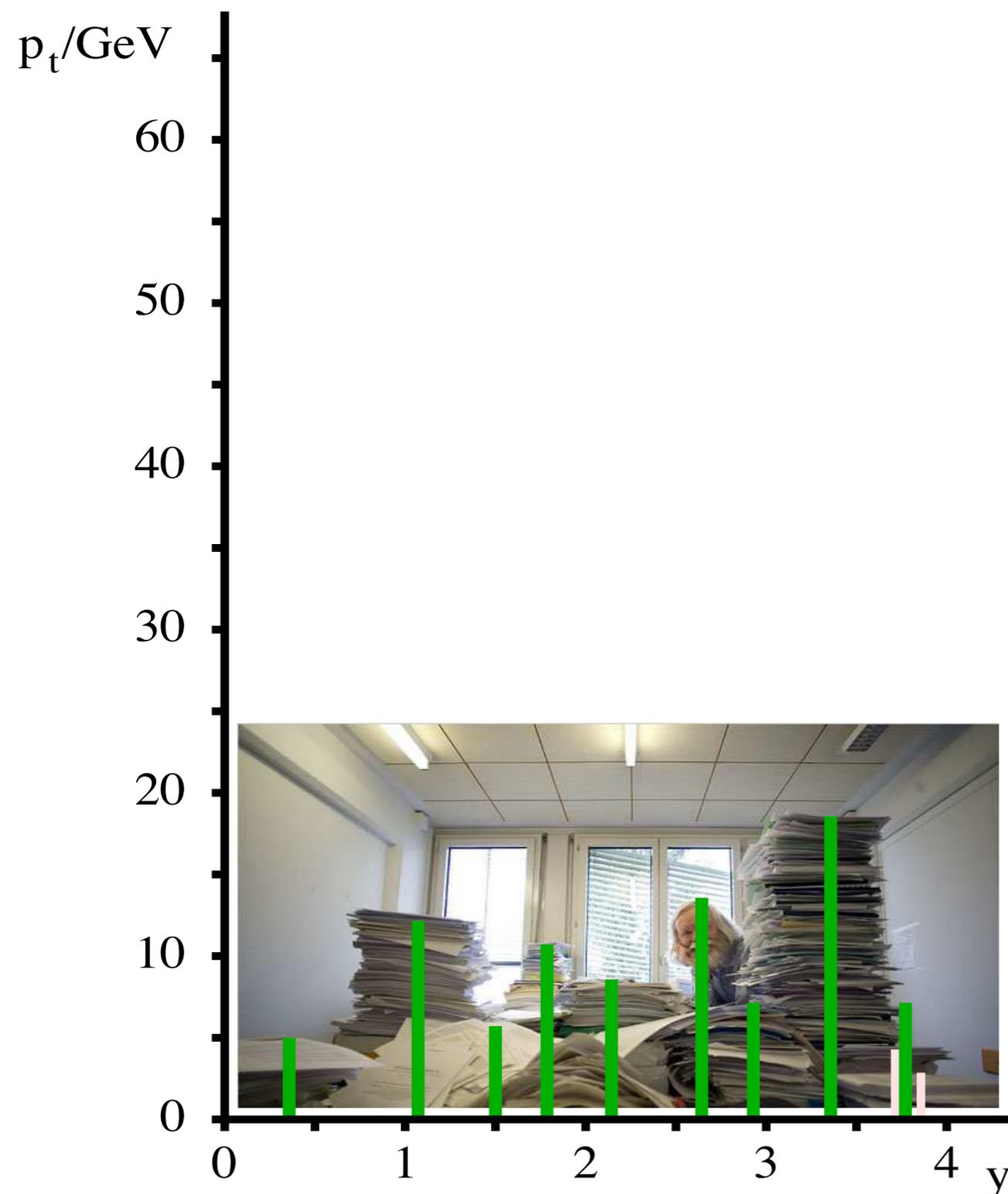
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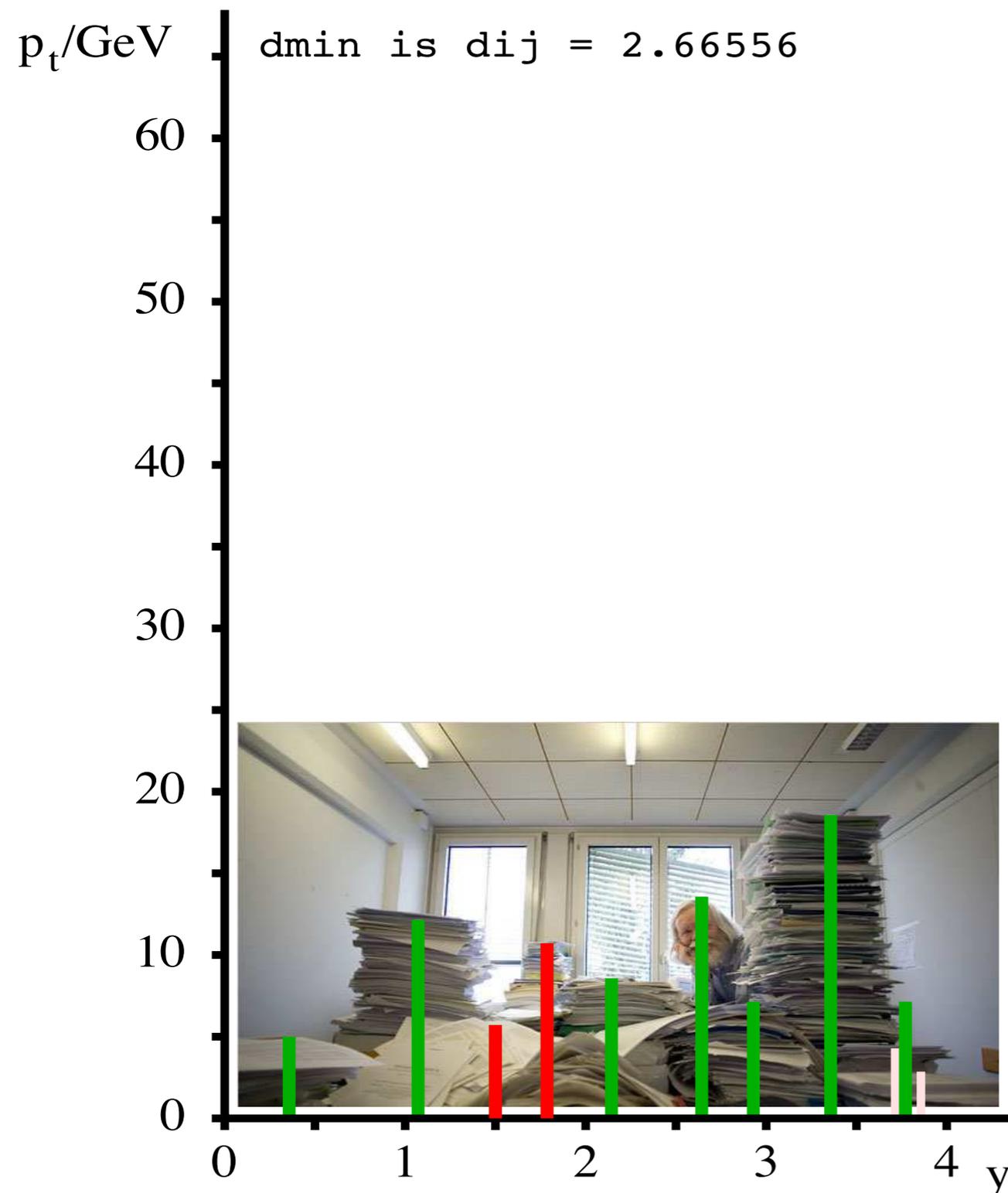
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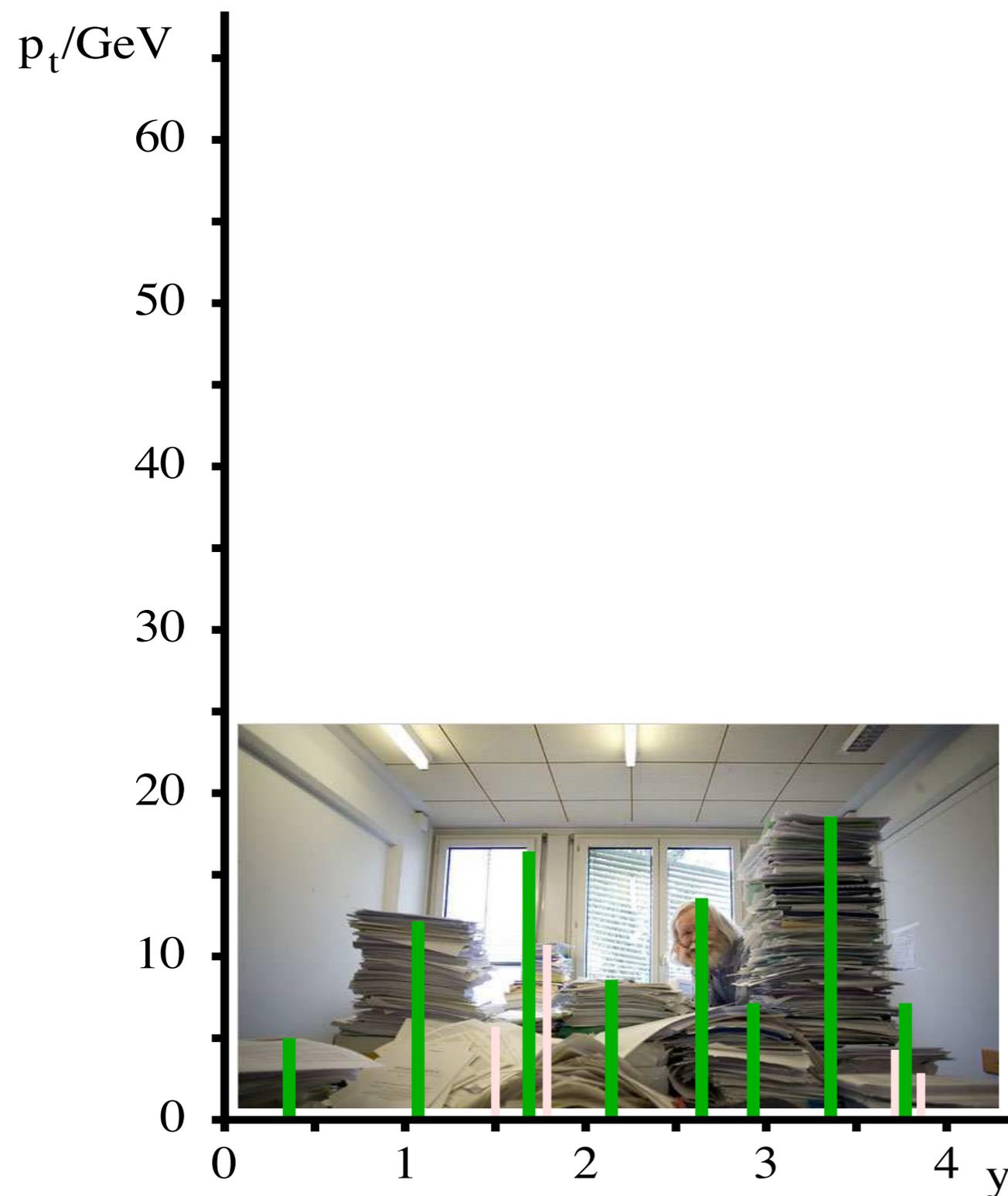
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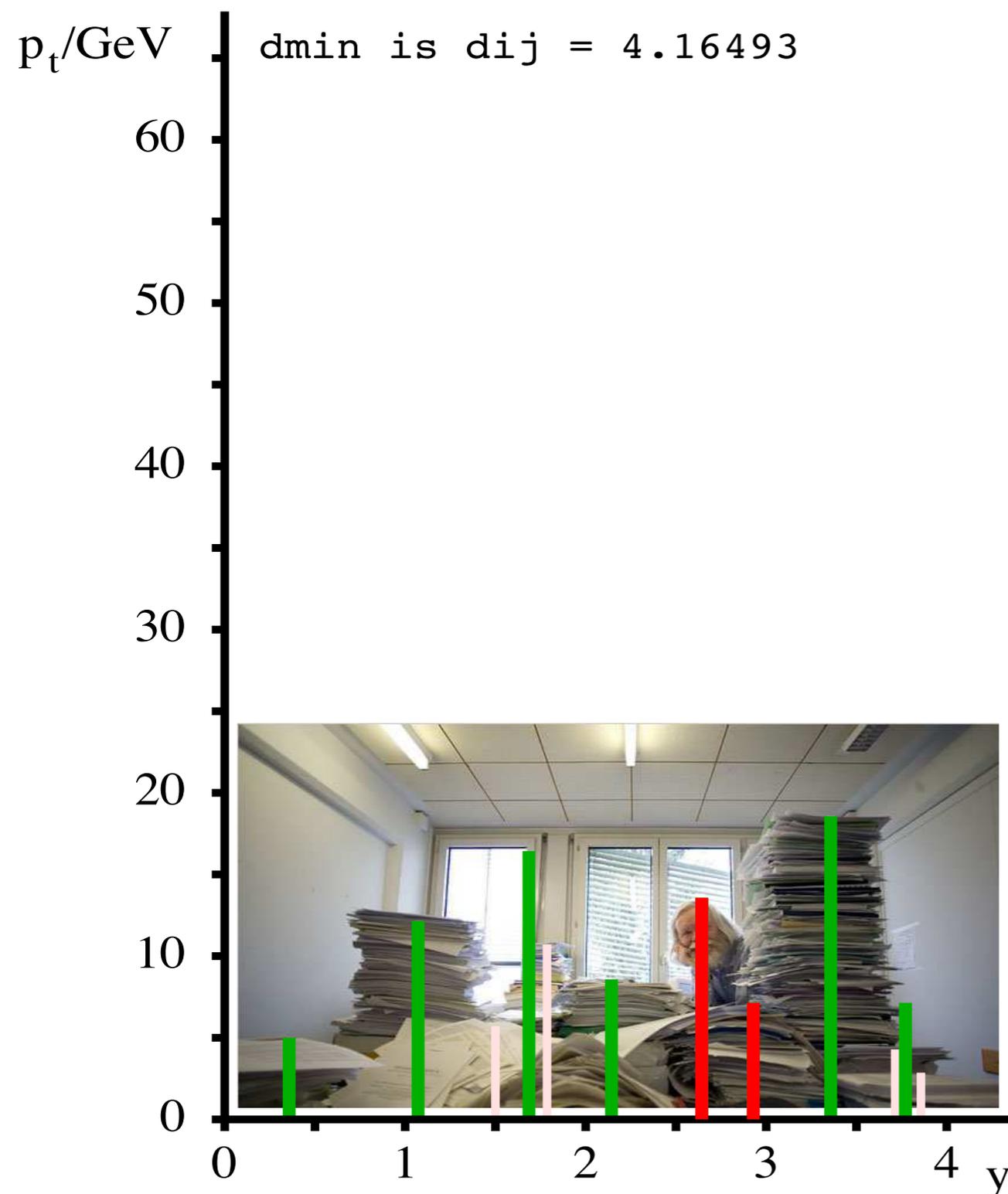
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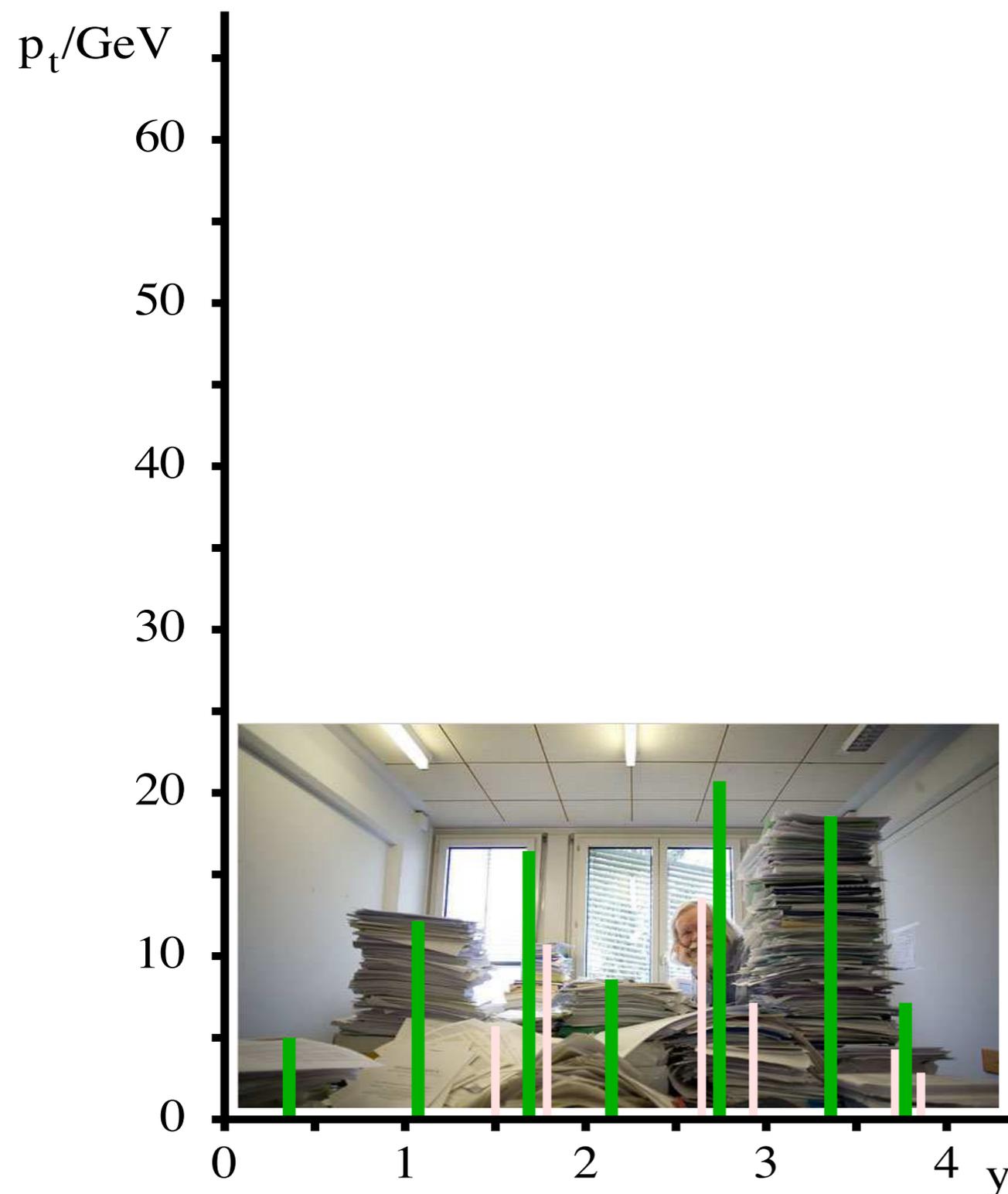
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$\phi$  assumed 0 for all towers



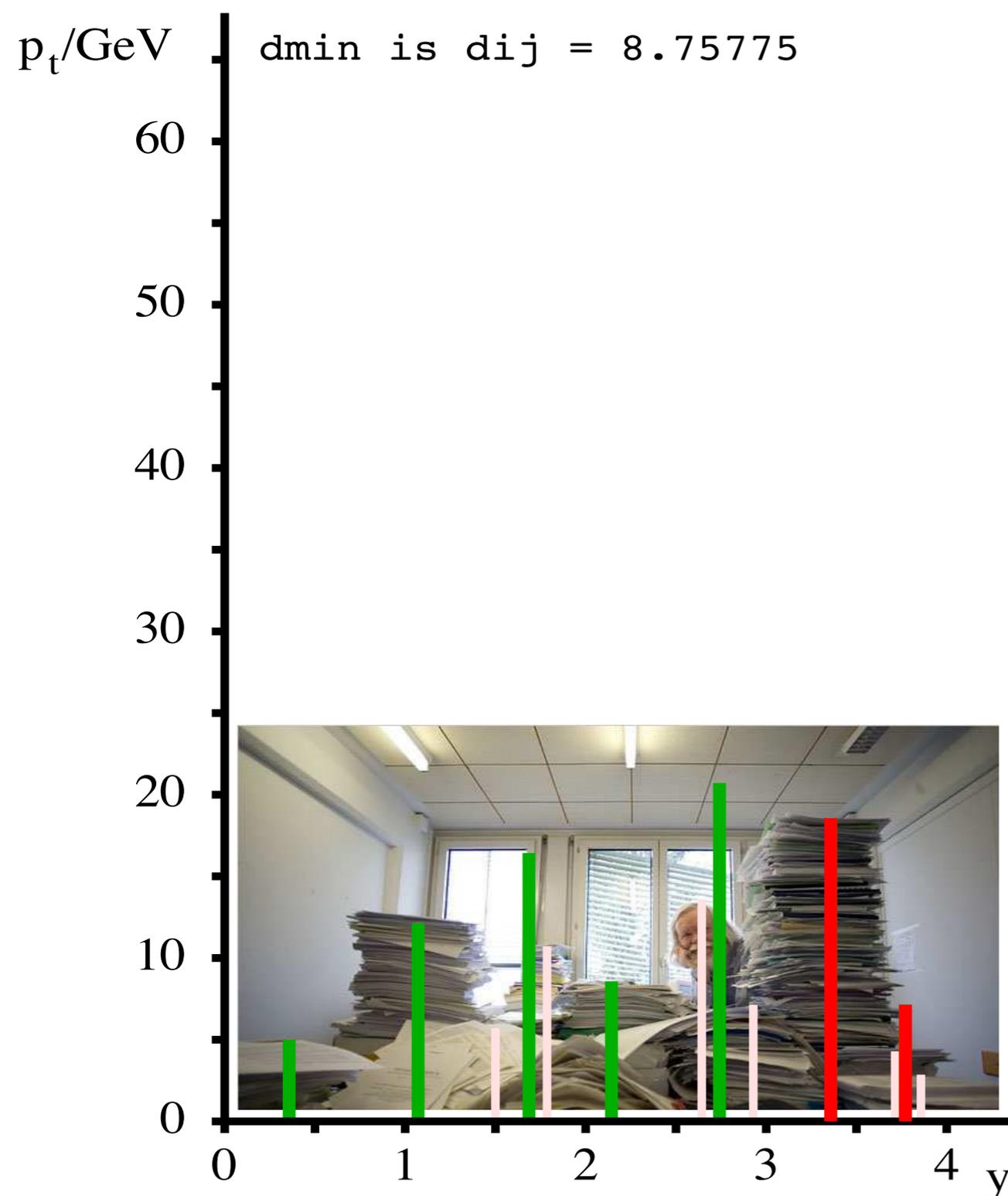
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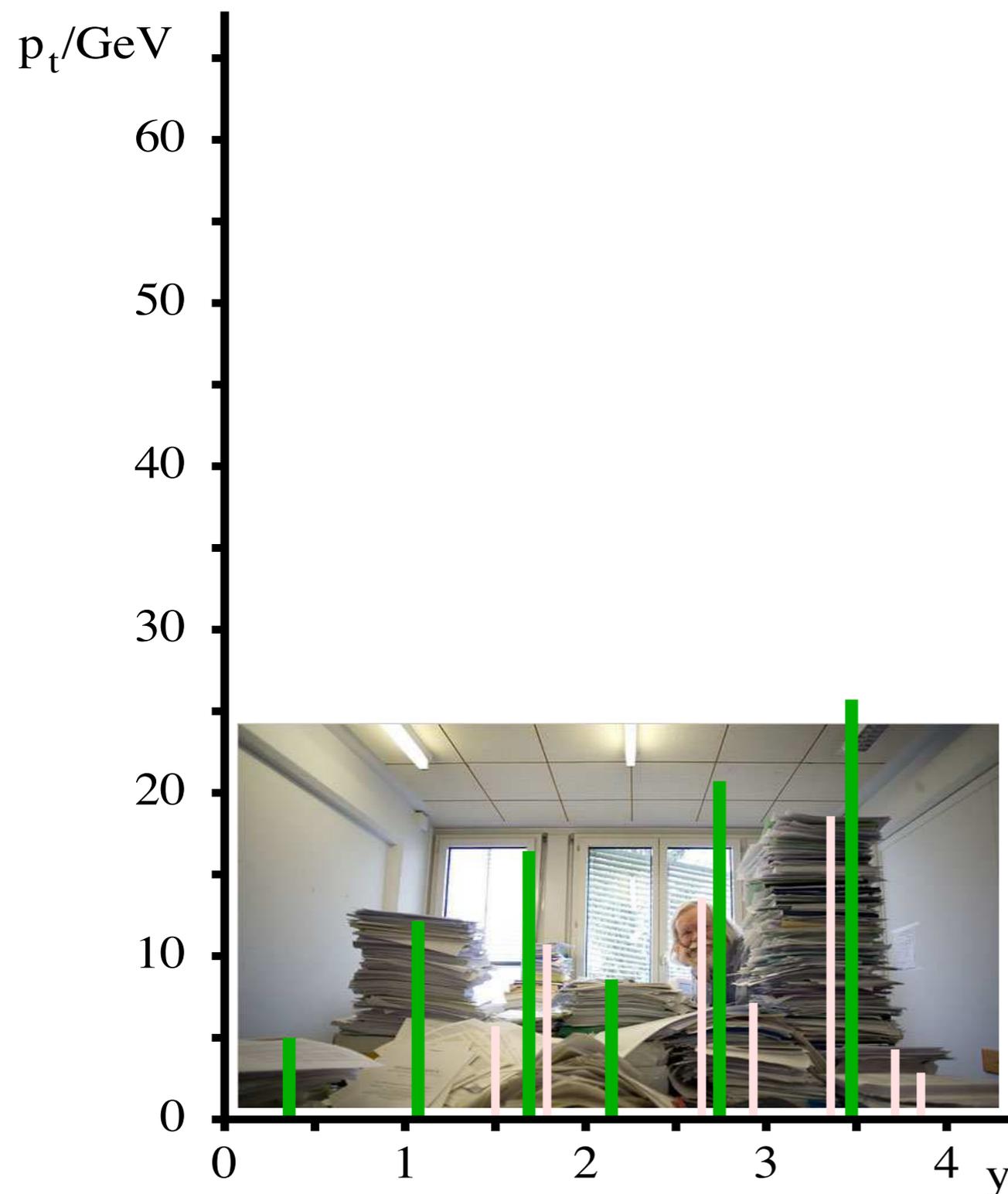
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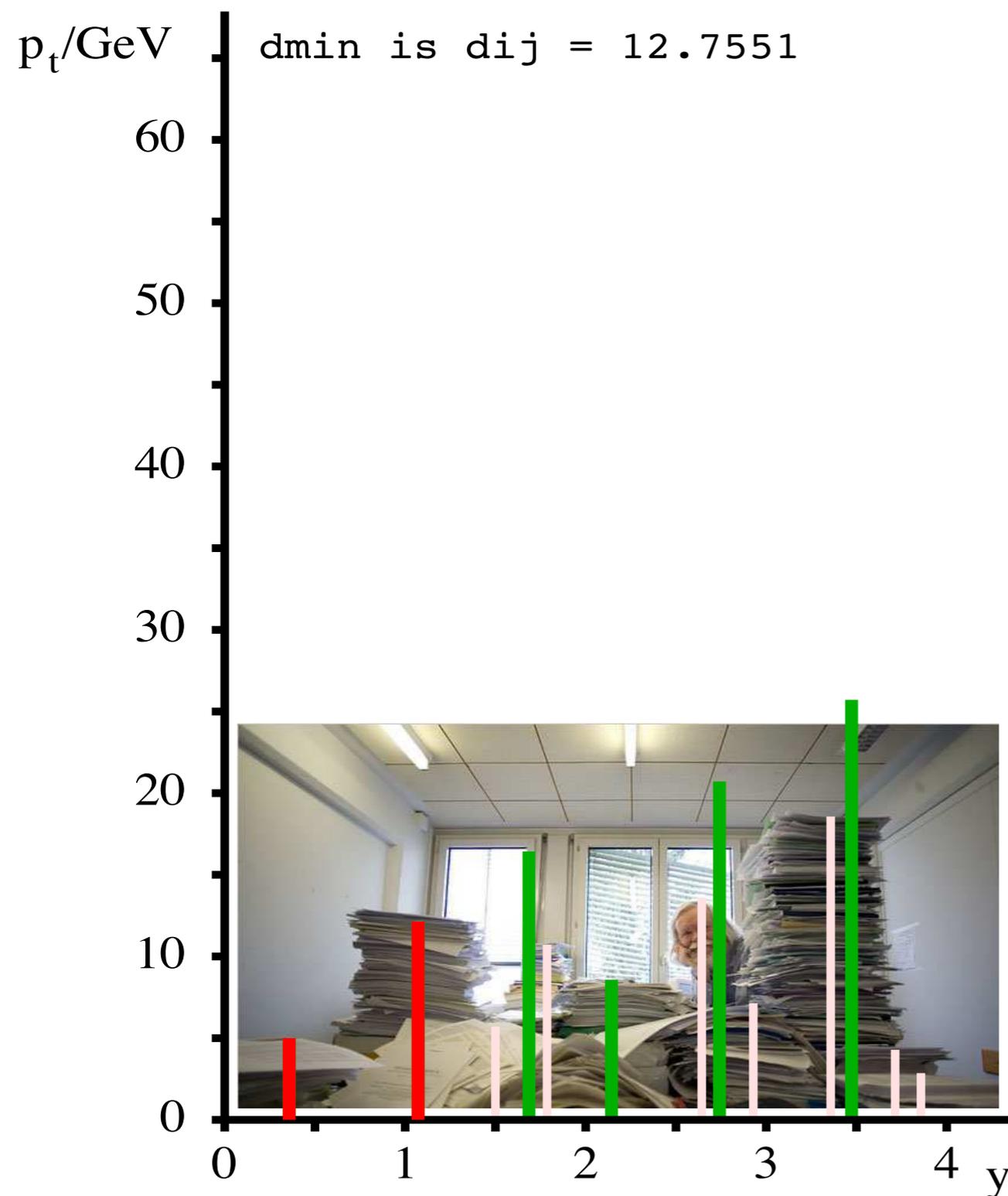
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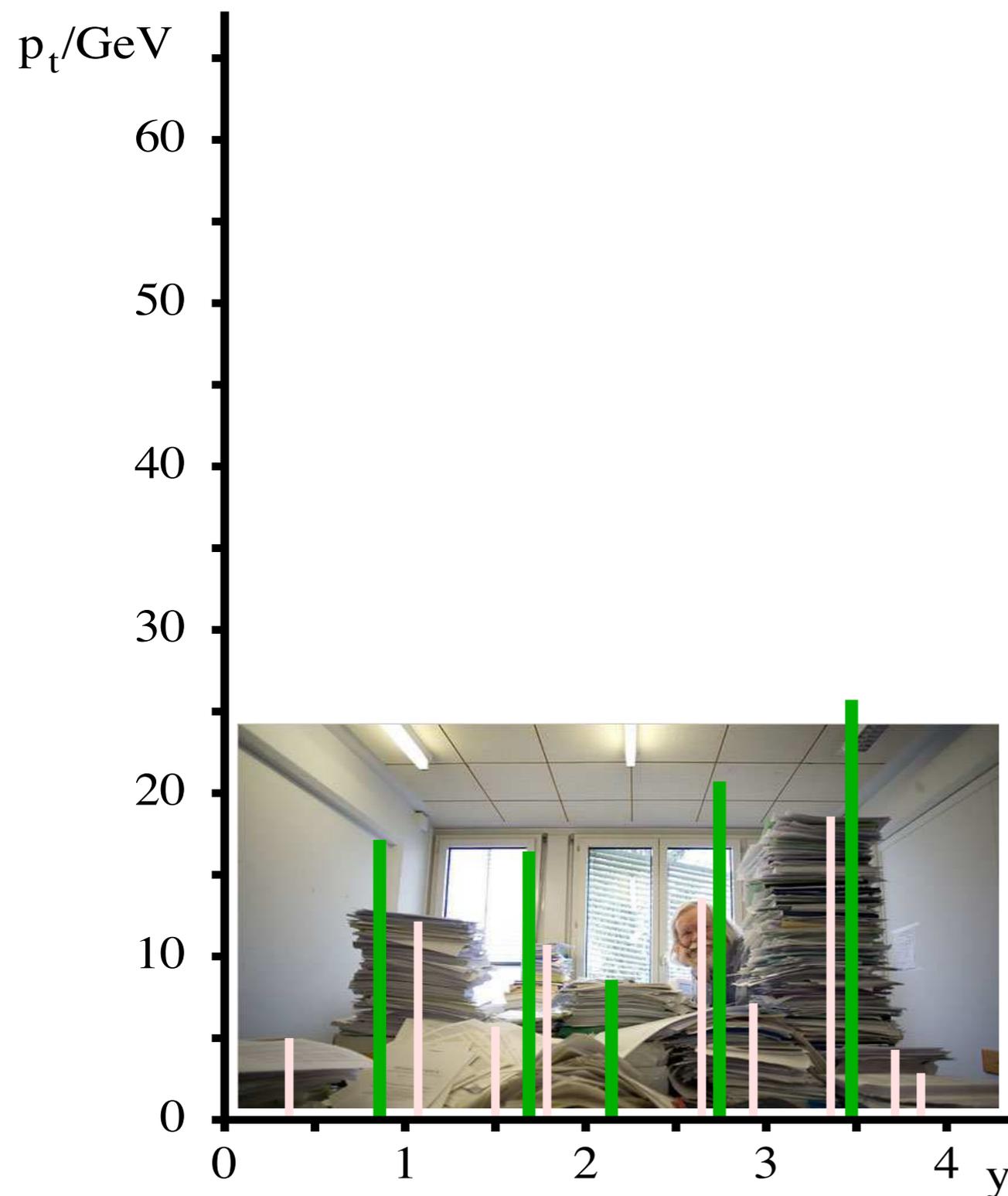
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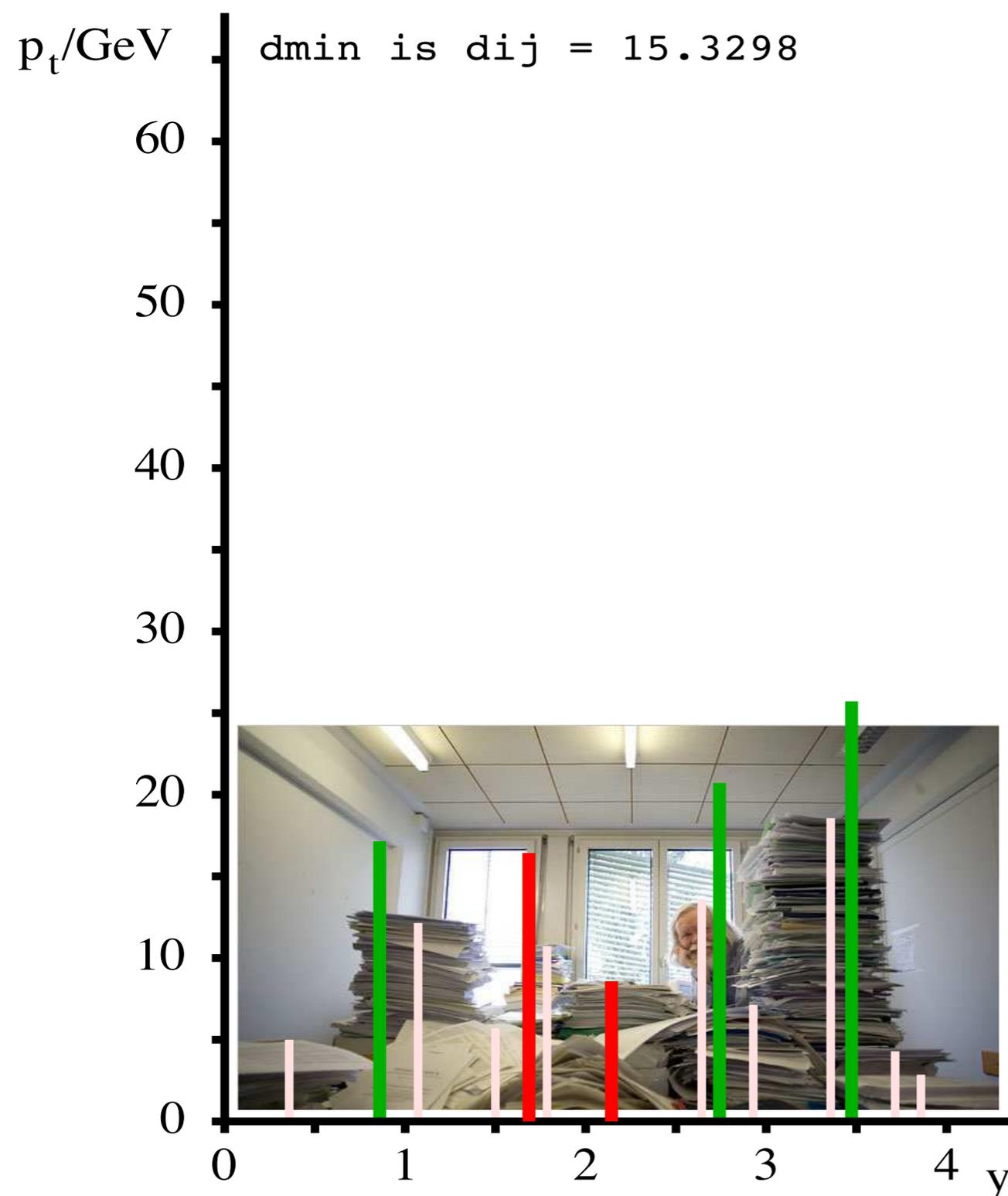
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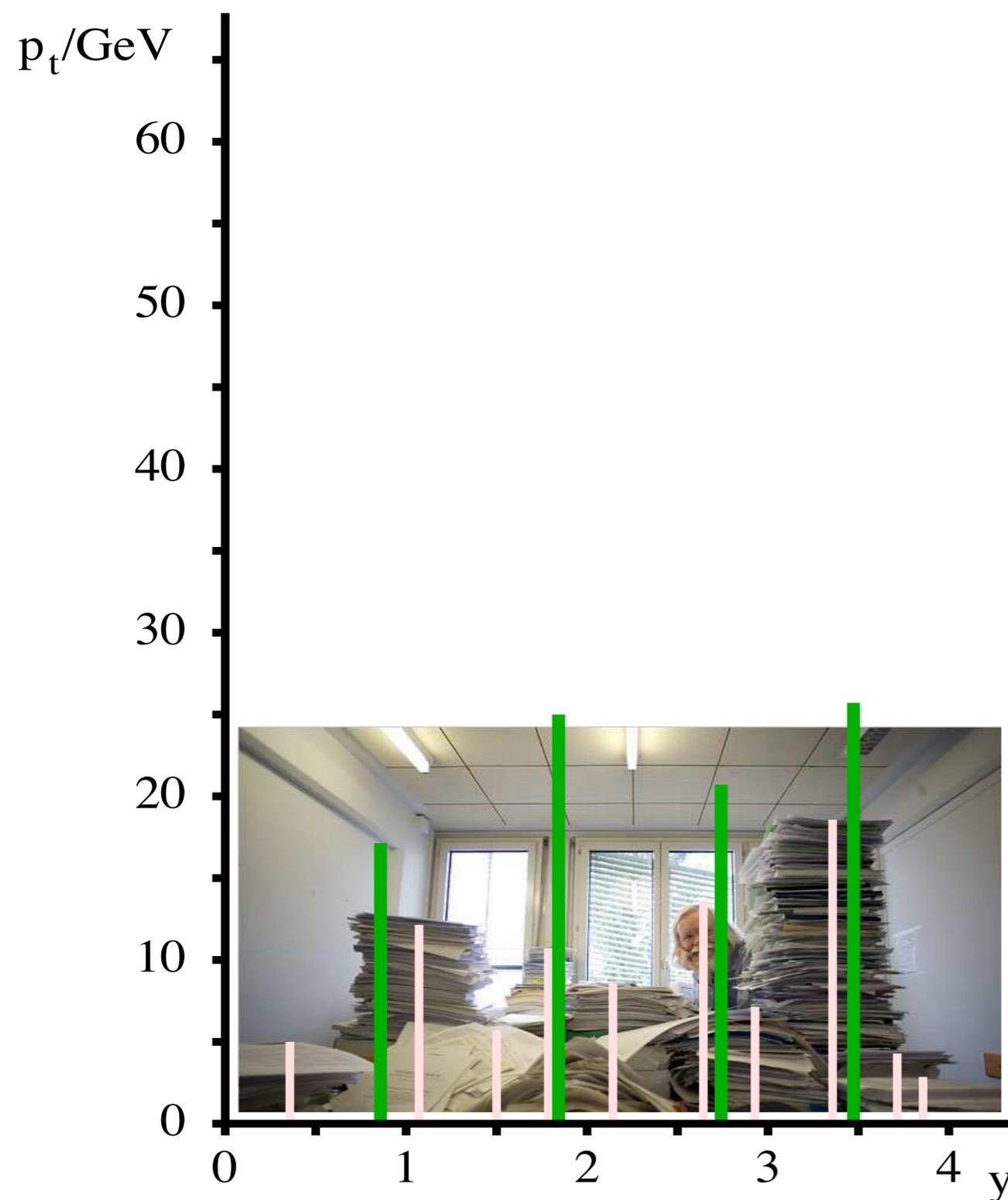
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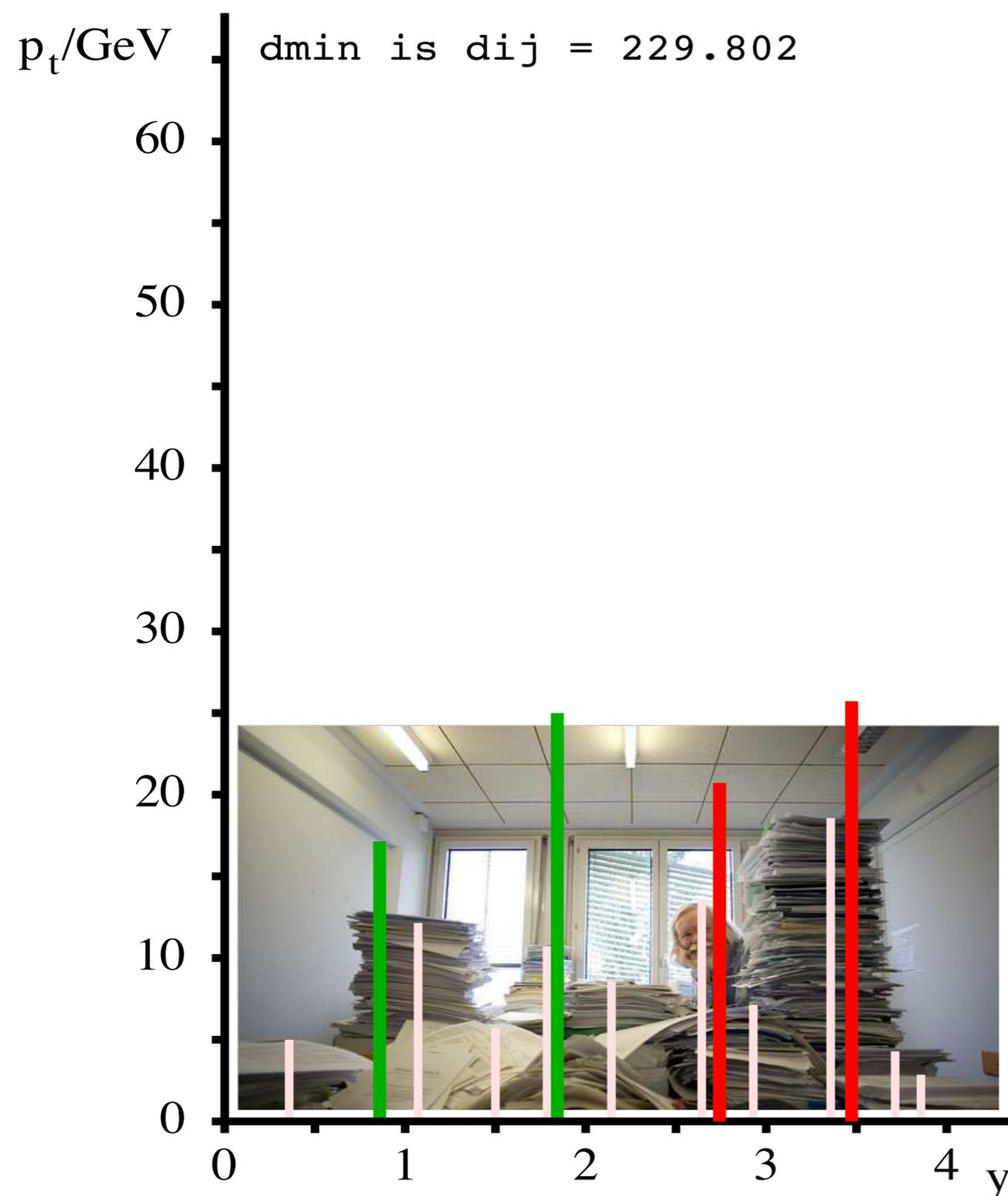
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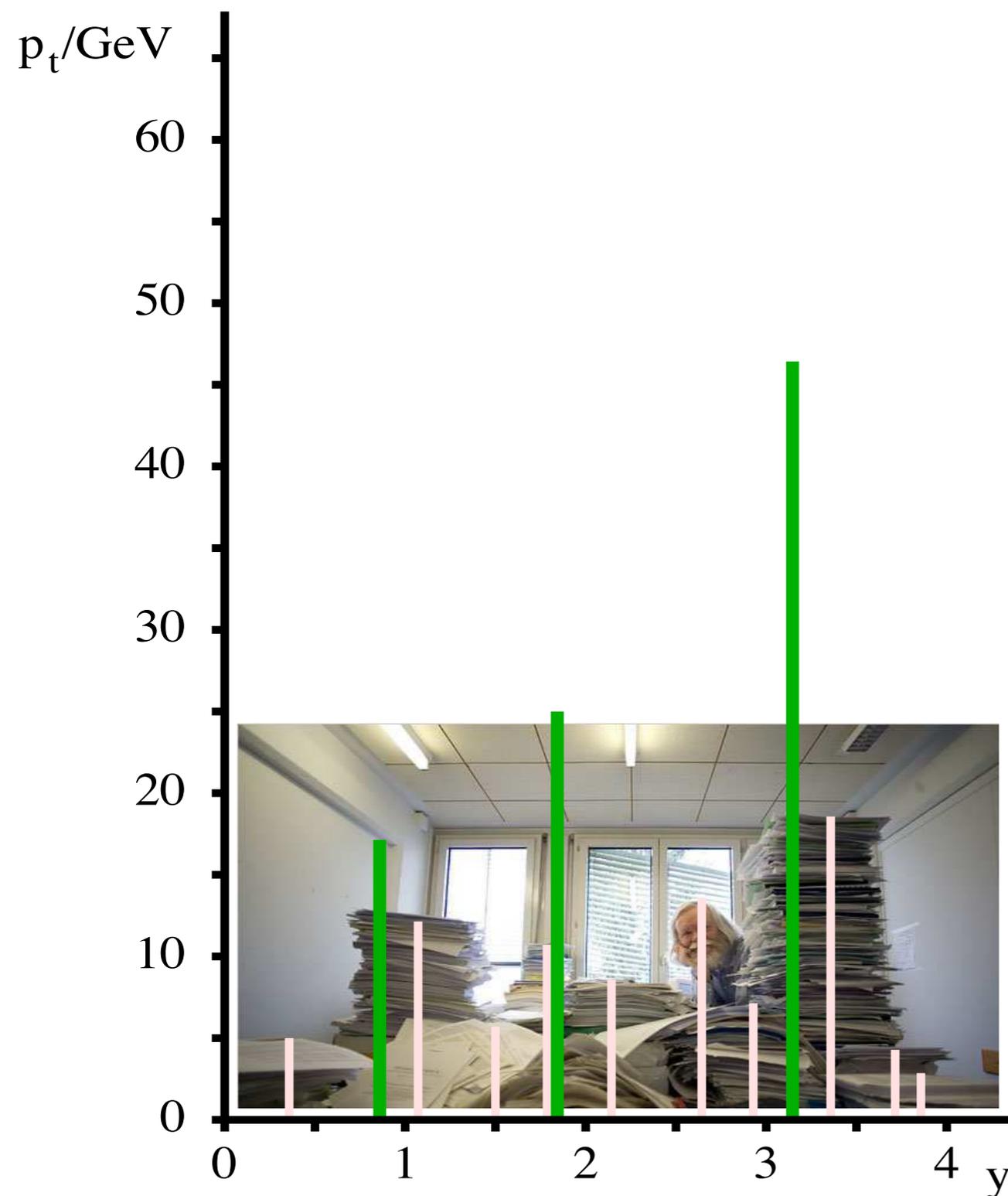
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$\phi$  assumed 0 for all towers



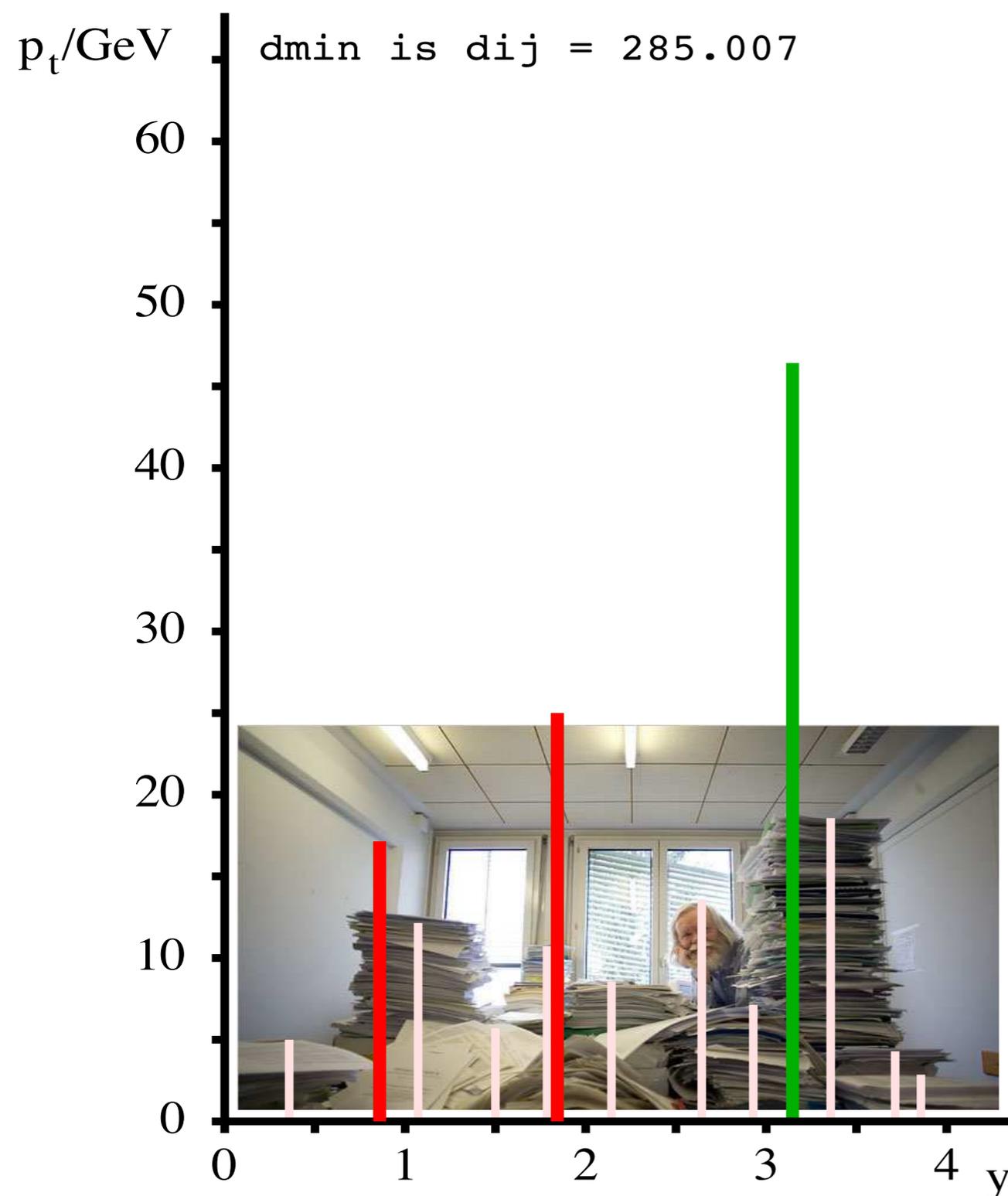
**$k_t$  alg.:** Find smallest of

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

- ▶ If  $d_{ij}$  recombine
- ▶ if  $d_{iB}$ ,  $i$  is a jet

Example clustering with  $k_t$  algorithm,  $R = 1.0$

$\phi$  assumed 0 for all towers



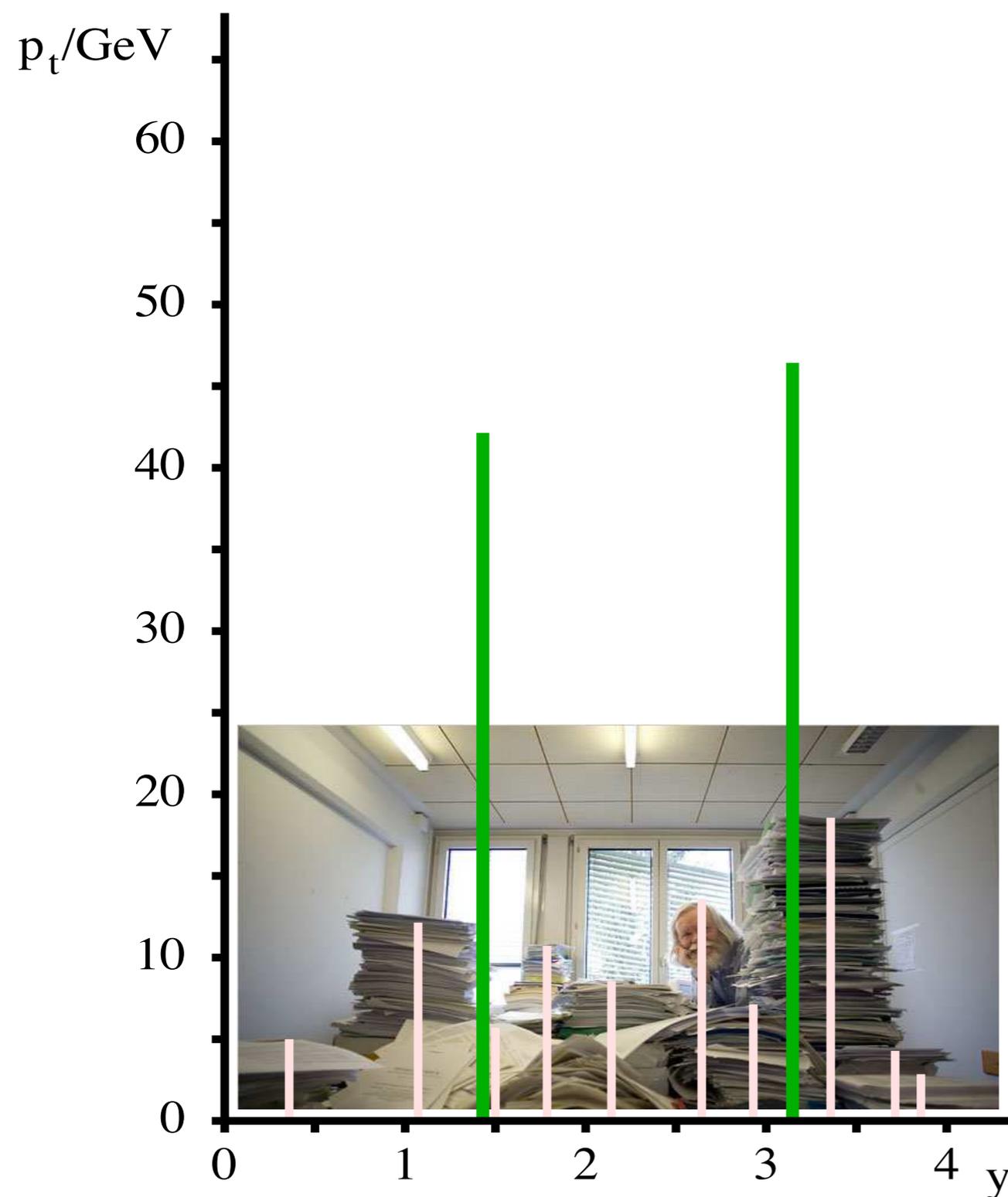
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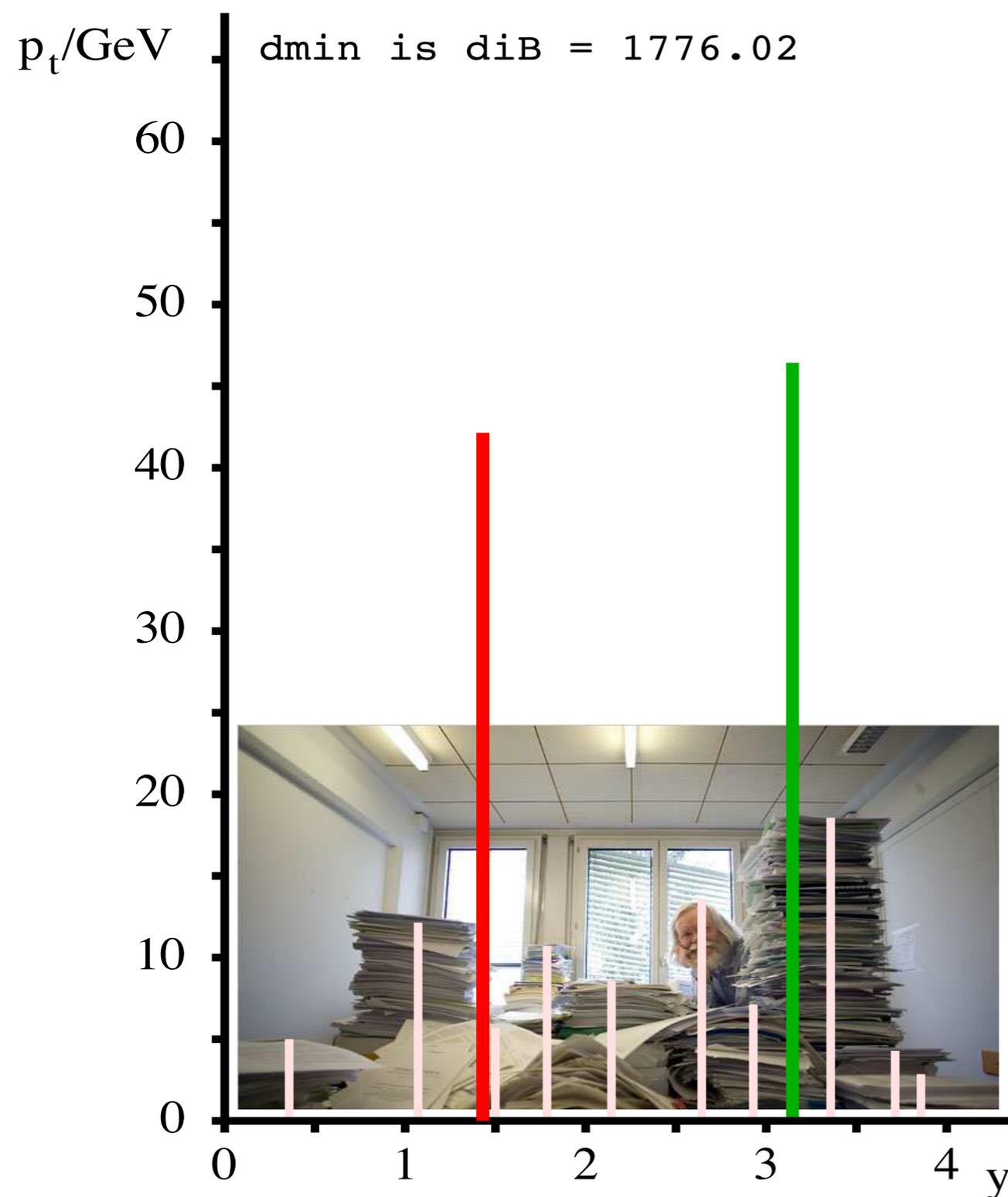
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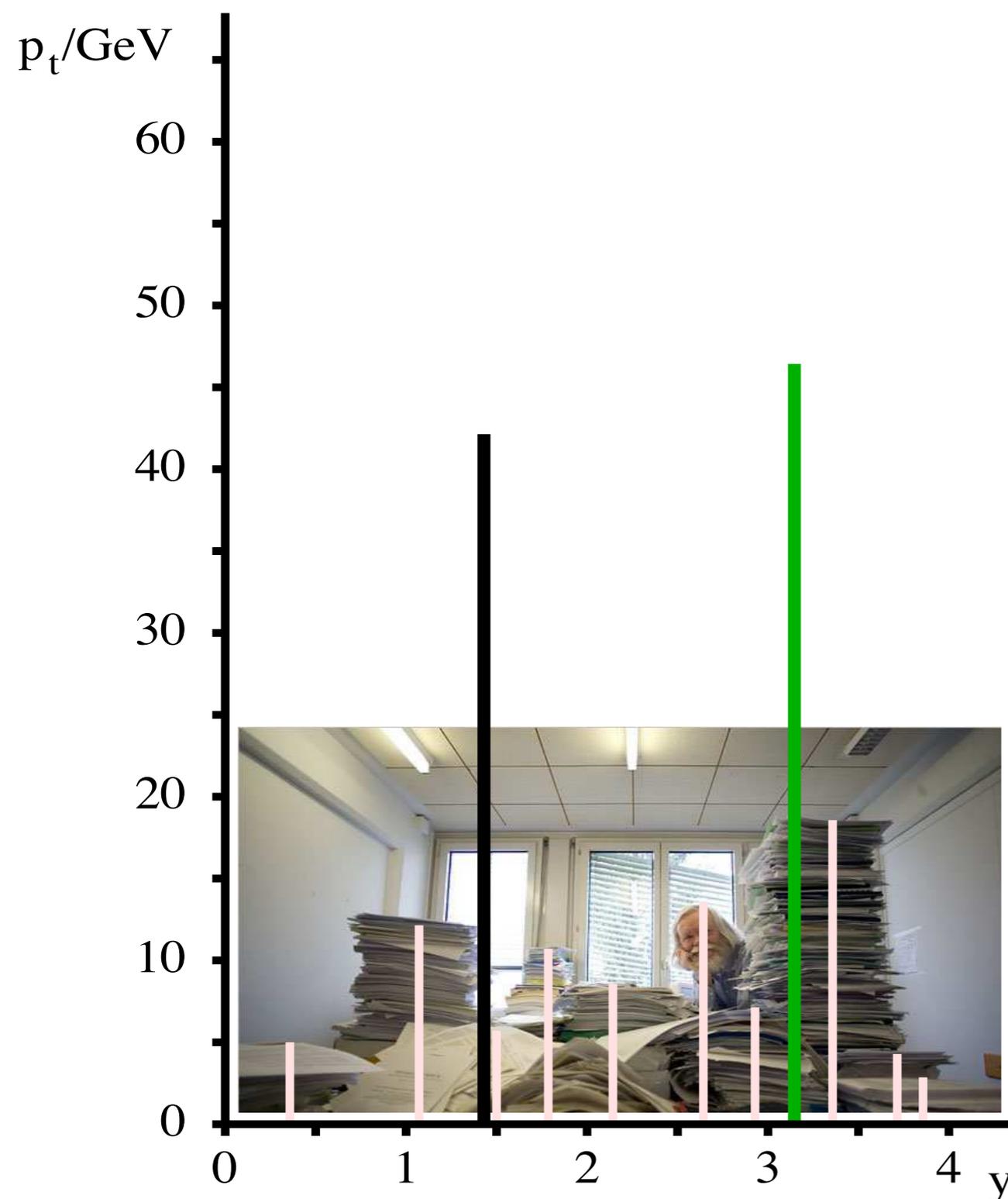
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Example clustering with  $k_t$  algorithm,  $R = 1.0$

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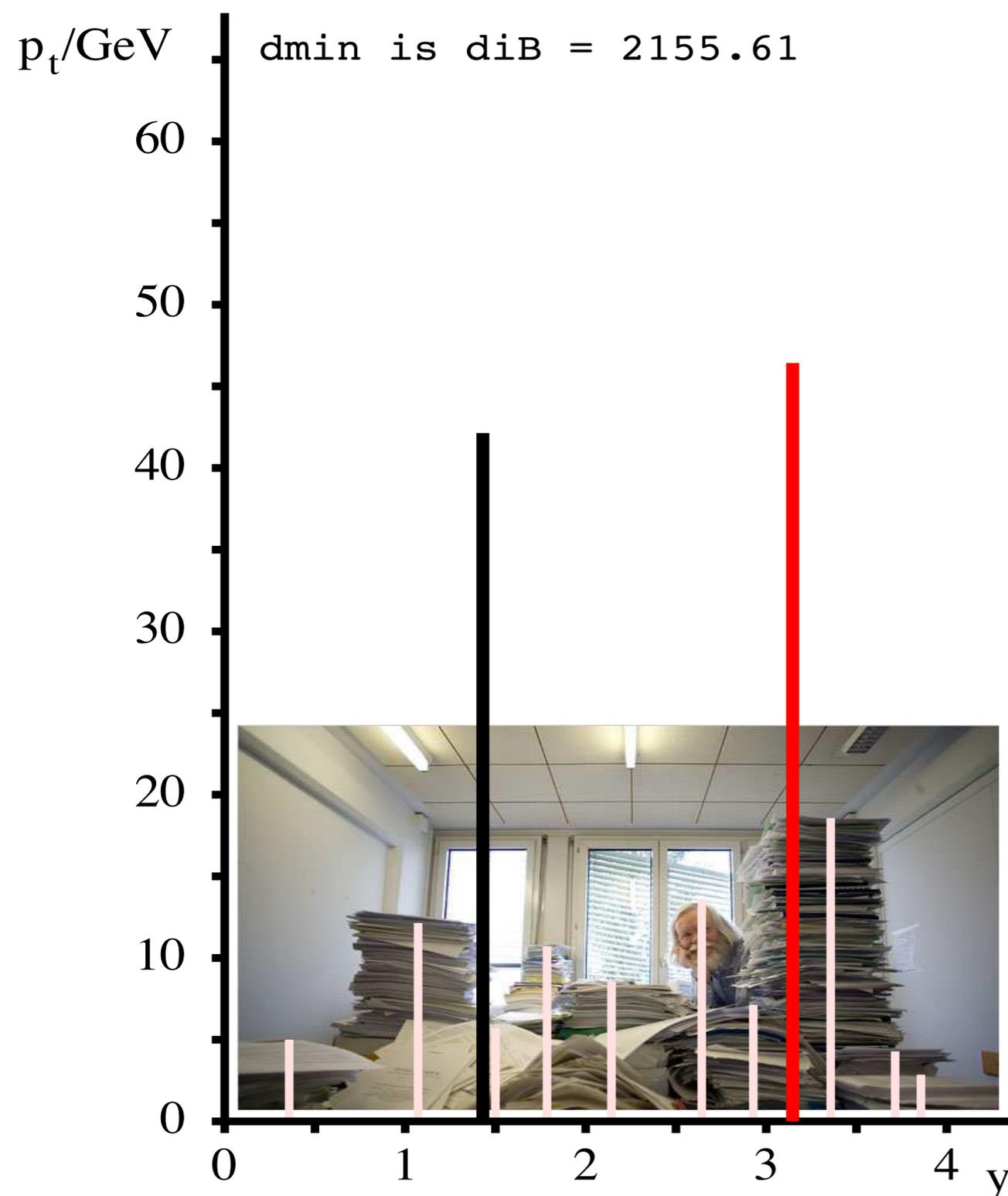
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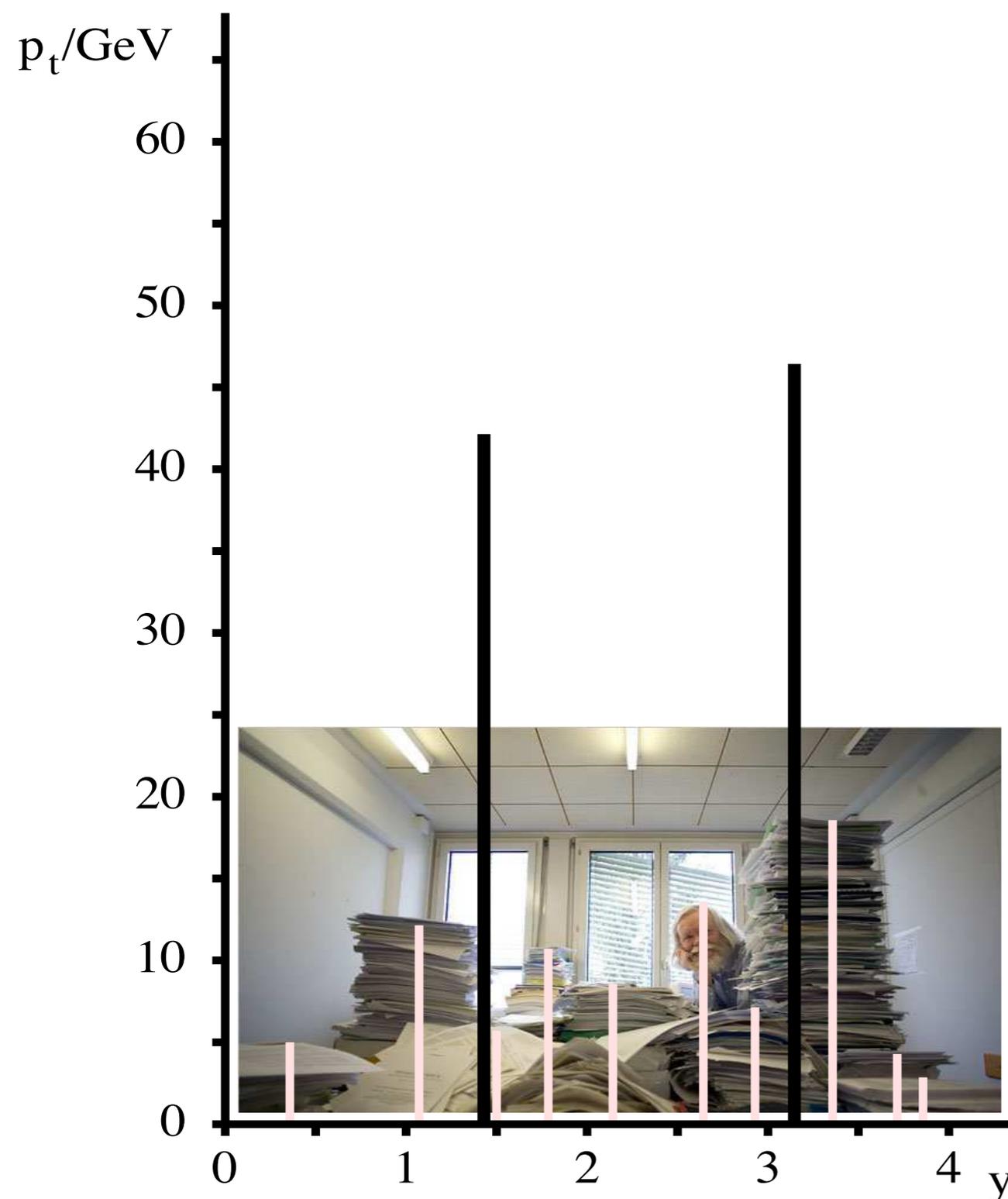
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Example clustering with  $k_t$  algorithm,  $R = 1.0$

$\phi$  assumed 0 for all towers

# Sequential recombination variants

## Cambridge/Aachen: the simplest of hadron-collider algorithms

- Recombine pair of objects closest in  $\Delta R_{ij}$
- Repeat until all  $\Delta R_{ij} > R$  — remaining objects are jets

Dokshitzer, Leder, Moretti, Webber '97 (Cambridge): more involved  $e^+e^-$  form

Wobisch & Wengler '99 (Aachen): simple inclusive hadron-collider form

One still applies a  $p_{t,\min}$  cut to the jets, as for inclusive  $k_t$

C/A privileges the collinear divergence of QCD;  
it 'ignores' the soft one

Anti- $k_t$ : formulated similarly to inclusive  $k_t$ , but with

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

Cacciari, GPS & Soyez '08 [+Delsart unpublished]

Anti- $k_t$  privileges the collinear divergence of QCD and disfavours clustering between pairs of soft particles

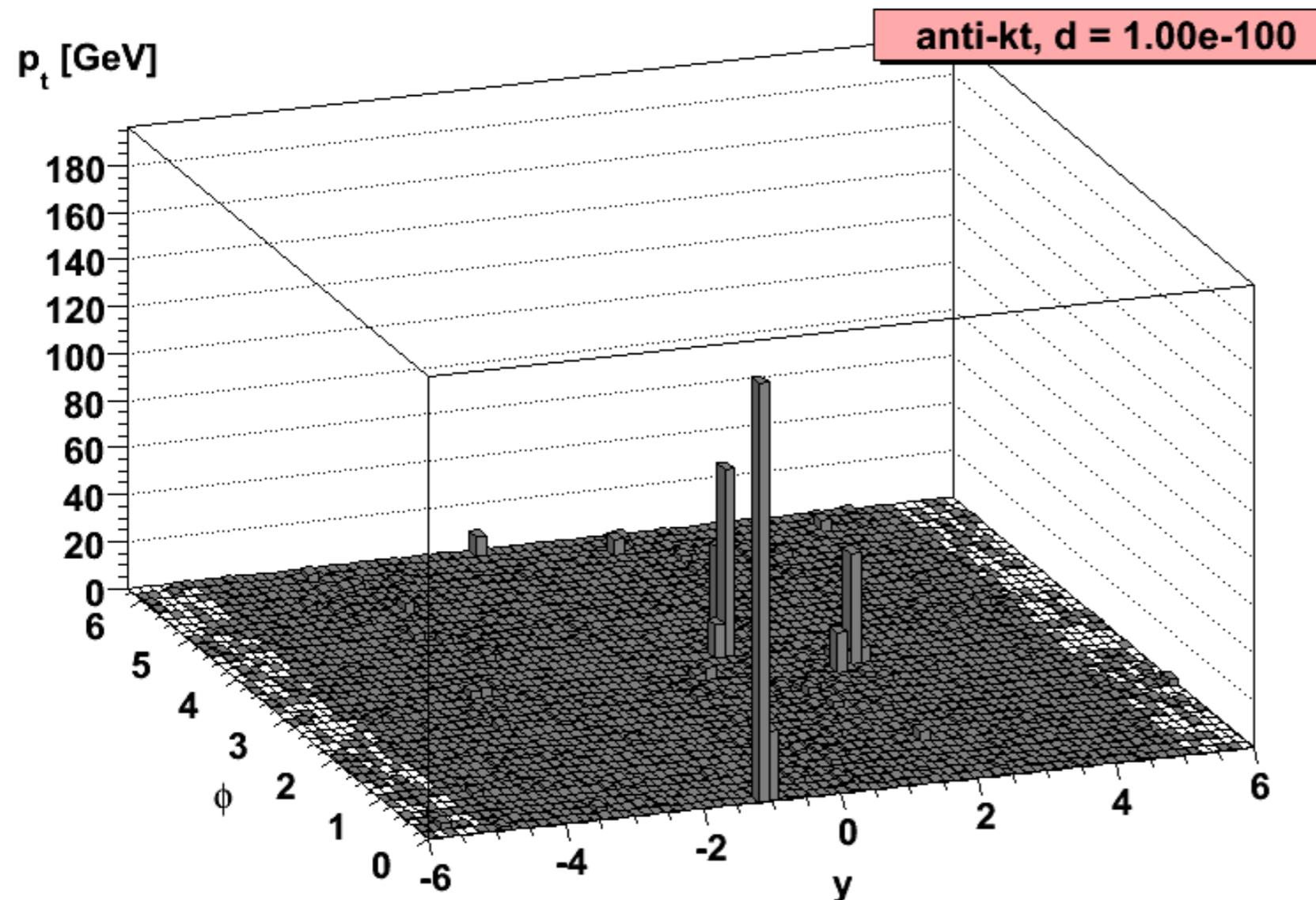
Most pairwise clusterings involve at least one hard particle

Clustering grows  
around hard cores

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

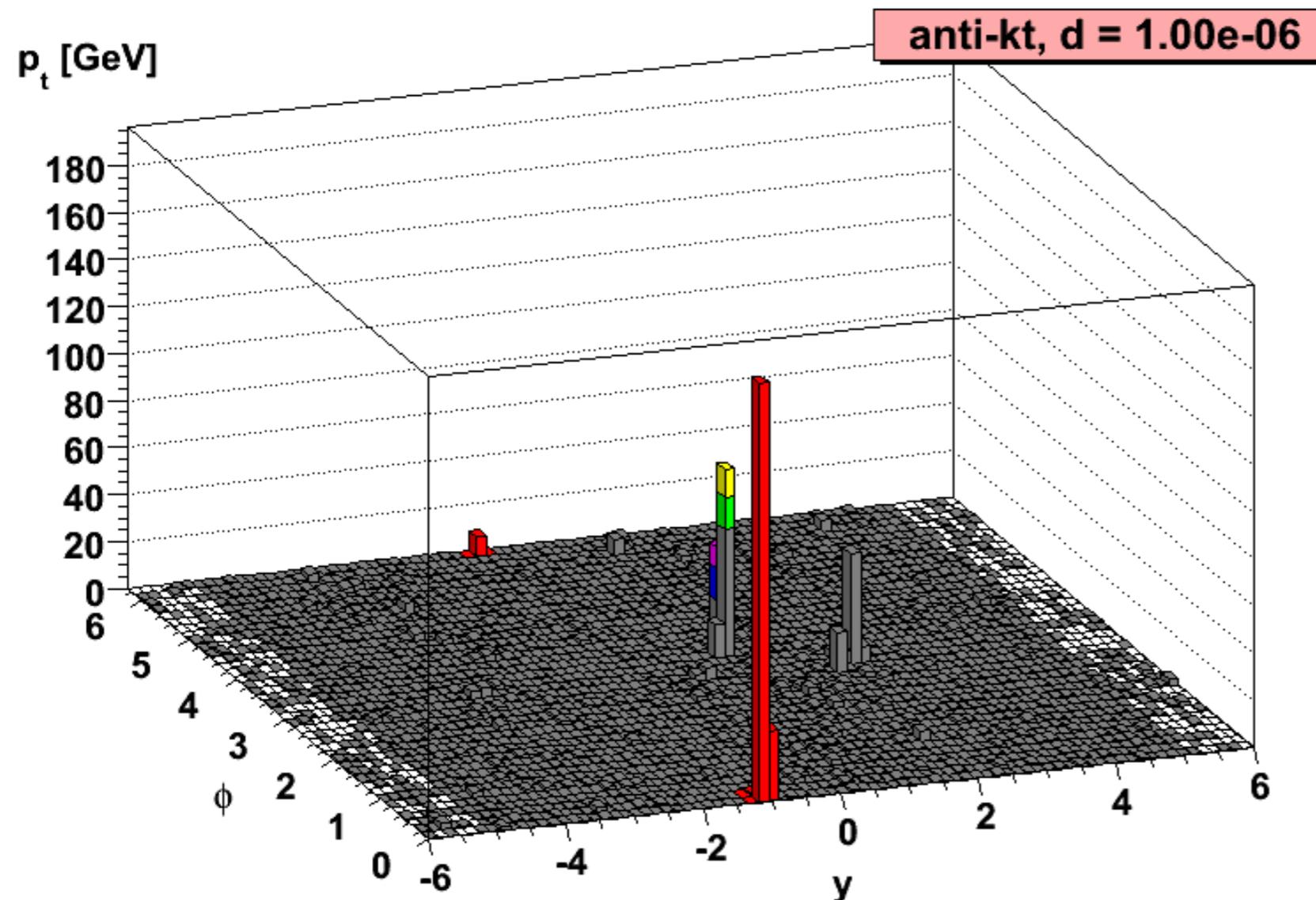
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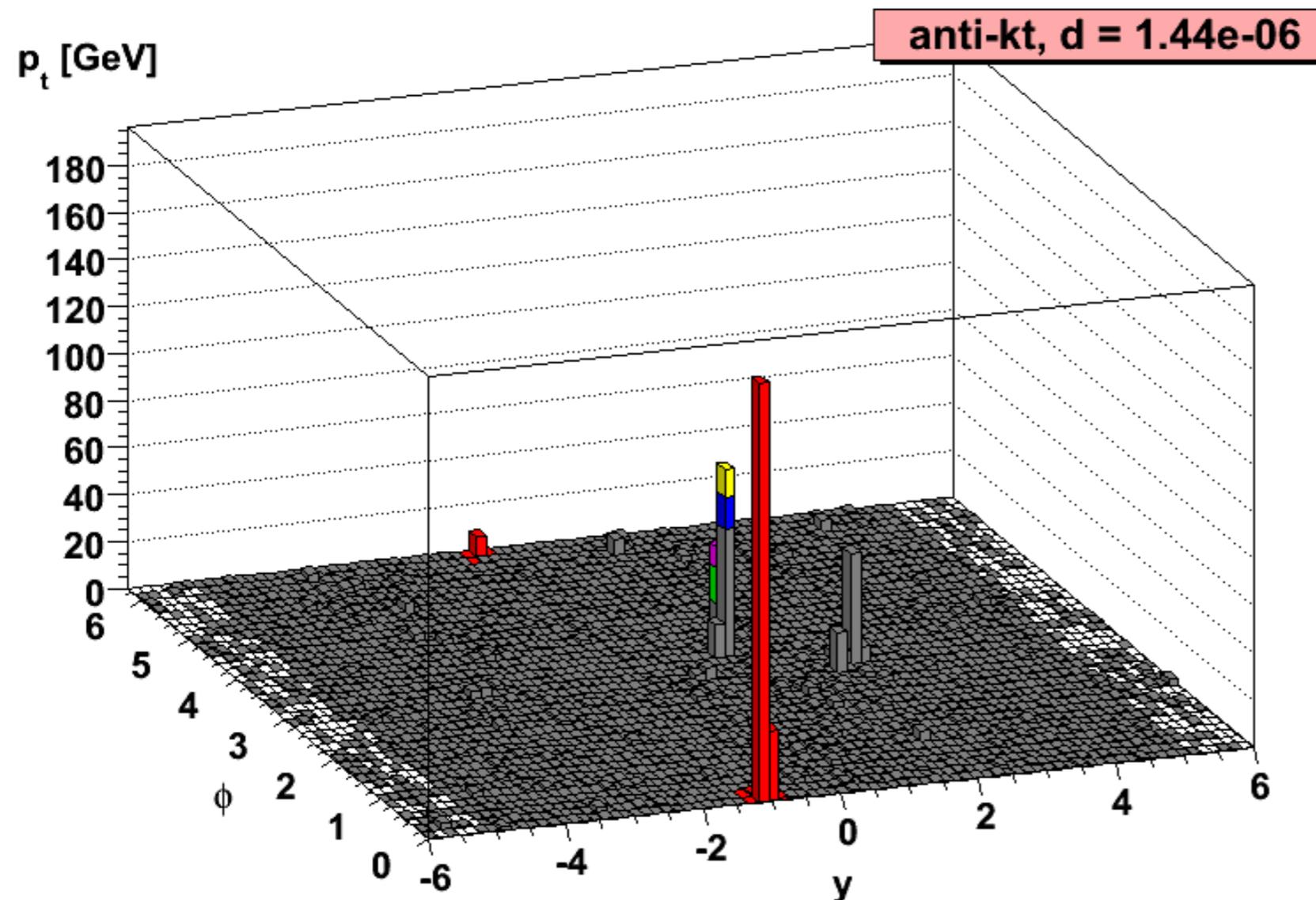
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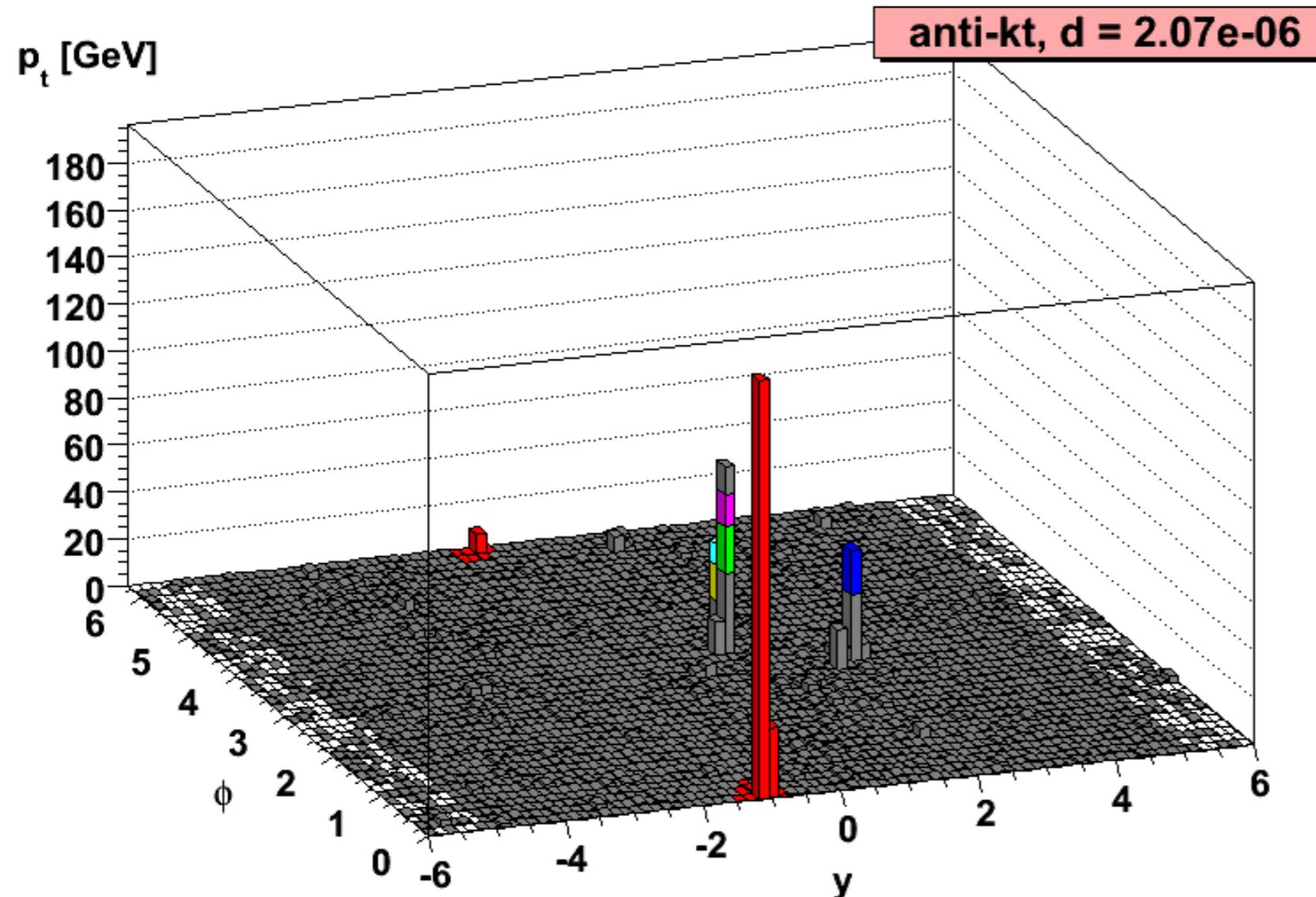
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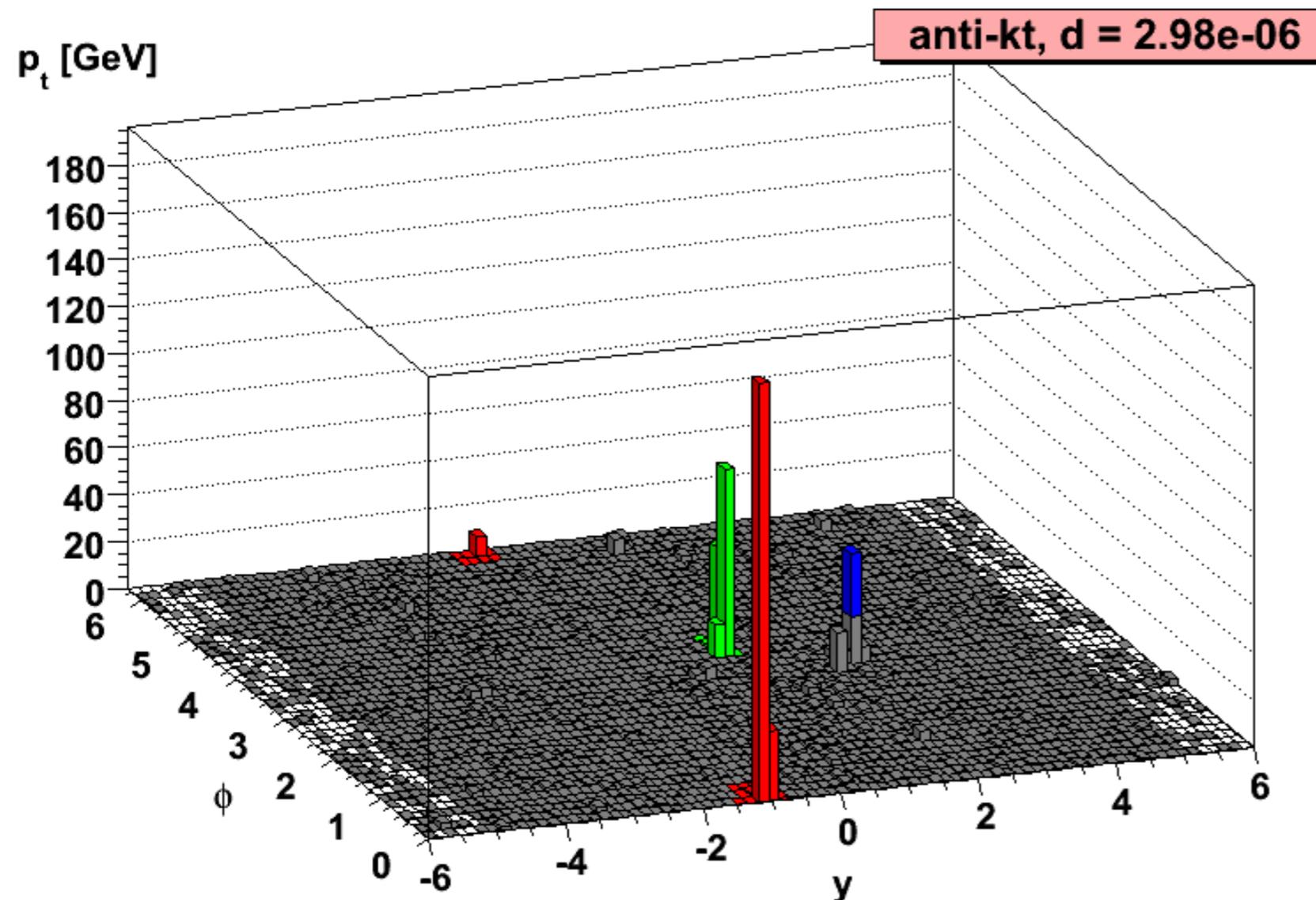
Clustering grows around hard cores

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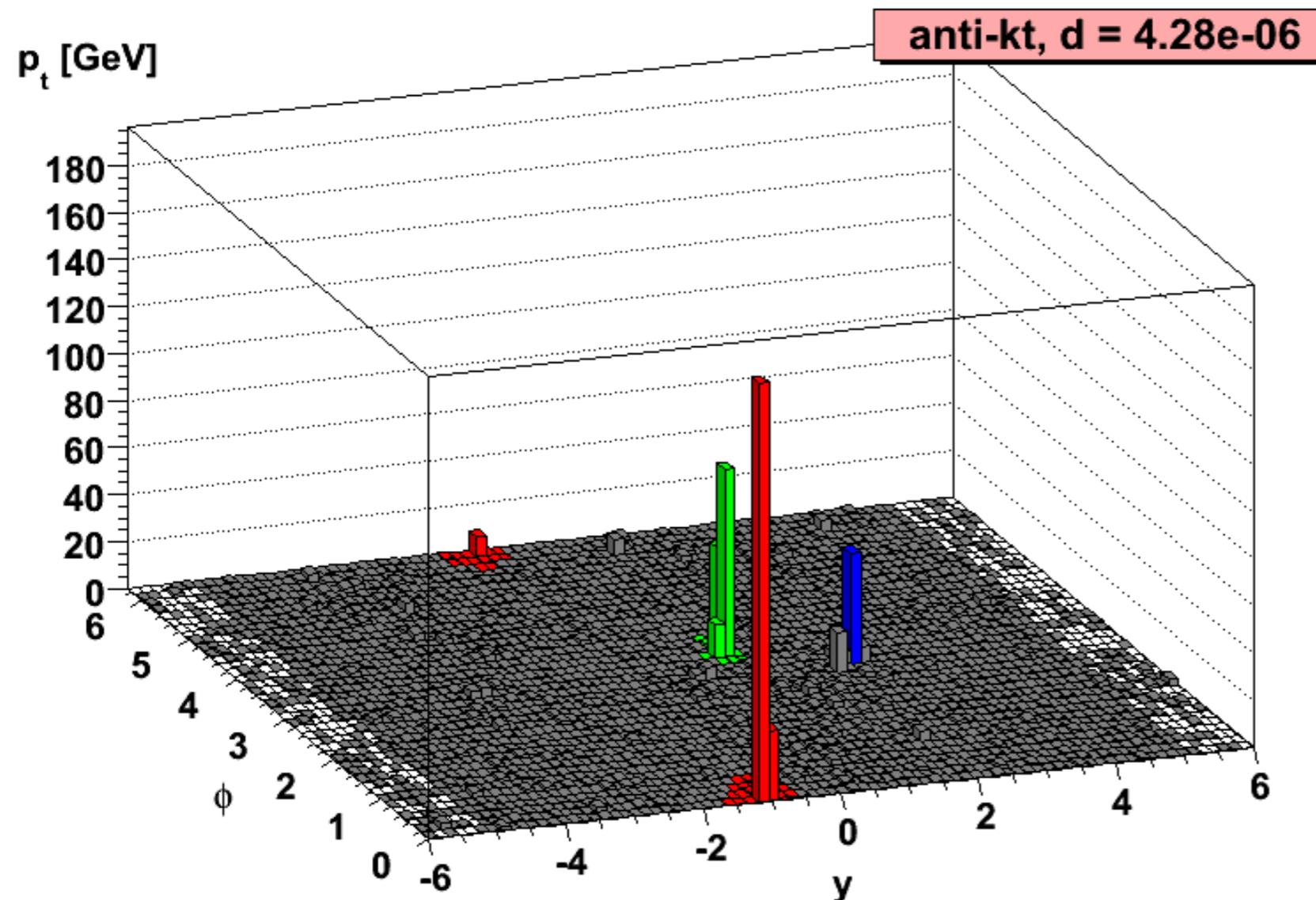
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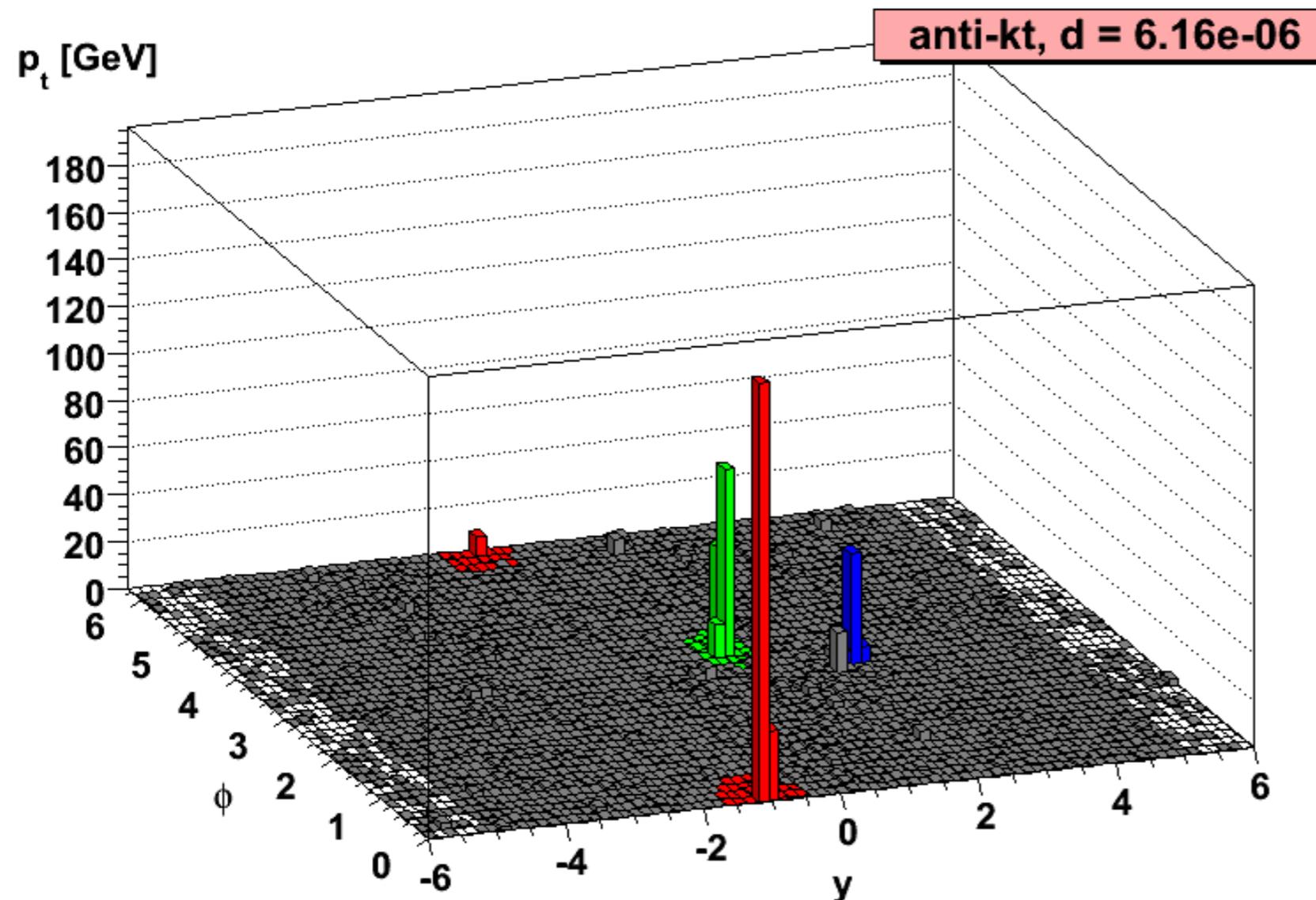
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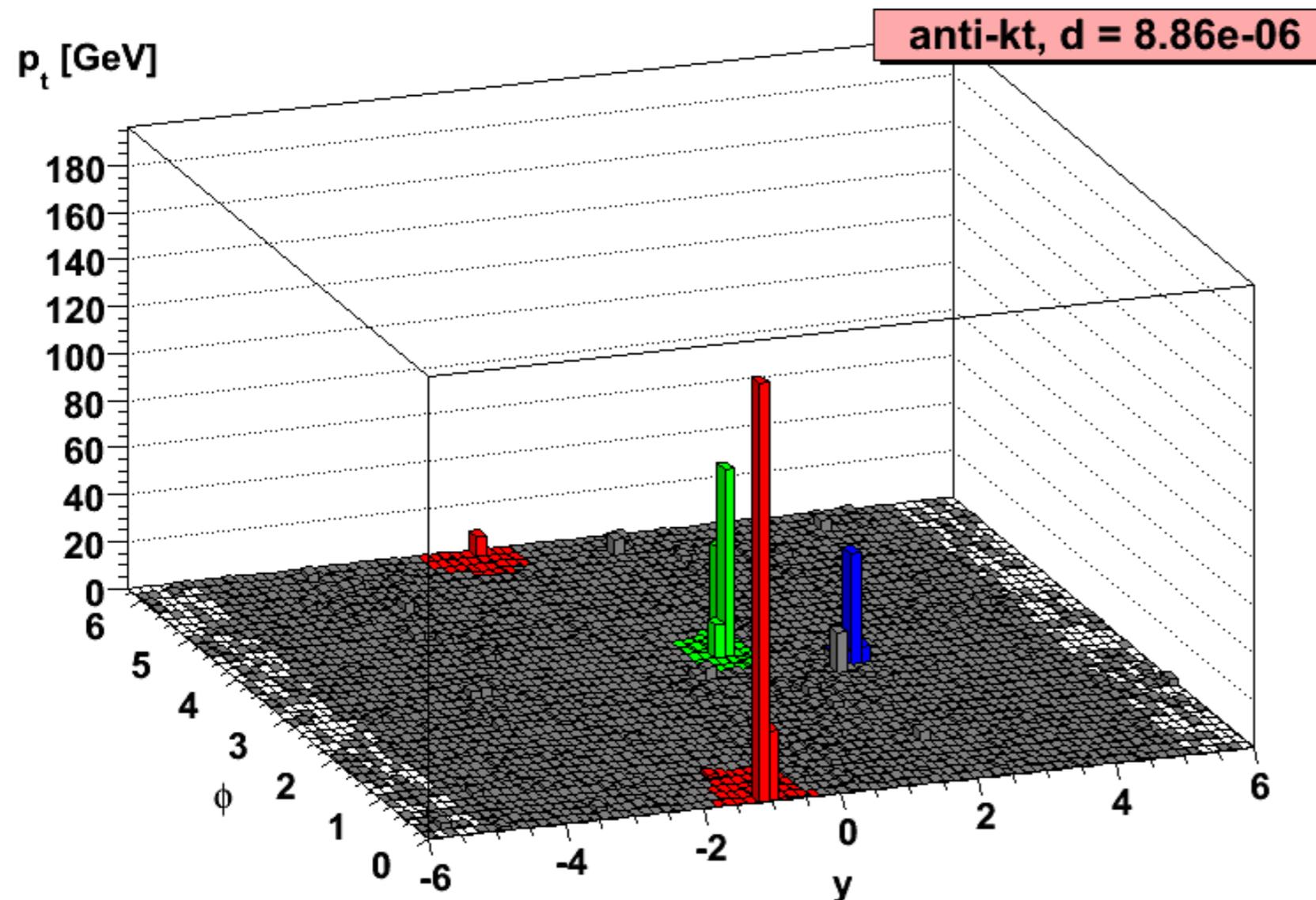
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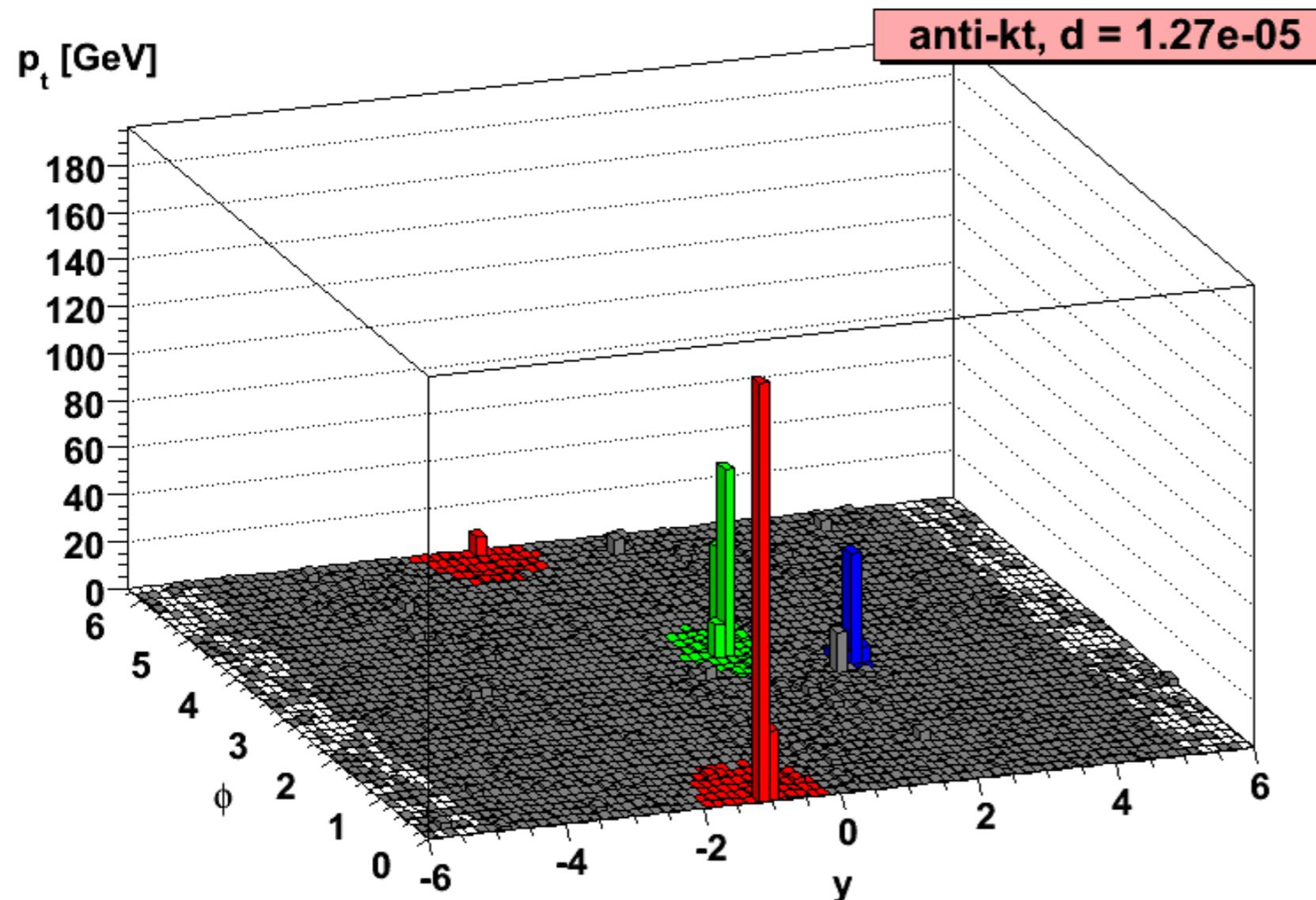
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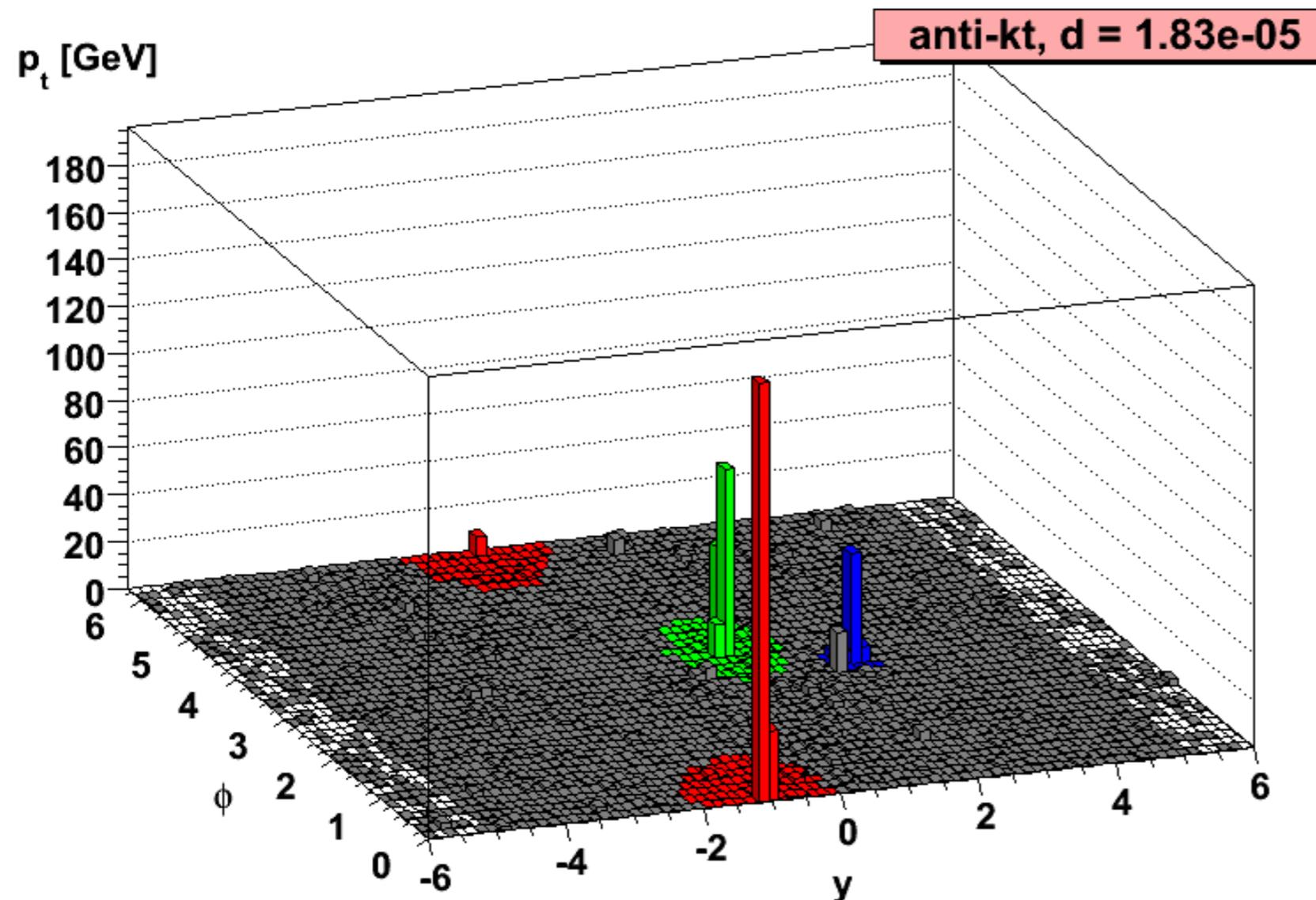
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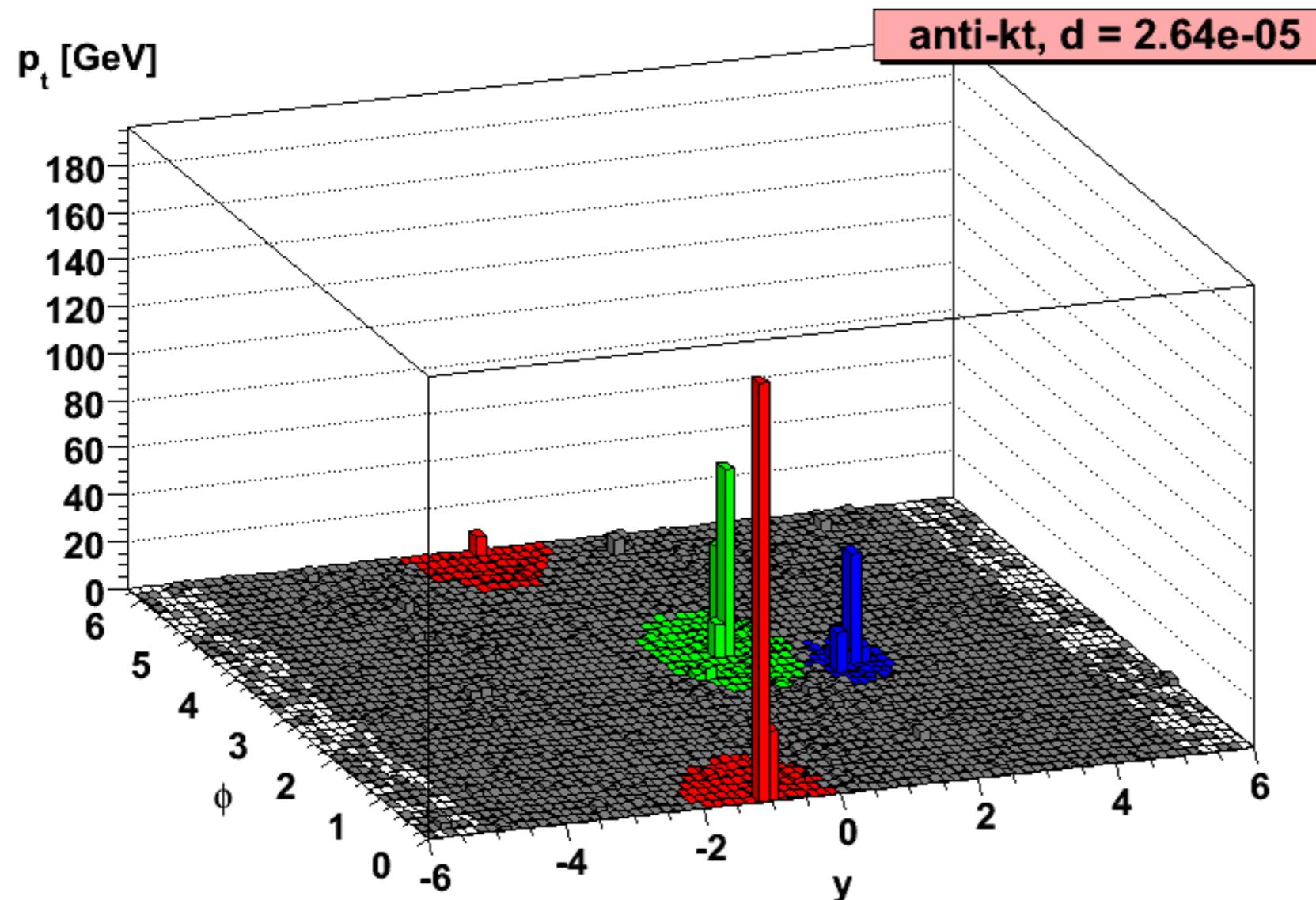
Clustering grows around hard cores

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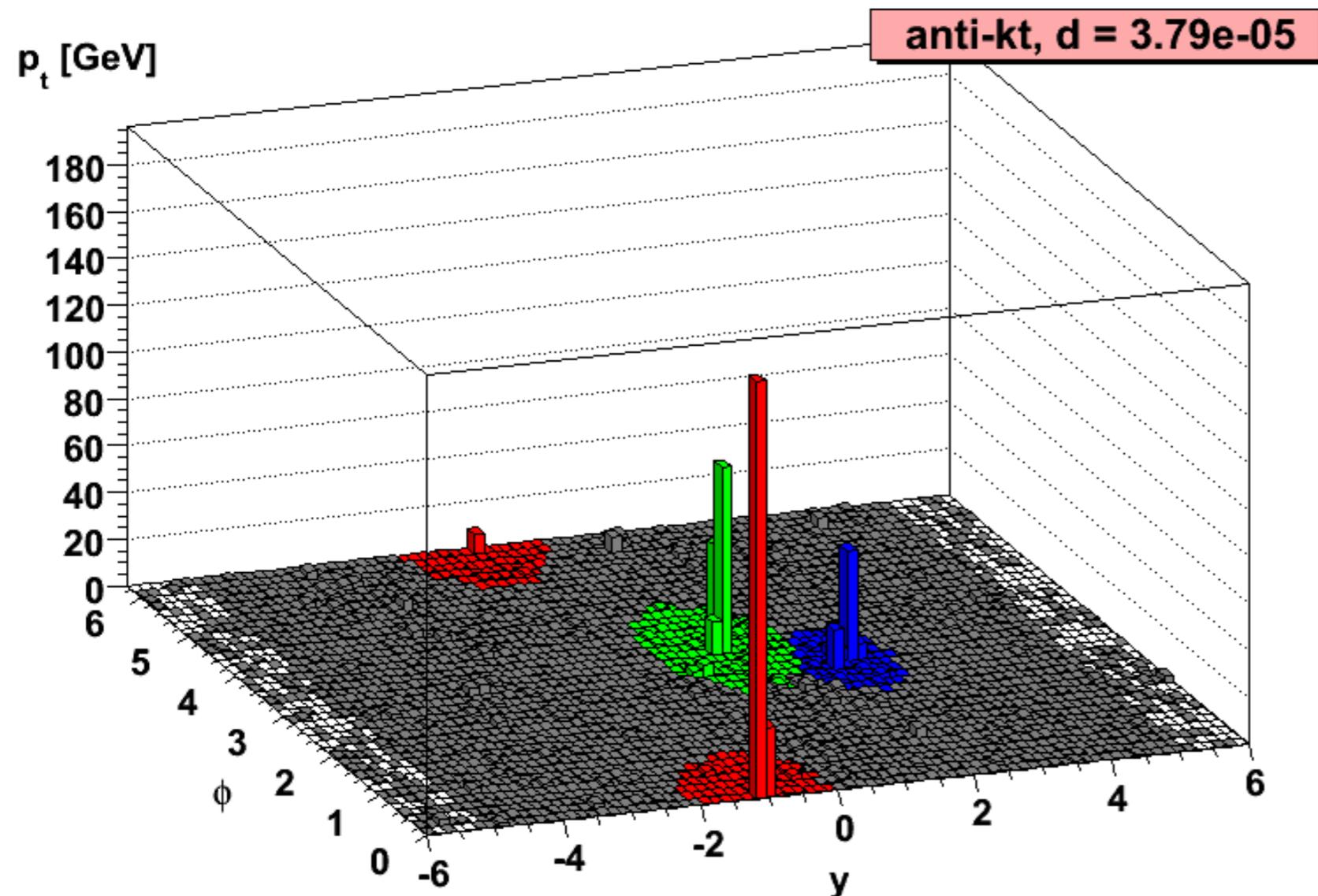
Clustering grows around hard cores

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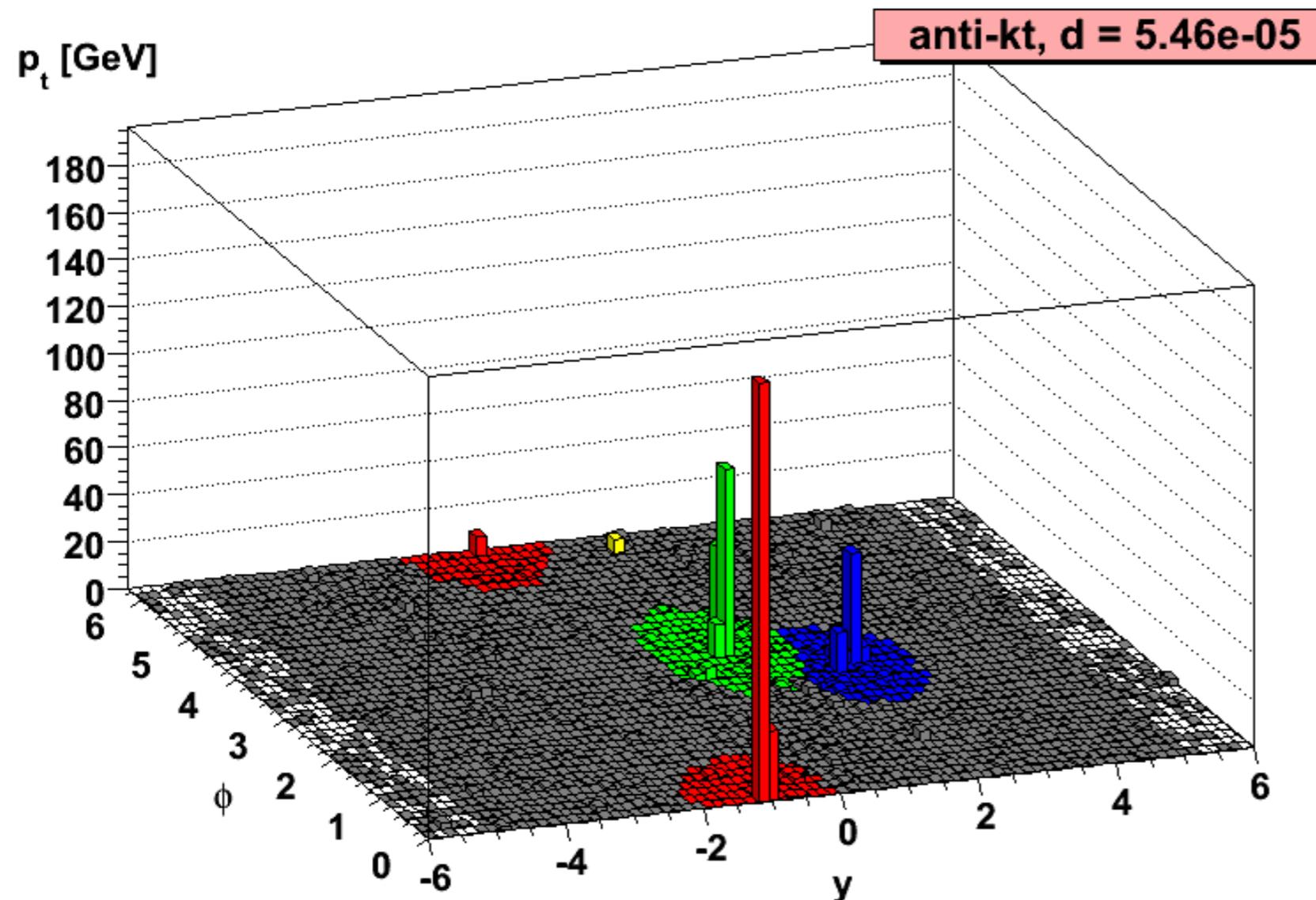
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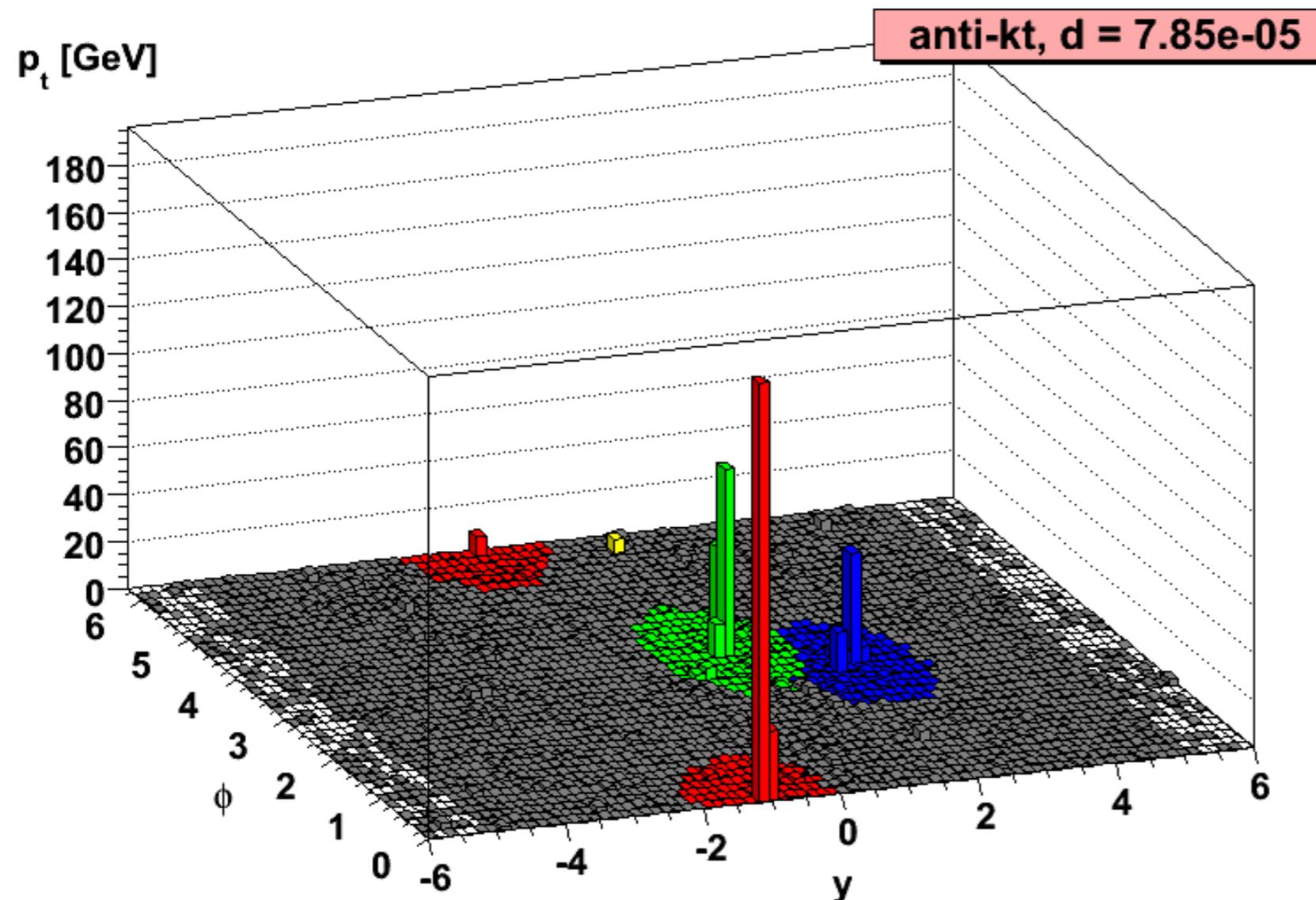
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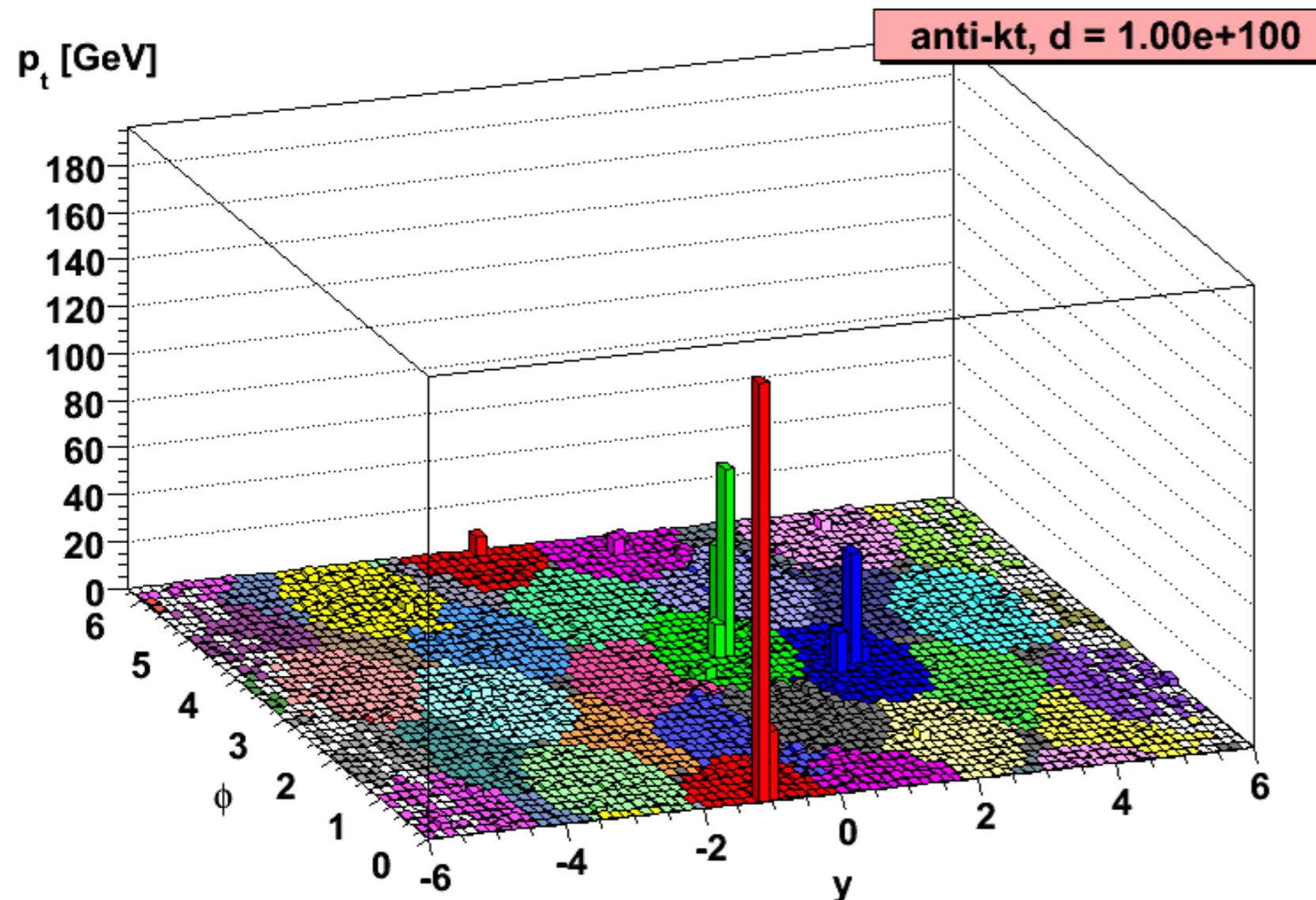
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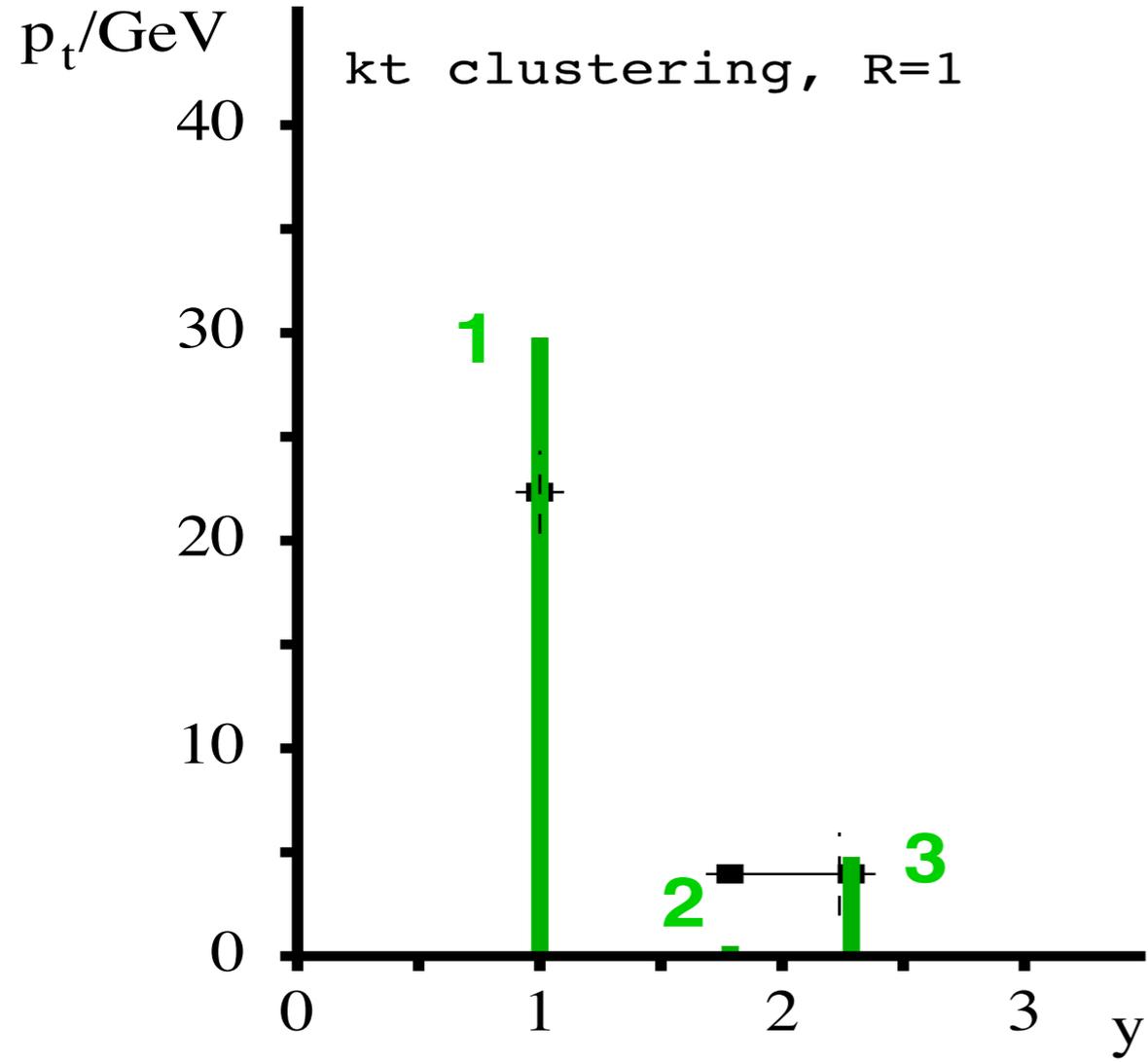


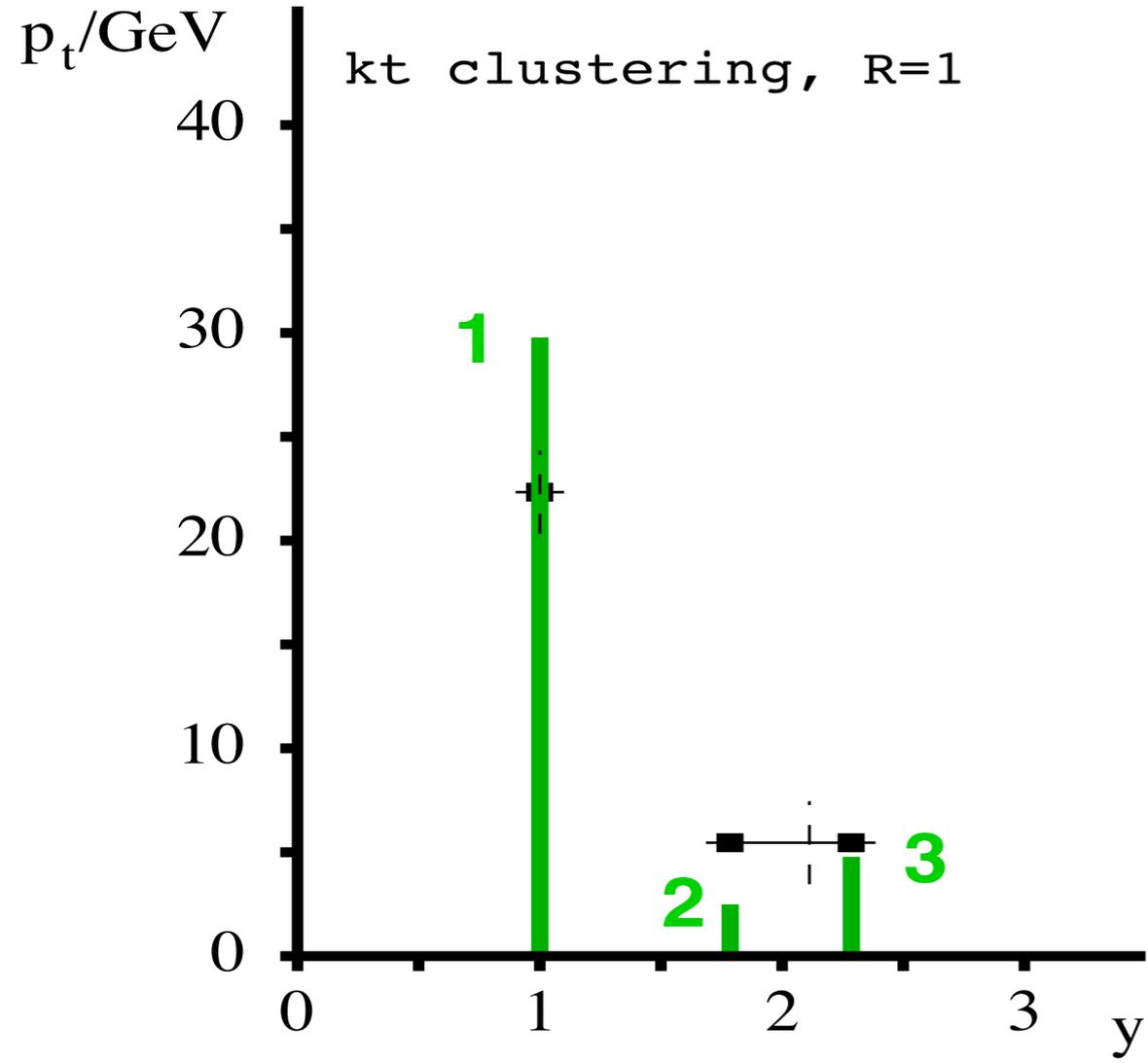
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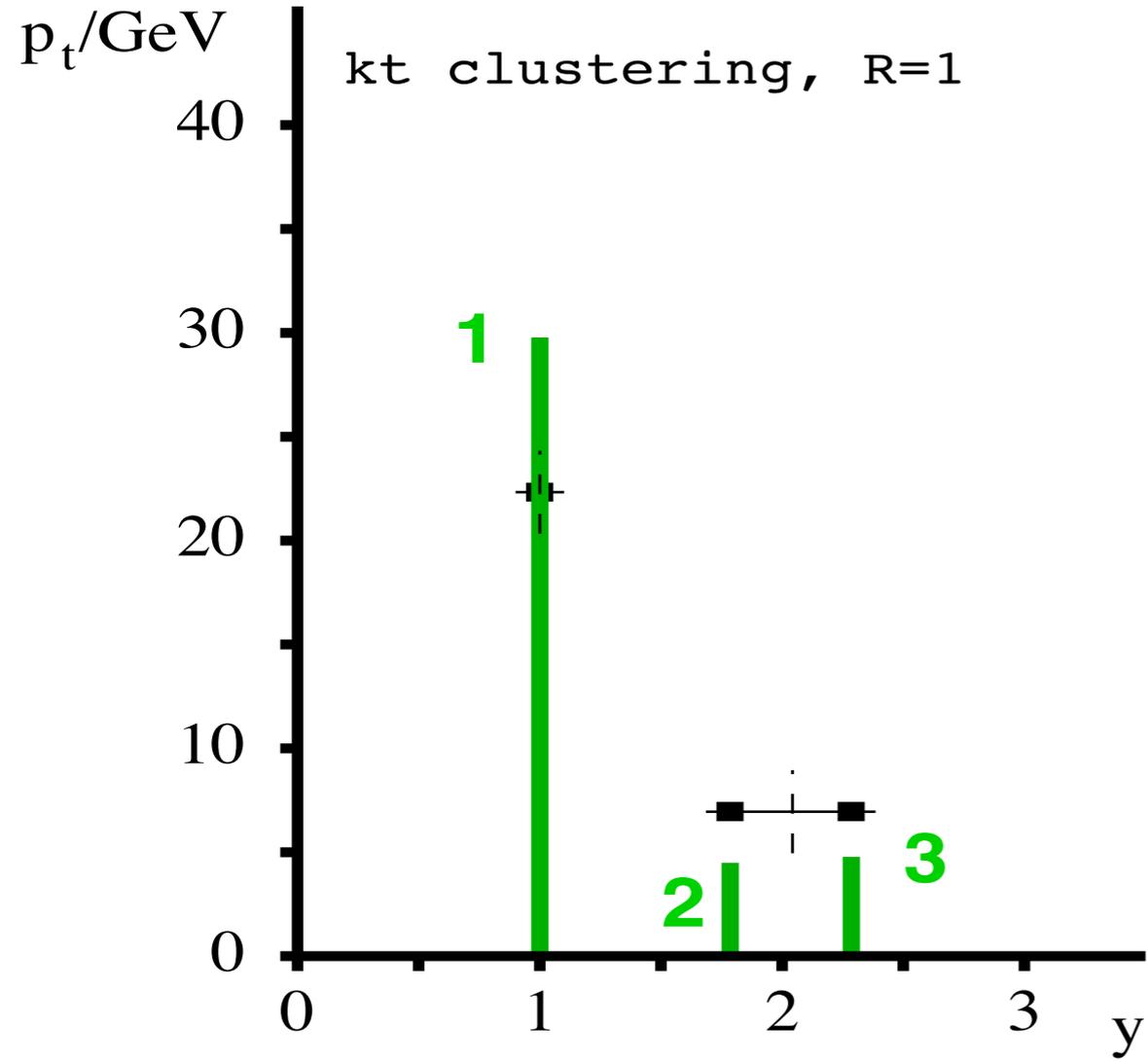
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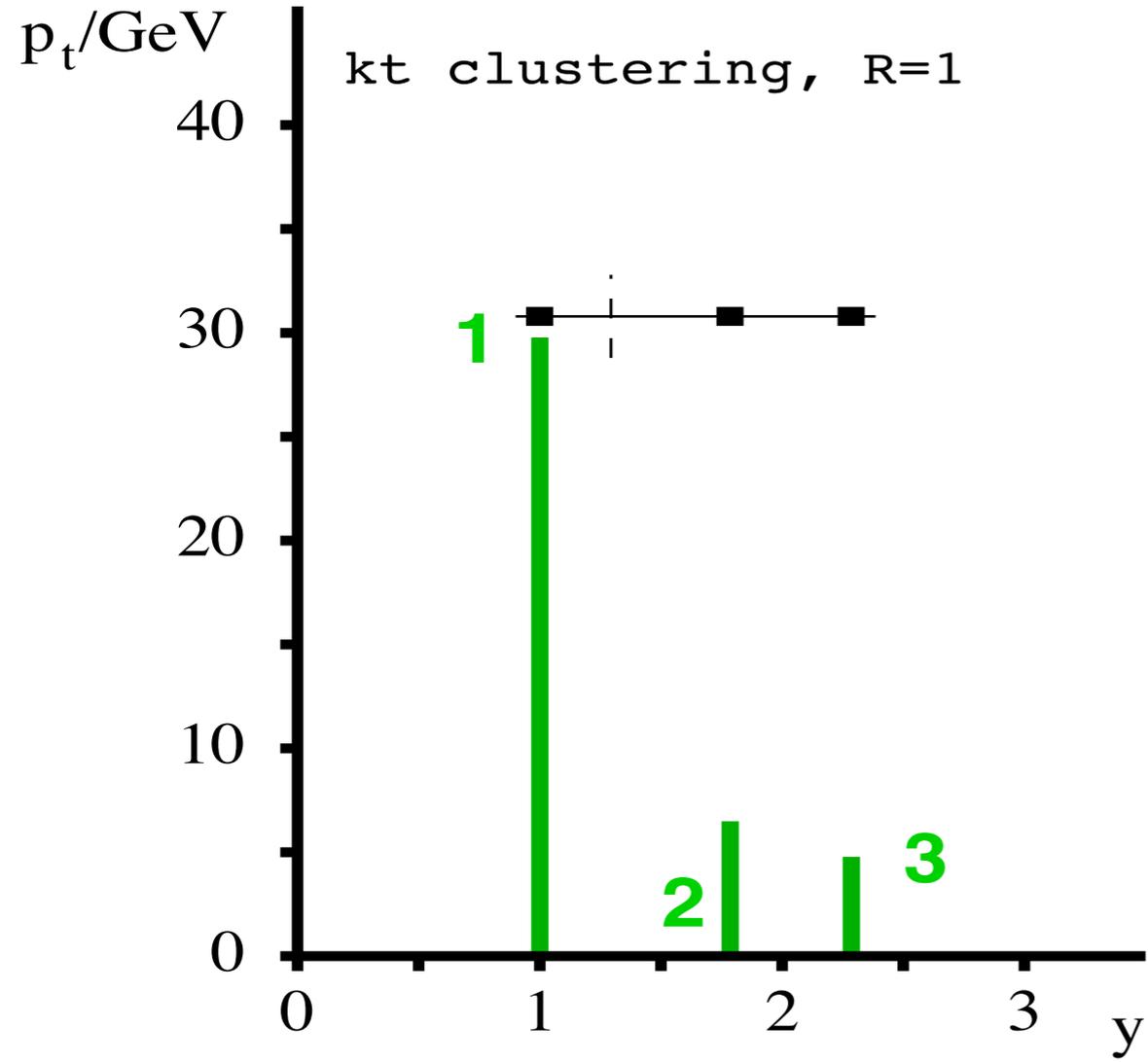


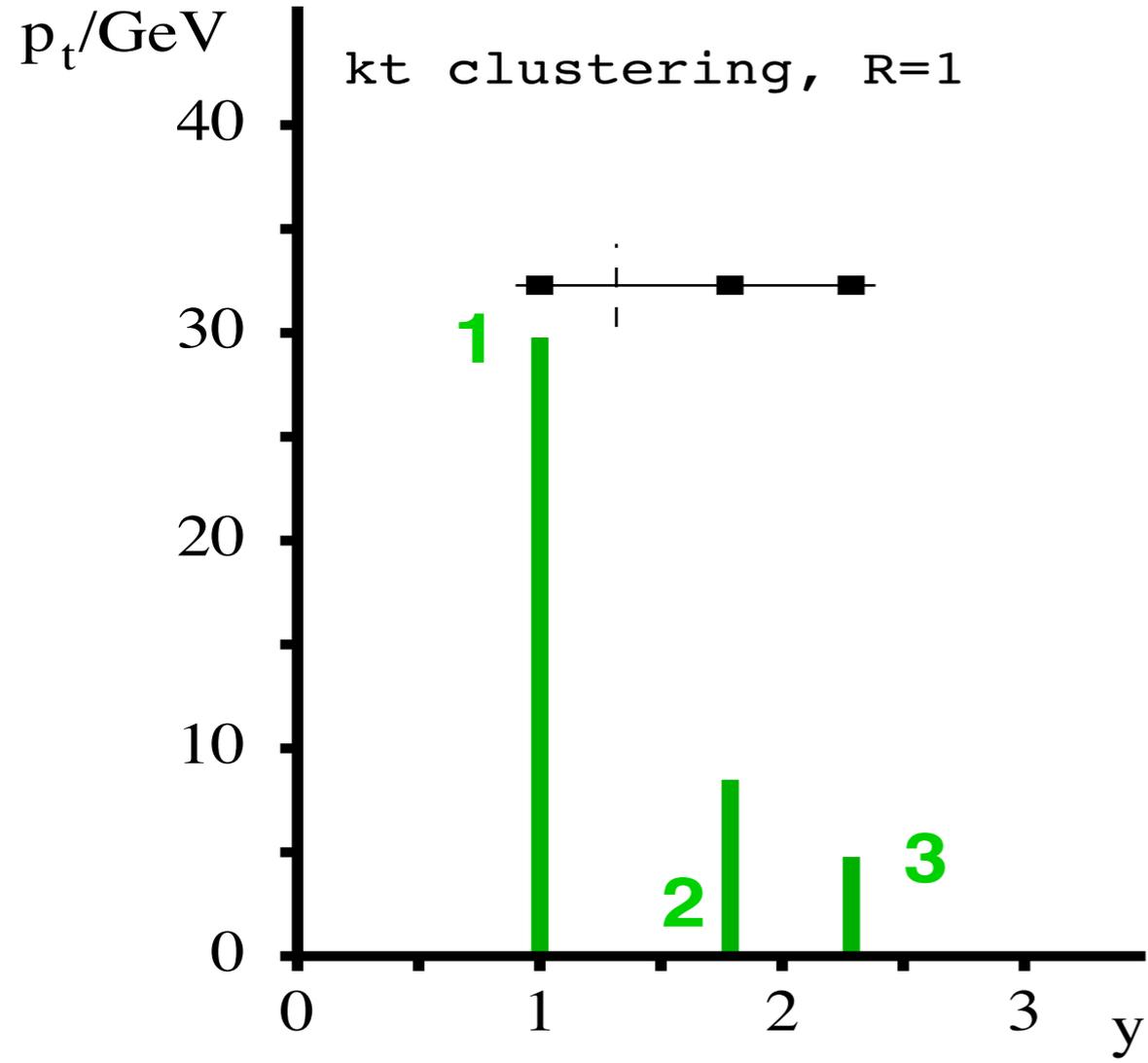
**Anti- $k_t$  gives circular jets ("cone-like") in a way that's infrared safe**

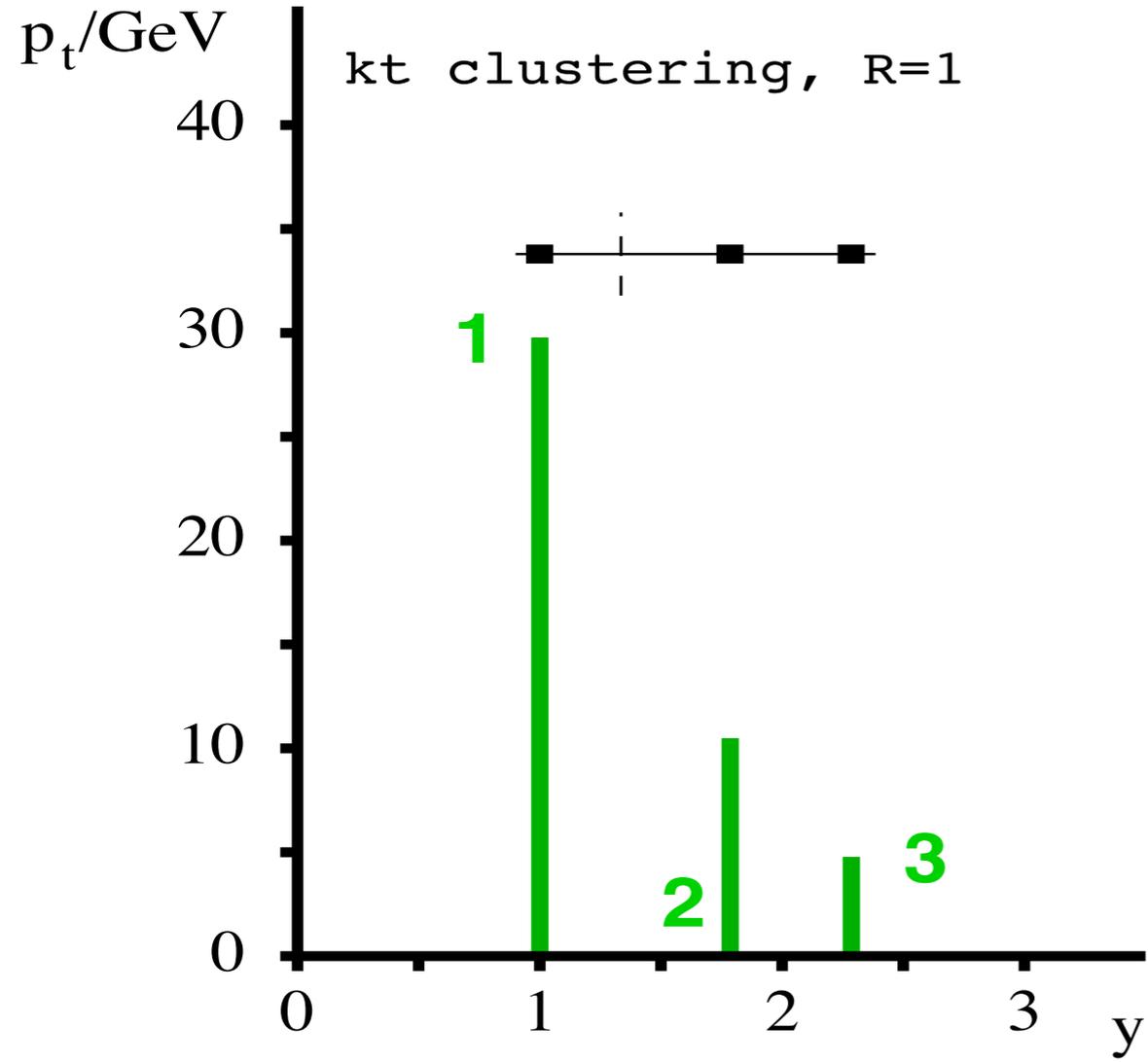


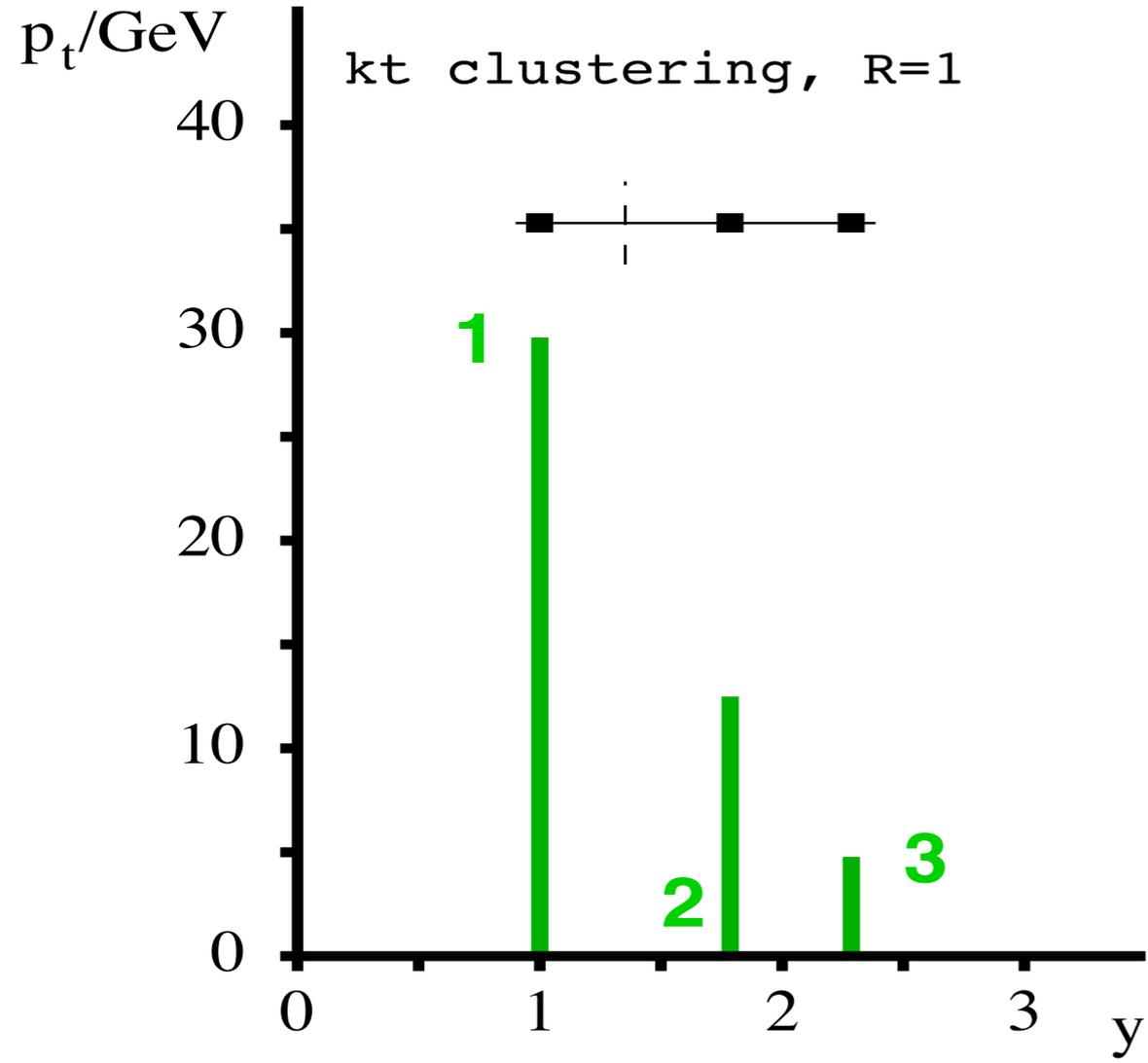


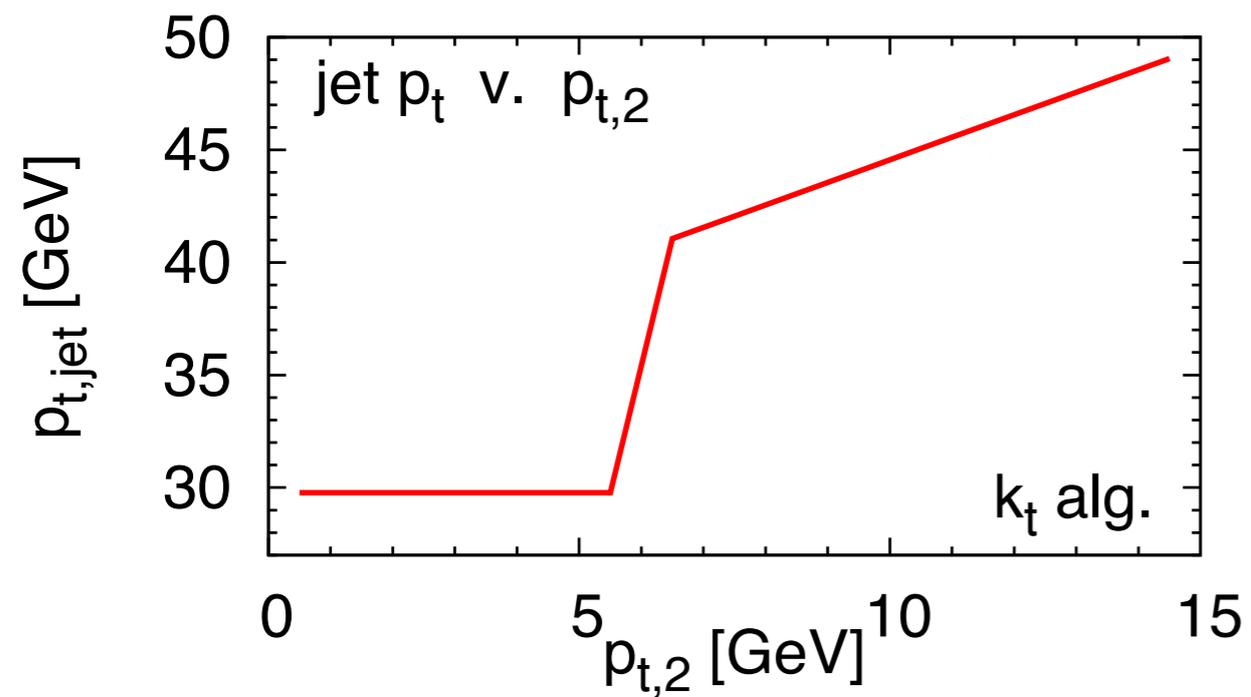
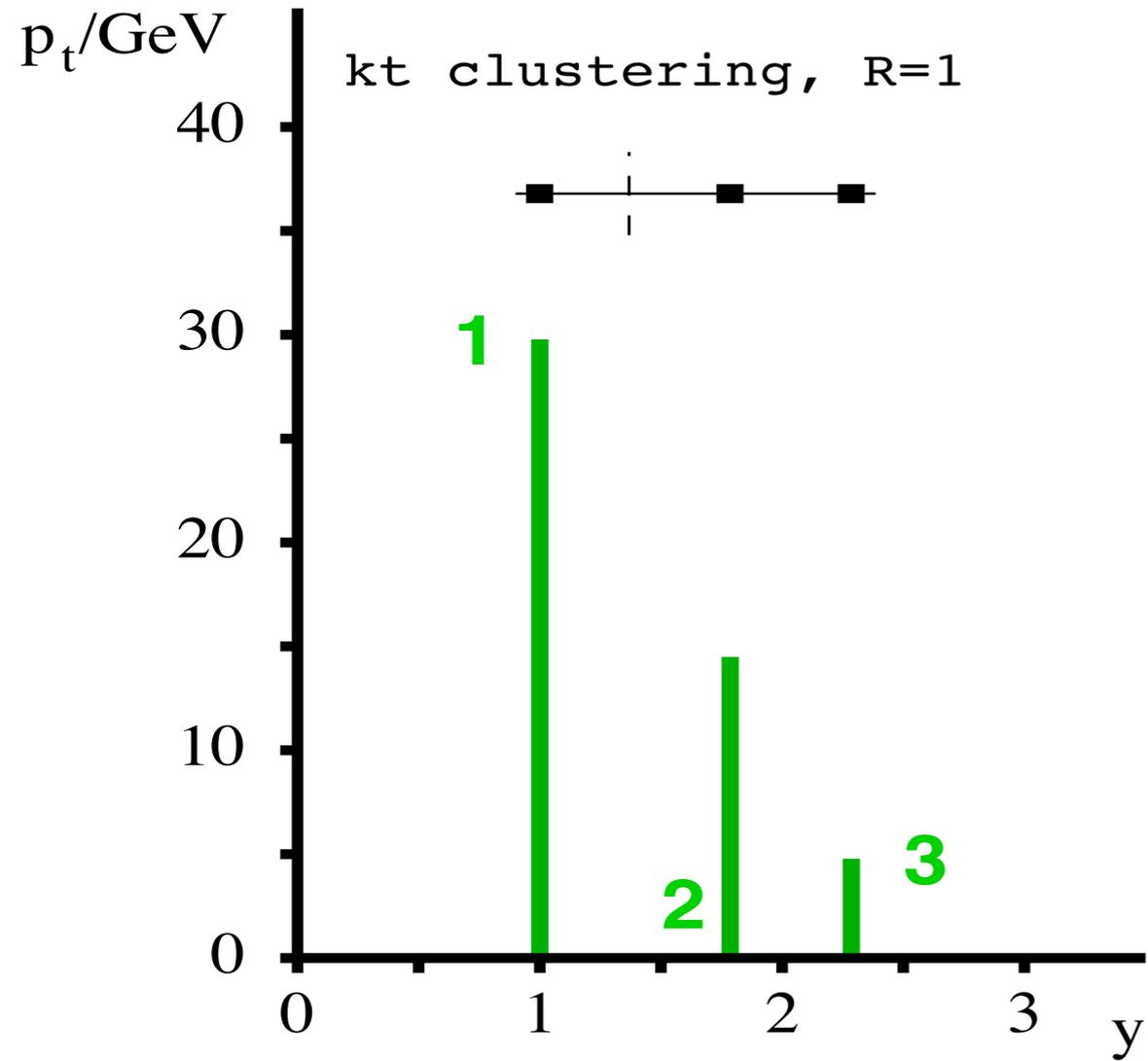




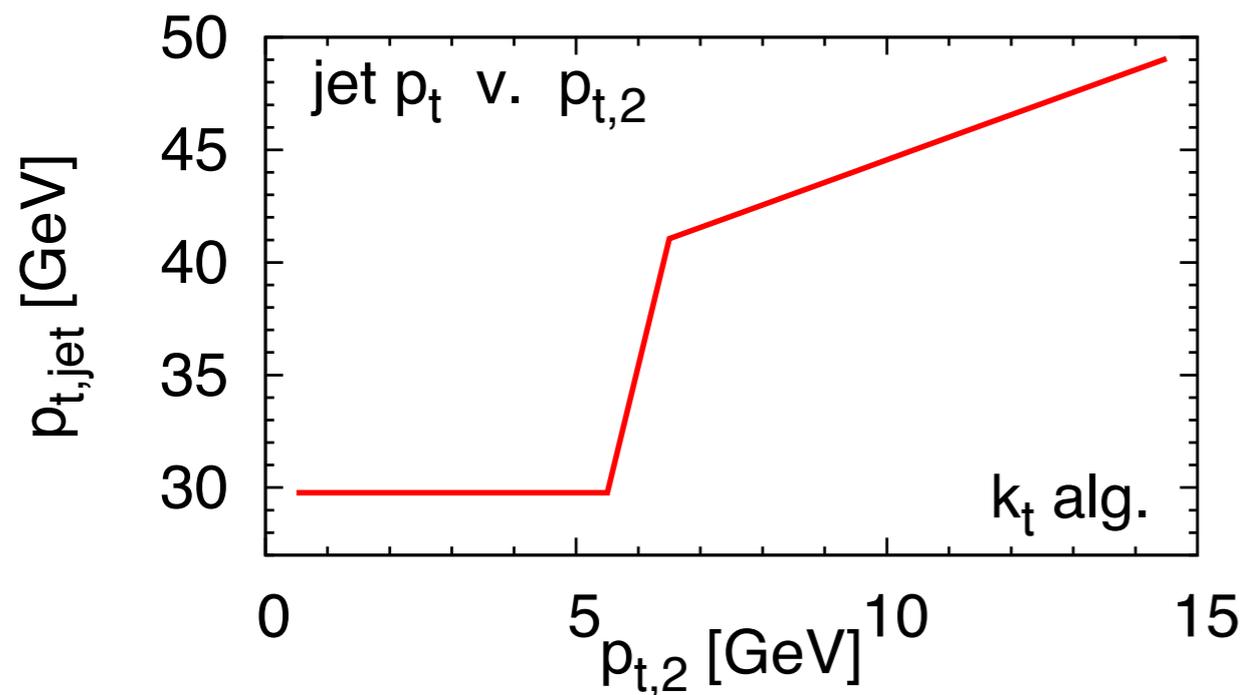
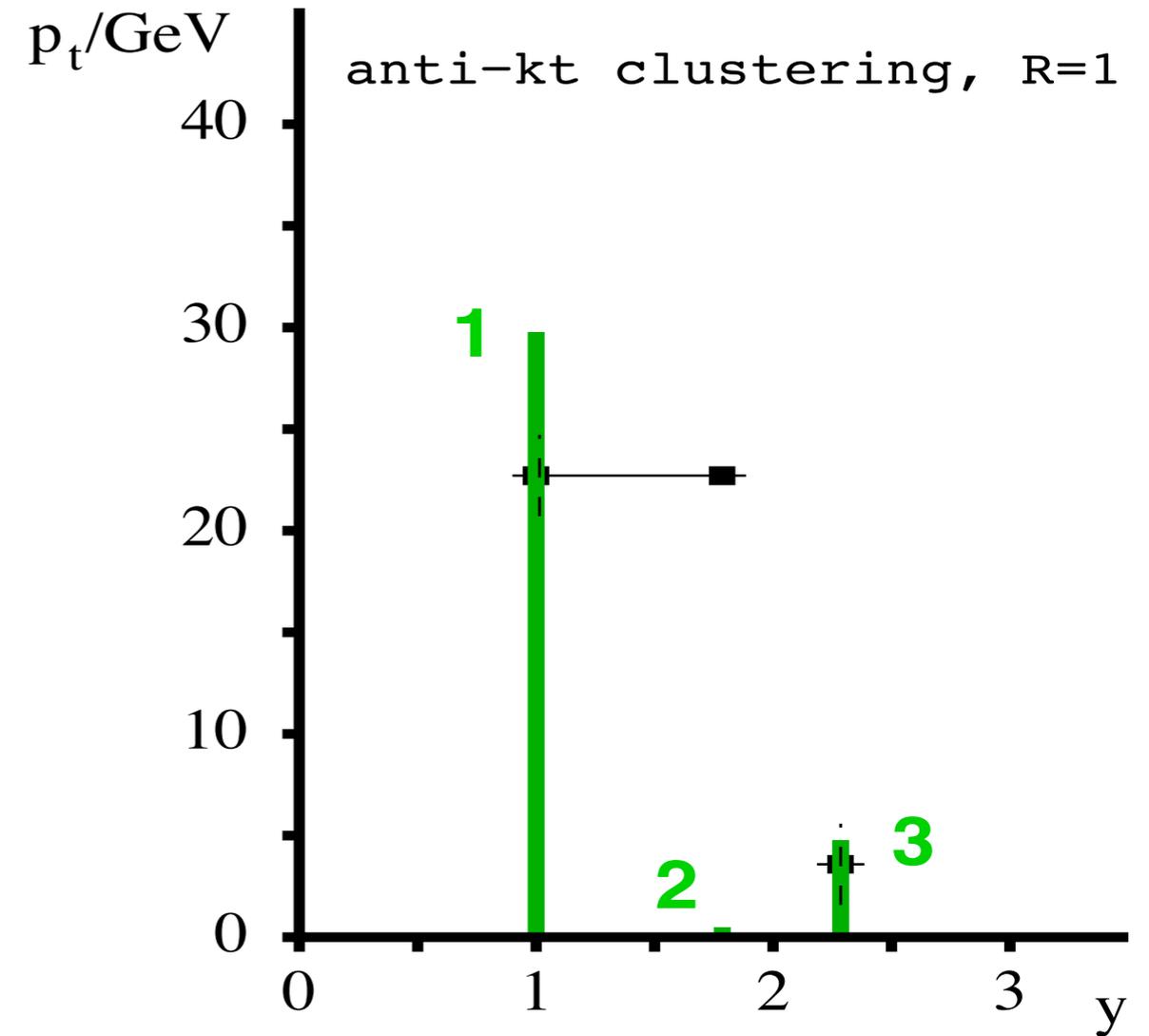
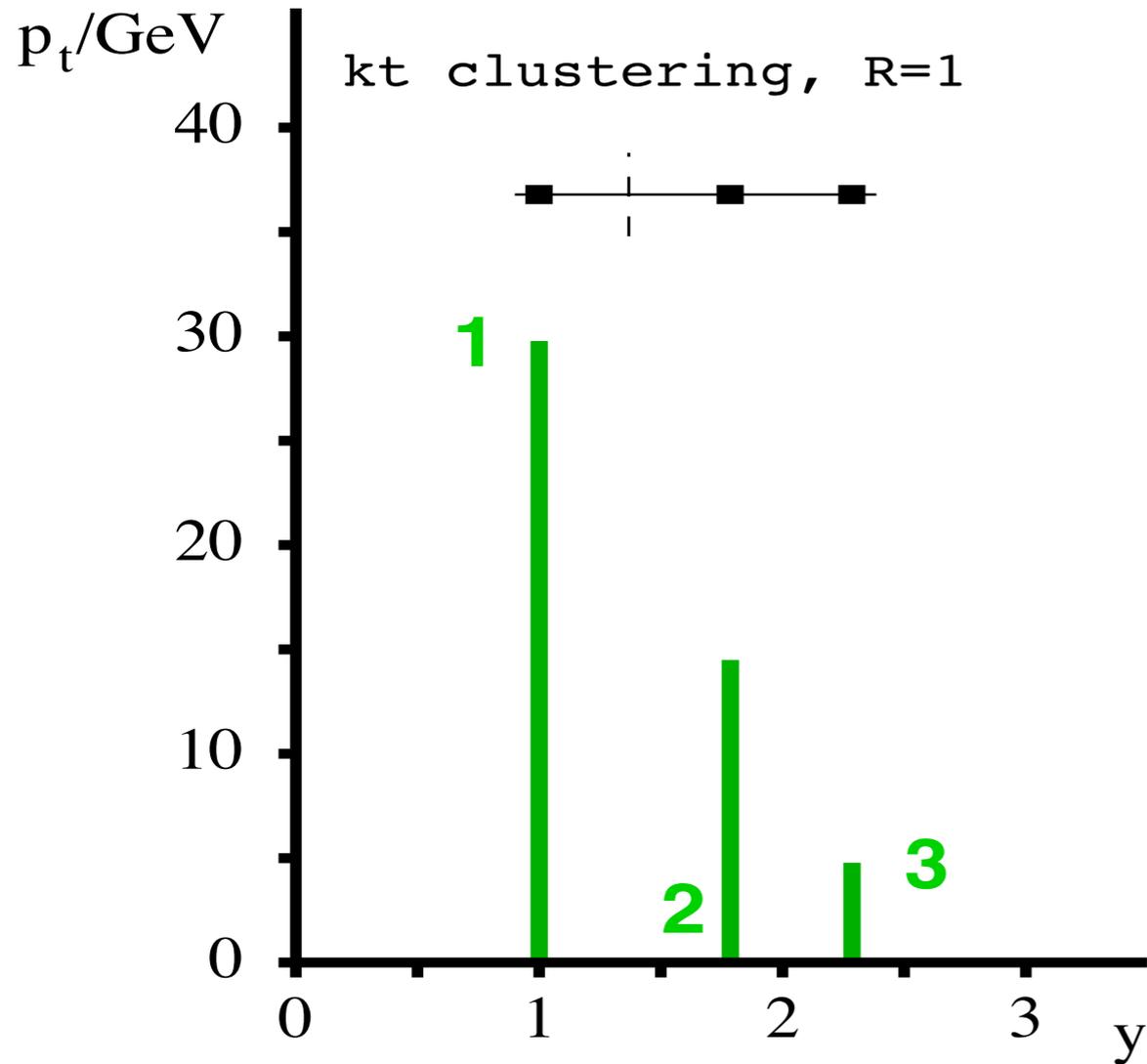




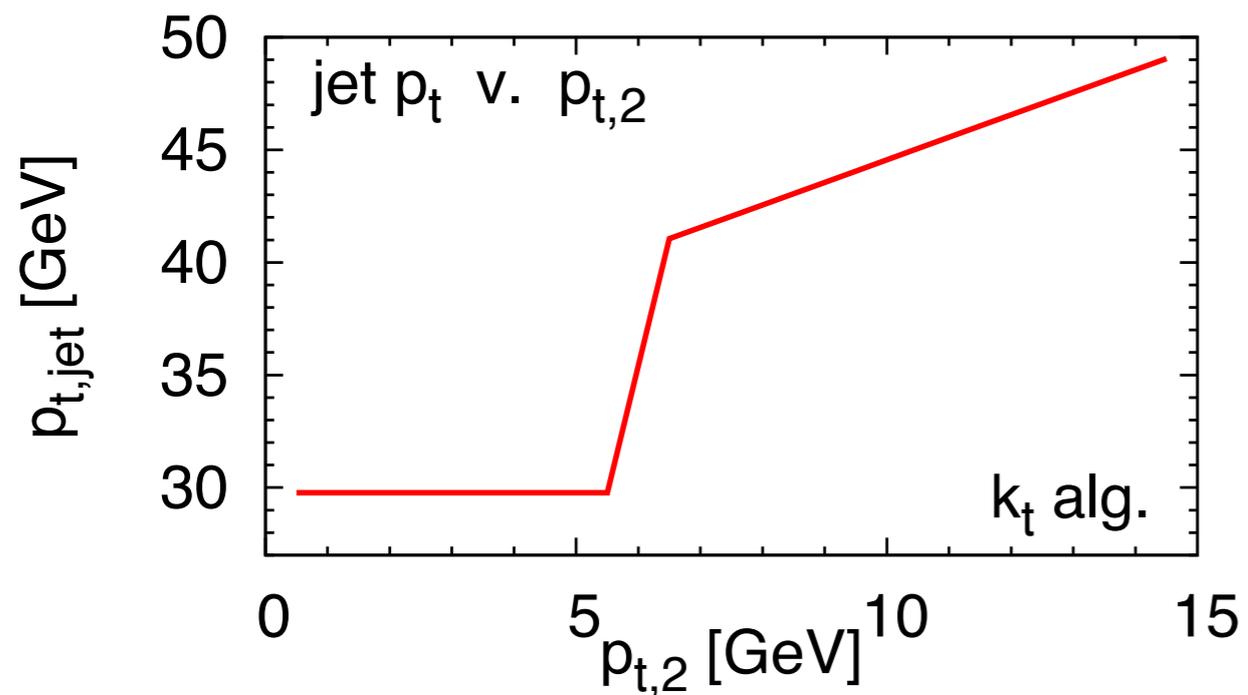
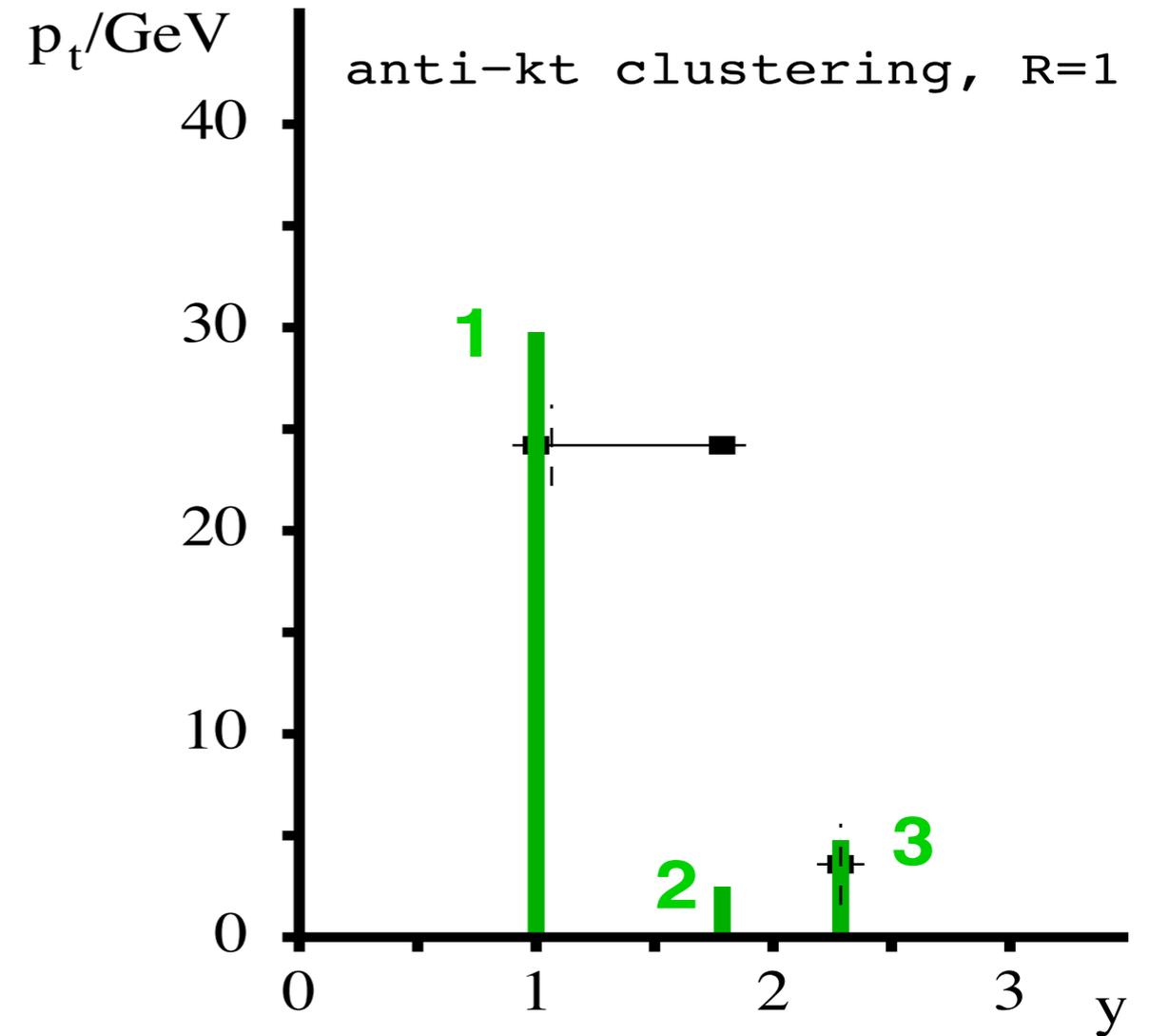
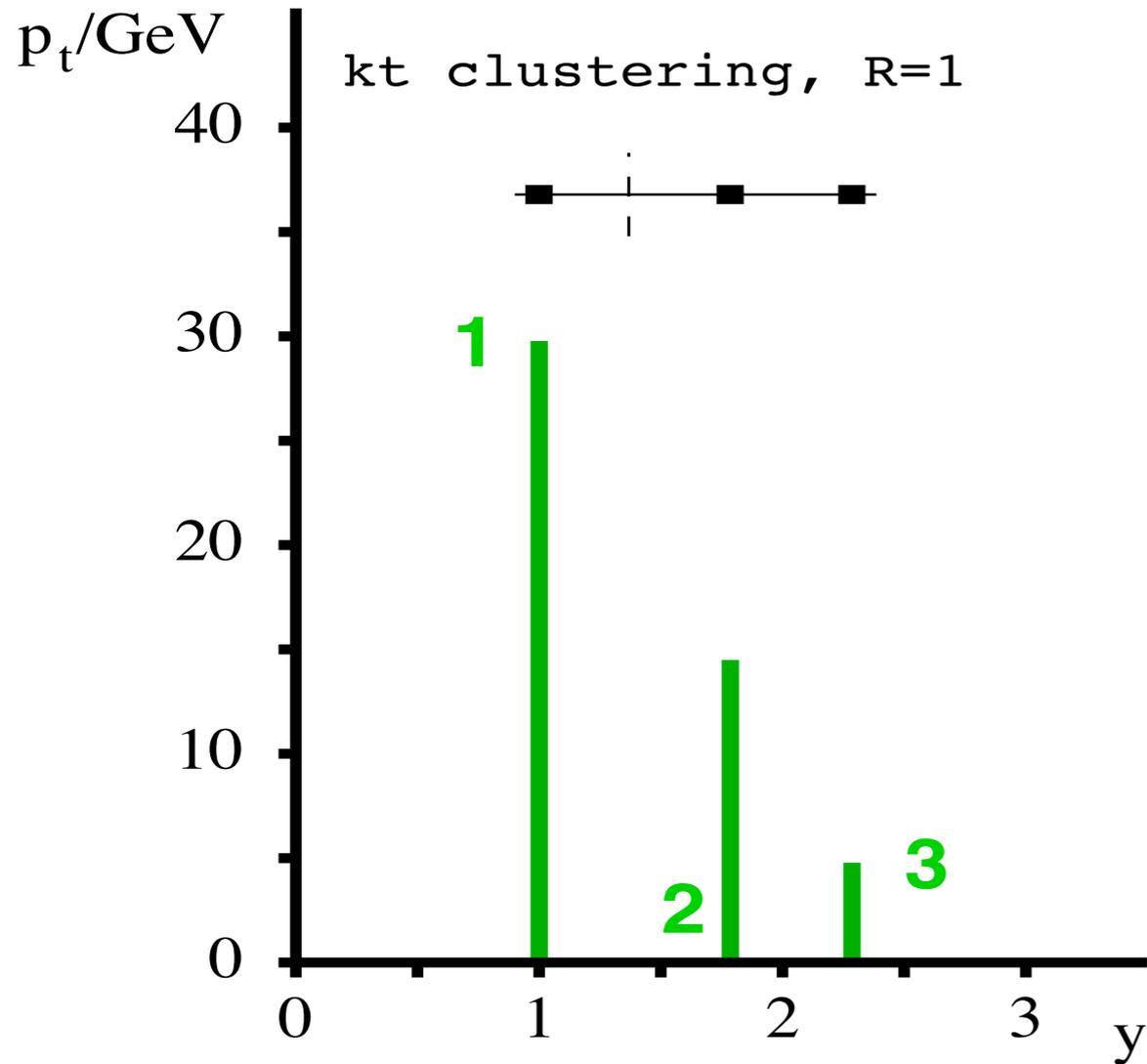




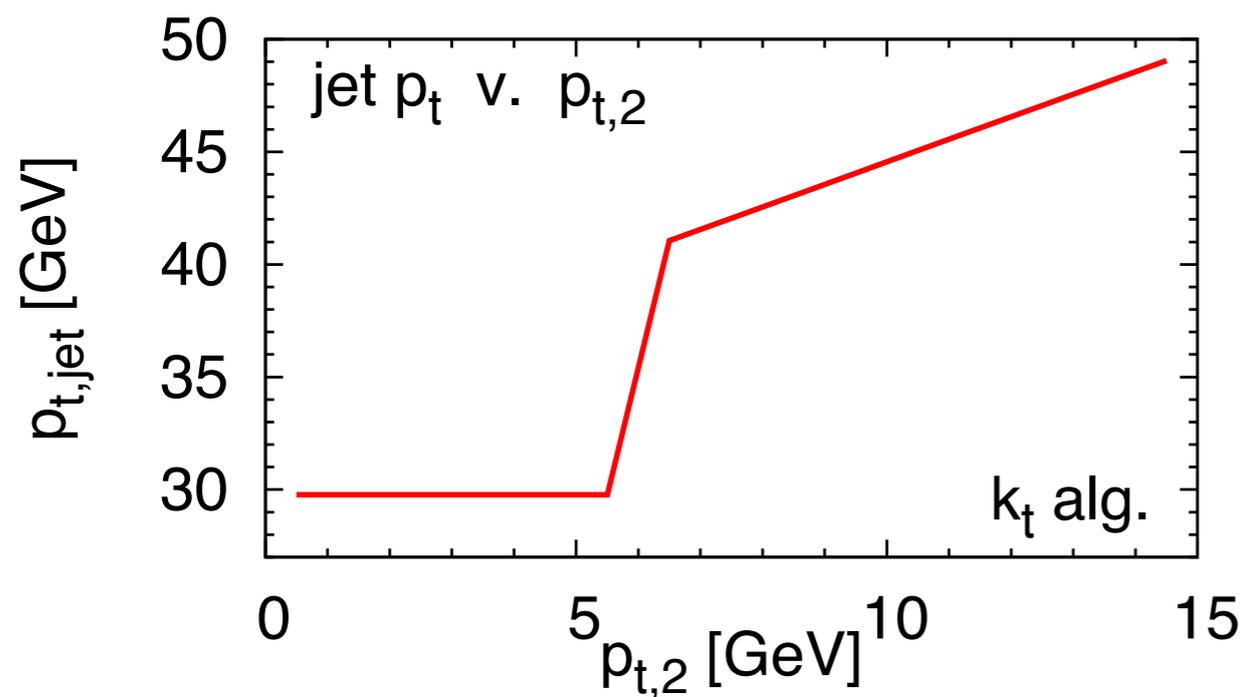
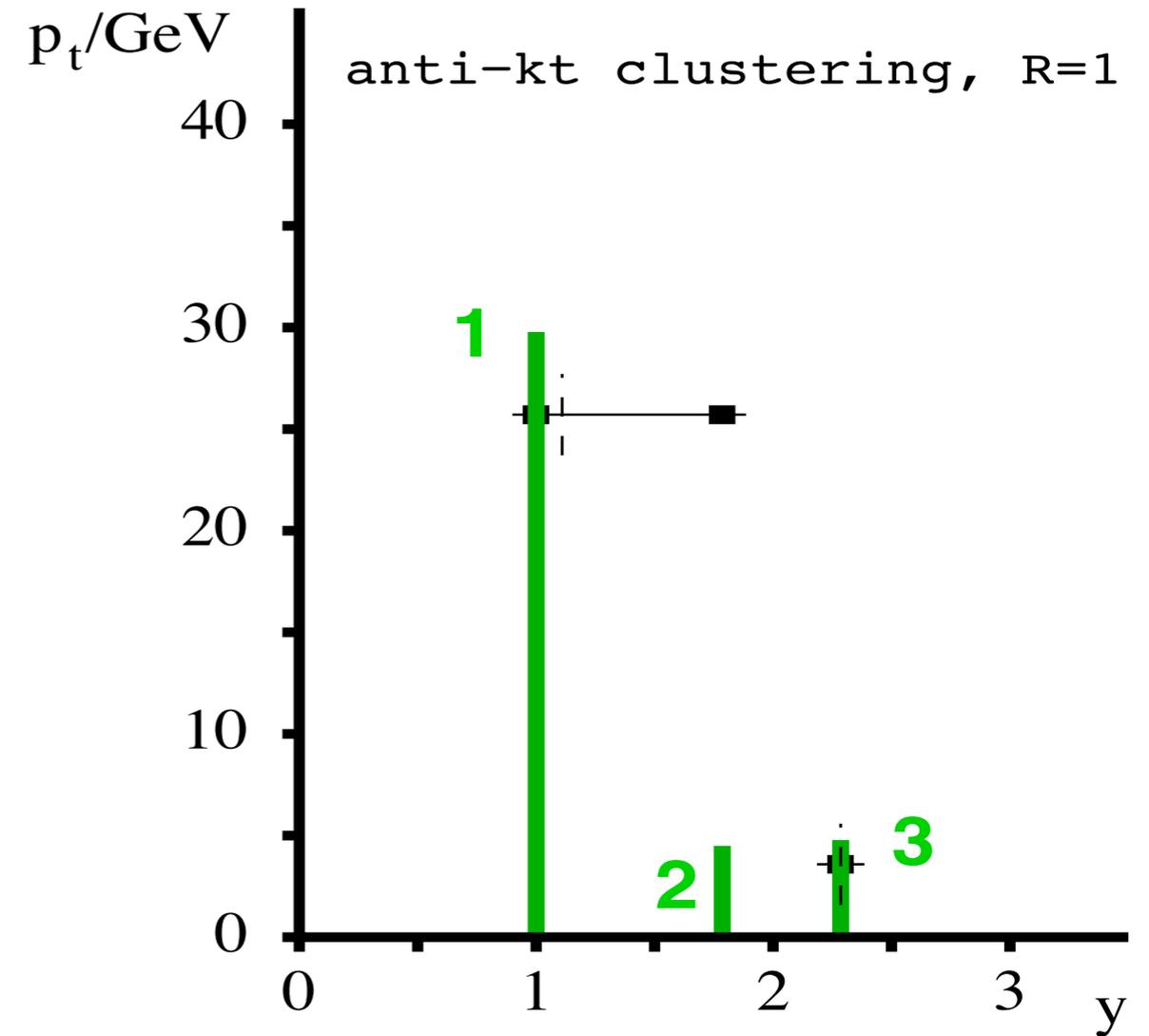
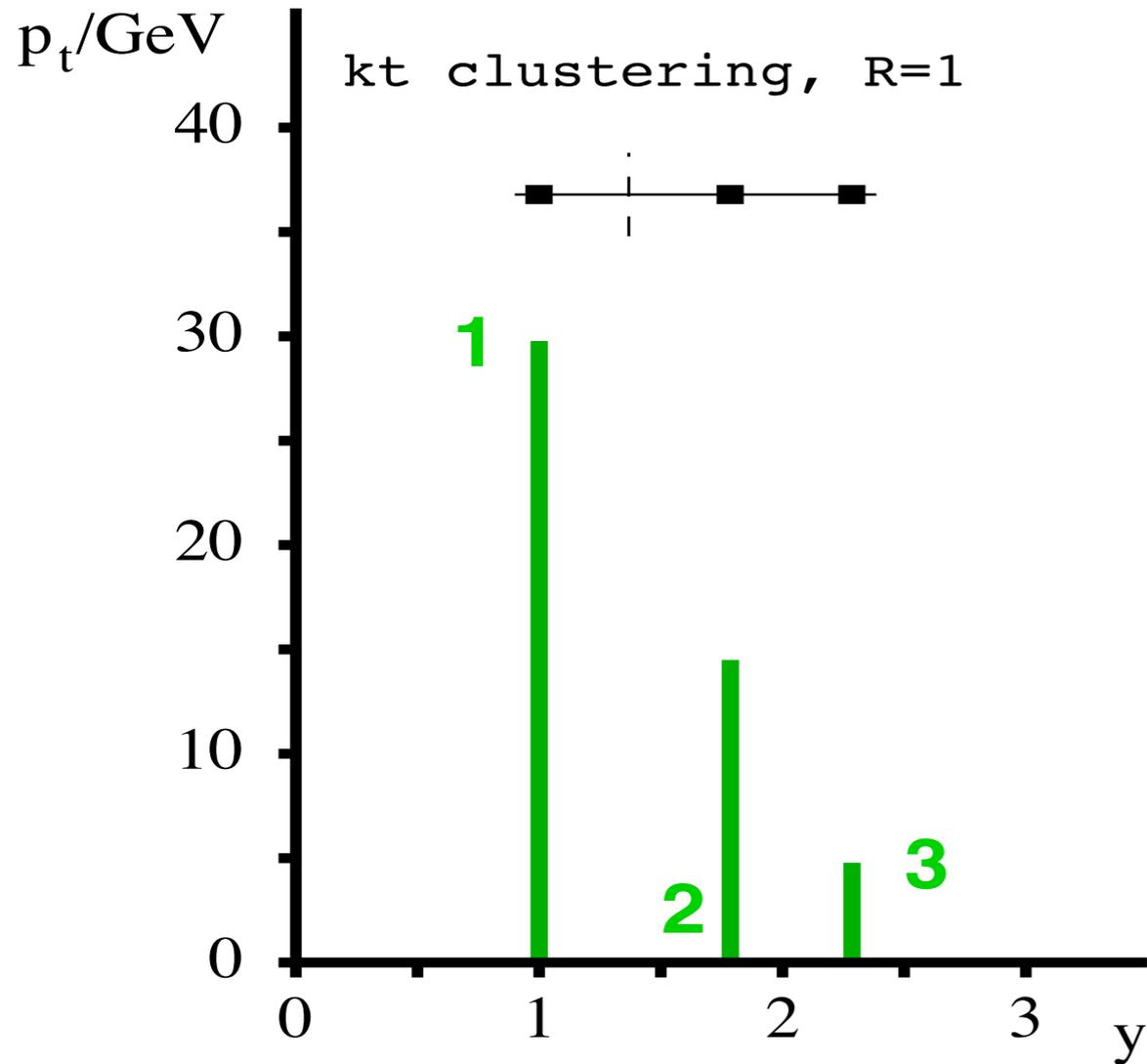
# Linearity: $k_t$ v. anti- $k_t$



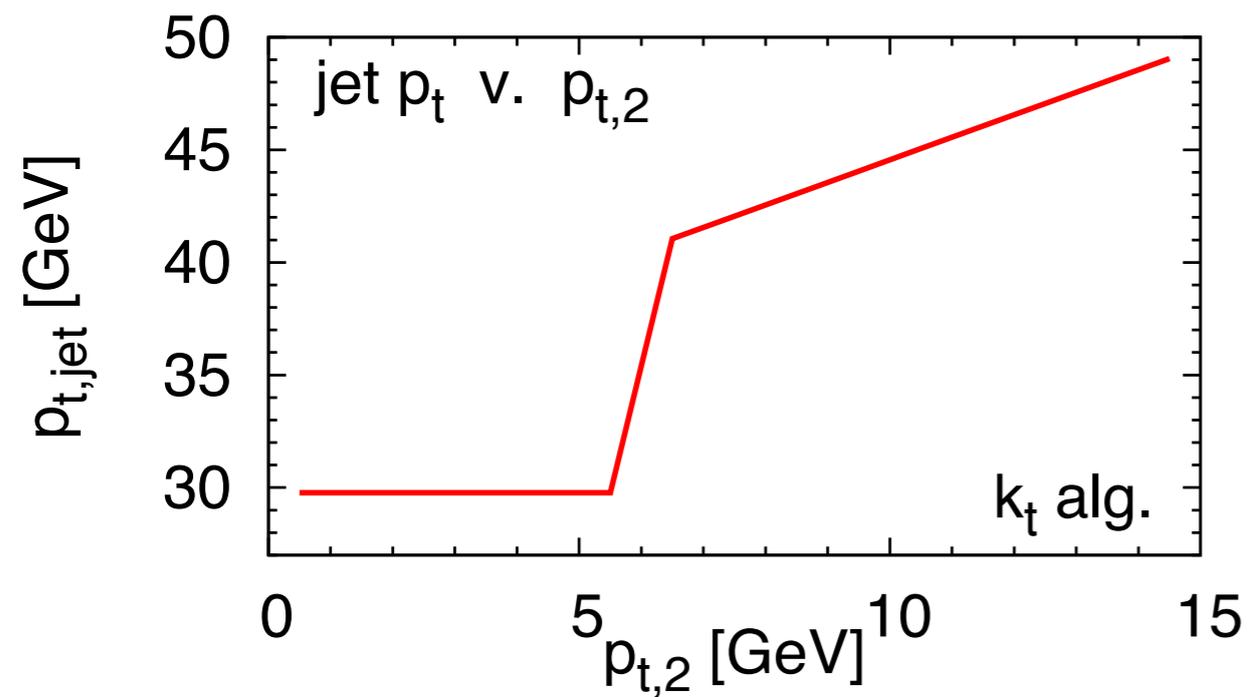
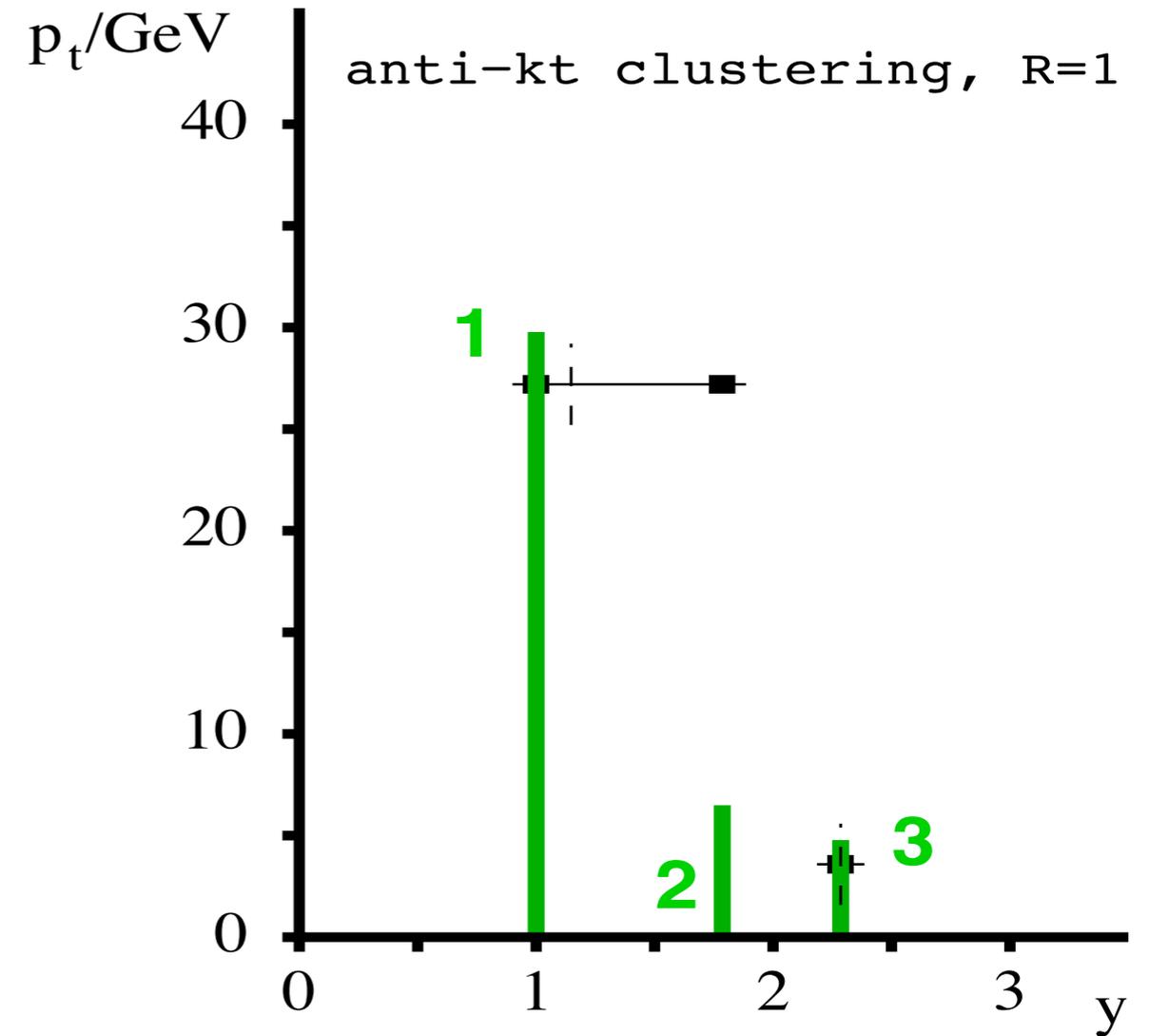
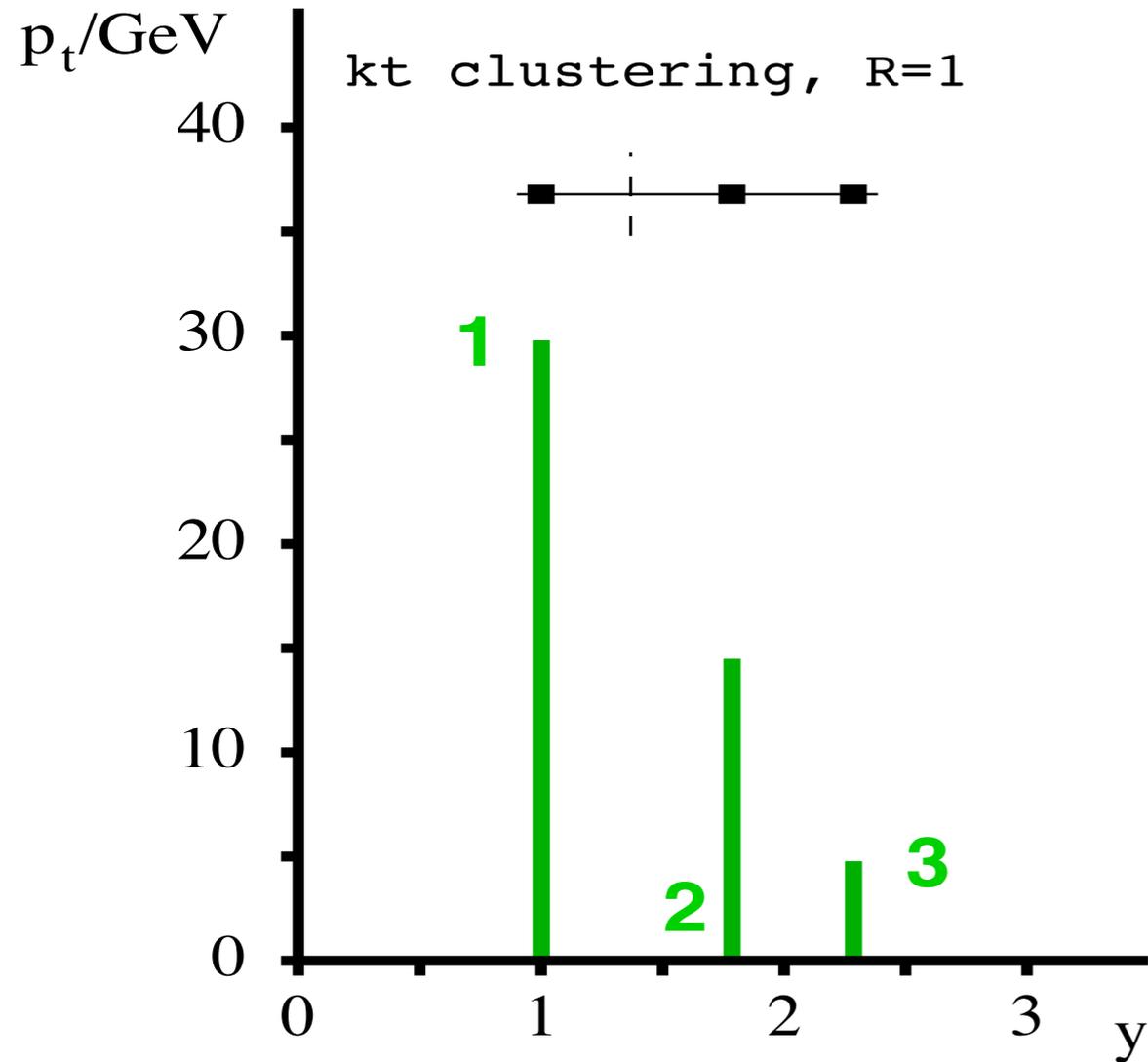
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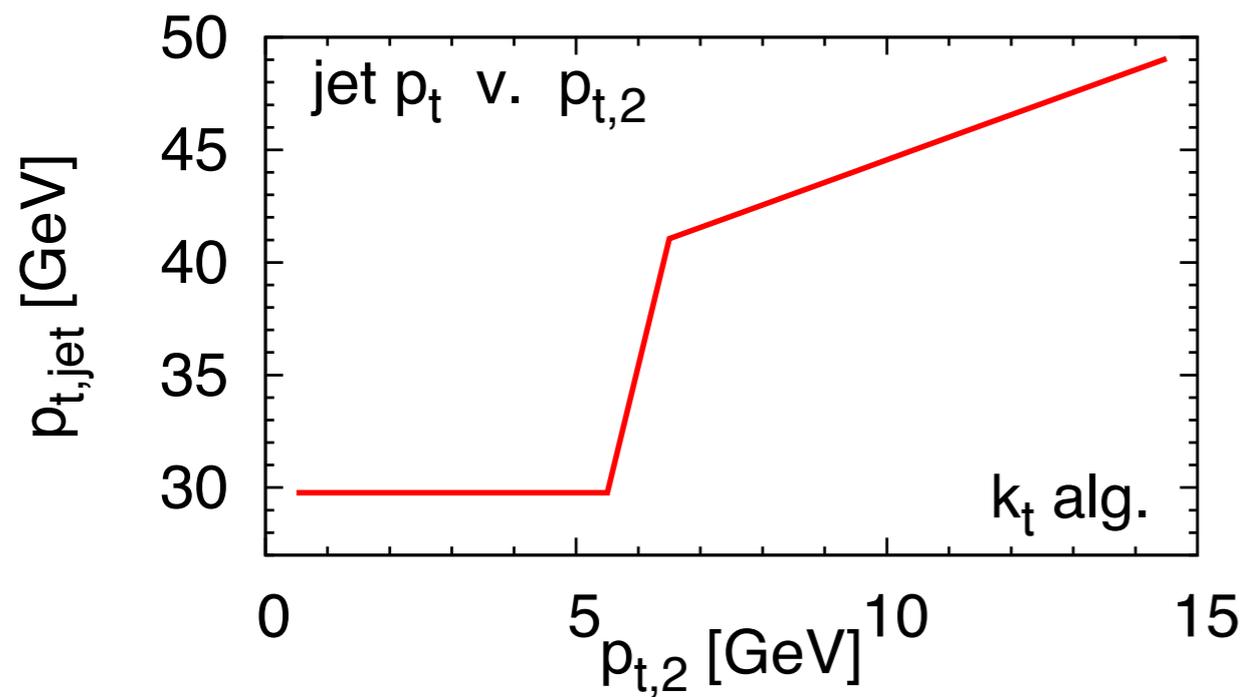
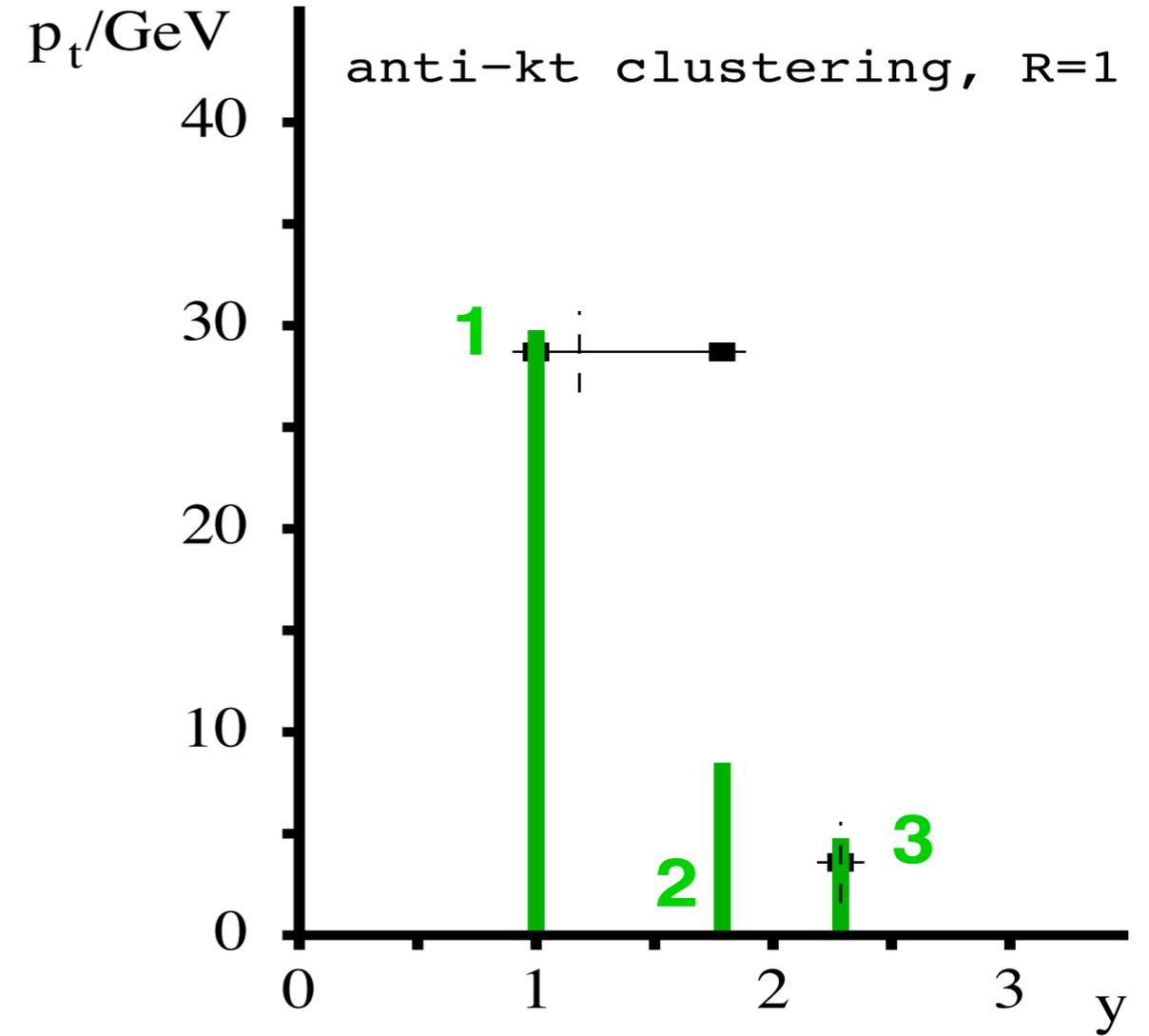
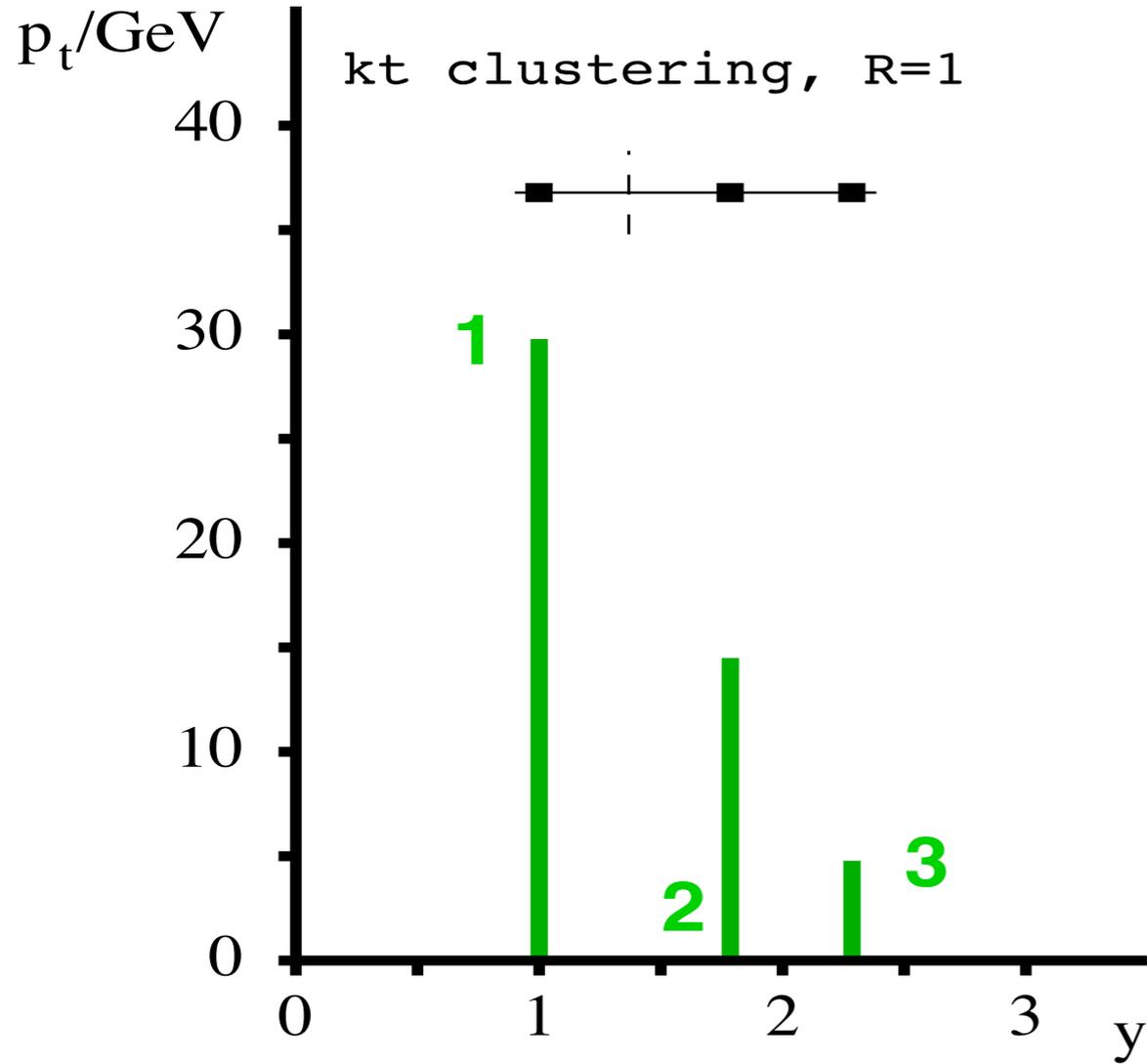
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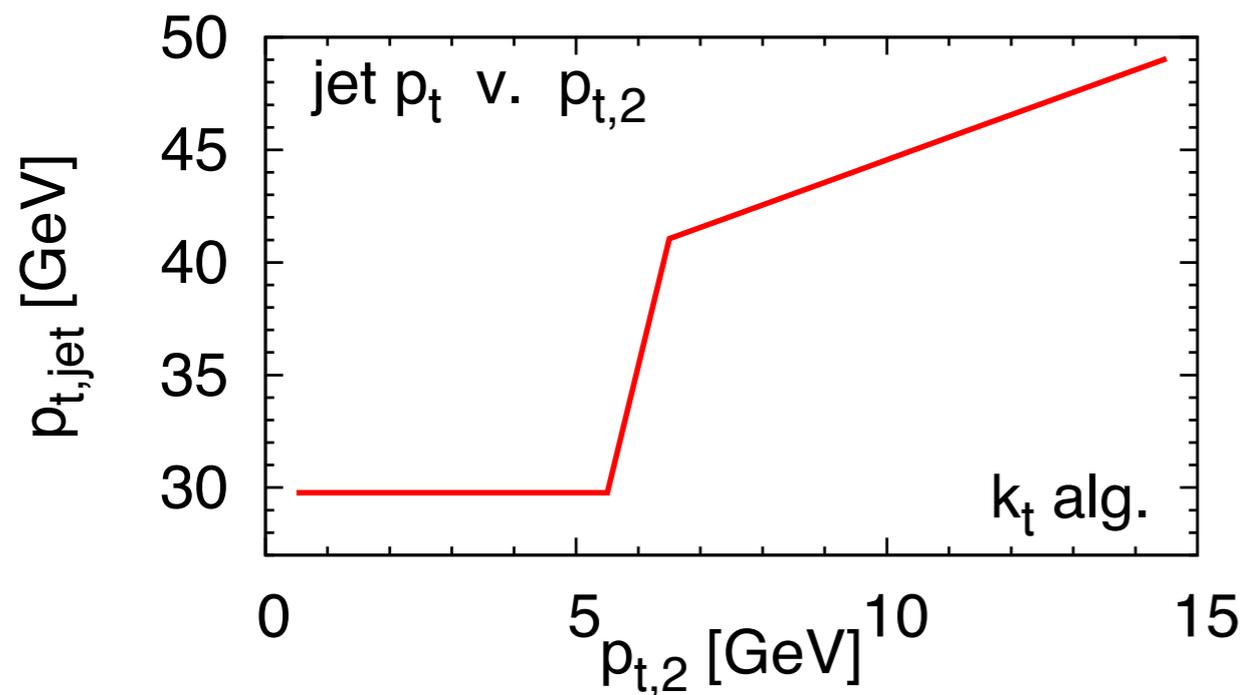
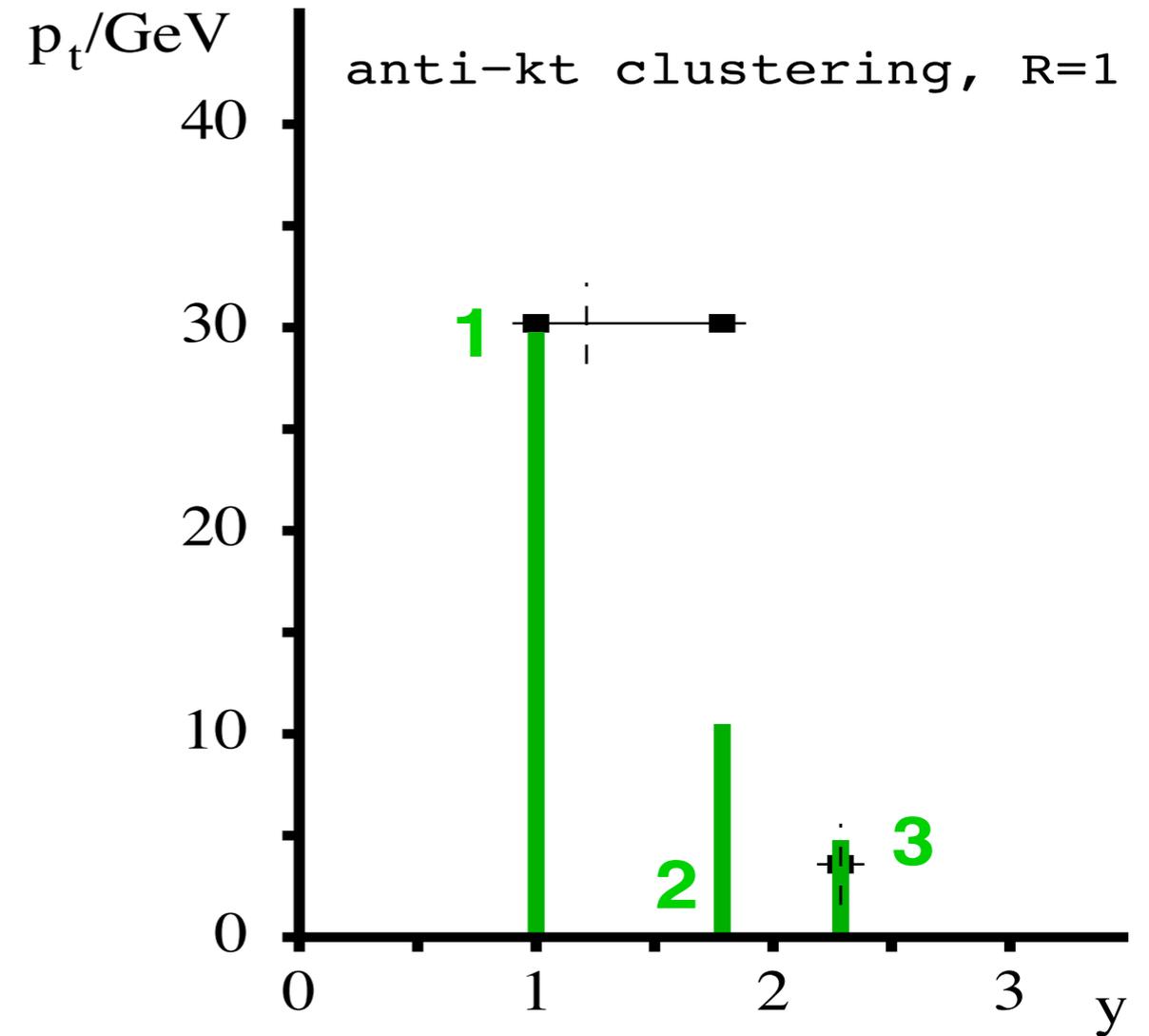
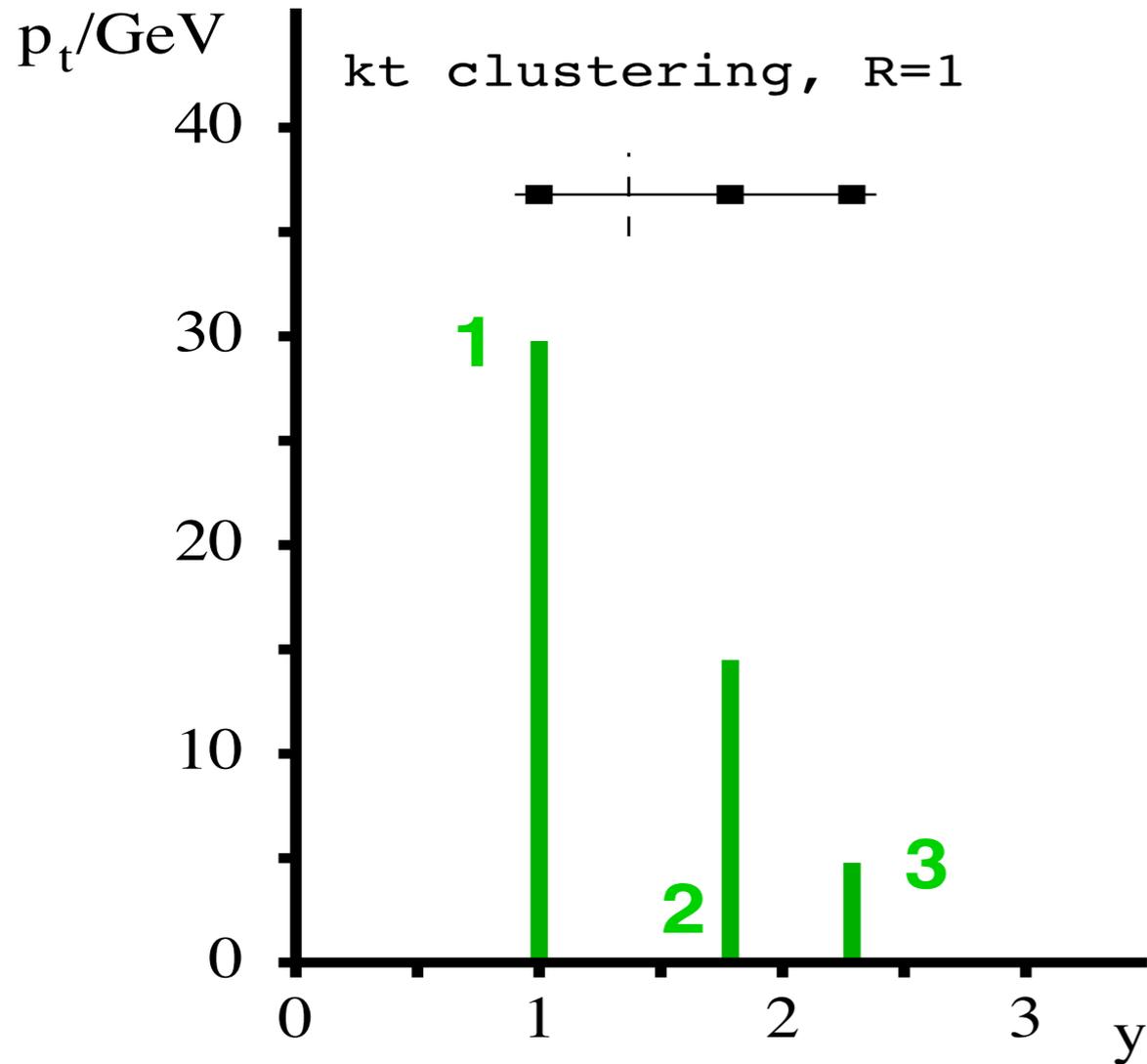
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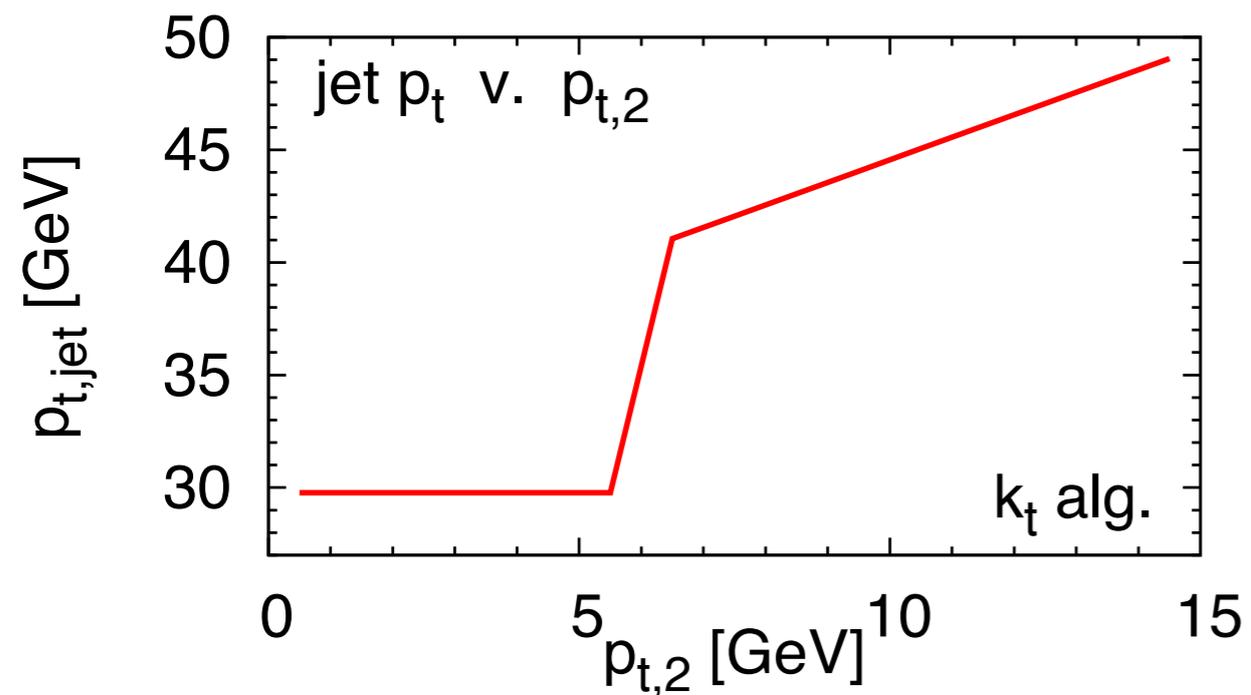
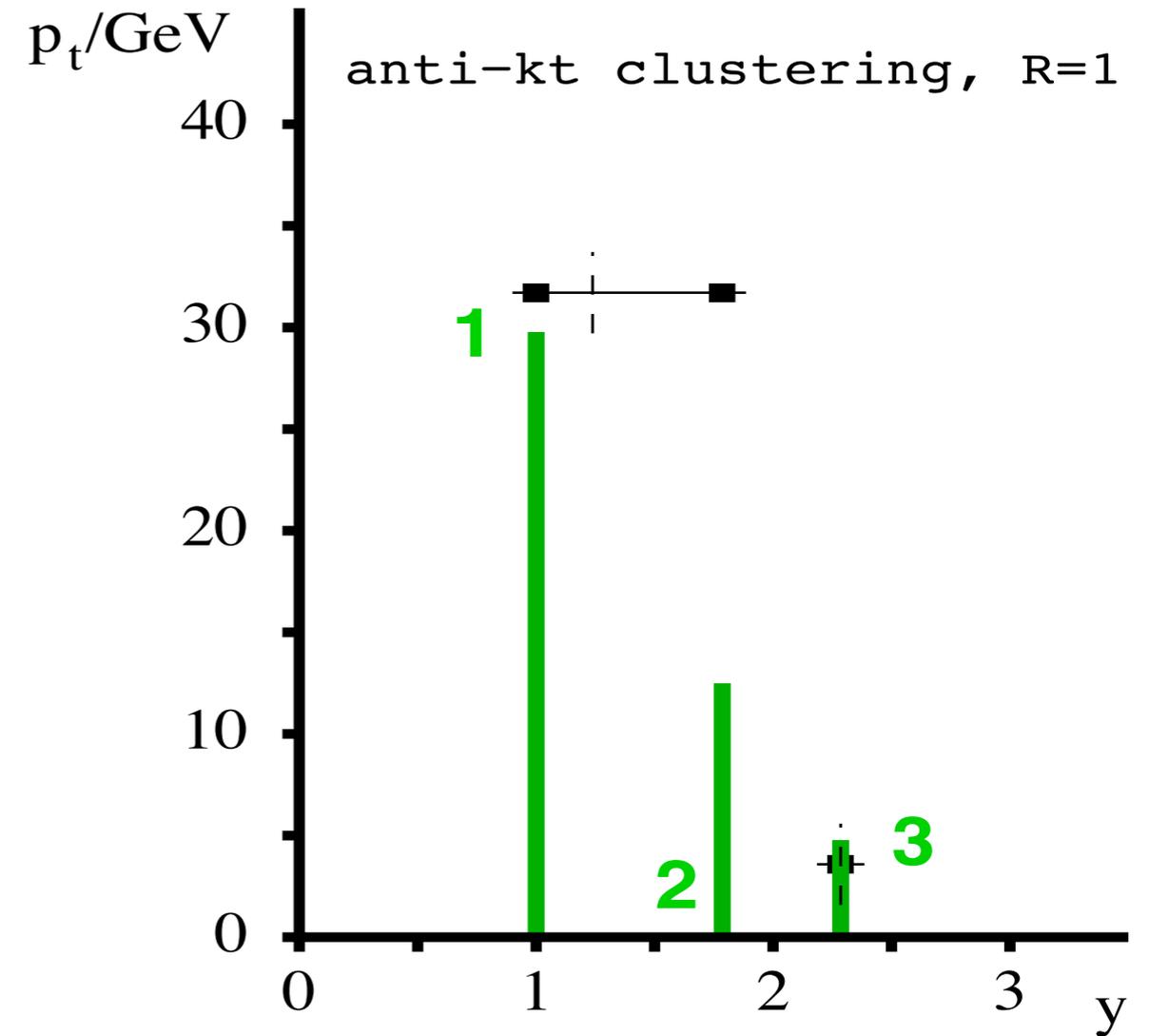
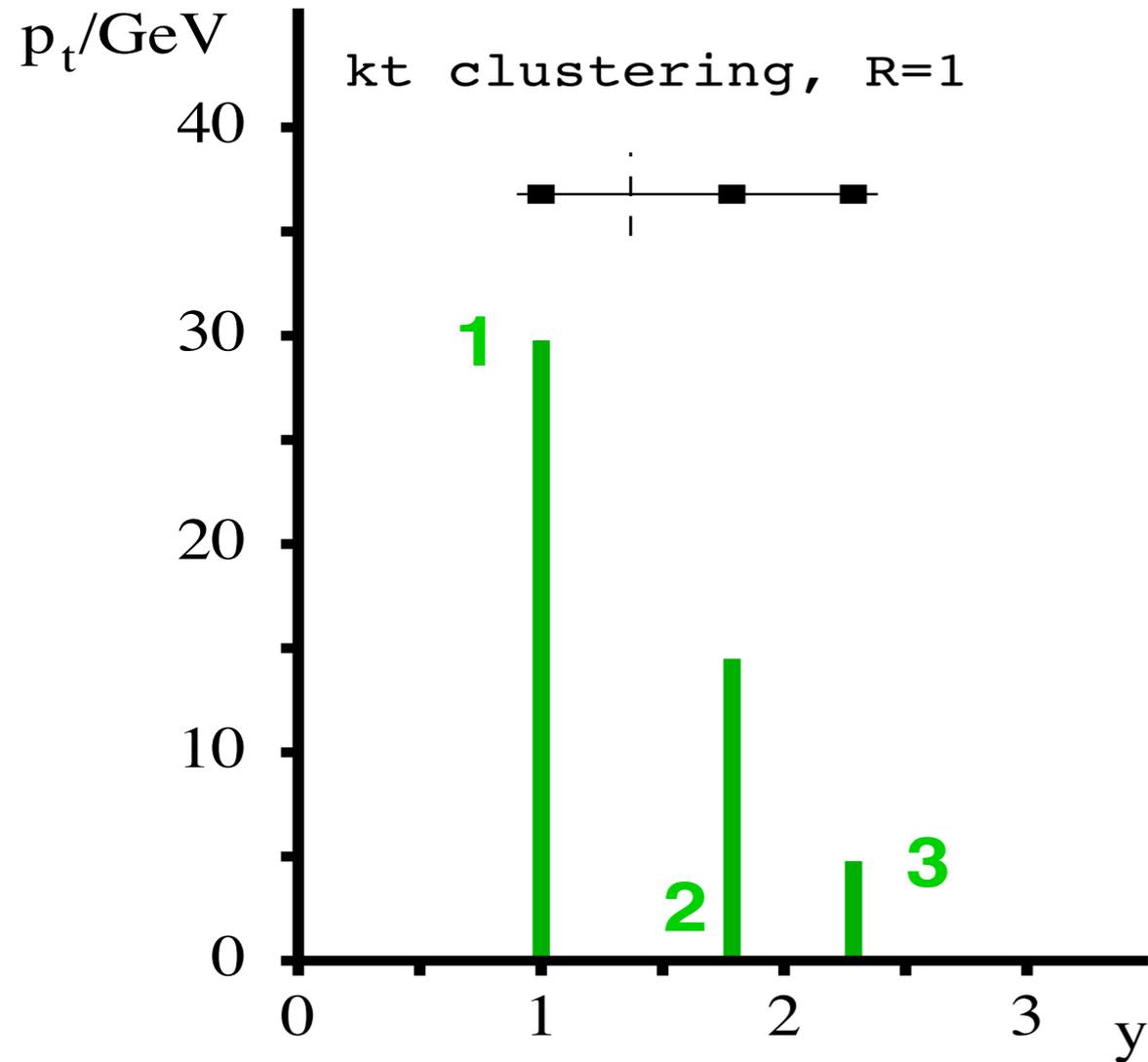
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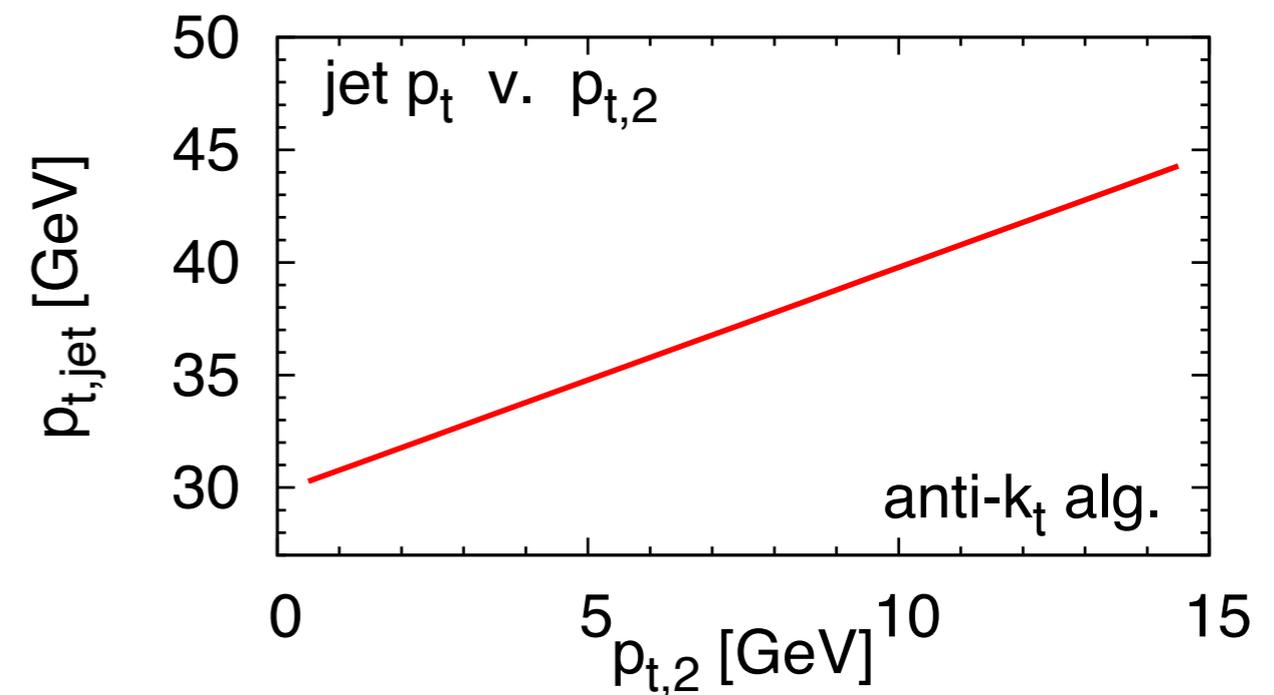
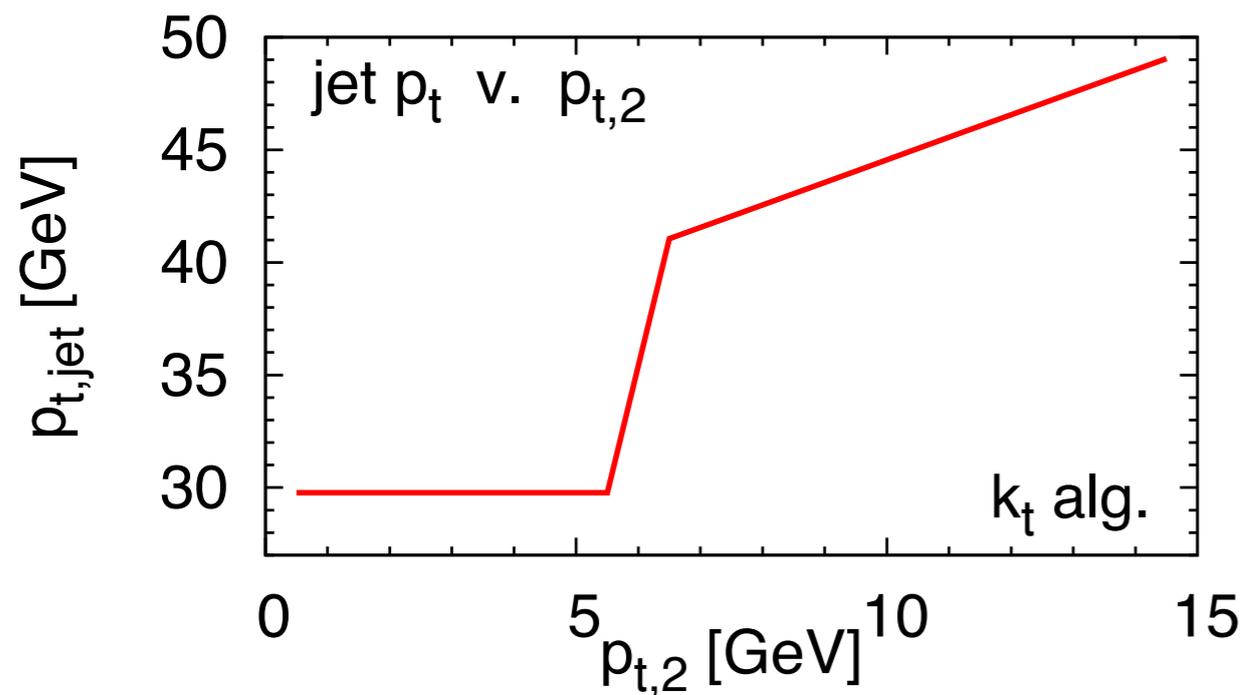
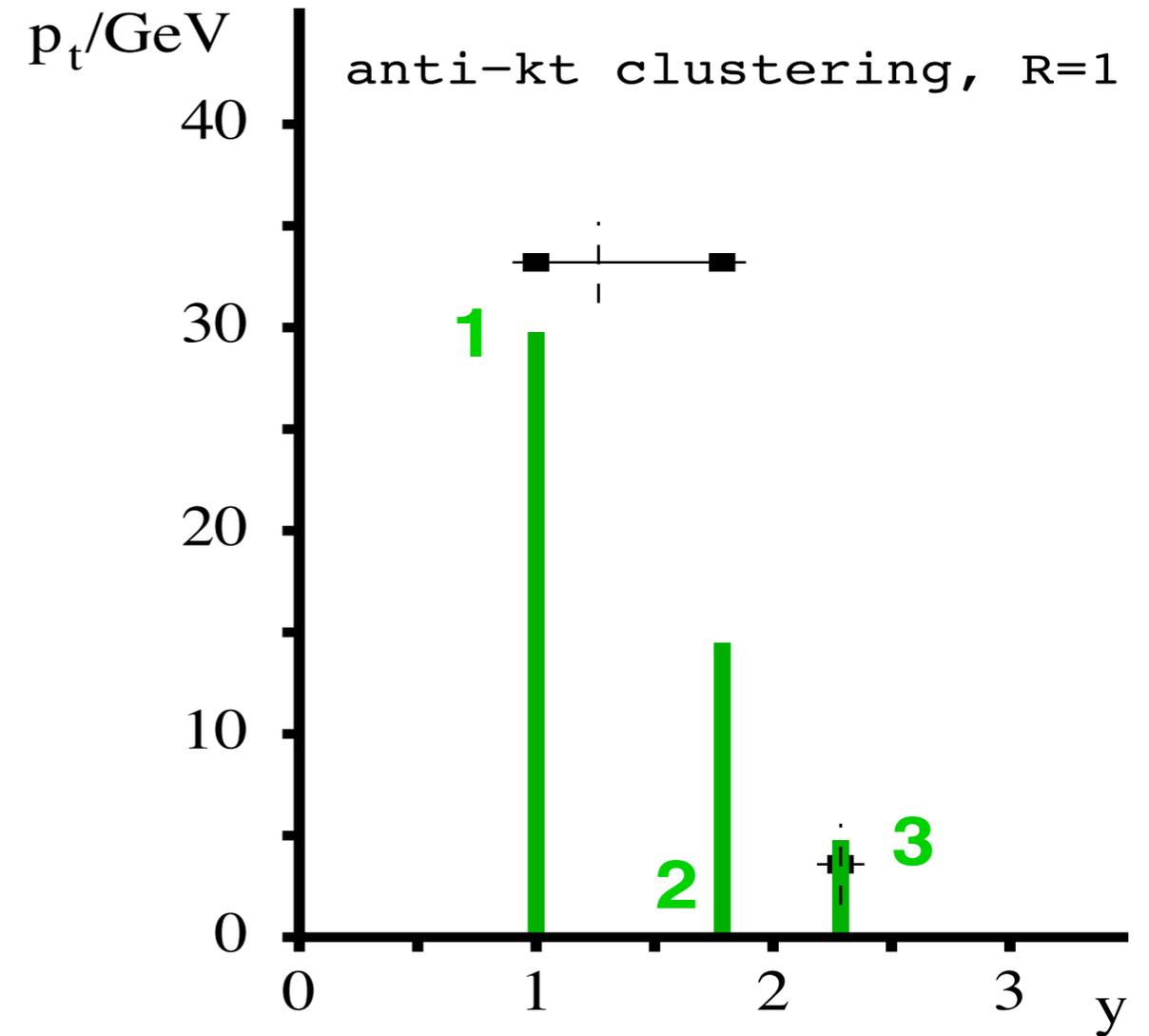
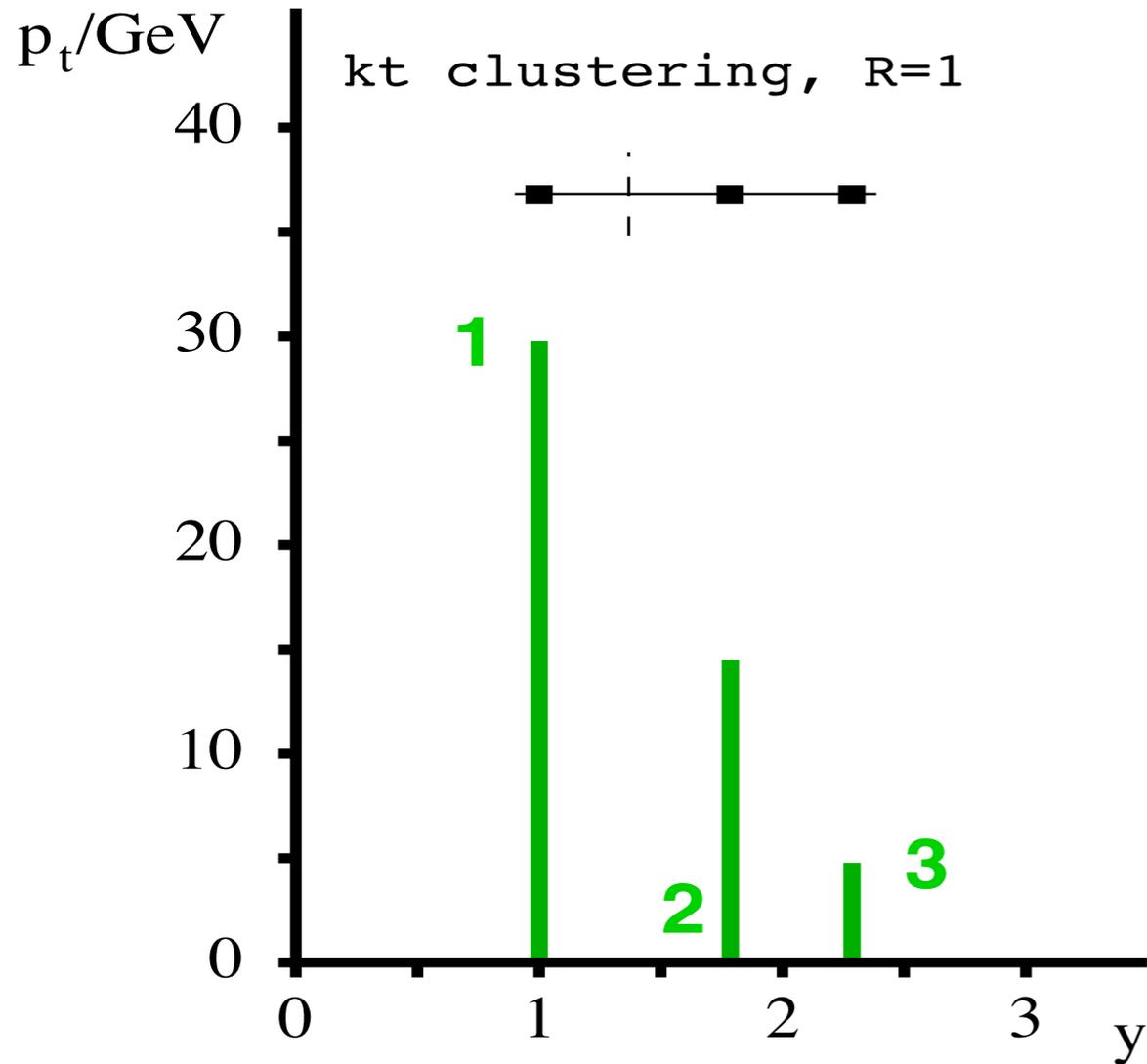
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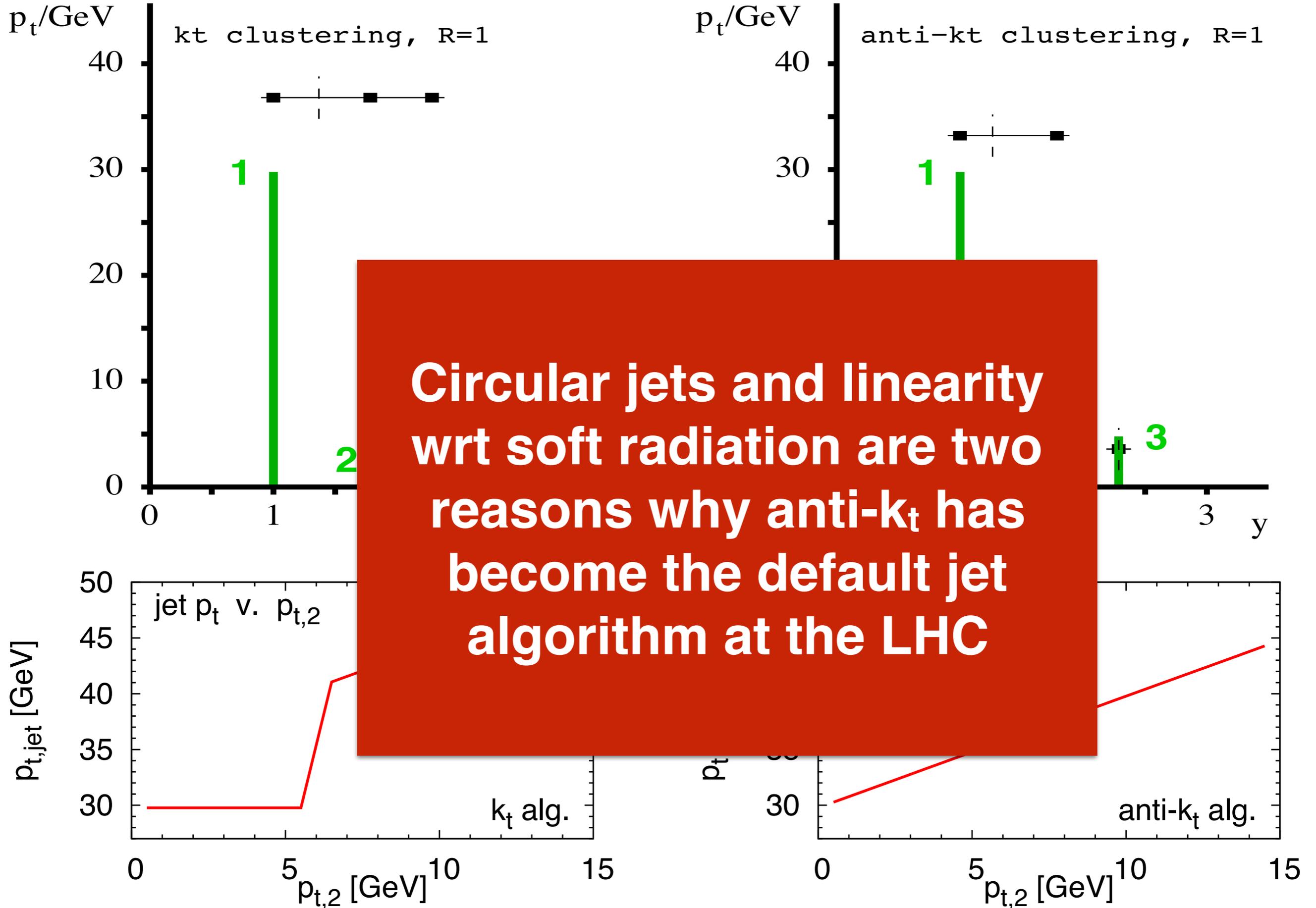


# Linearity: $k_t$ v. anti- $k_t$



# Linearity: $k_t$ v. anti- $k_t$





**Circular jets and linearity wrt soft radiation are two reasons why anti- $k_t$  has become the default jet algorithm at the LHC**

```
// specify a jet definition
double R = 0.4
JetDefinition jet_def(antikt_algorithm, R);
```

jet\_algorithm can be any one of the four IRC safe pp-collider algorithms, or also a variety of e<sup>+</sup>e<sup>-</sup> algorithms, both native and plugins

```
// specify the input particles
vector<PseudoJet> input_particles = . . .;
```

```
// specify a jet definition
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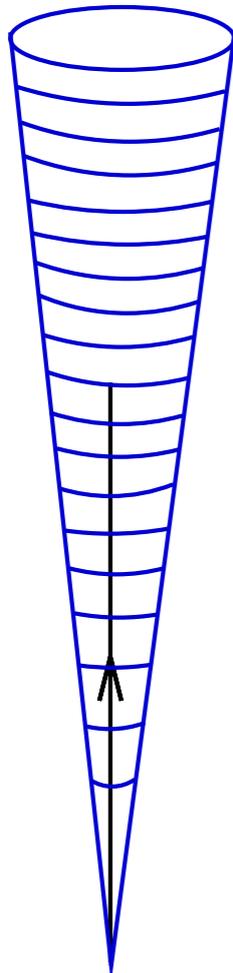
```
// specify the input particles
vector<PseudoJet> input_particles = . . .;
```

```
// extract the jets
vector<PseudoJet> jets = jet_def(input_particles);

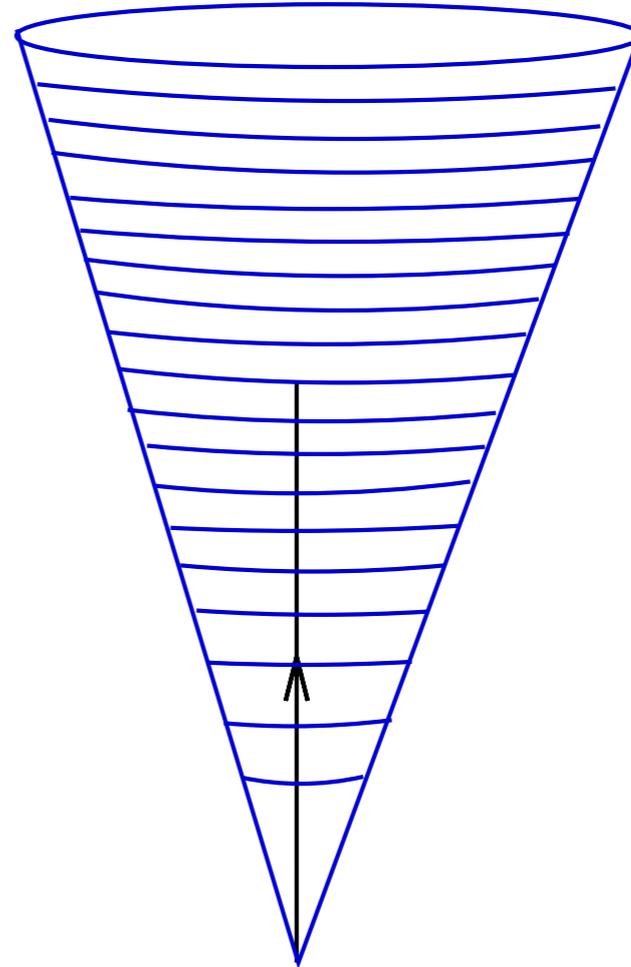
// pt of hardest jet
double pt_hardest = jets[0].pt();

// constituents of hardest jet
vector<PseudoJet> constituents = jets[0].constituents();
```

## Small jet radius

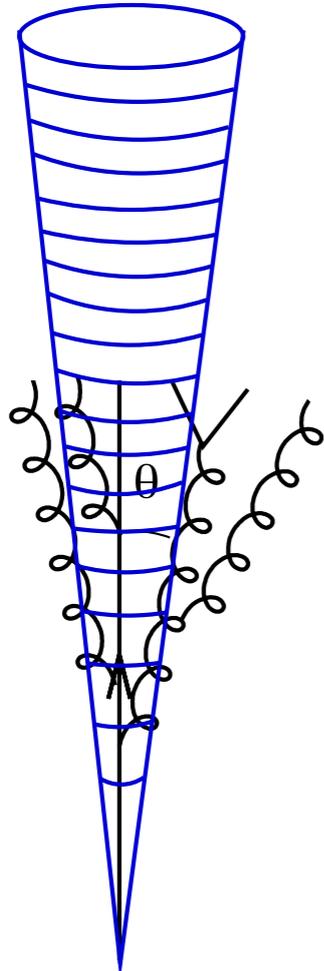


## Large jet radius

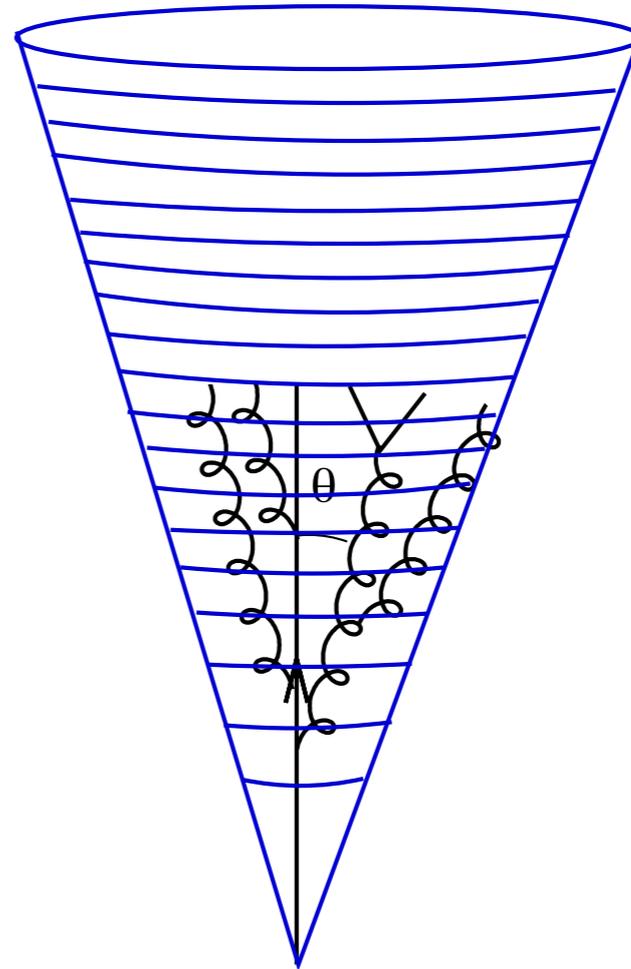


single parton @ LO: **jet radius irrelevant**

## Small jet radius

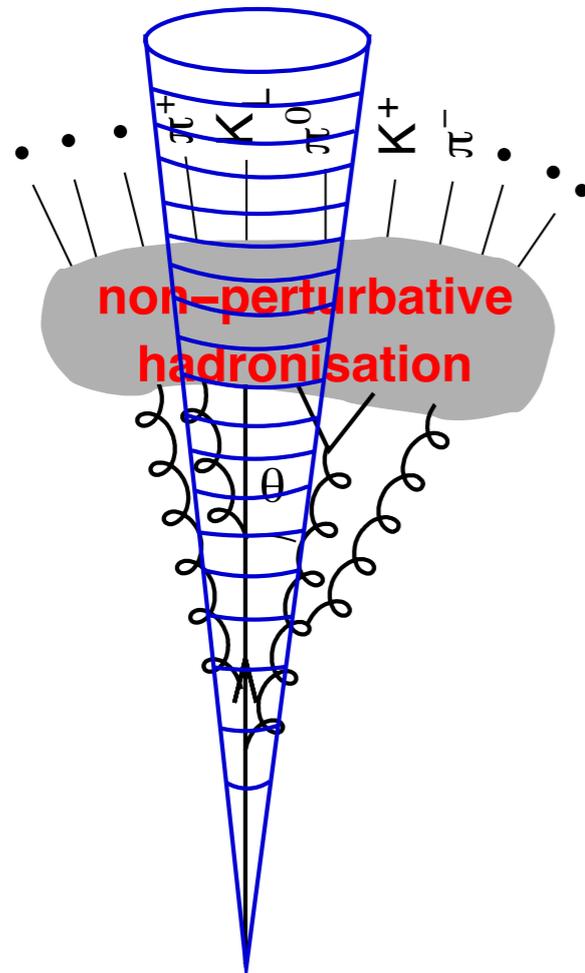


## Large jet radius

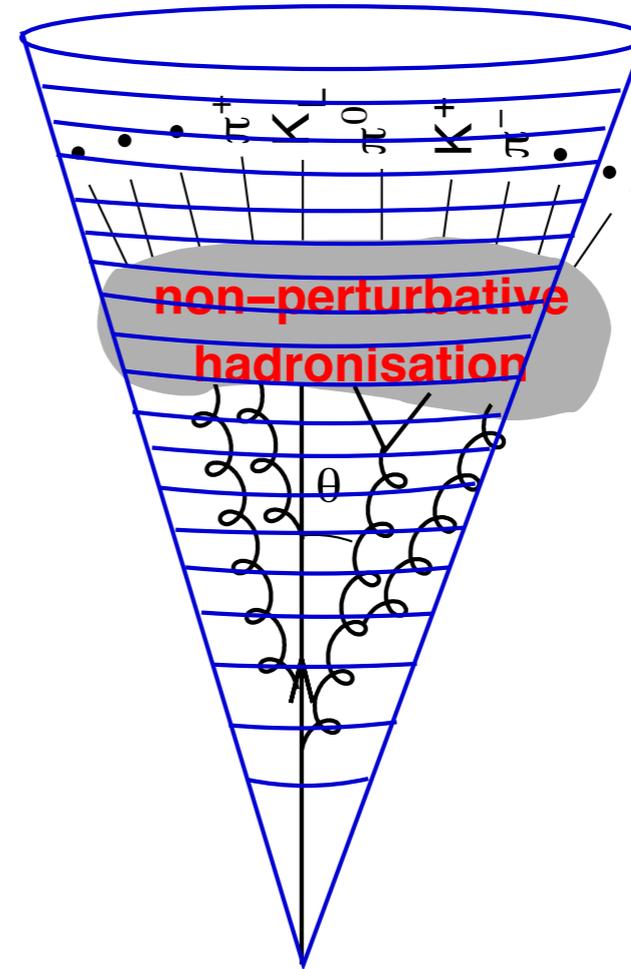


perturbative fragmentation: **large jet radius better**  
(it captures more)

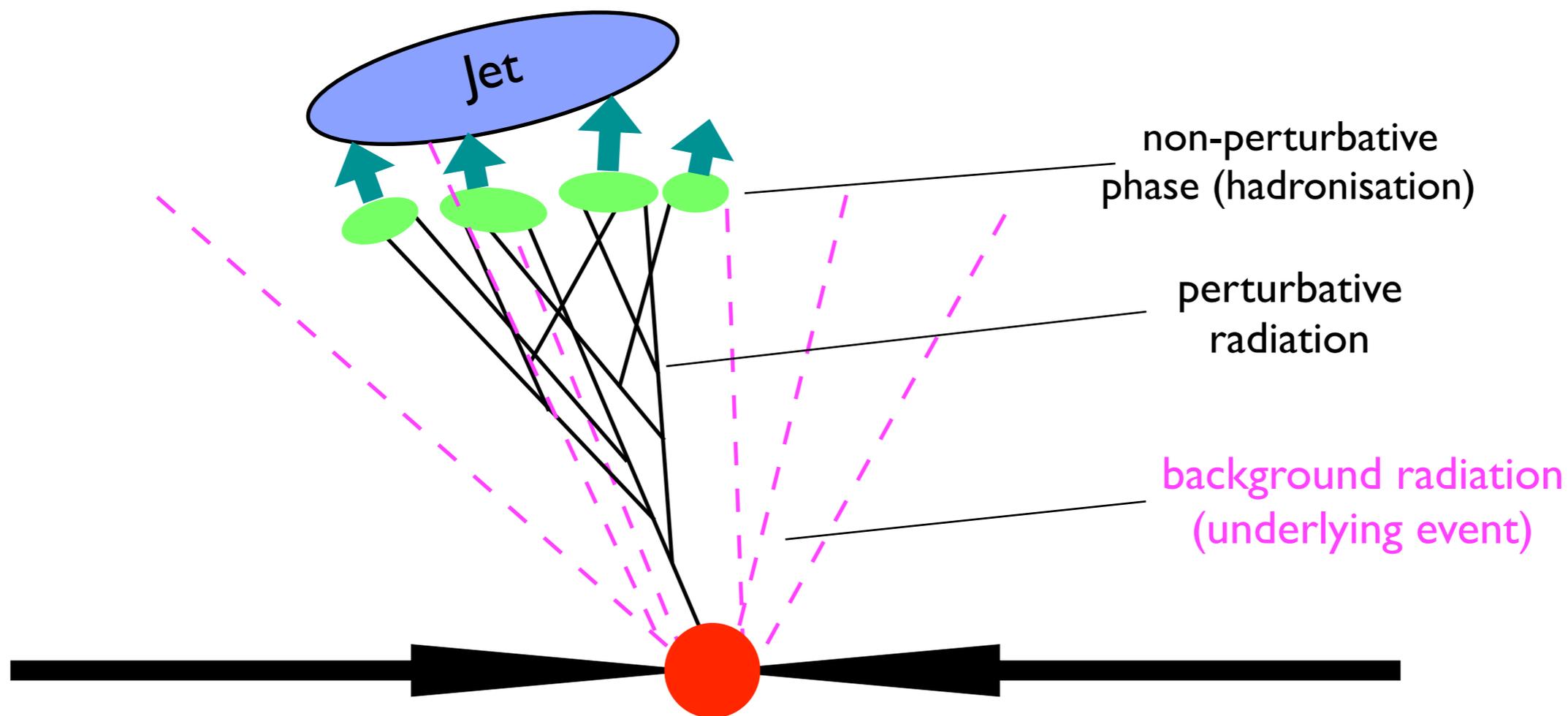
## Small jet radius

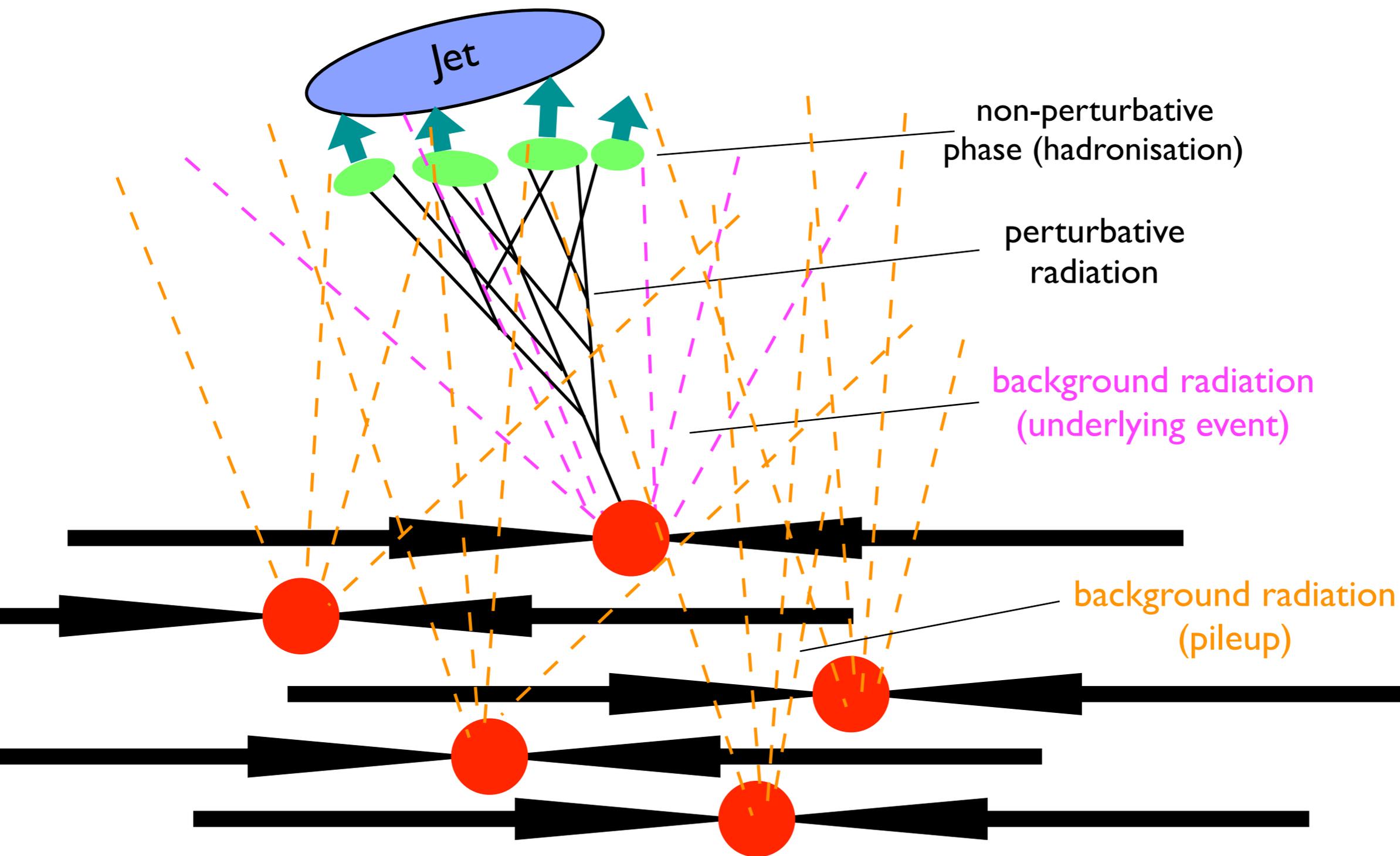


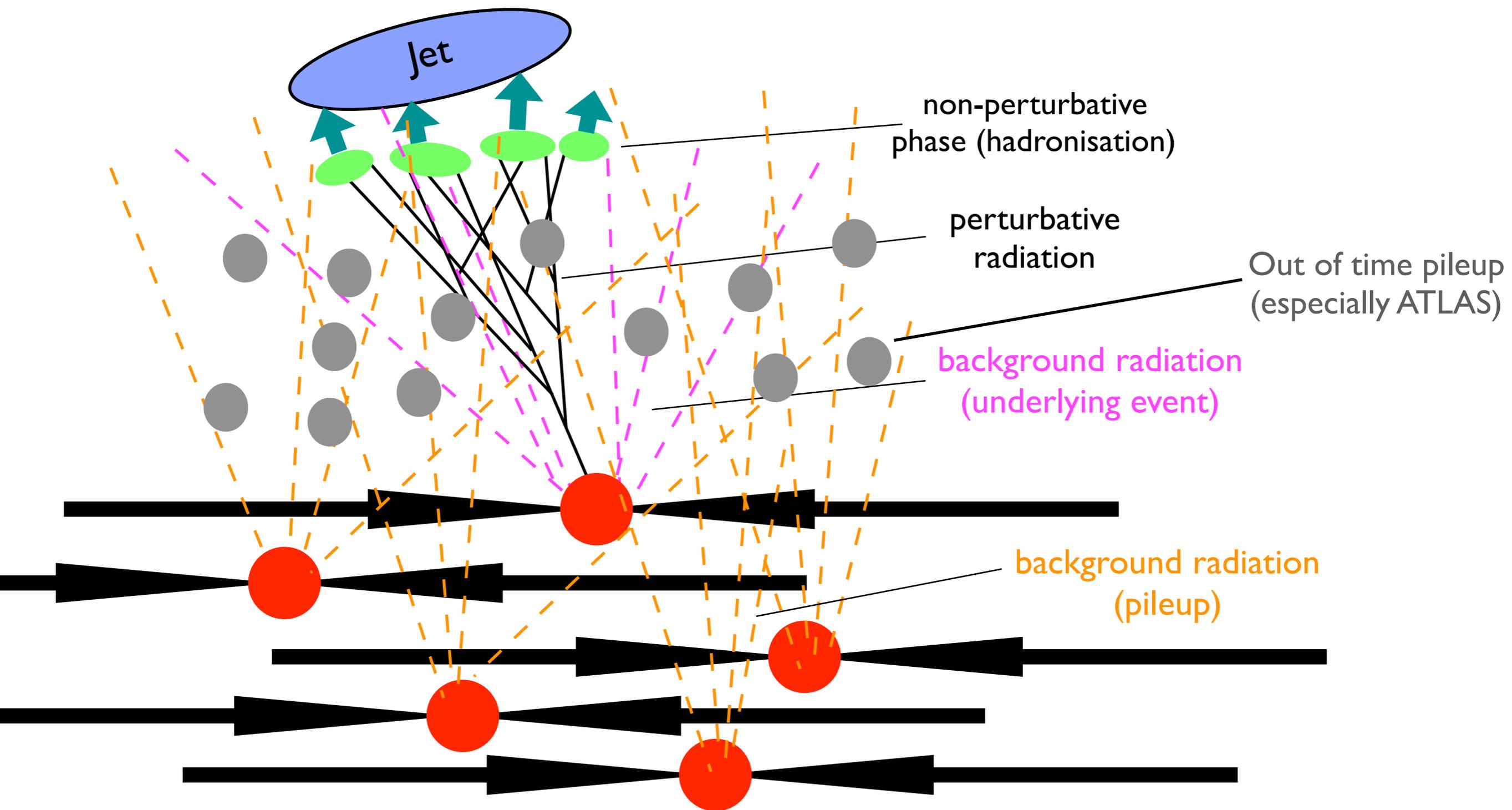
## Large jet radius



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



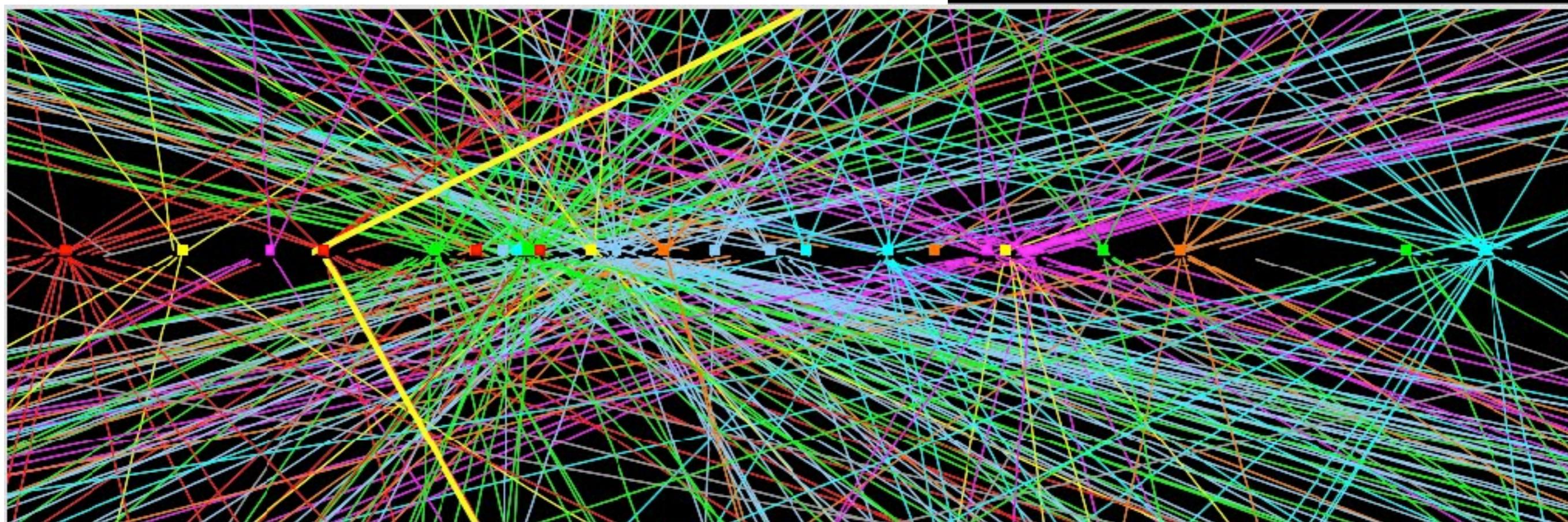
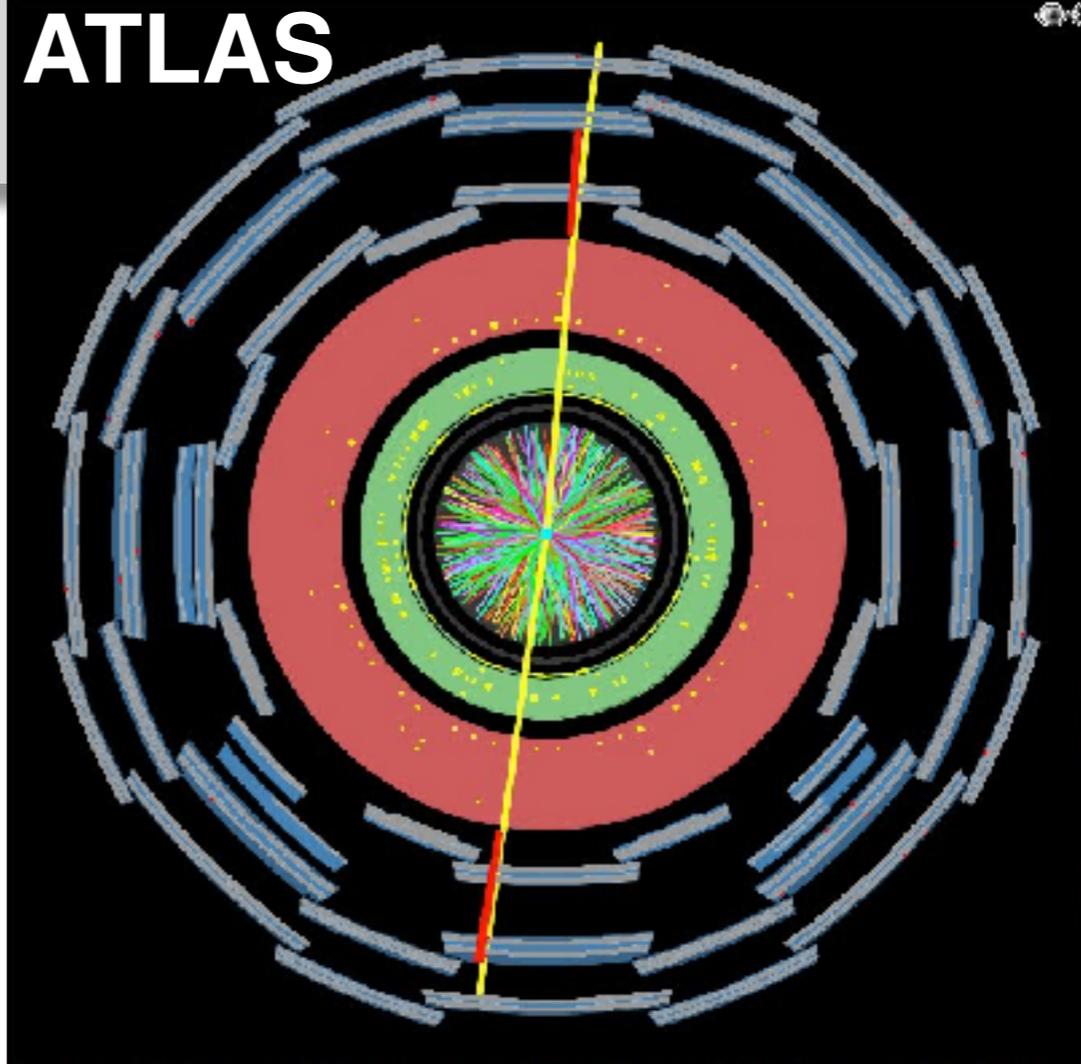




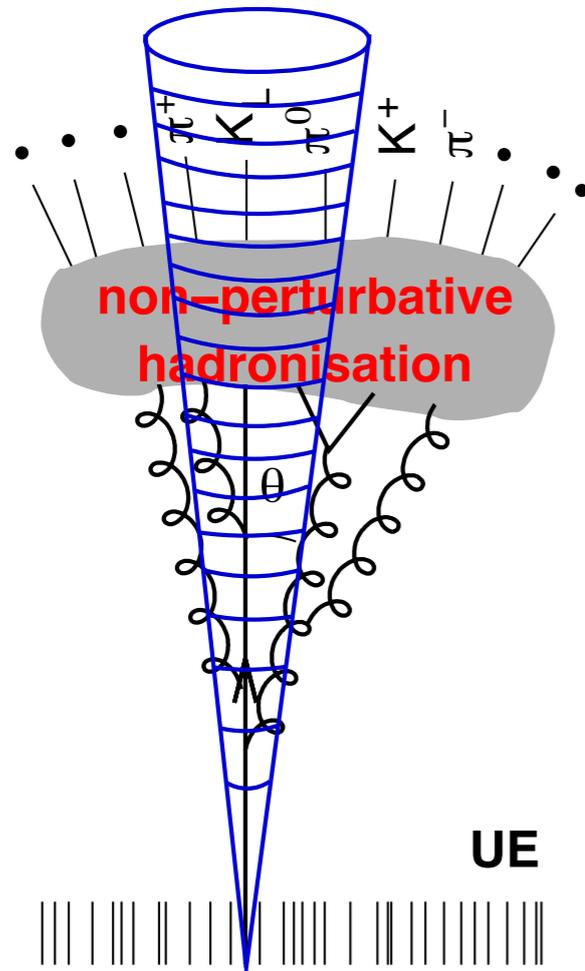
# Pileup for real

a few cm

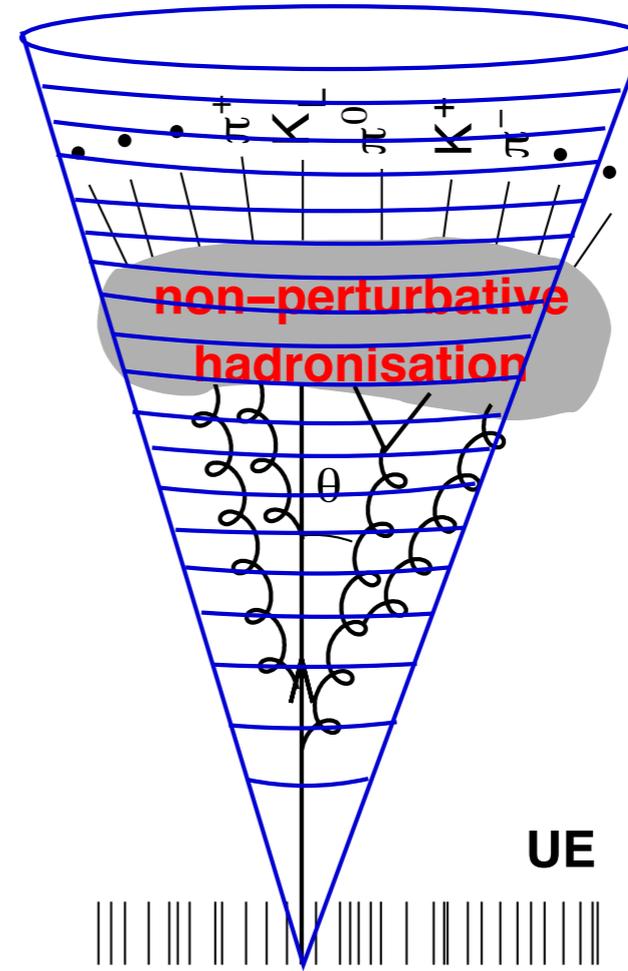
~ 20 m



## Small jet radius

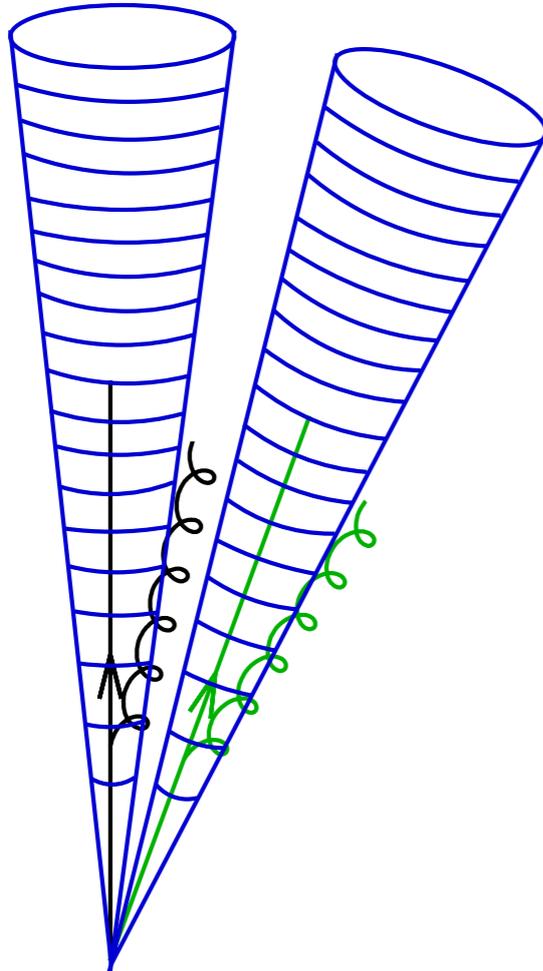


## Large jet radius

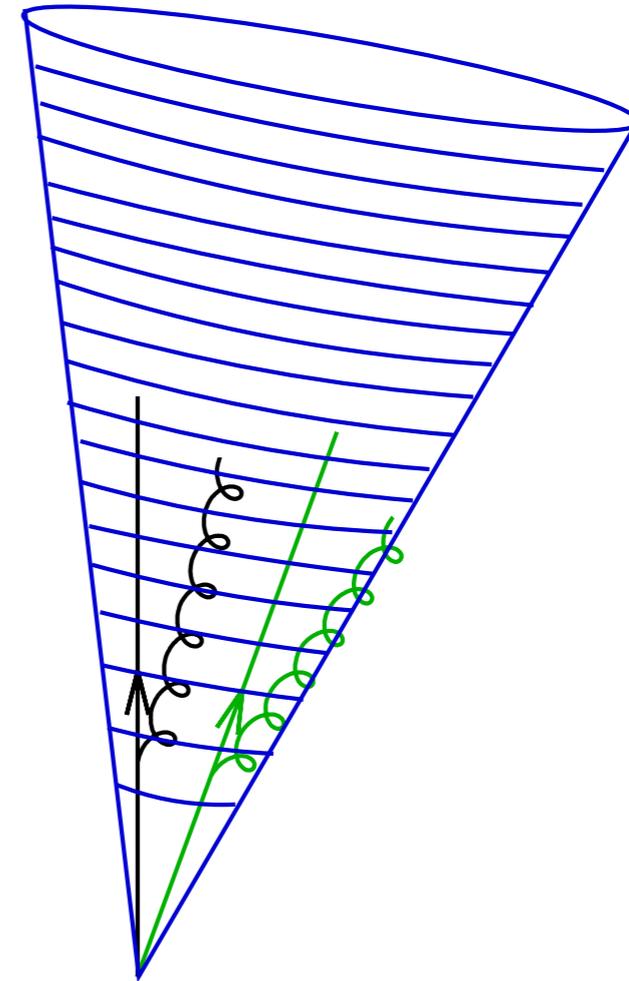


underlying ev. & pileup “noise”: **small jet radius better**  
(it captures less)

**Small jet radius**



**Large jet radius**



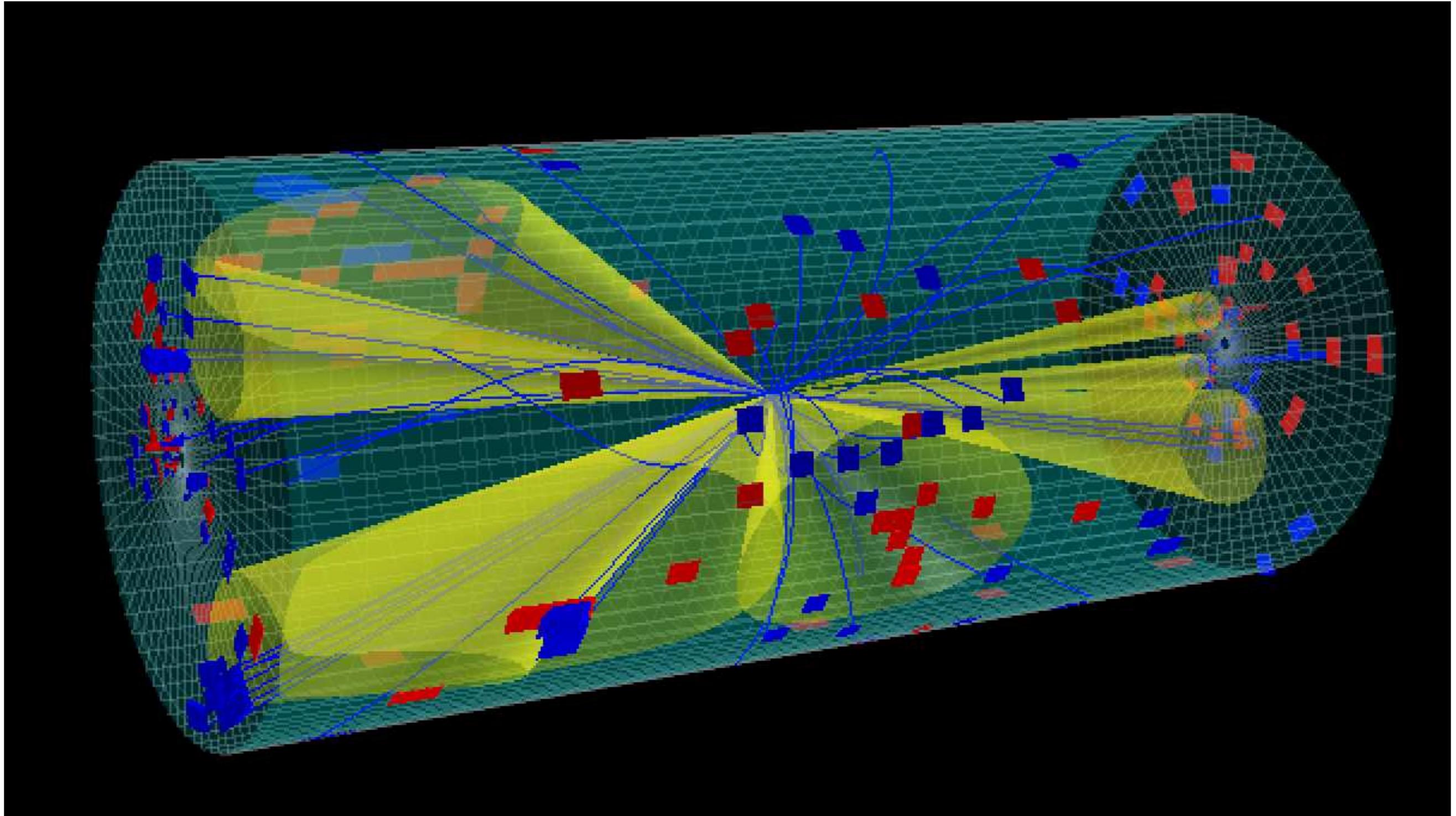
multi-hard-parton events: **small jet radius better**  
(it resolves partons more effectively)

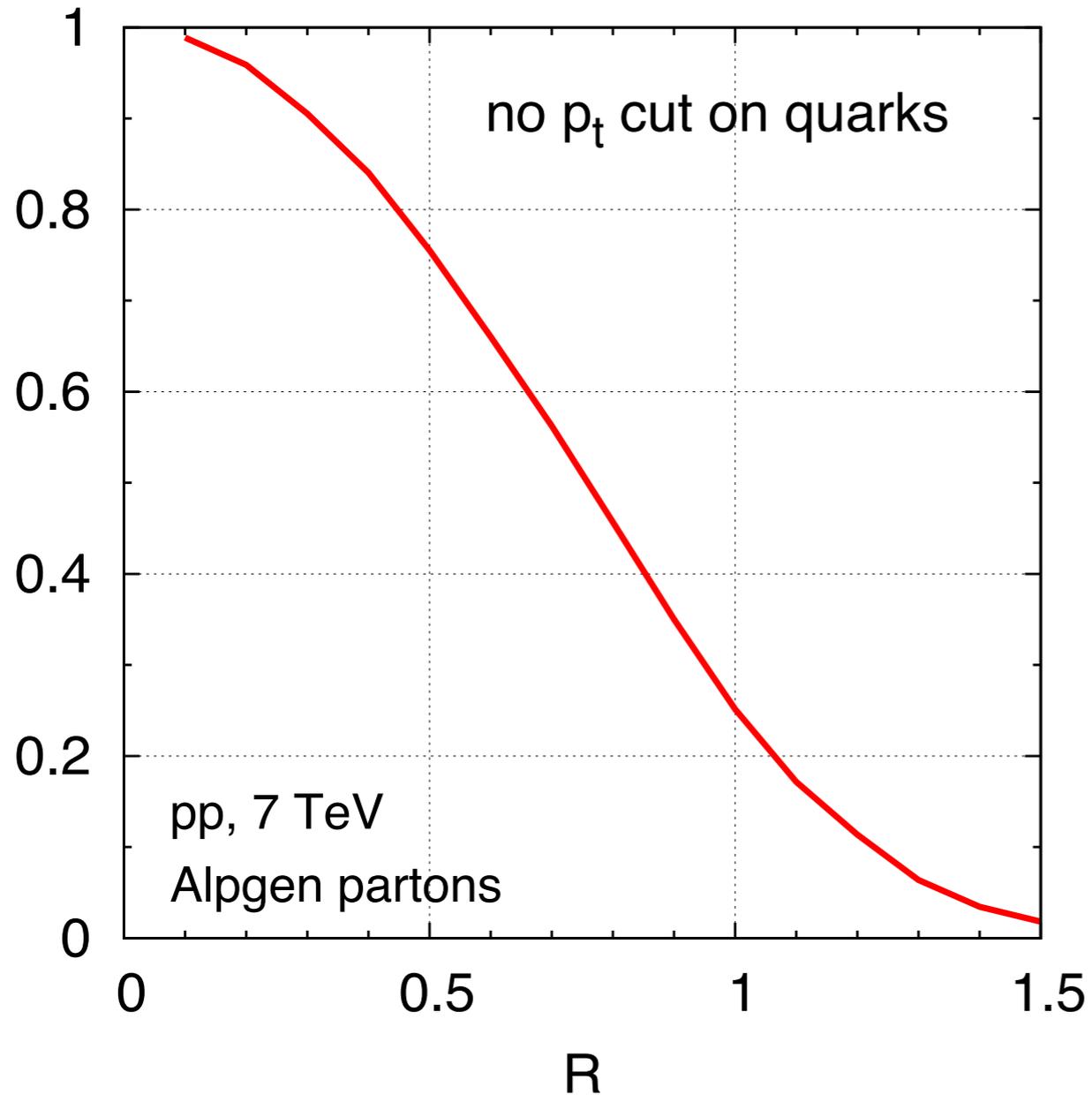
Can we capture all quarks and gluons?

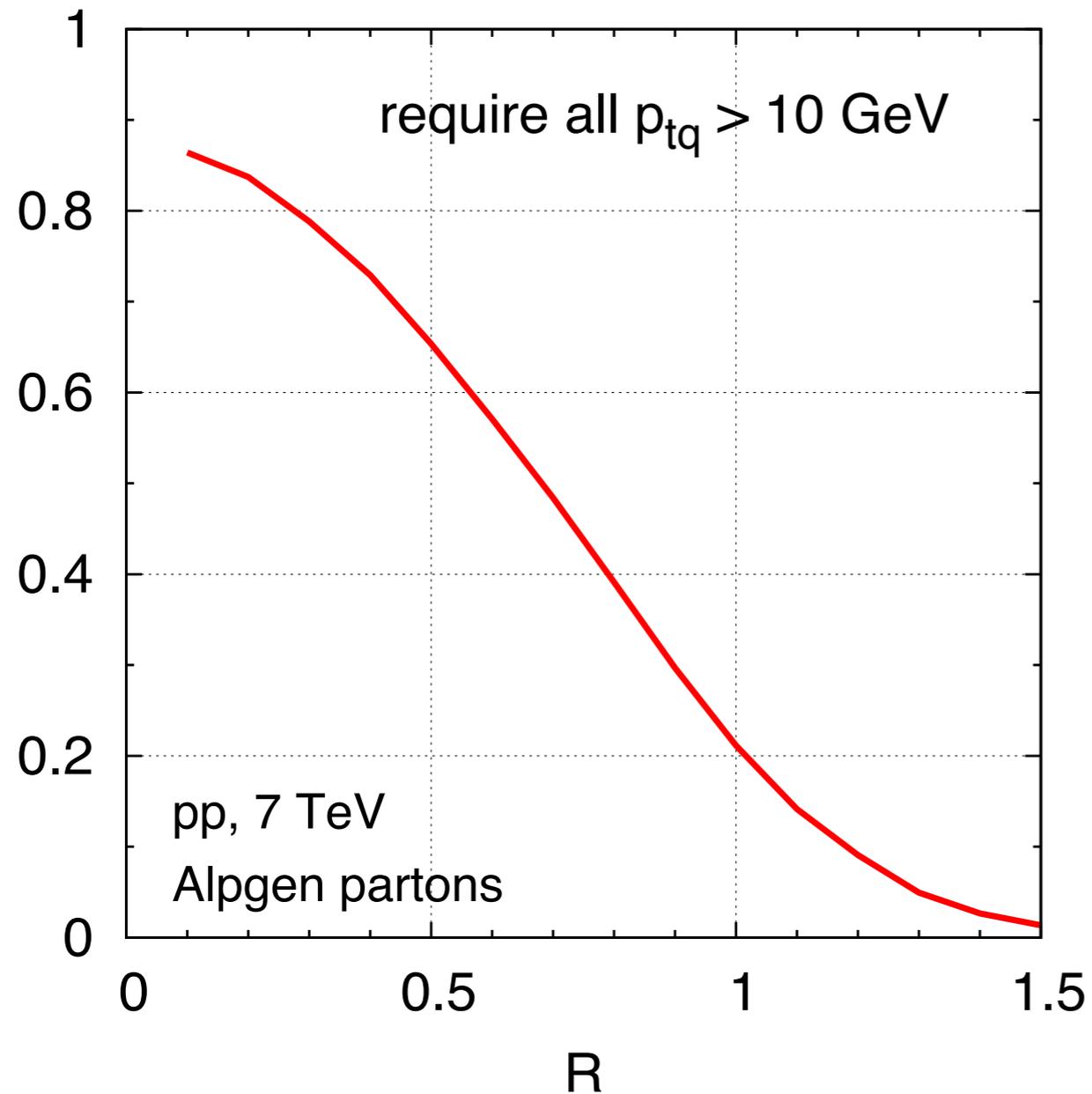
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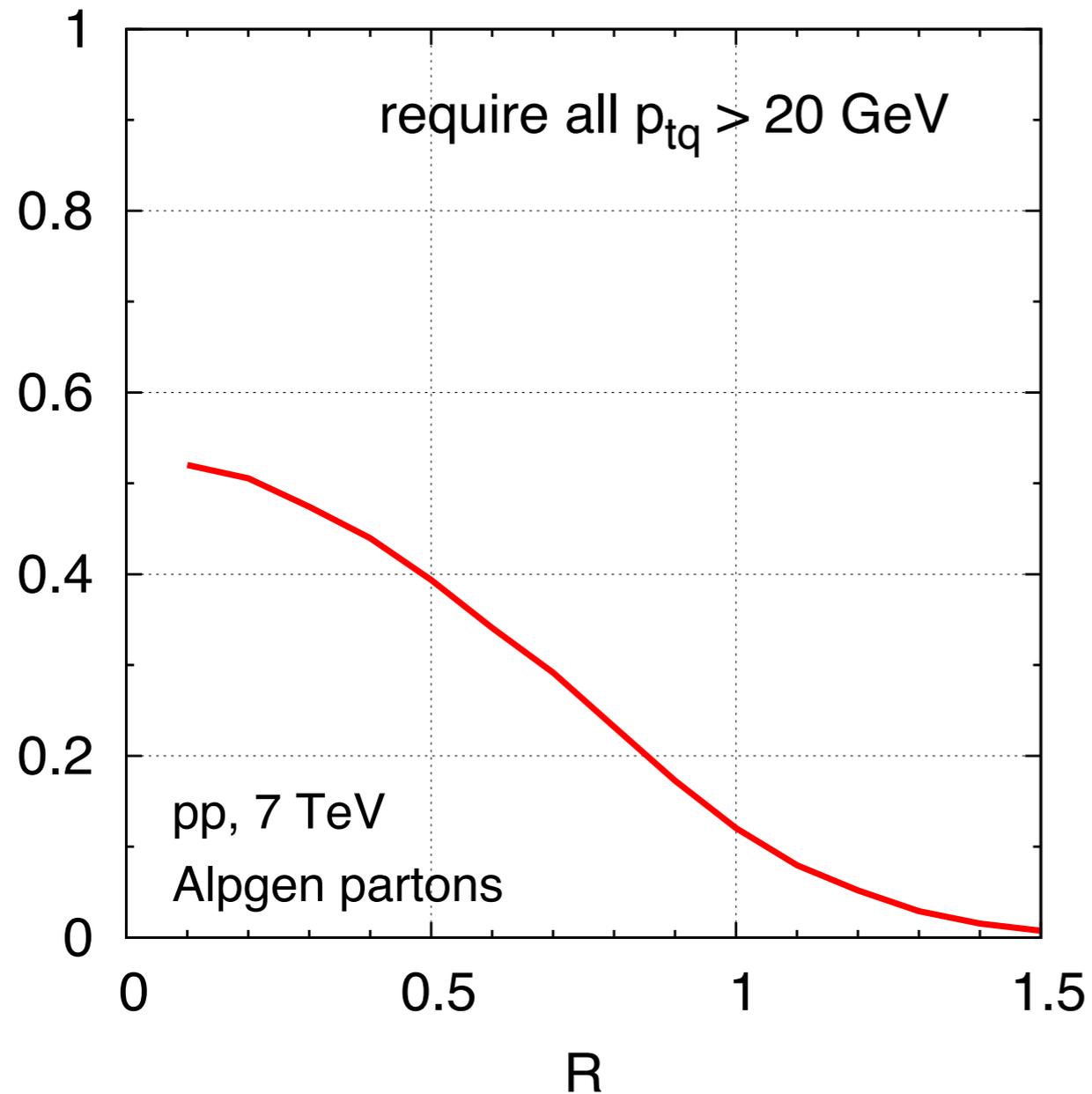
$$pp \rightarrow t\bar{t}$$

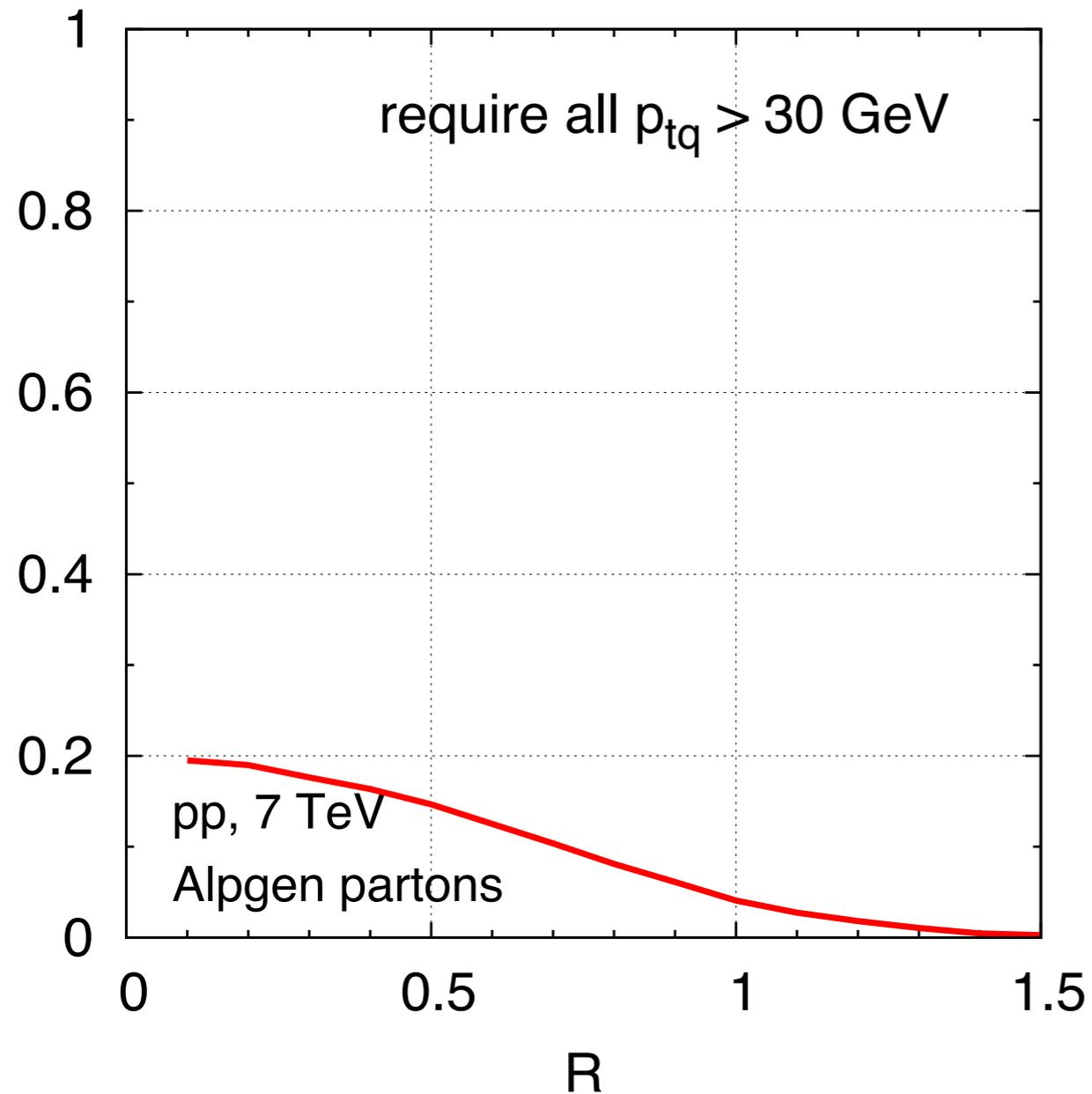
simulated with Pythia, displayed with Delphes

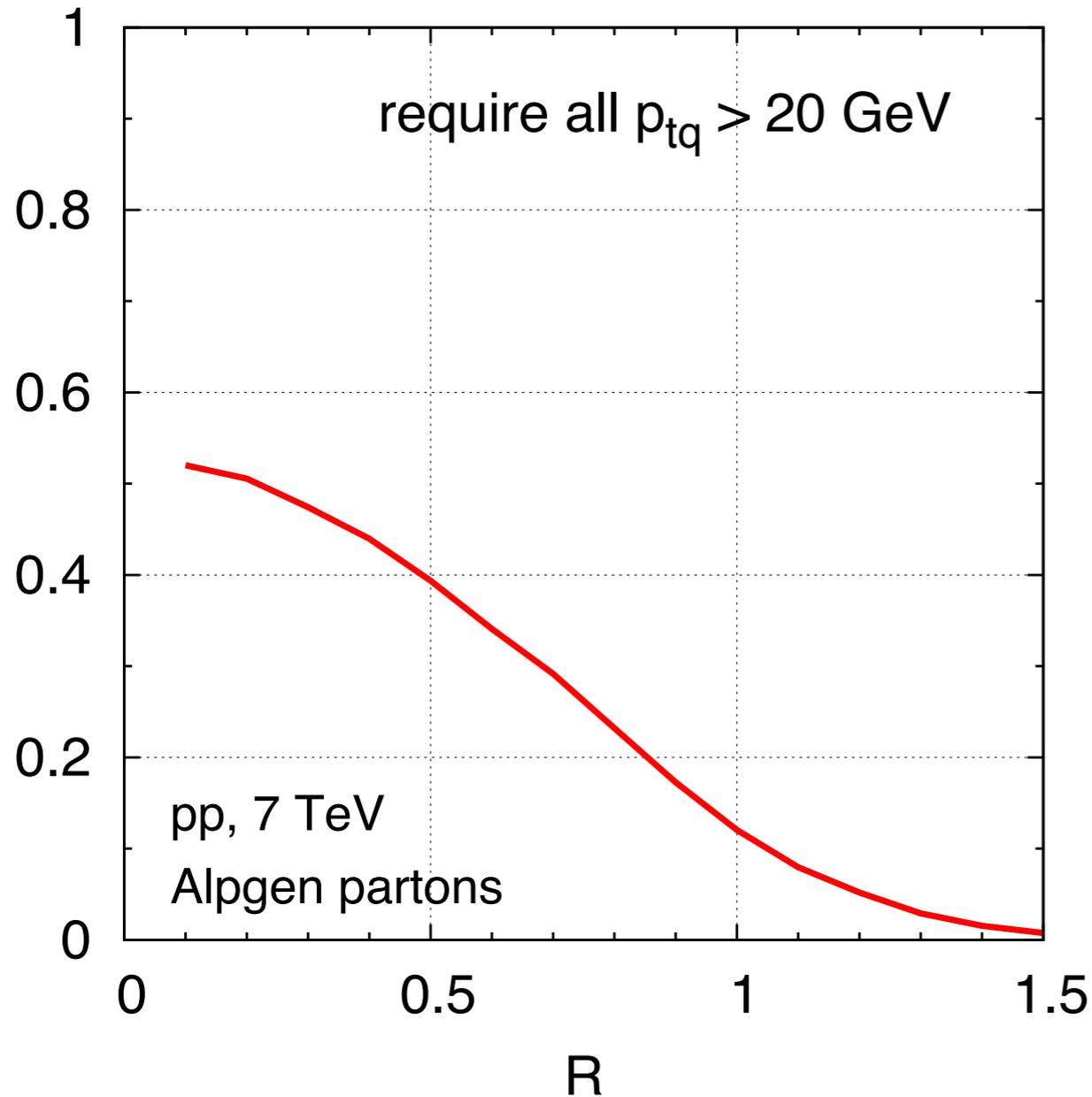


**Alpgen  $pp \rightarrow t\bar{t} \rightarrow 6q$** fraction of  $pp \rightarrow t\bar{t} \rightarrow 6q$  events with all  $R_{qq} > R$ 

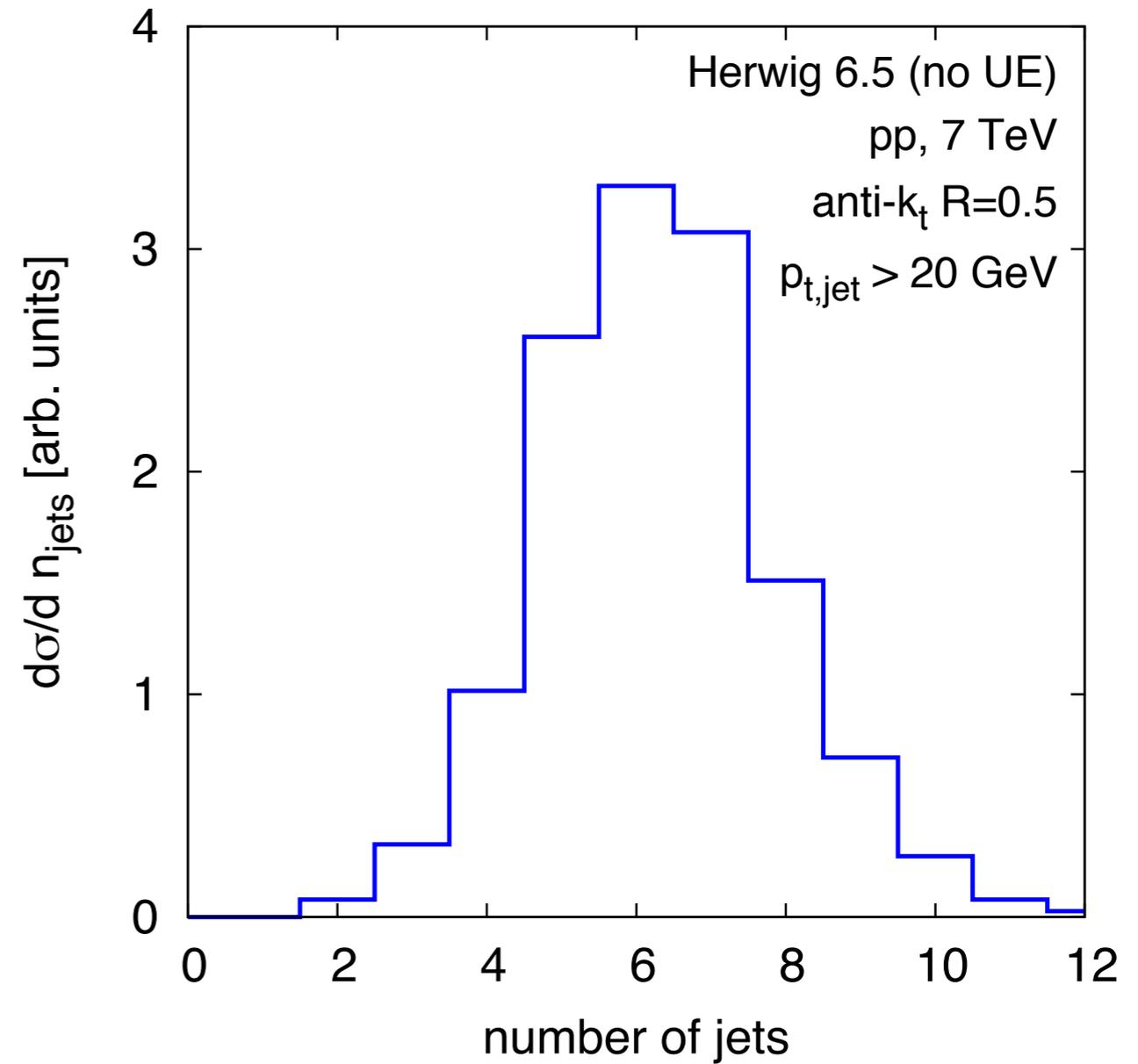
**Alpgen  $pp \rightarrow t\bar{t} \rightarrow 6q$** fraction of  $pp \rightarrow t\bar{t} \rightarrow 6q$  events with all  $R_{qq} > R$ 

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

Distribution of number of jets



# Two things that make jets@LHC special

The large hierarchy of scales

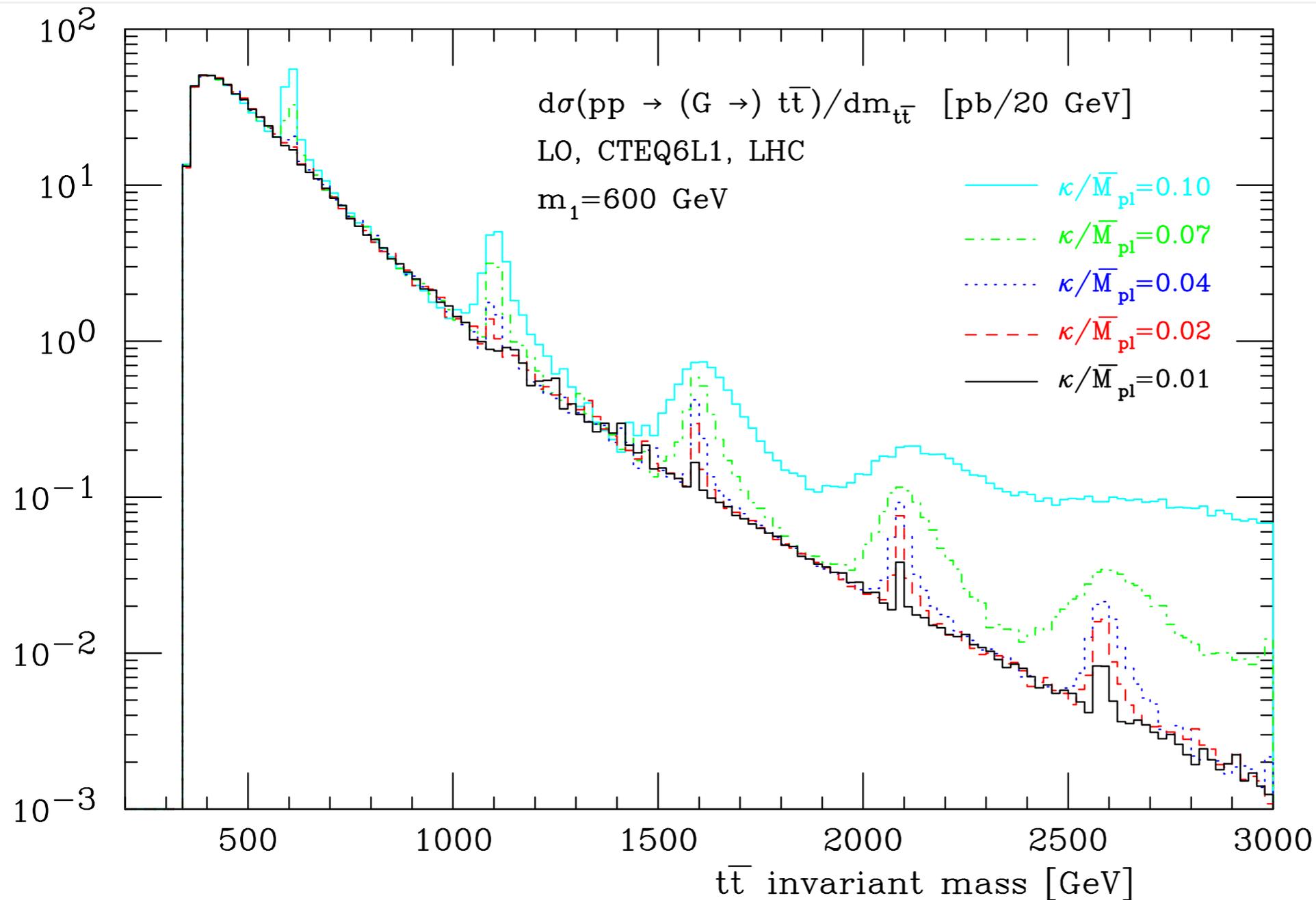
$$\sqrt{s} \gg M_{EW}$$

The huge pileup

$$n_{pileup} \sim 20 - 40 \quad (\rightarrow 140 \text{ at HL-LHC})$$

*[These involve two opposite extremes: low  $p_t$  and high  $p_t$ , which nevertheless talk to each other]*

# e.g. $t\bar{t}$ resonances

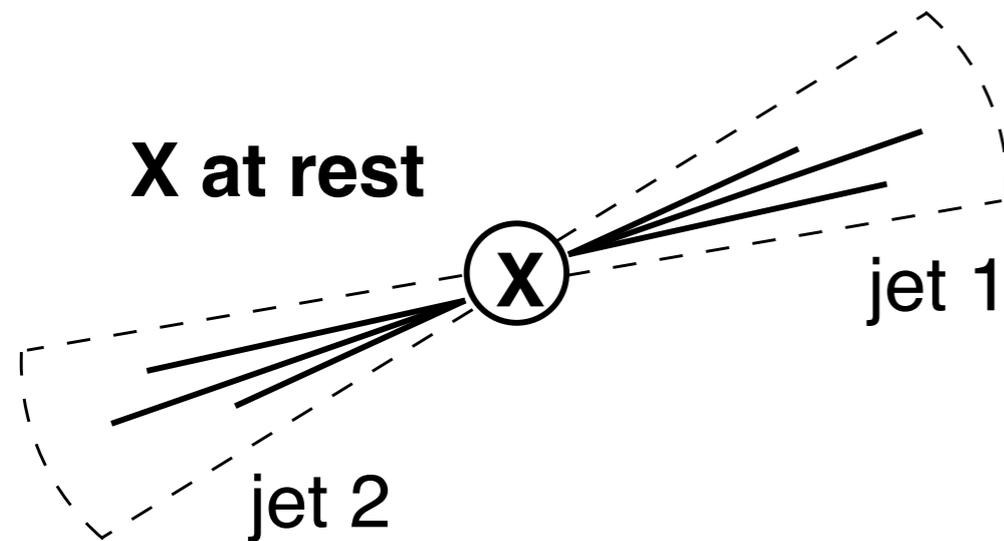


RS KK resonances  $\rightarrow t\bar{t}$ , from Frederix & Maltoni, 0712.2355

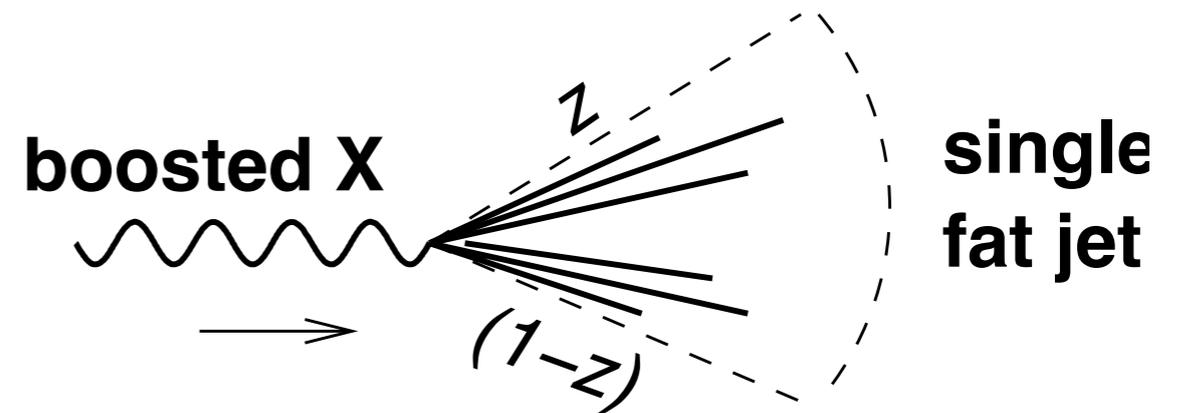
NB: QCD dijet spectrum is  $\sim 10^3$  times  $t\bar{t}$

# Boosted EW scale objects

Normal analyses: two quarks from  $X \rightarrow q\bar{q}$  reconstructed as two jets



**High- $p_t$  regime: EW object X is boosted, decay is collimated,  $q\bar{q}$  both in same jet**



Happens for  $p_t \gtrsim 2m/R$   
 $p_t \gtrsim 320 \text{ GeV}$  for  $m = m_W$ ,  $R = 0.5$

Two widely used terms  
though there's not a  
consensus about  
what they mean

## Tagging

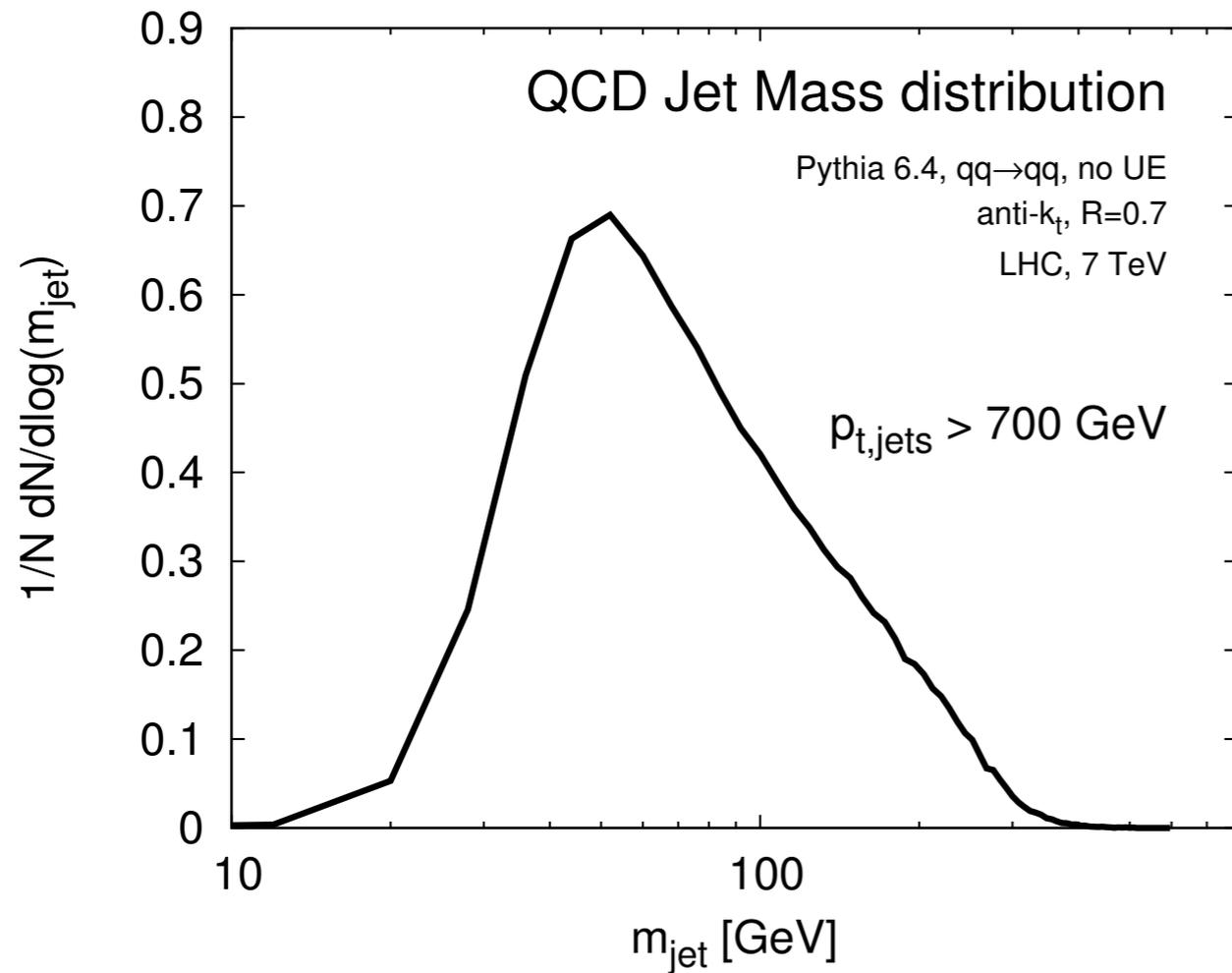
- reduces the background, leaves much of signal
- you can tag with underlying hard n-prong structure and based on radiation pattern

## Grooming

- improves signal mass resolution (removing pileup, etc.), without significantly changing background & signal event numbers

One core idea for  
tagging

# Inside the jet mass

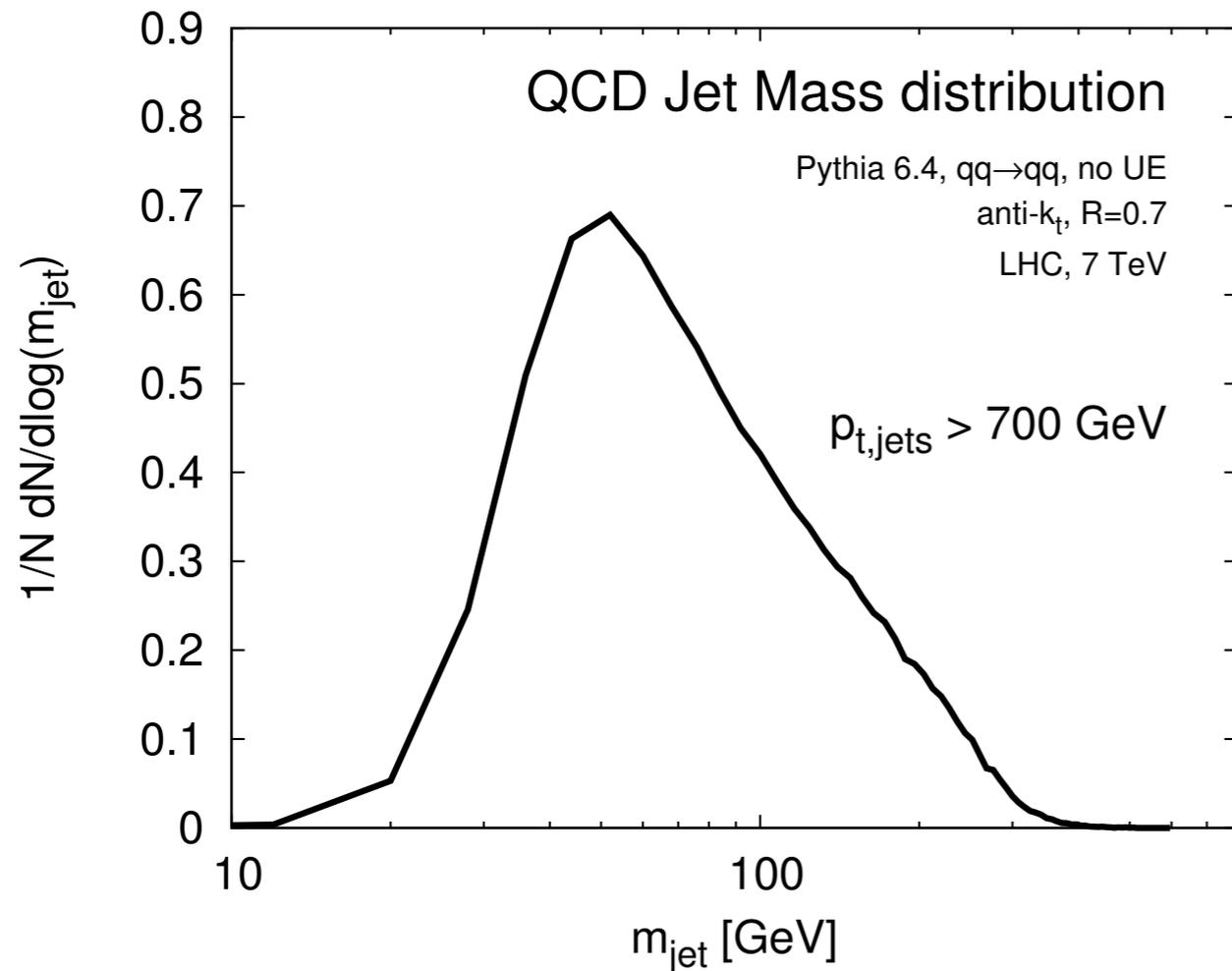


QCD jet mass distribution has the approximate

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

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

# Inside the jet mass



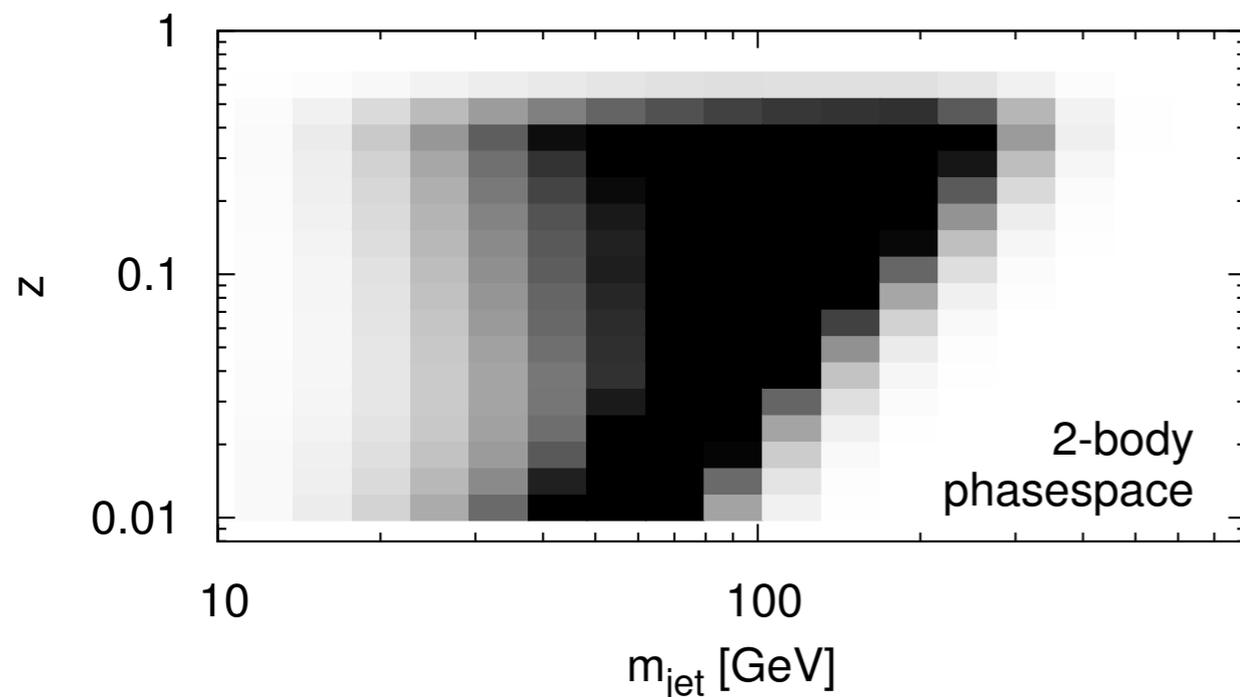
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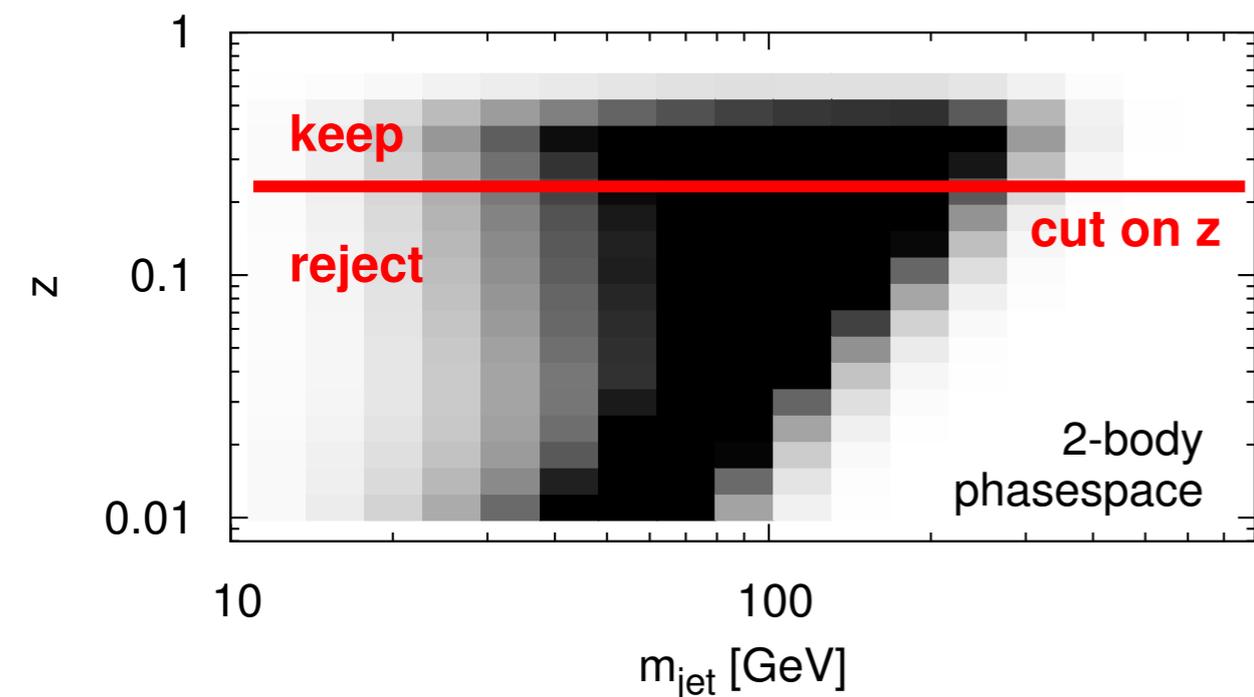
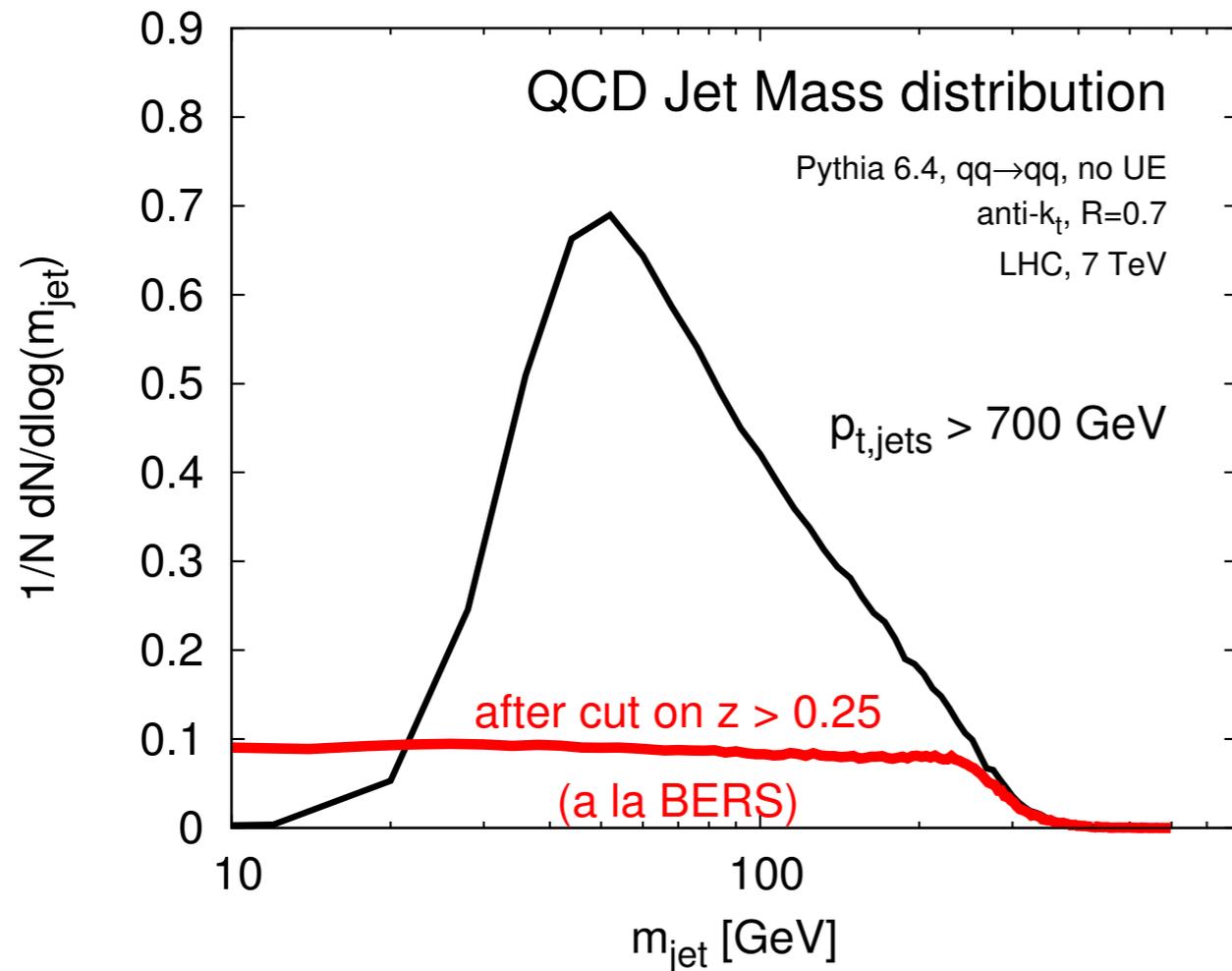
Work from '80s and '90s  
+ Almeida et al '08

The logarithm comes from integral over soft divergence of QCD:

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



# Inside the jet mass



QCD jet mass distribution has the approximate

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

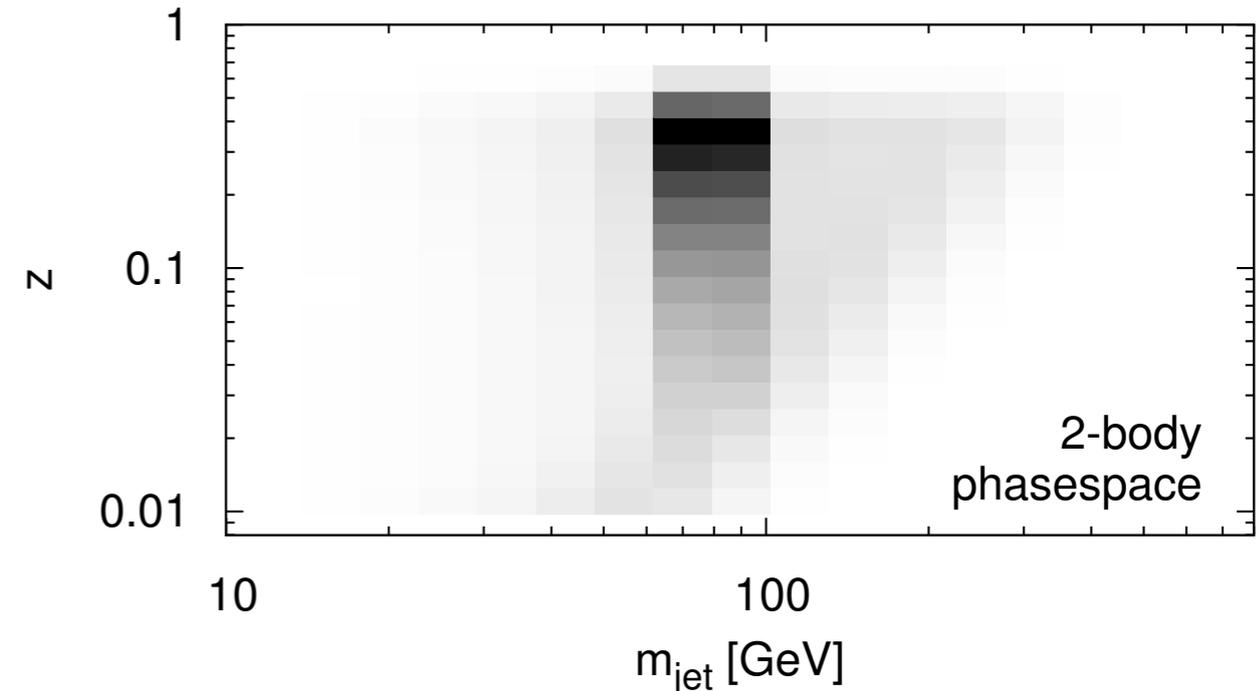
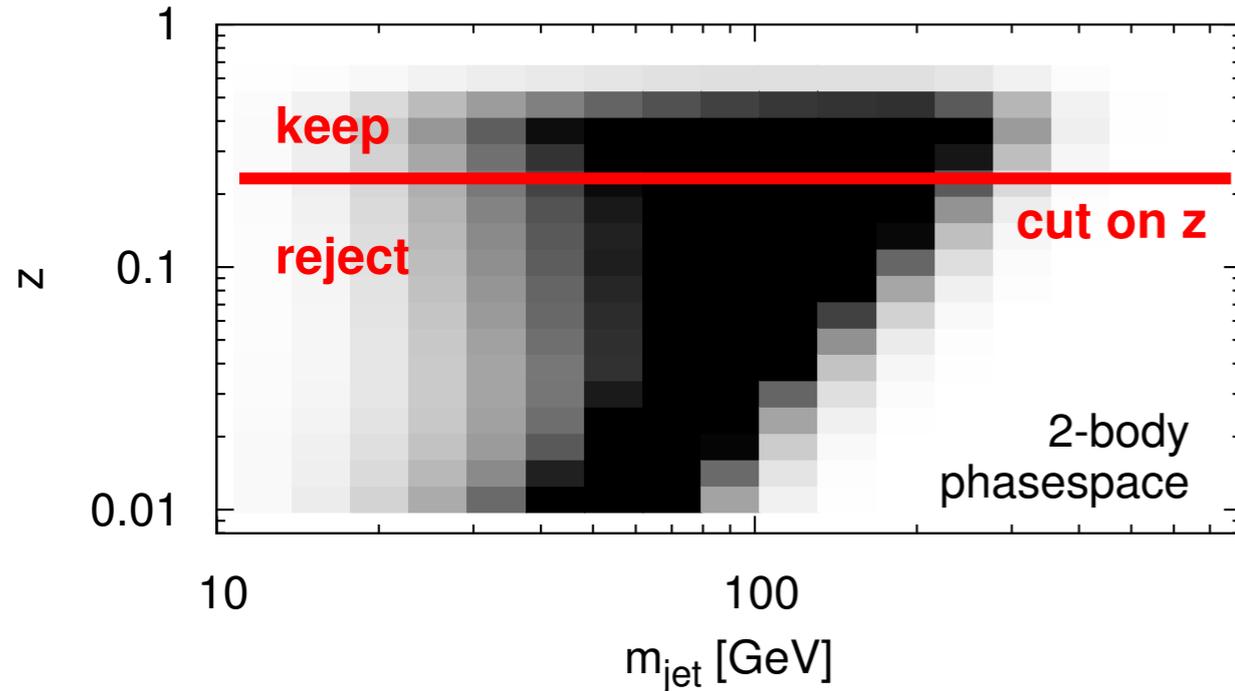
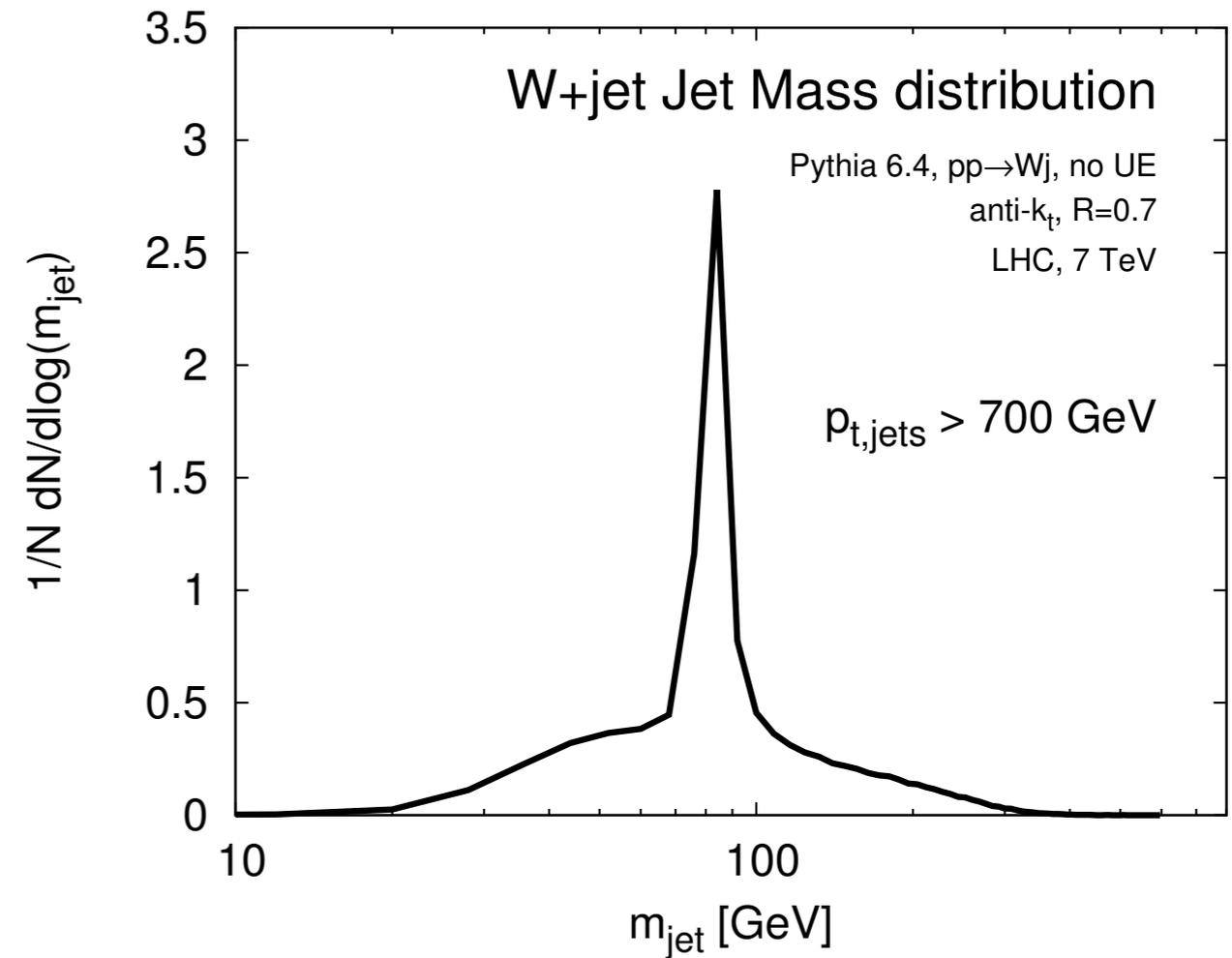
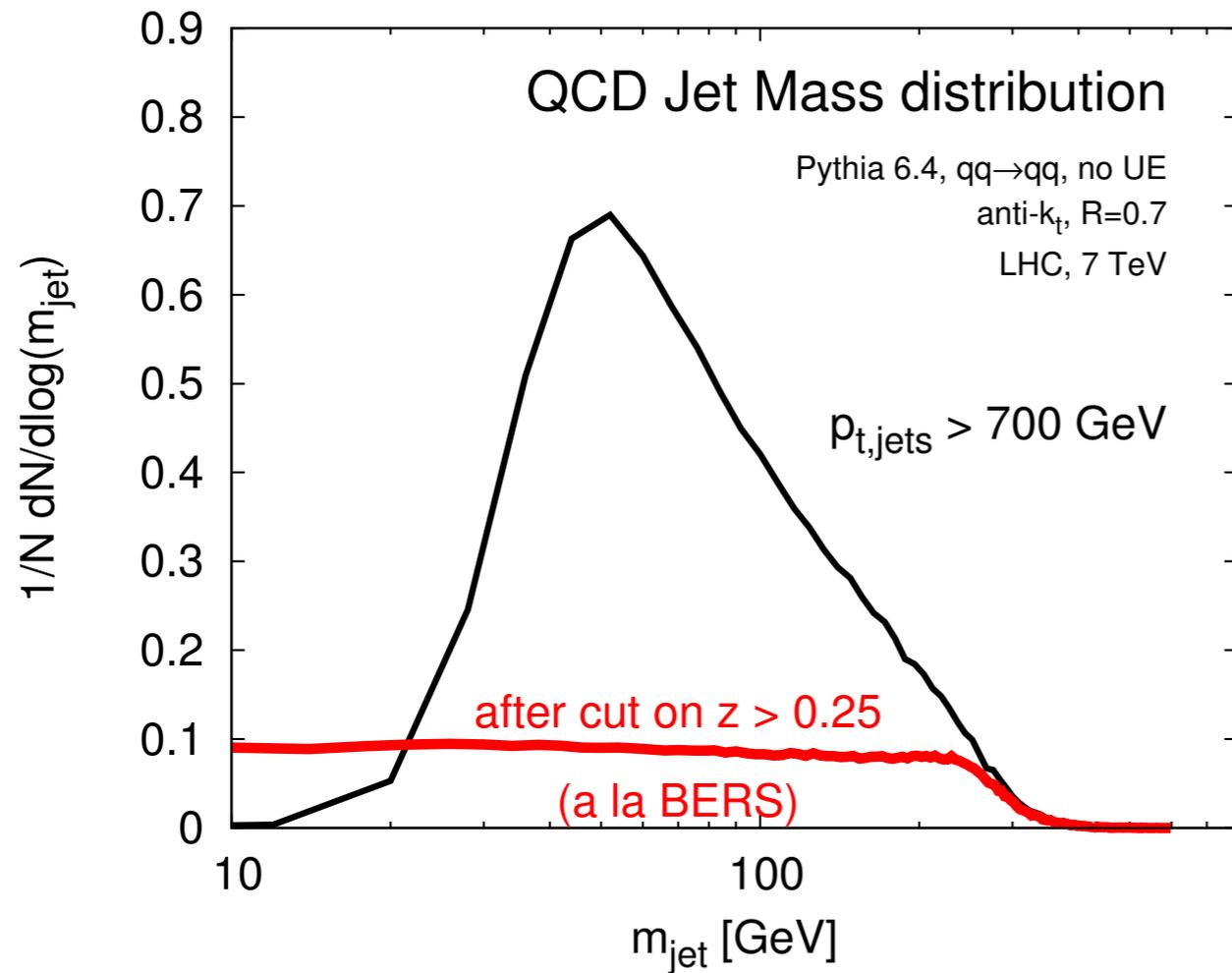
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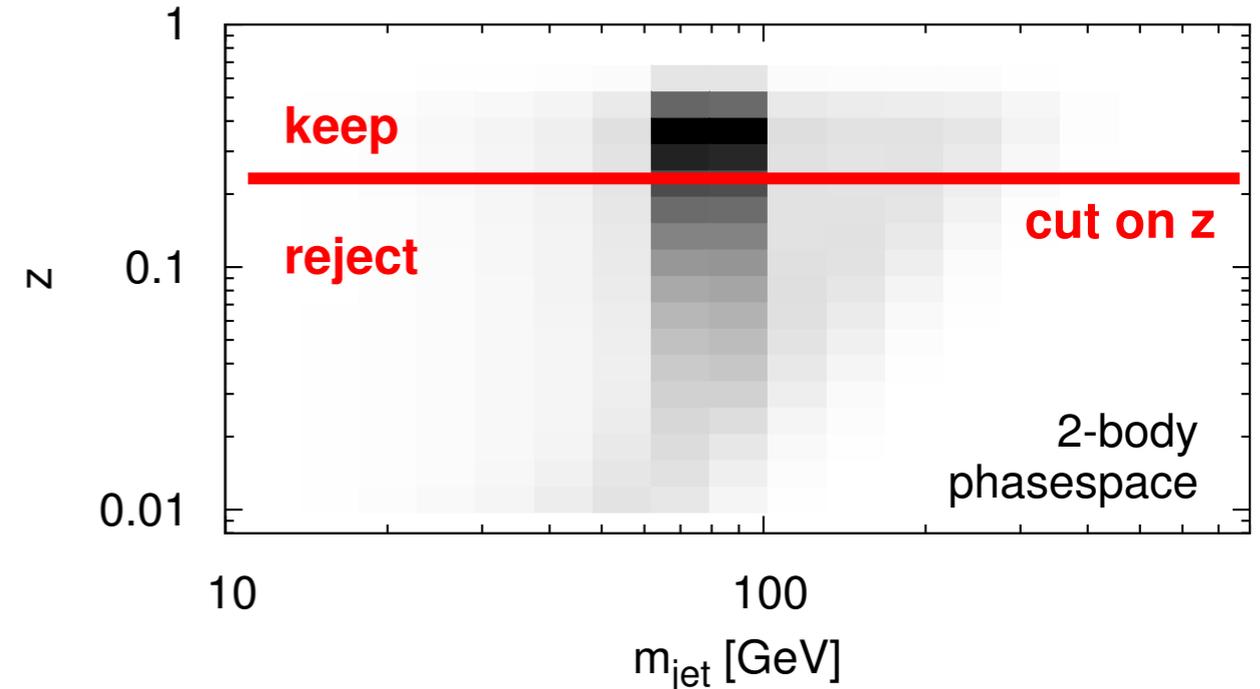
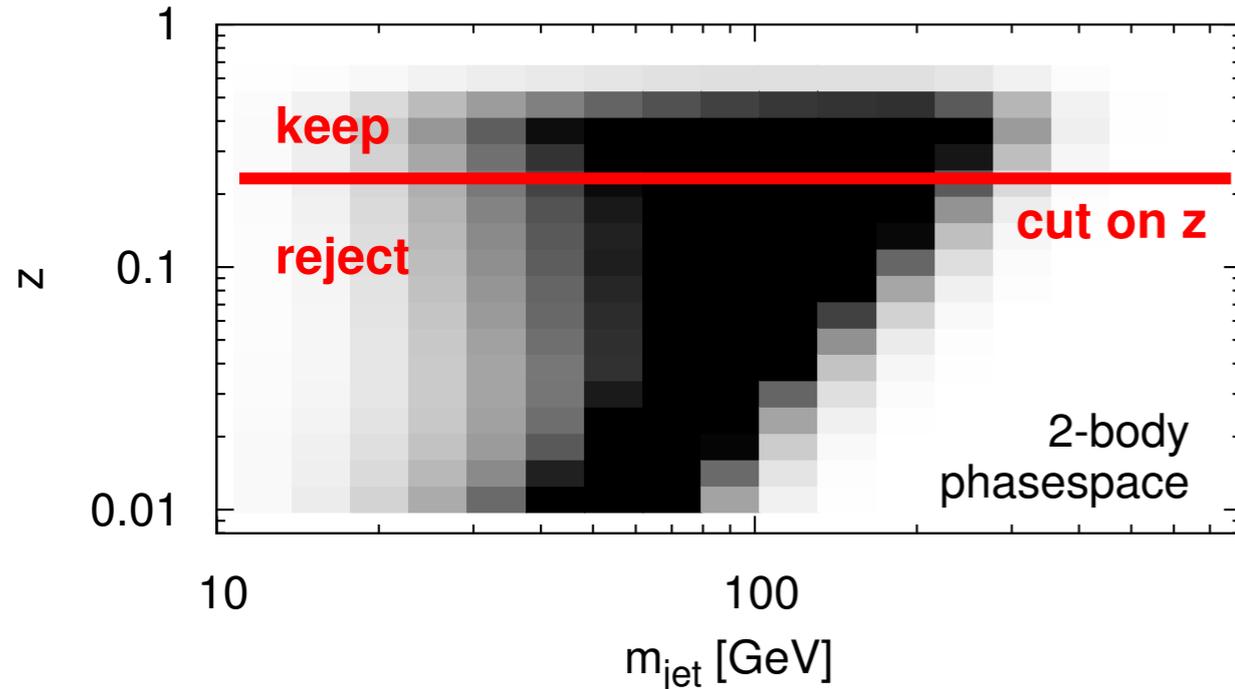
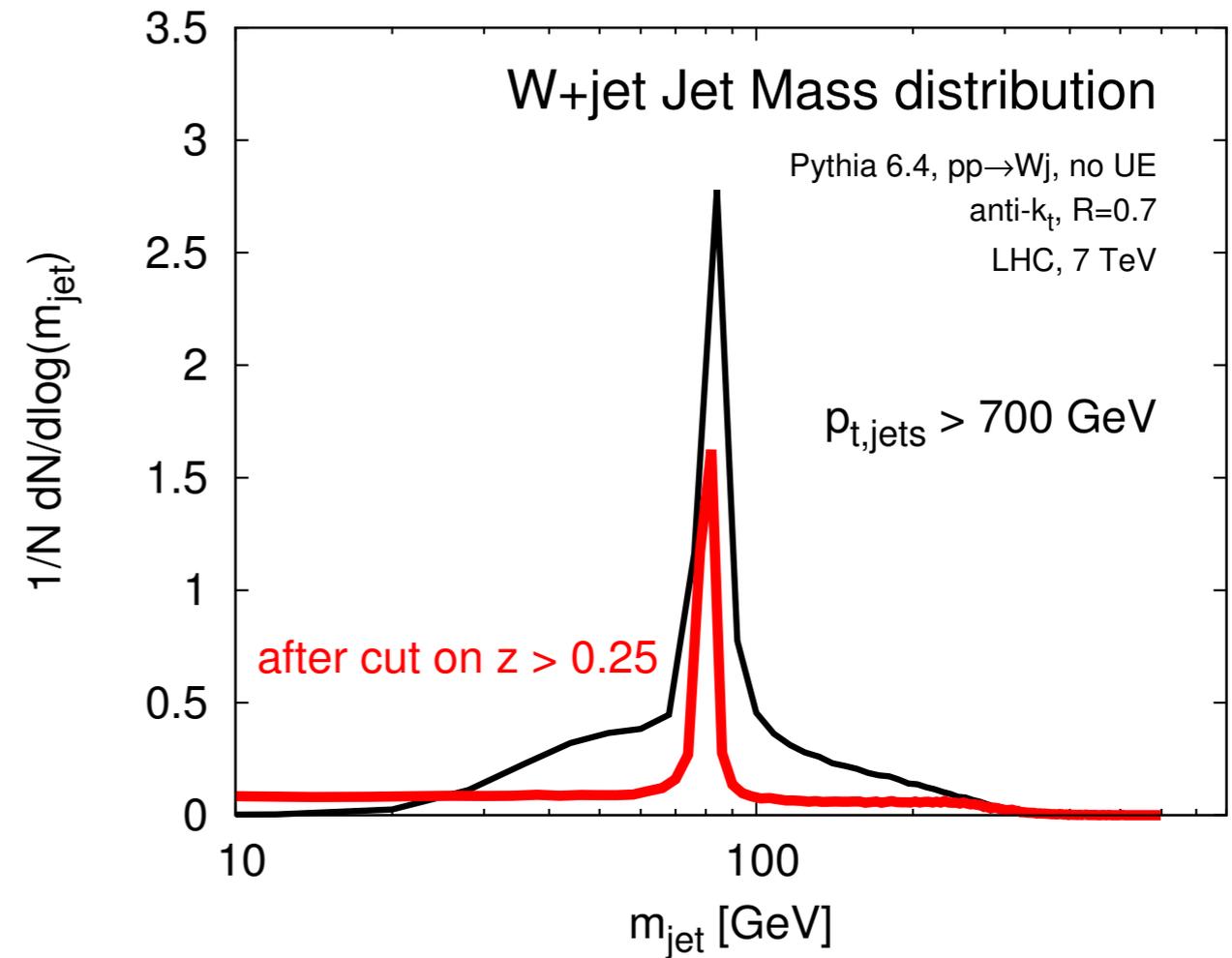
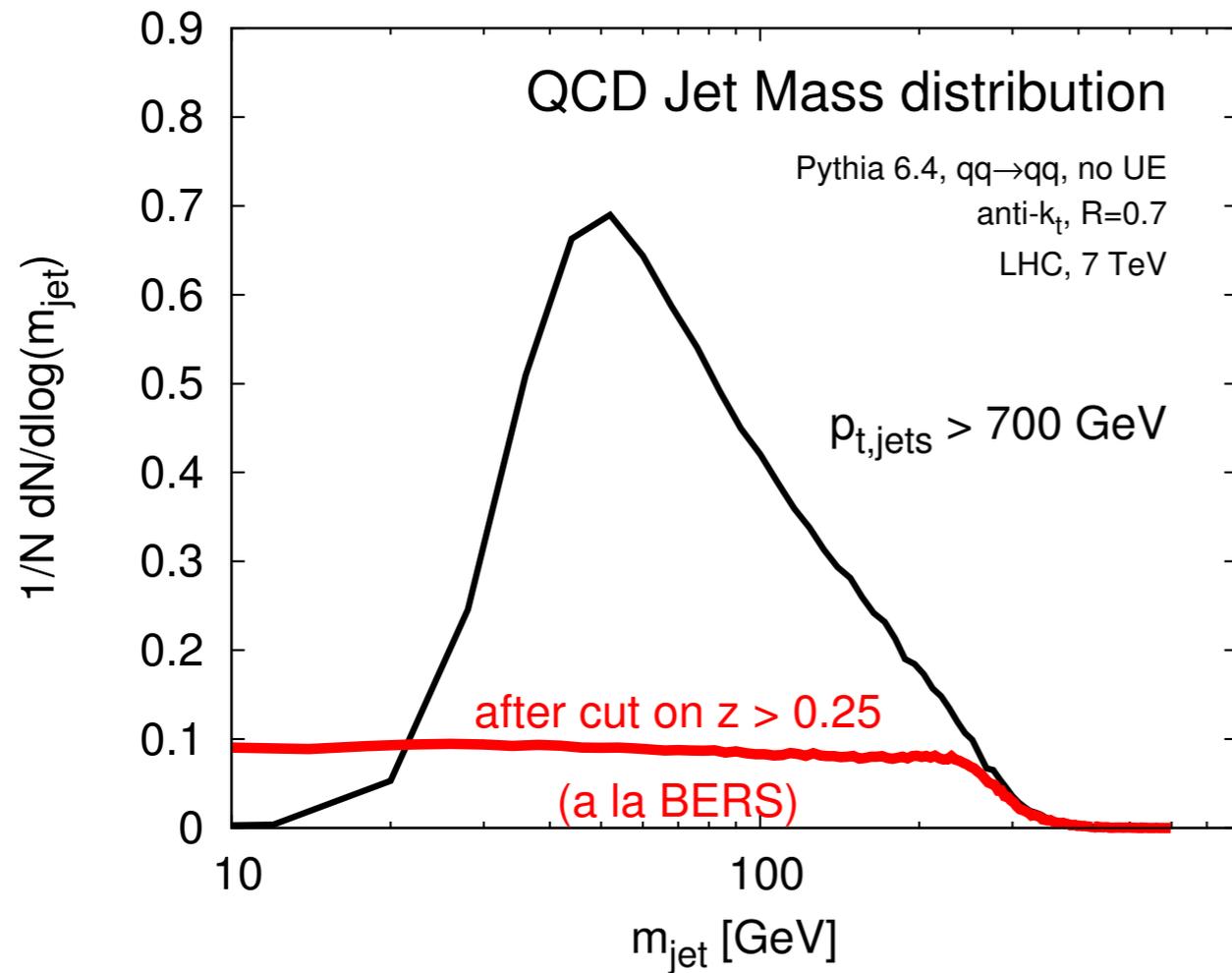
$$\int_{\frac{m^2}{p_t^2 R^2}}^{\frac{1}{2}} \frac{dz}{z}$$

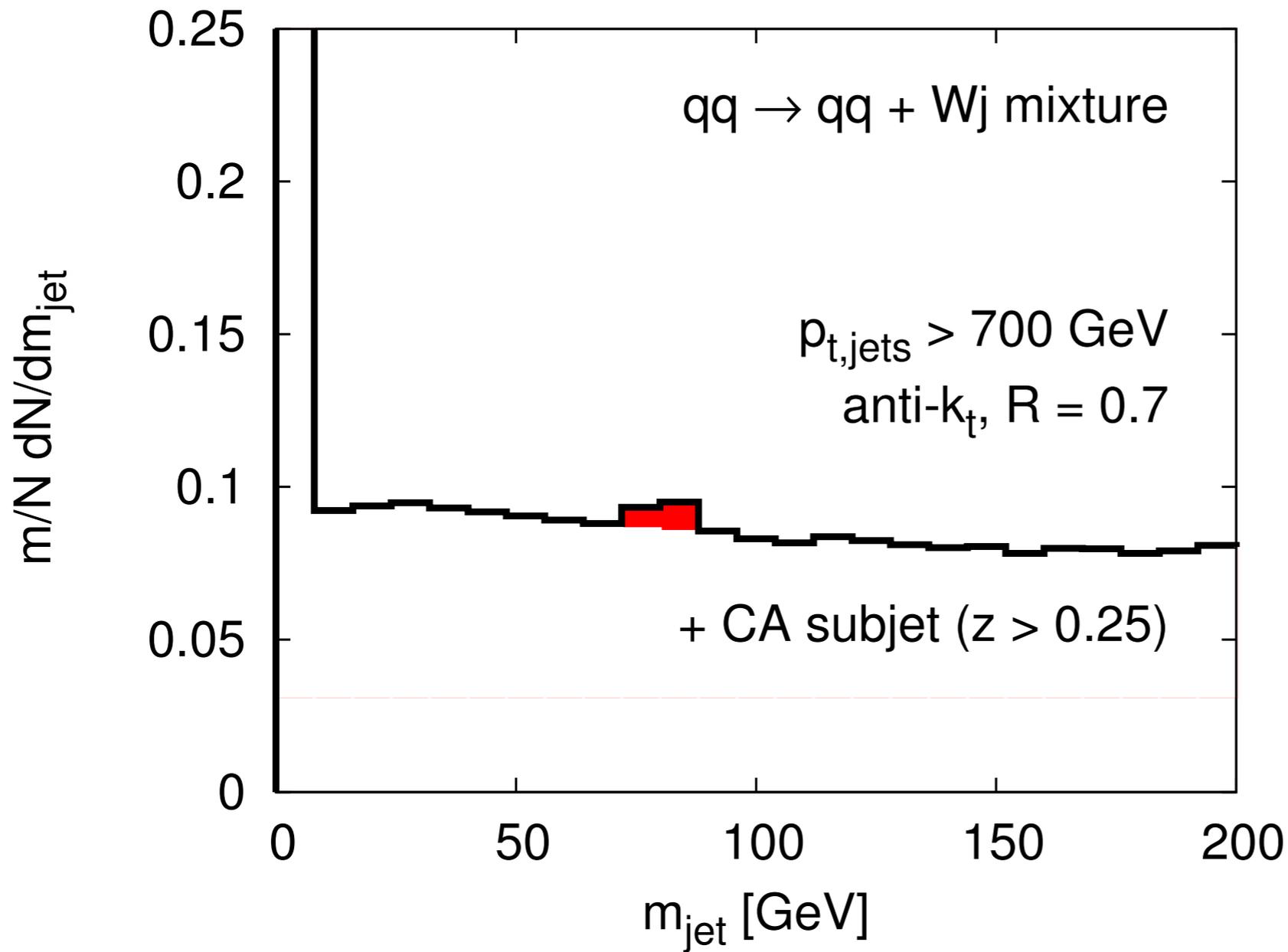
A hard cut on  $z$  reduces QCD background & simplifies its shape

# Inside the jet mass



# Inside the jet mass

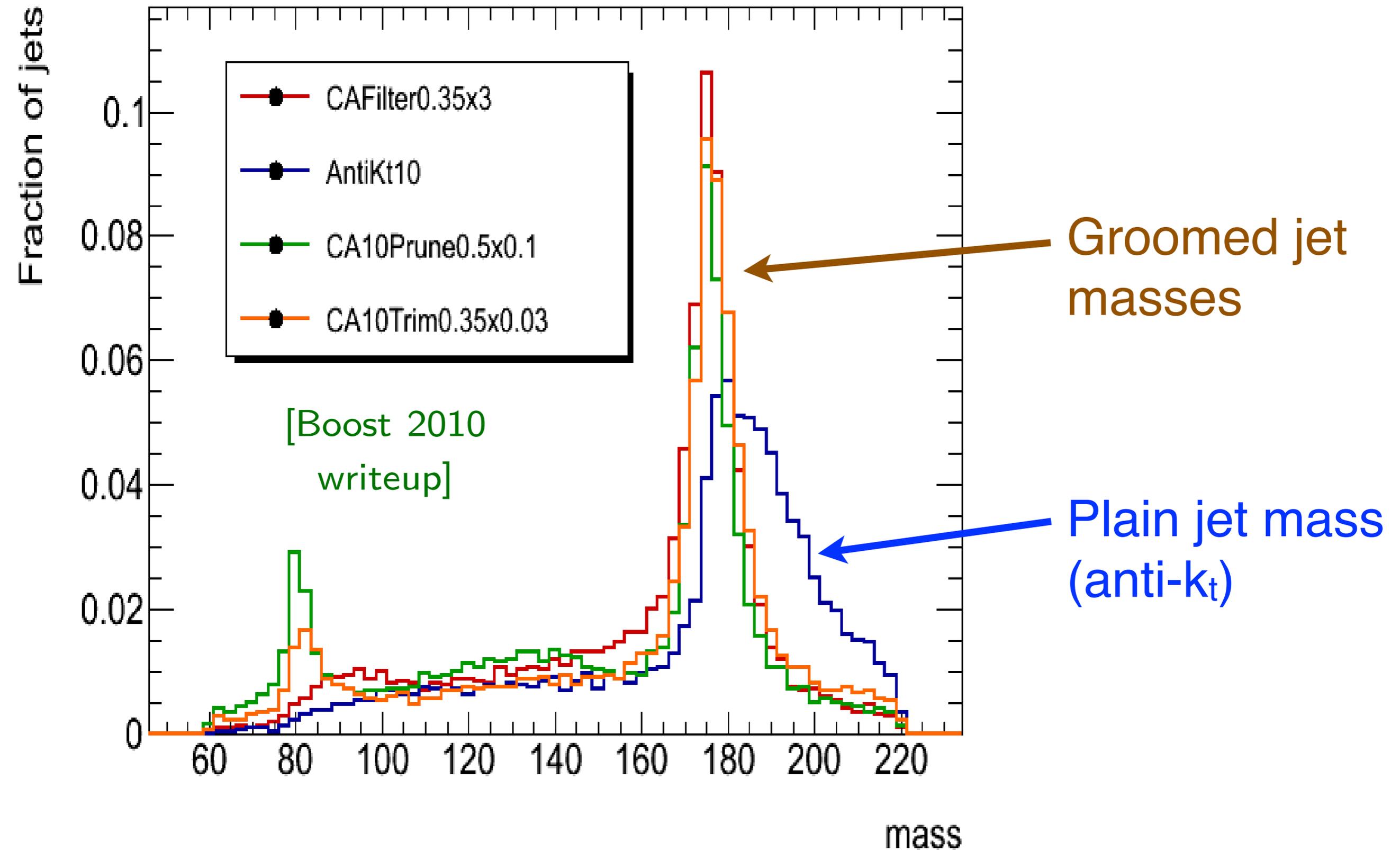




Signal + bkgd  
after cut on z

One core idea for  
grooming

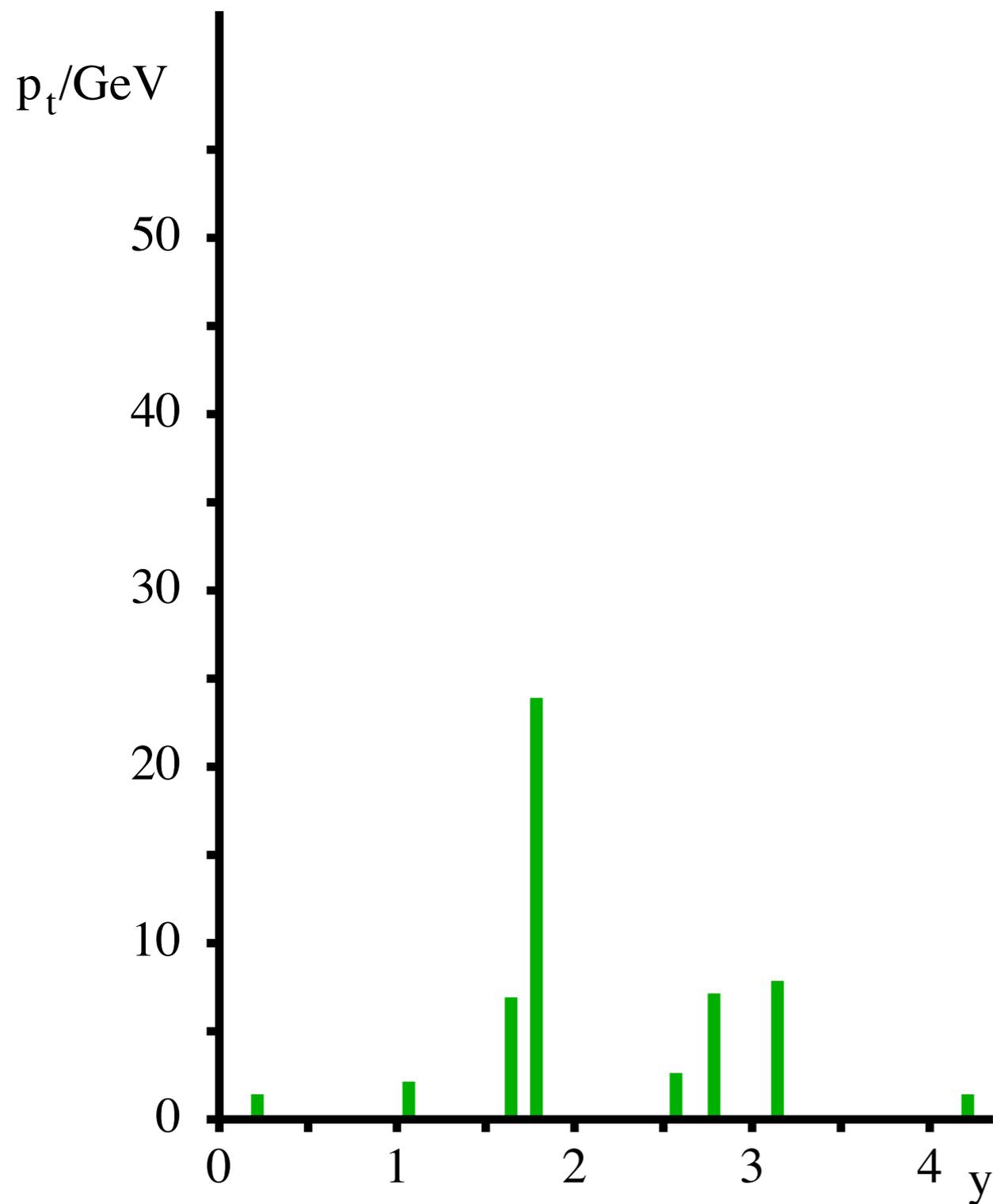
[see blackboard]



How do the tools work  
in practice?

# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm



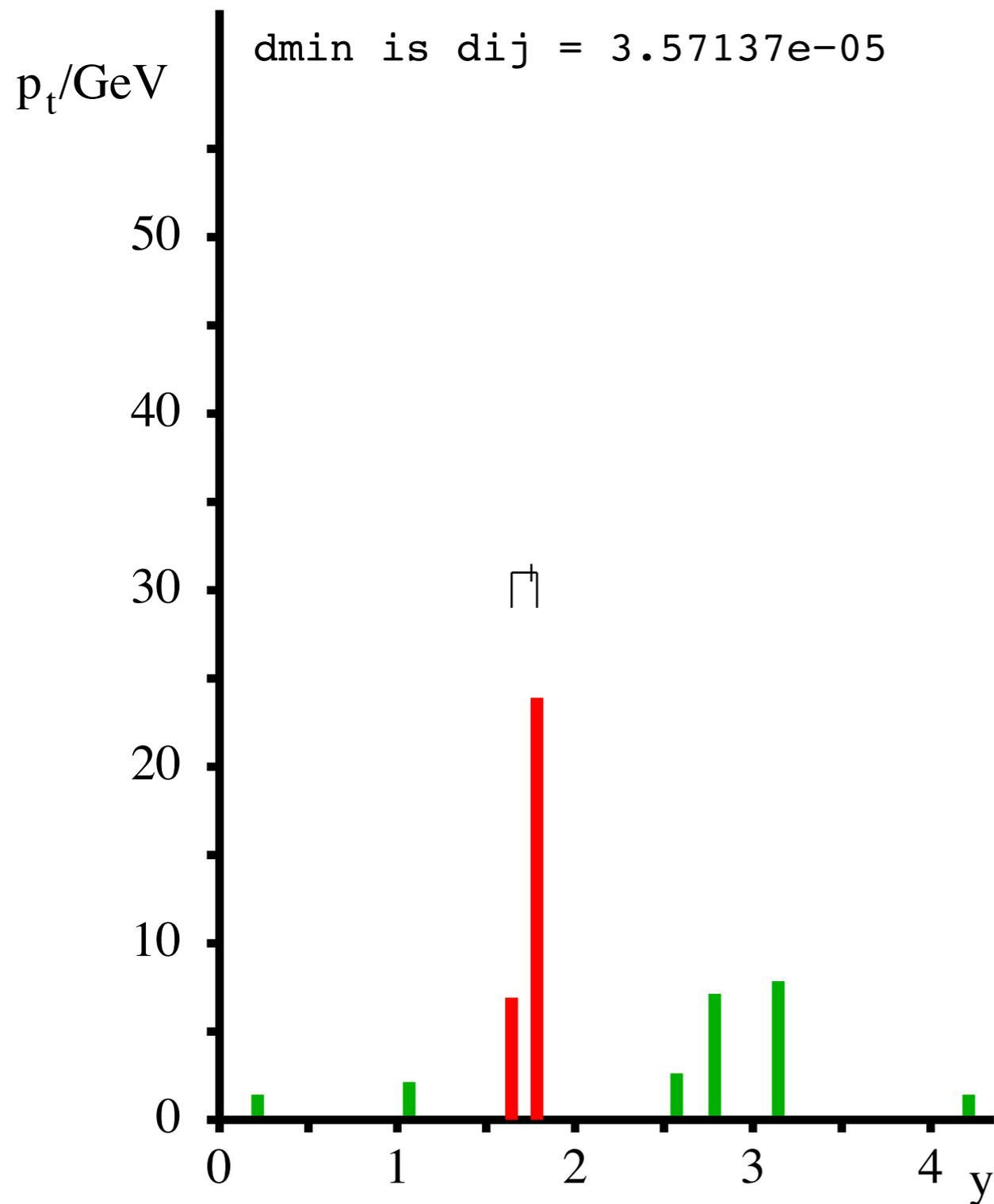
How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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  $d_{ij} = 3.57137e-05$

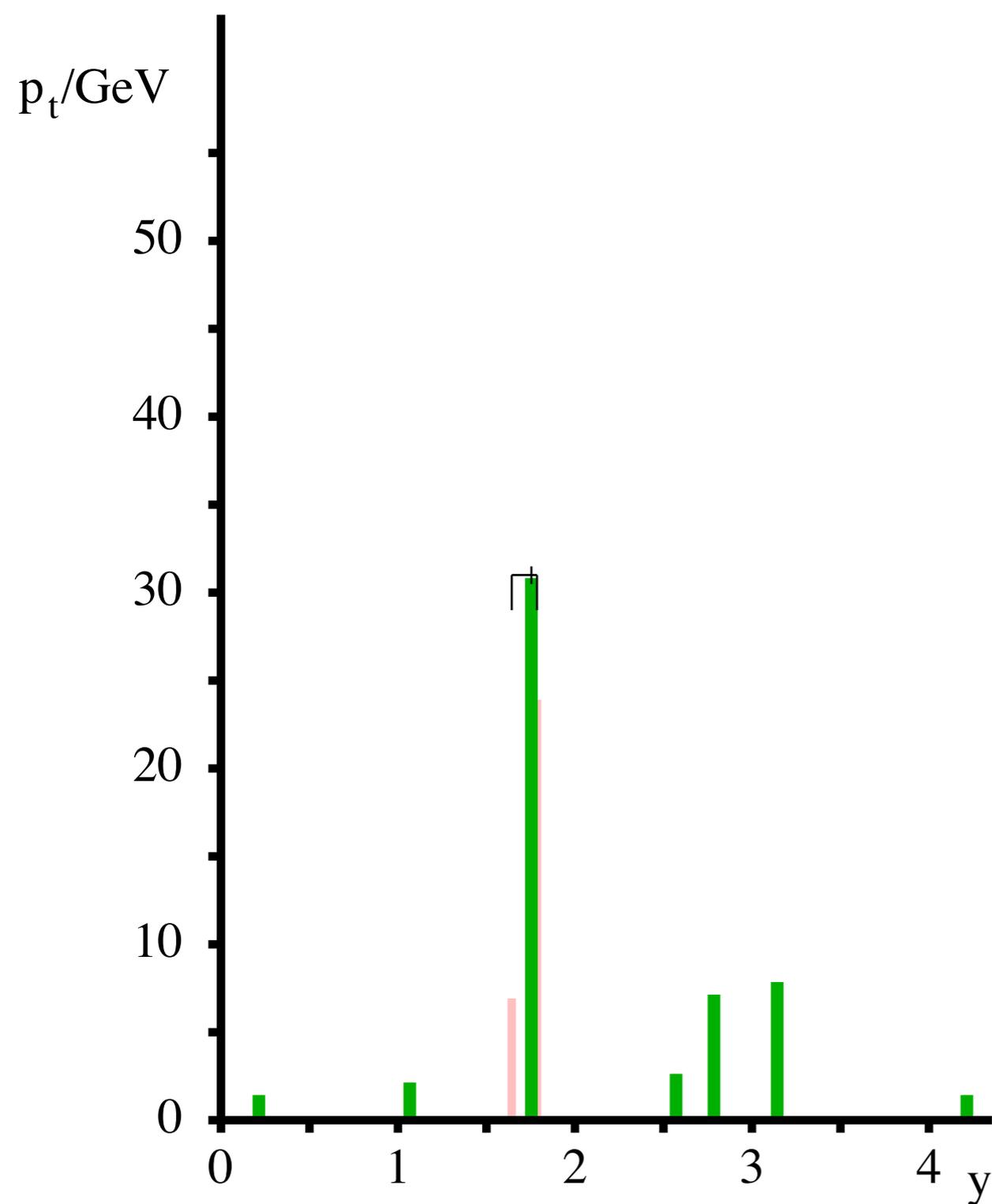


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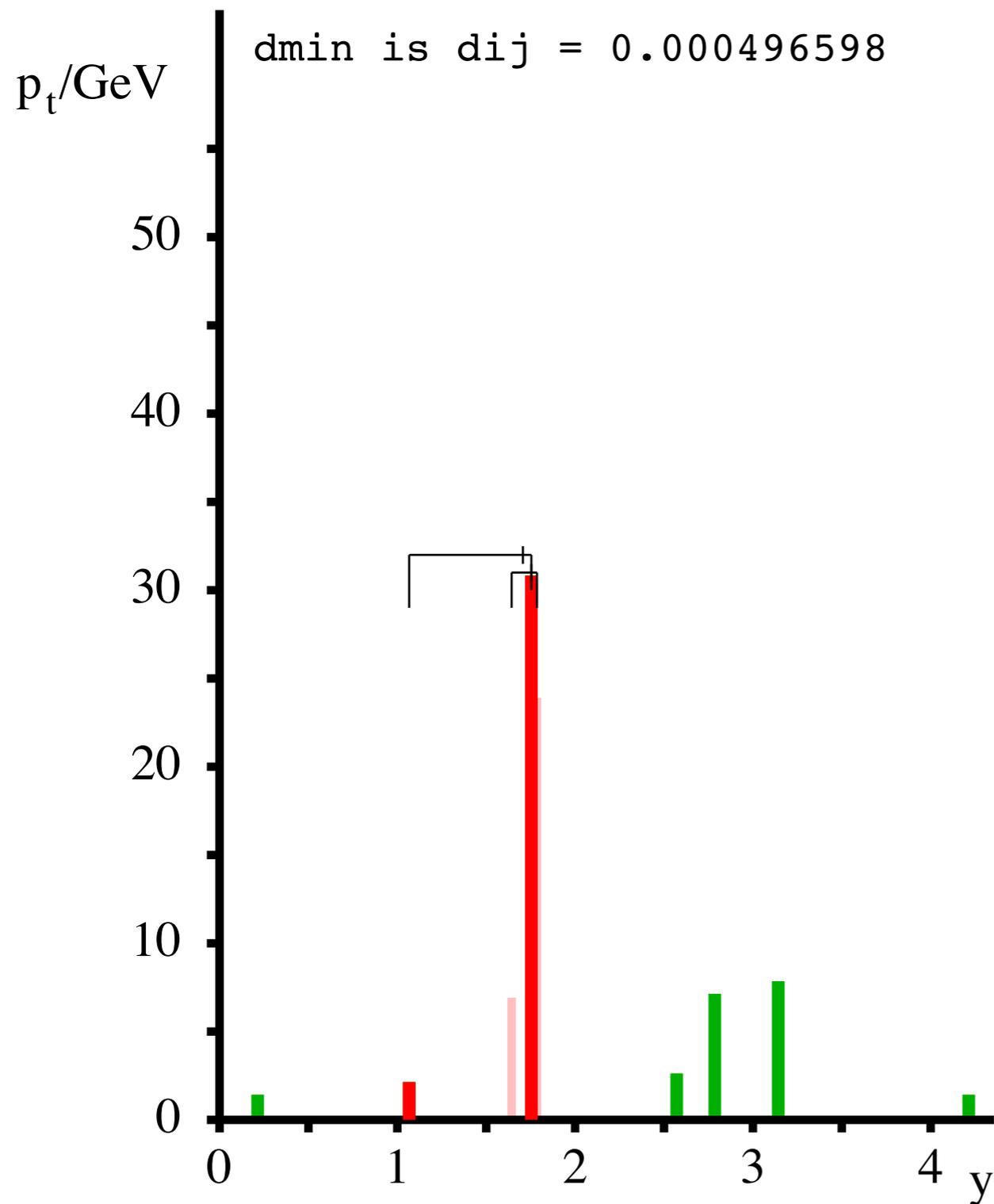
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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

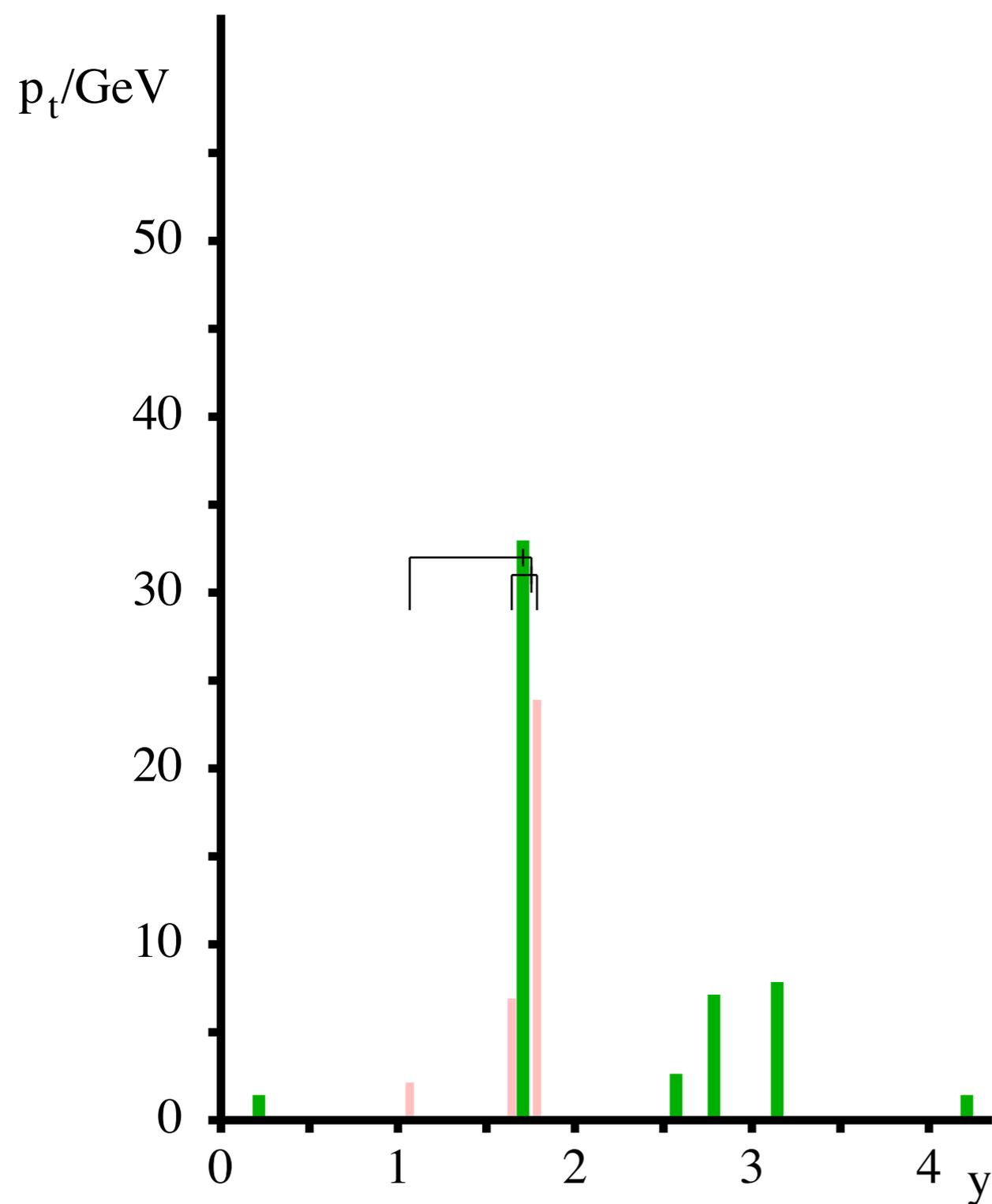


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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm



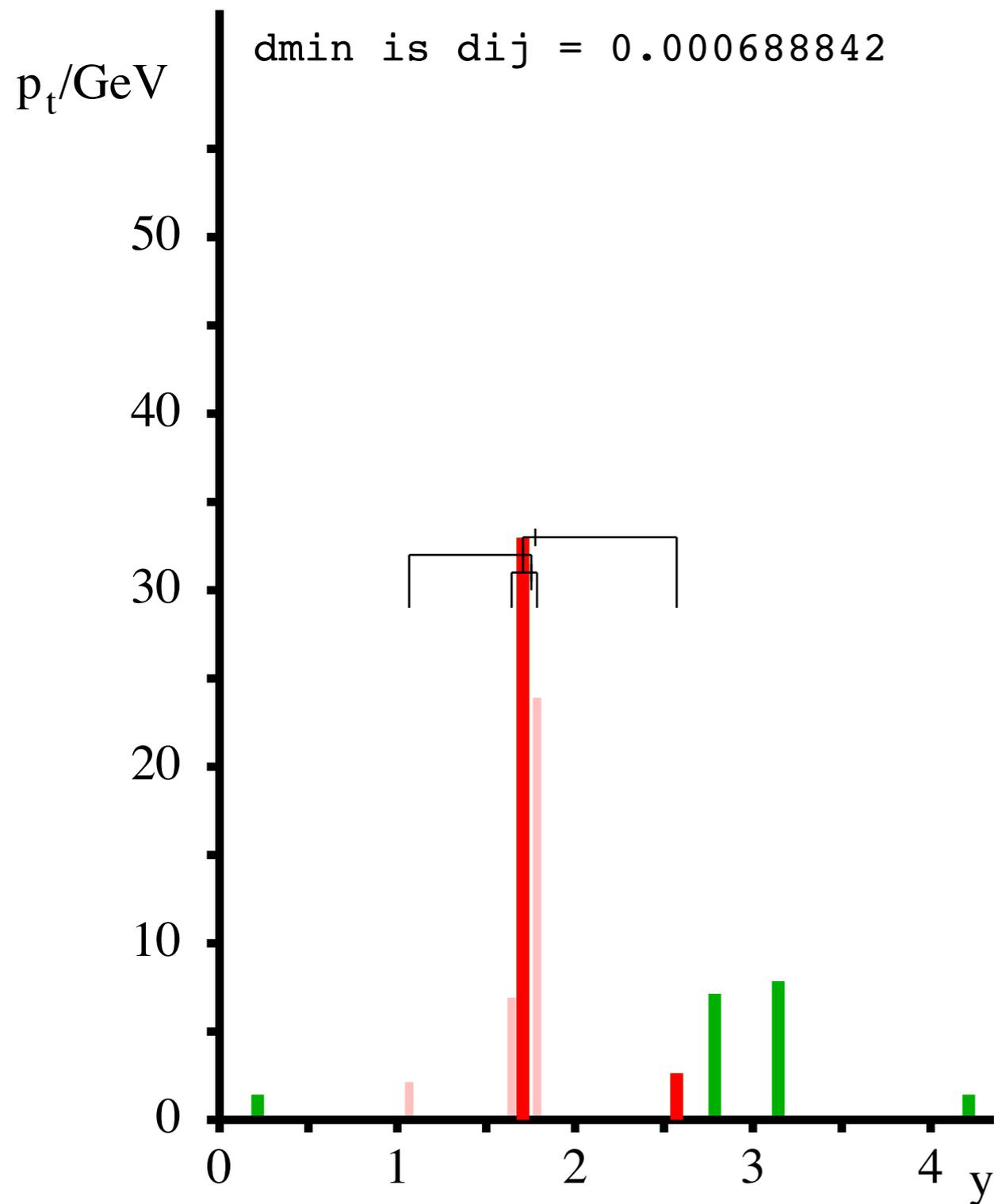
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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

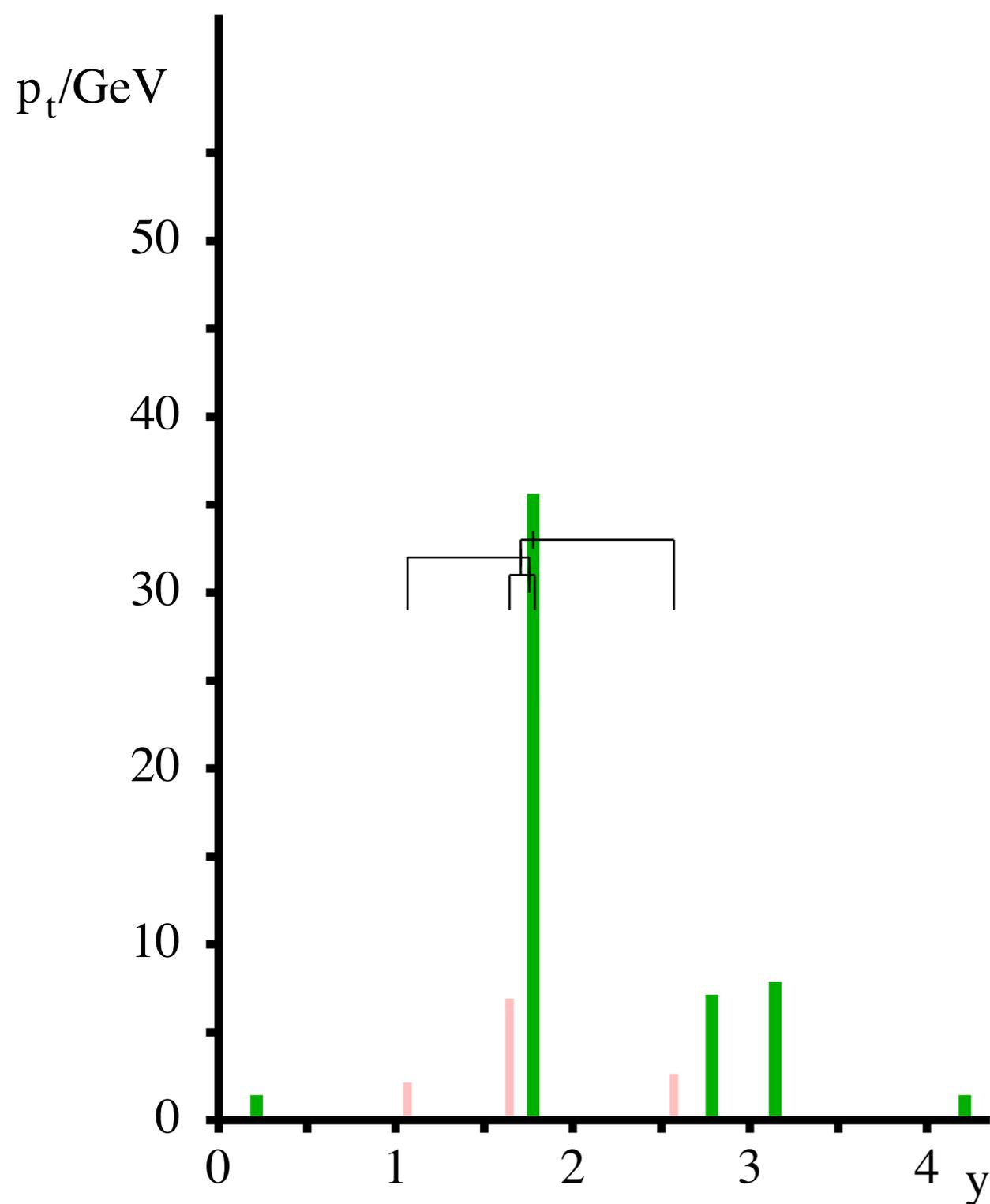


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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm



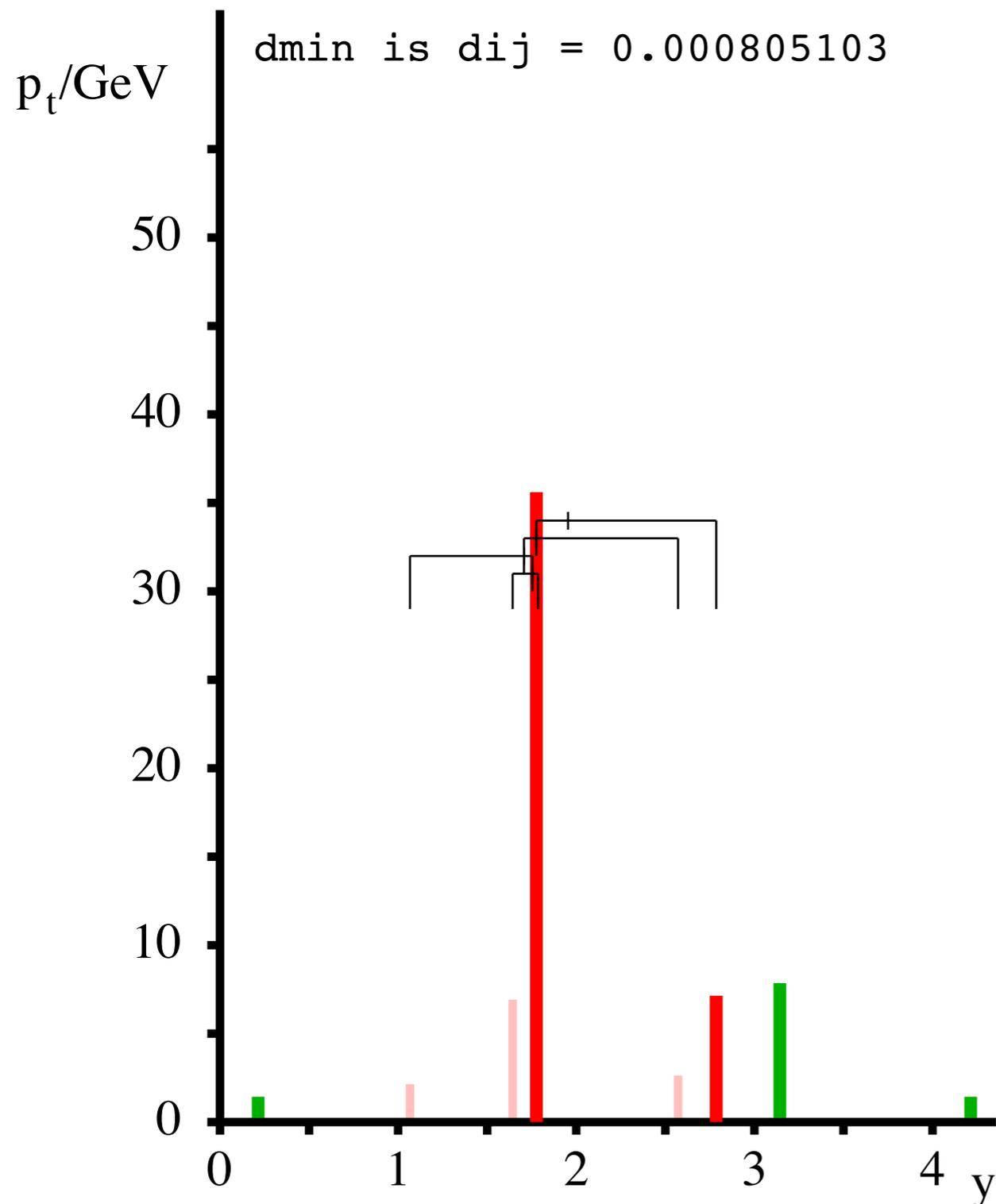
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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



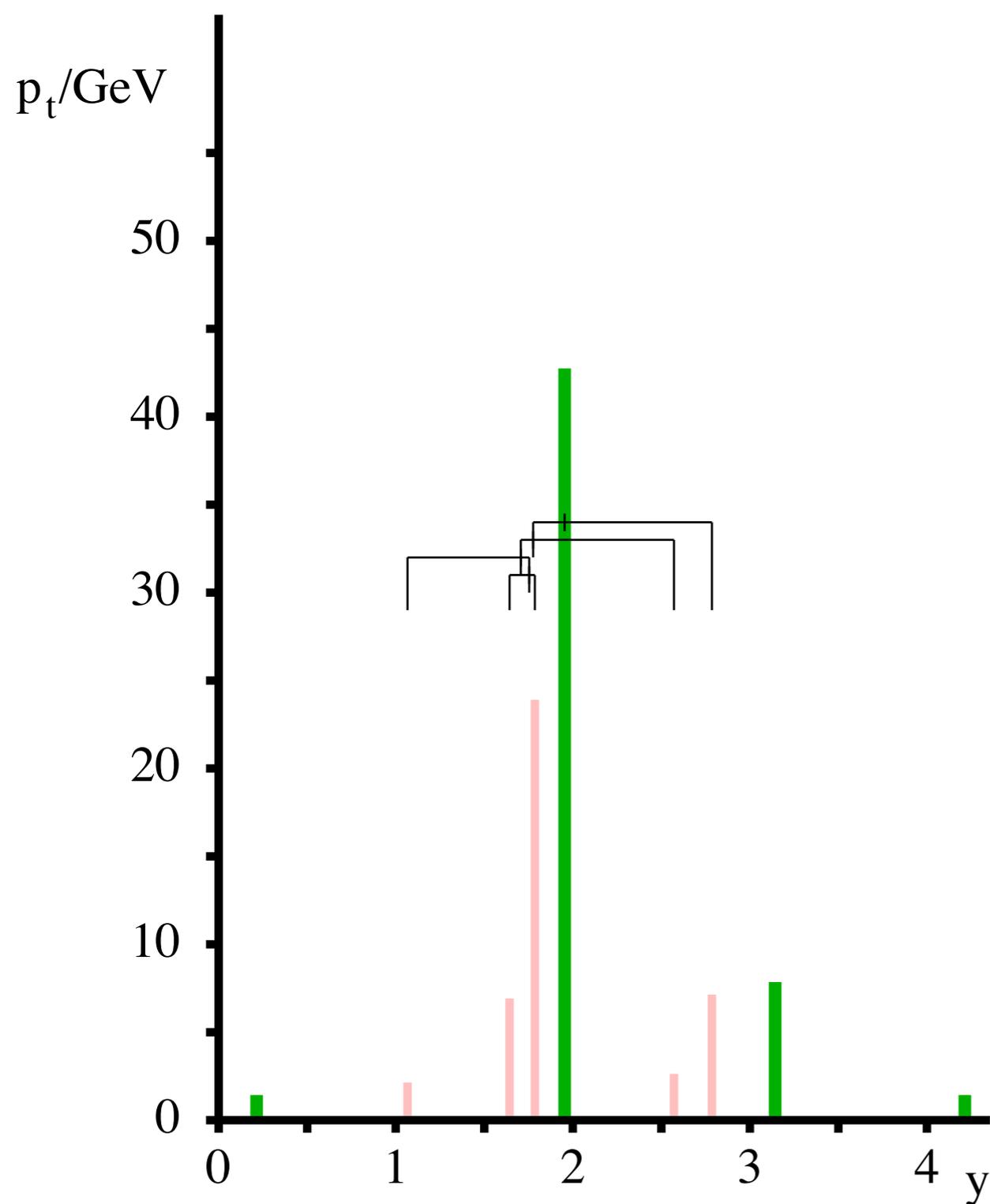
How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

*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



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

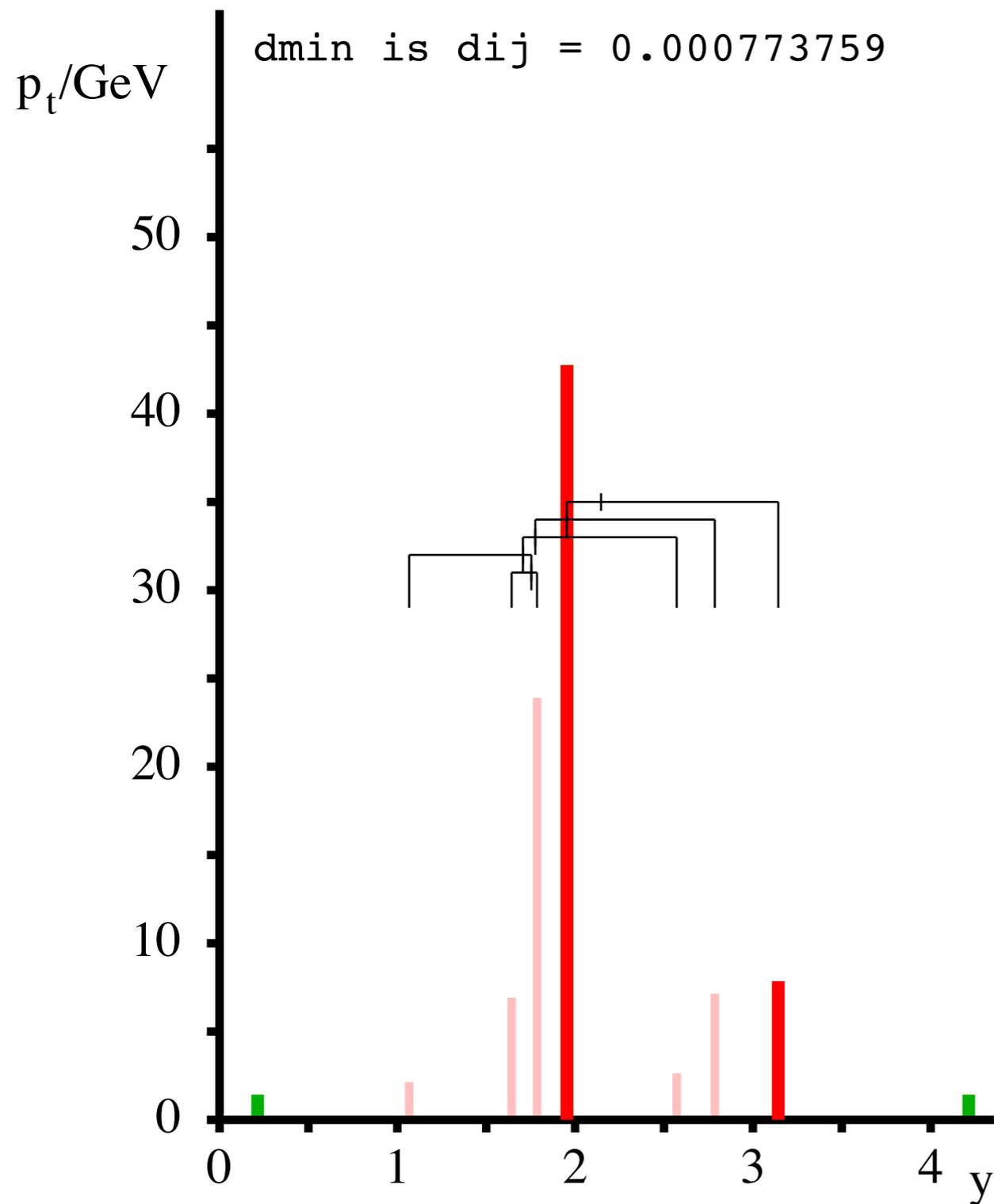
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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



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

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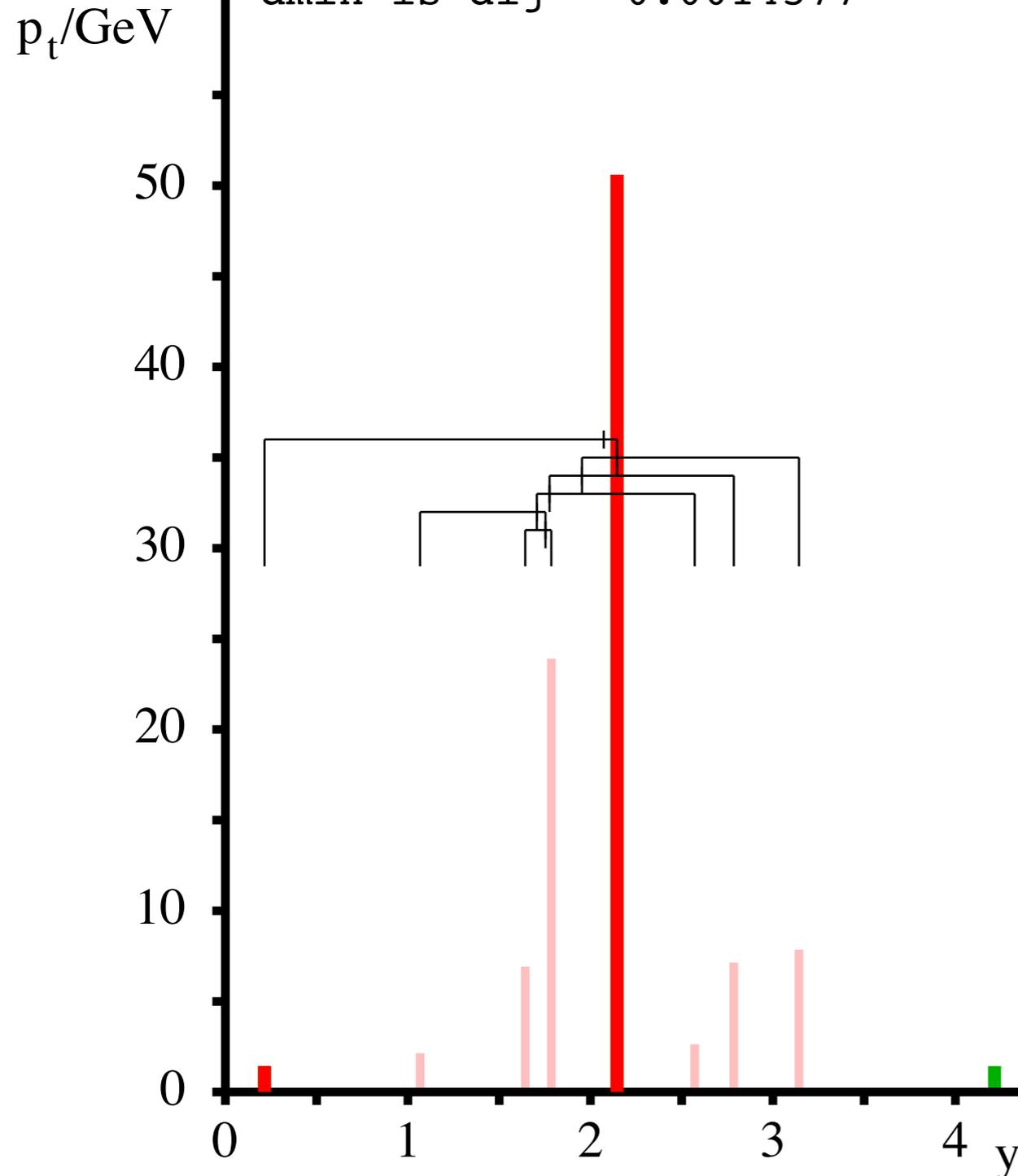
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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



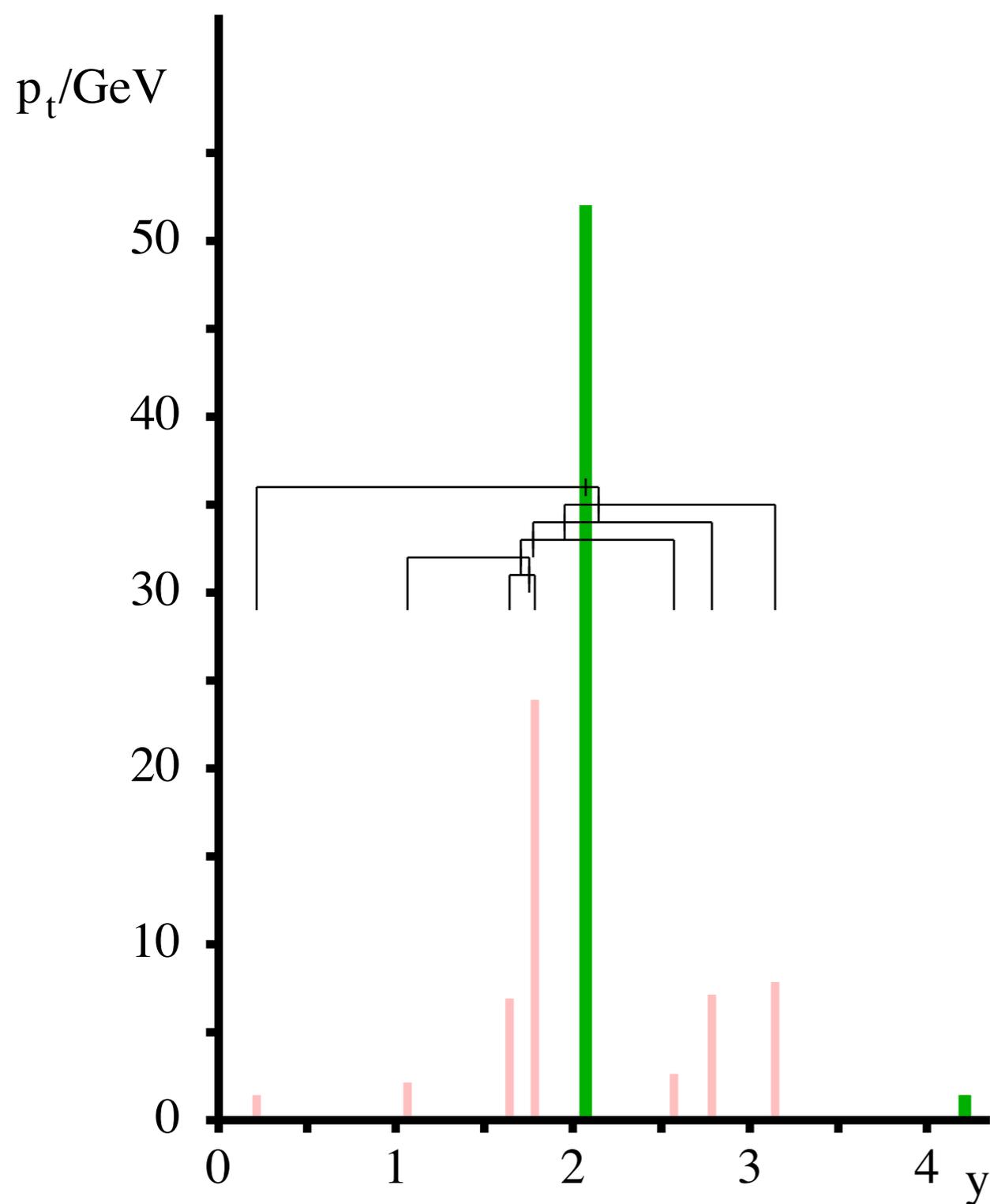
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm



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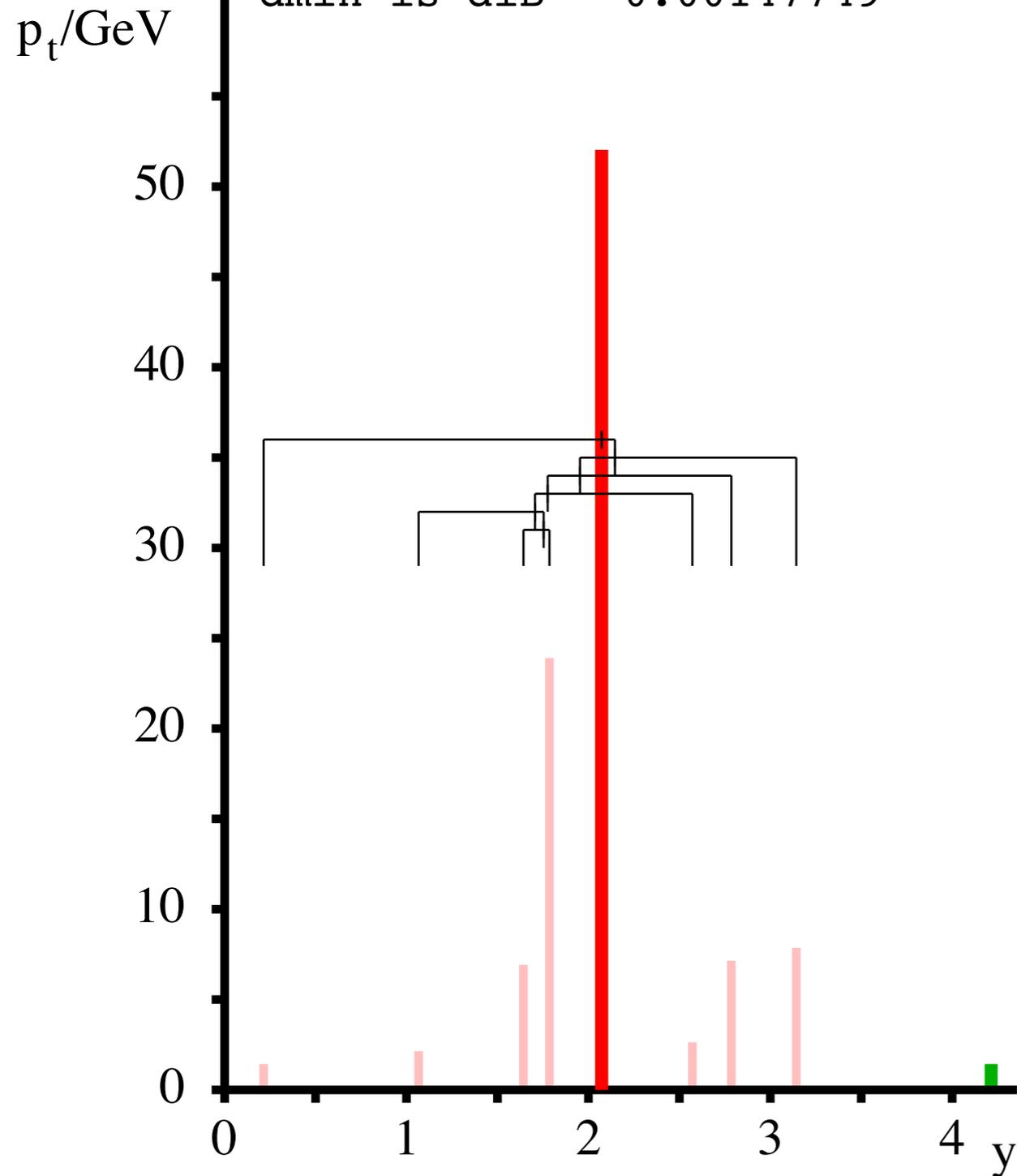
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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



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

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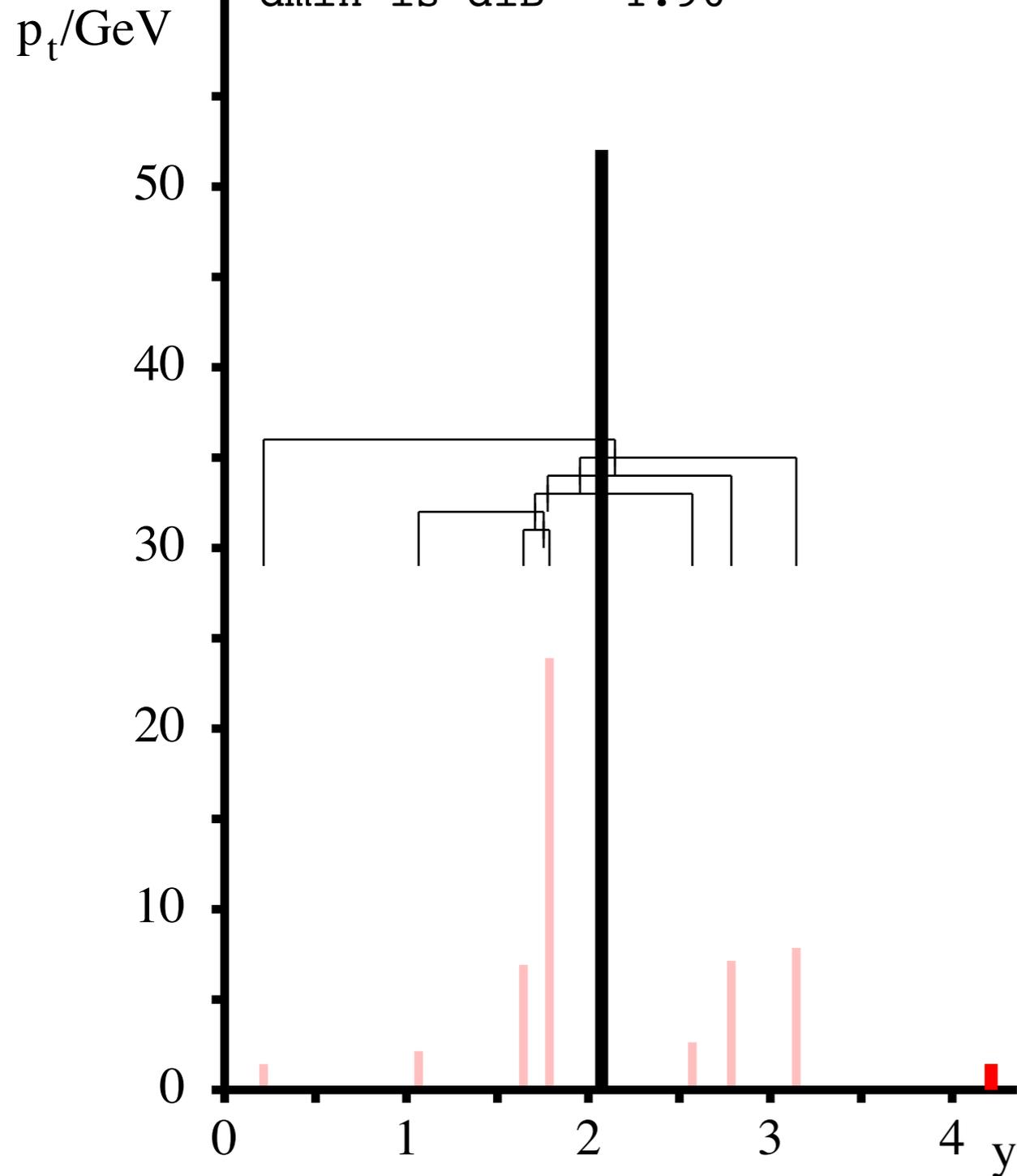
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# Identifying jet substructure: try out anti- $k_t$

## anti- $k_t$ algorithm

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



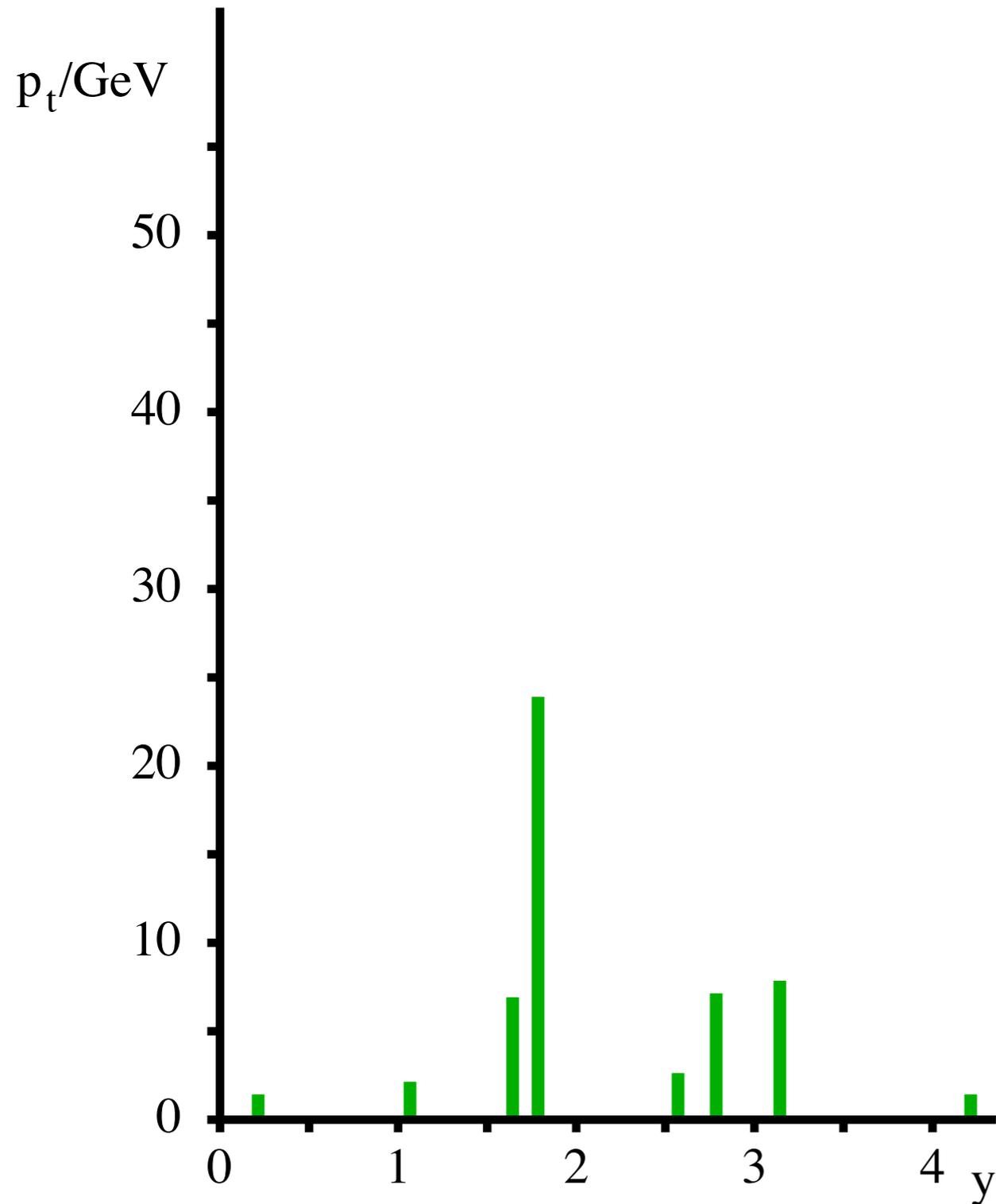
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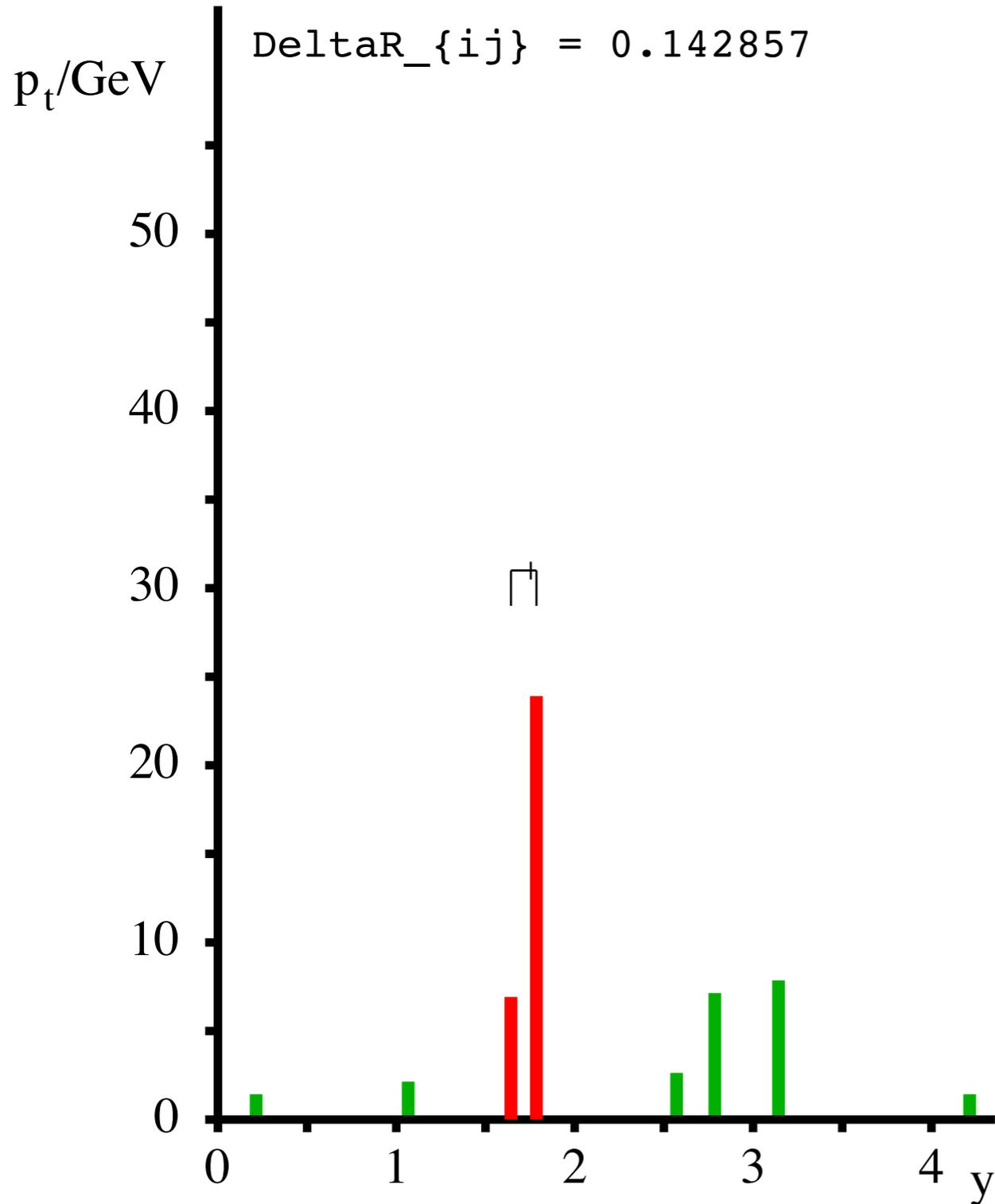


## Cambridge/Aachen algorithm



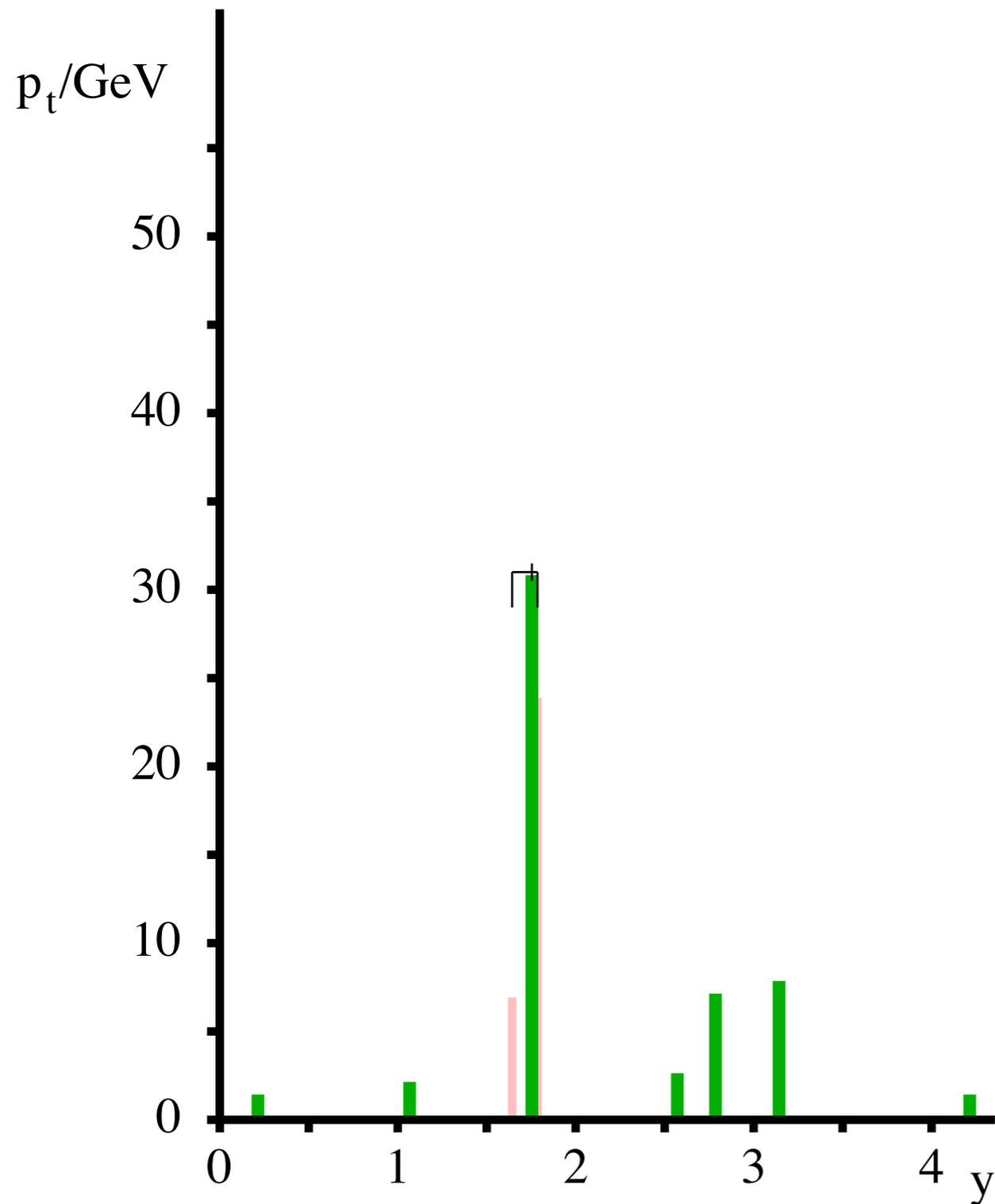
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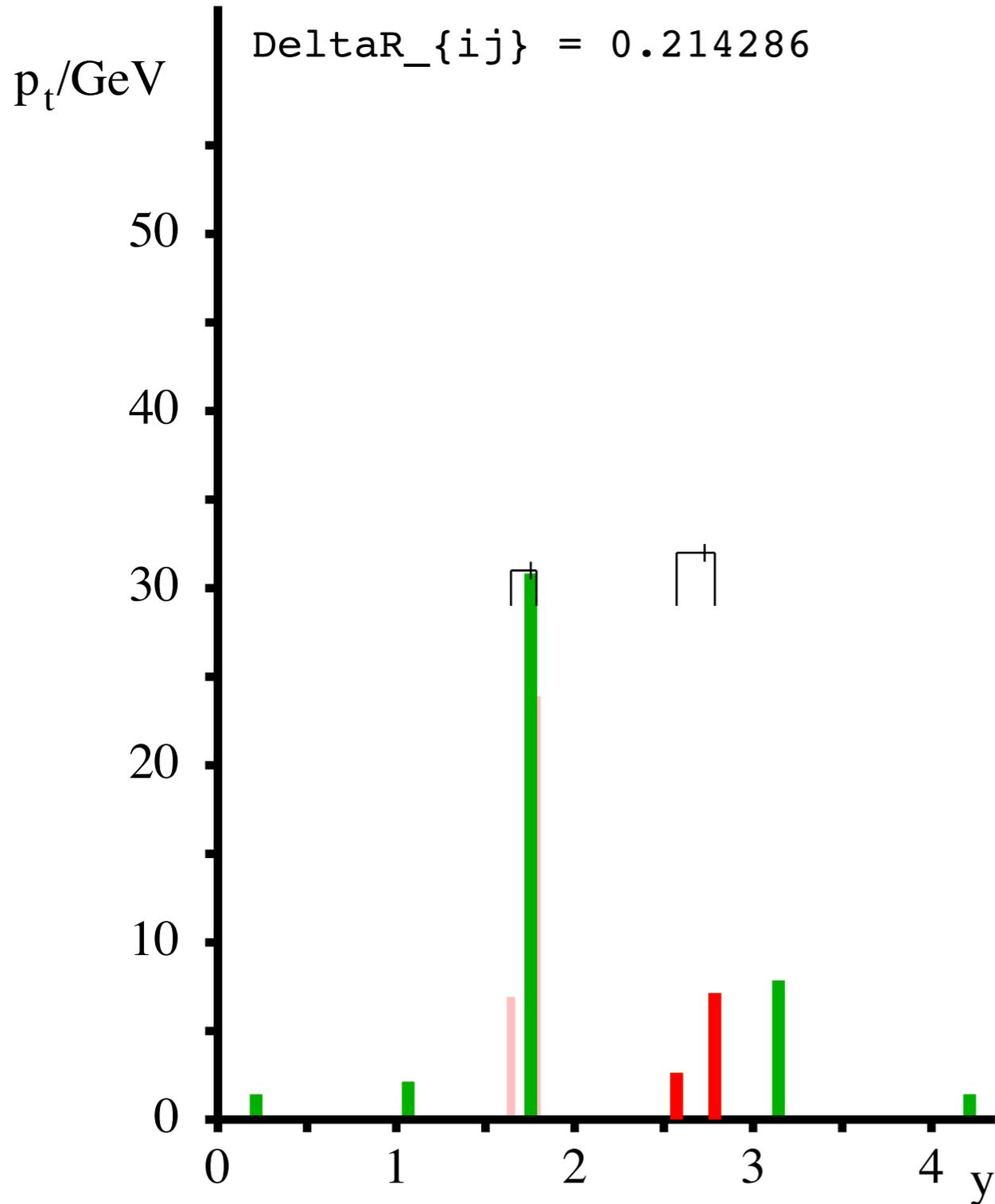
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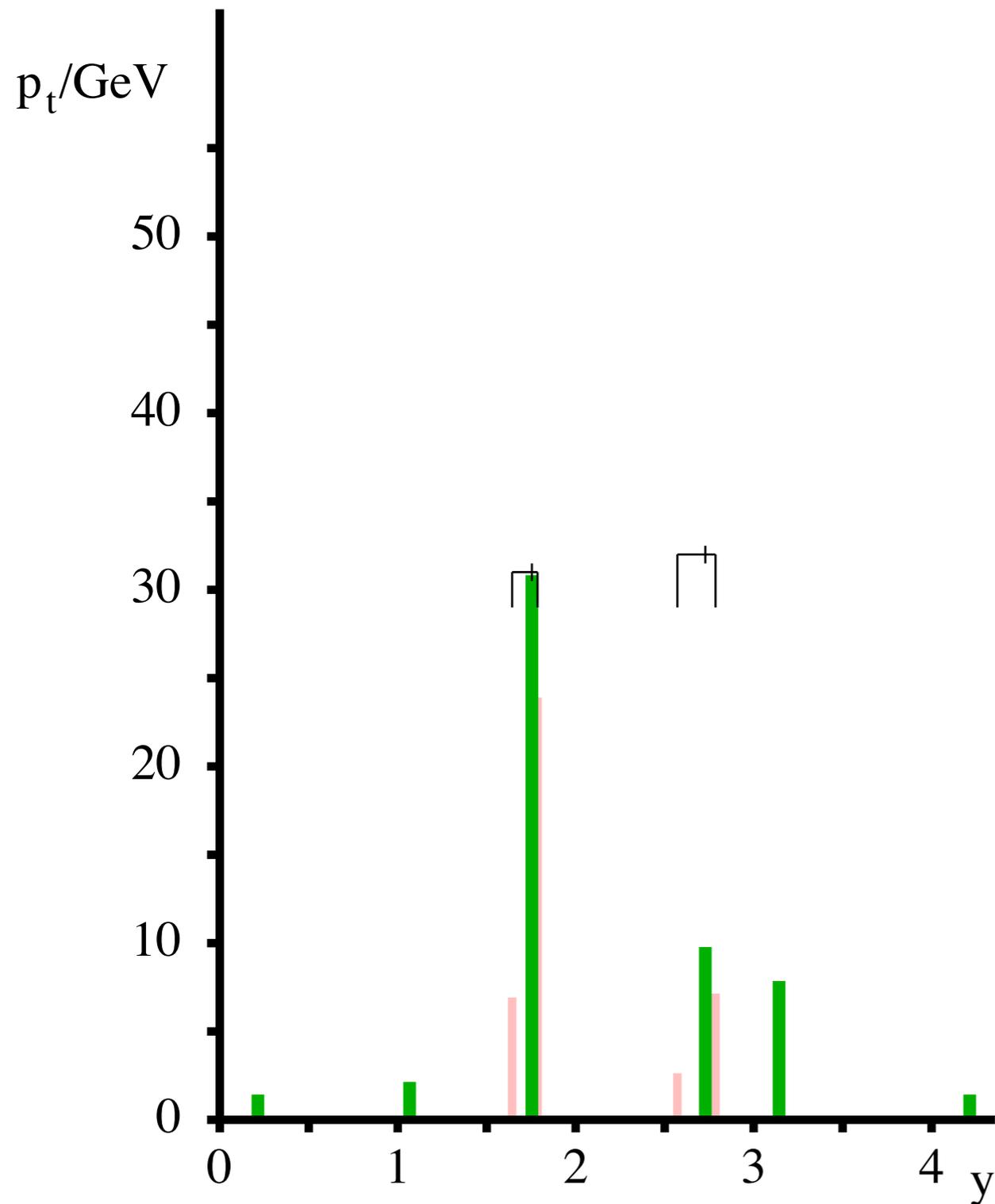
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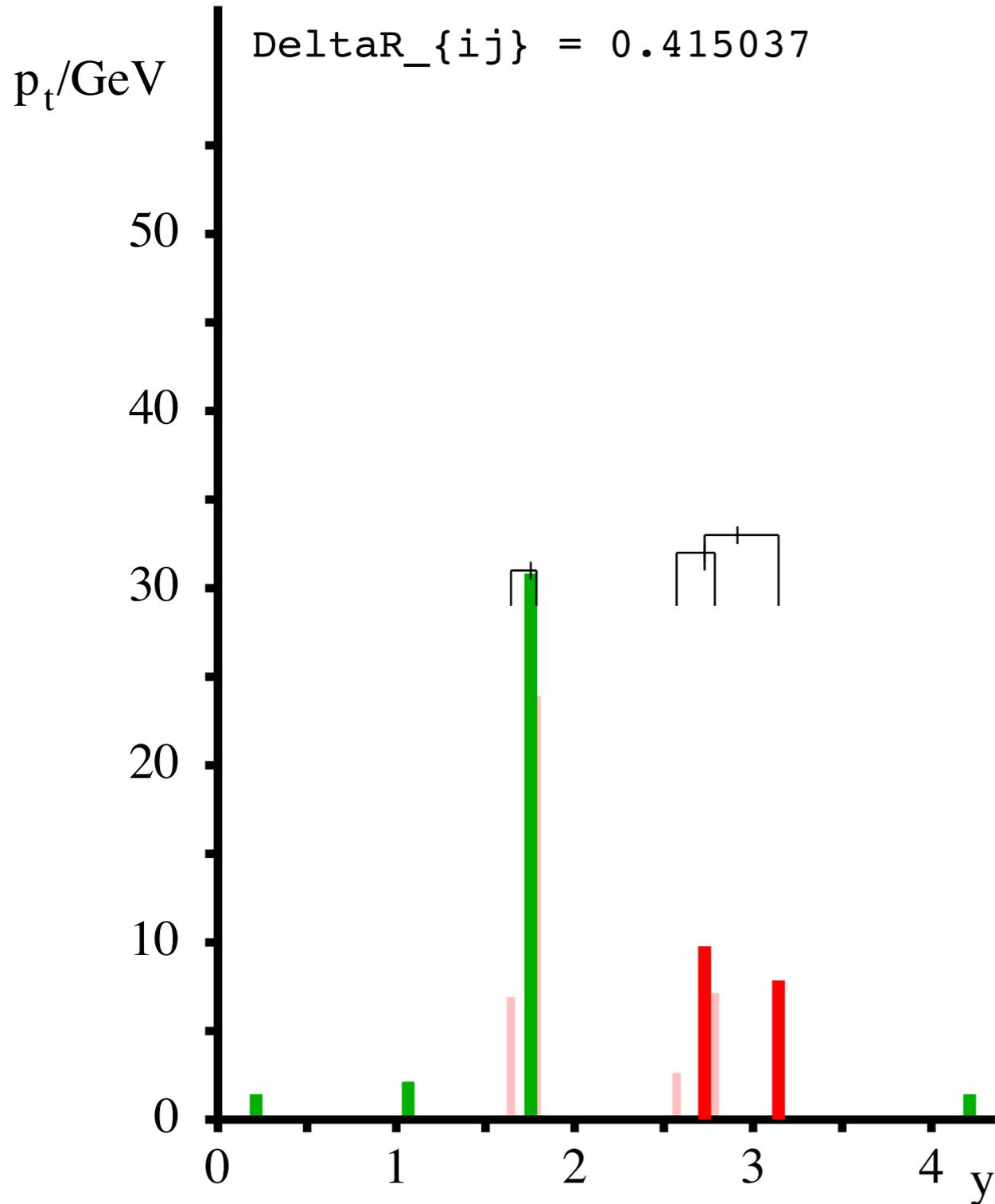
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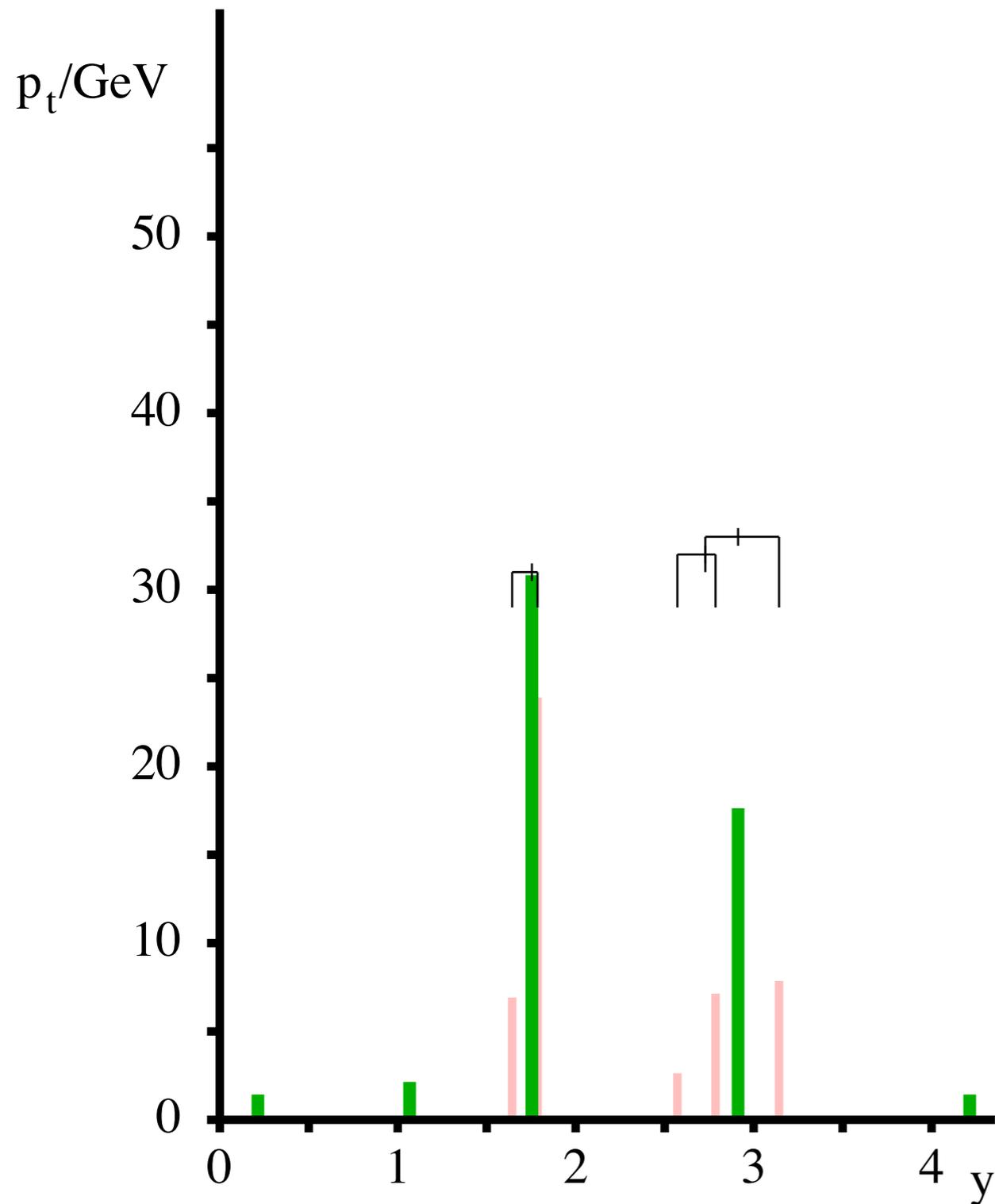
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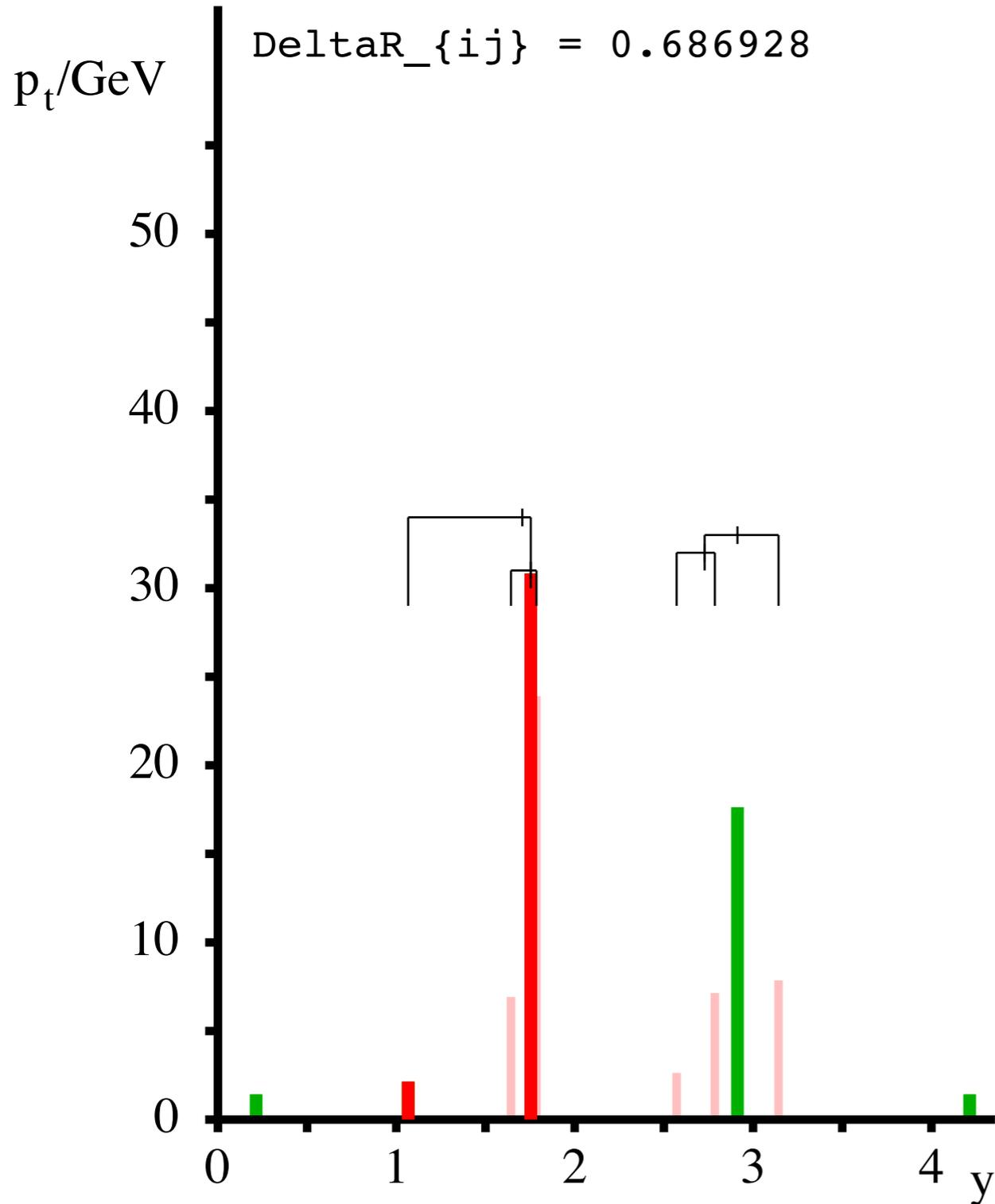
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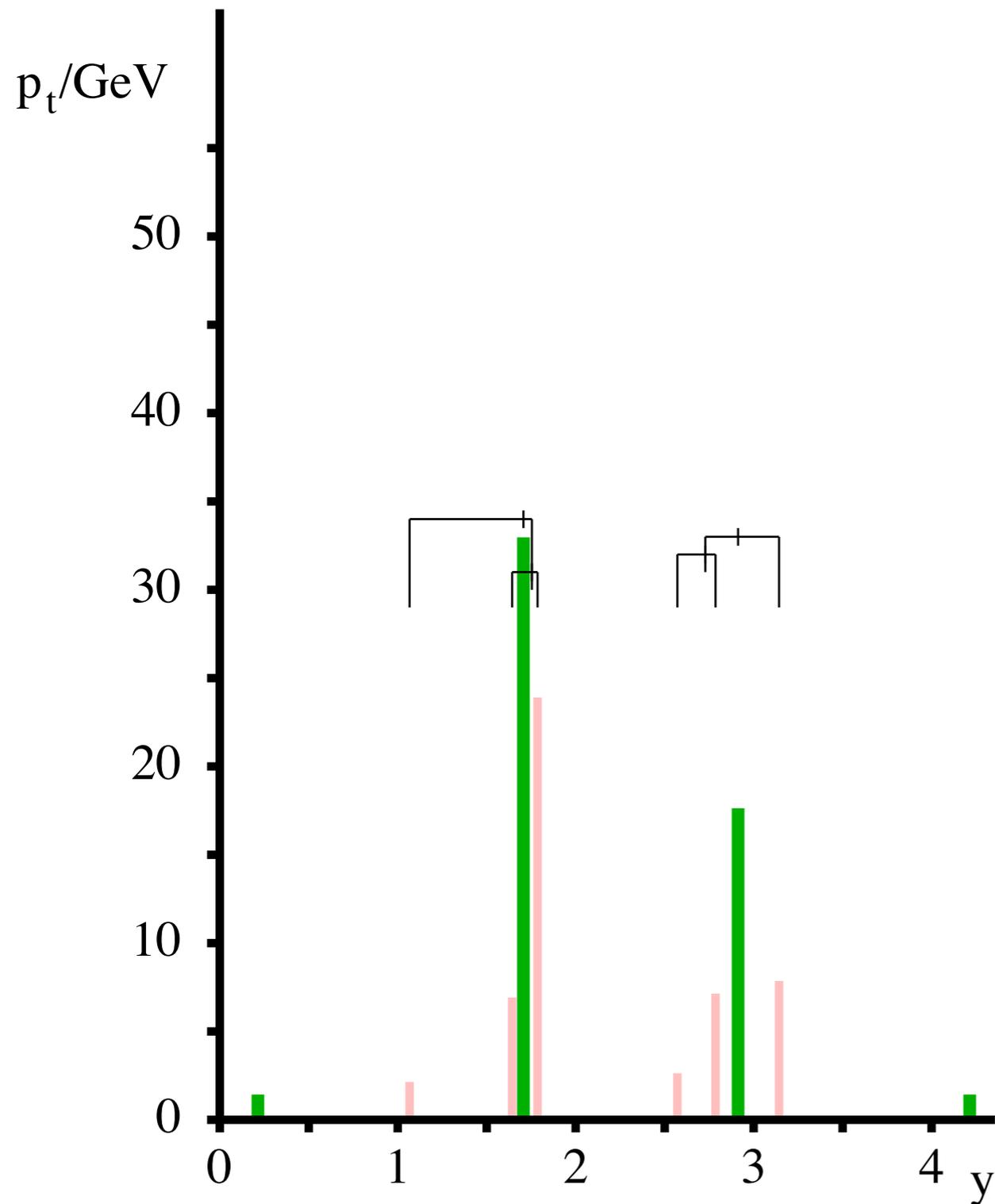
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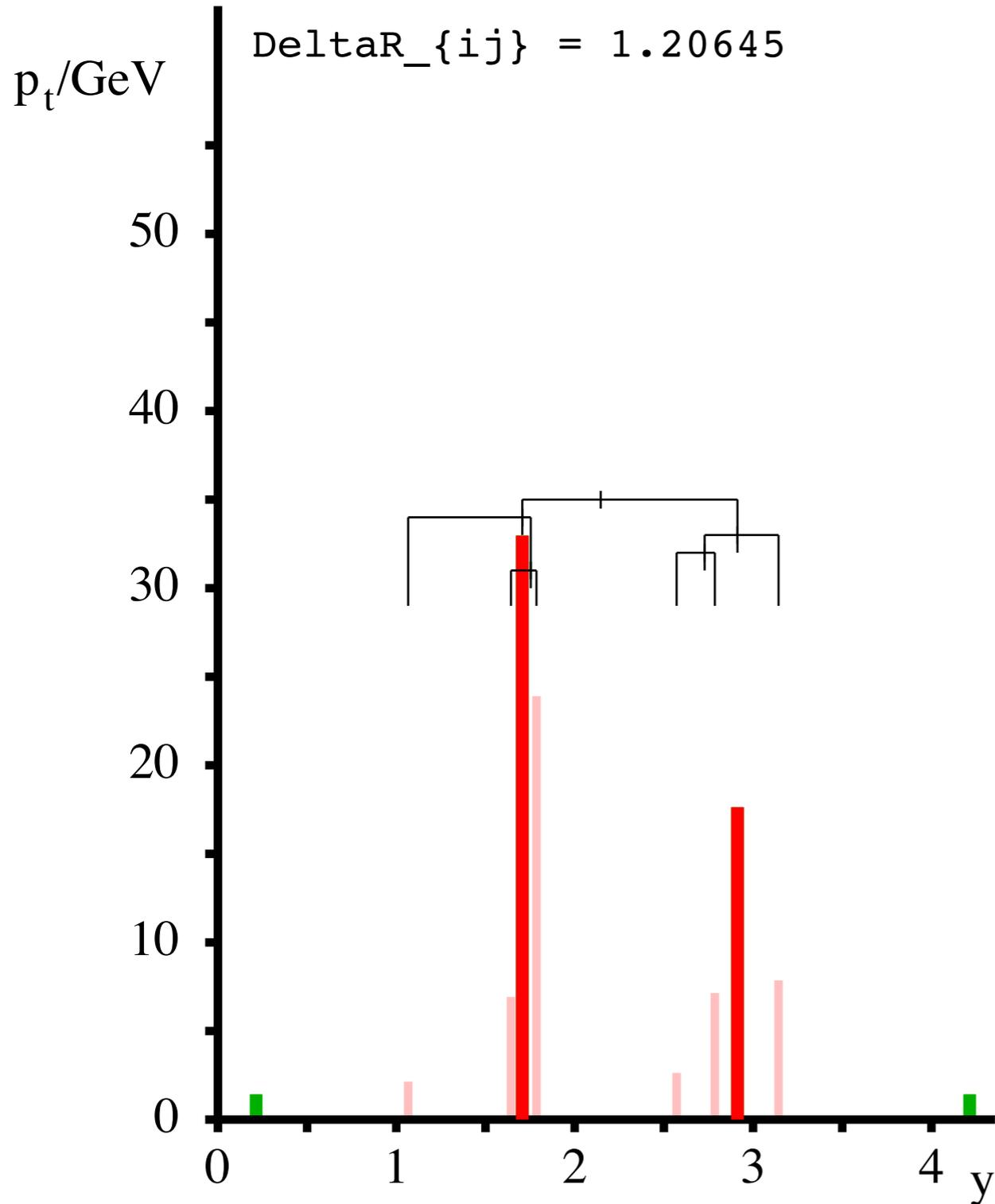
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**C/A identifies two hard blobs with limited soft contamination**

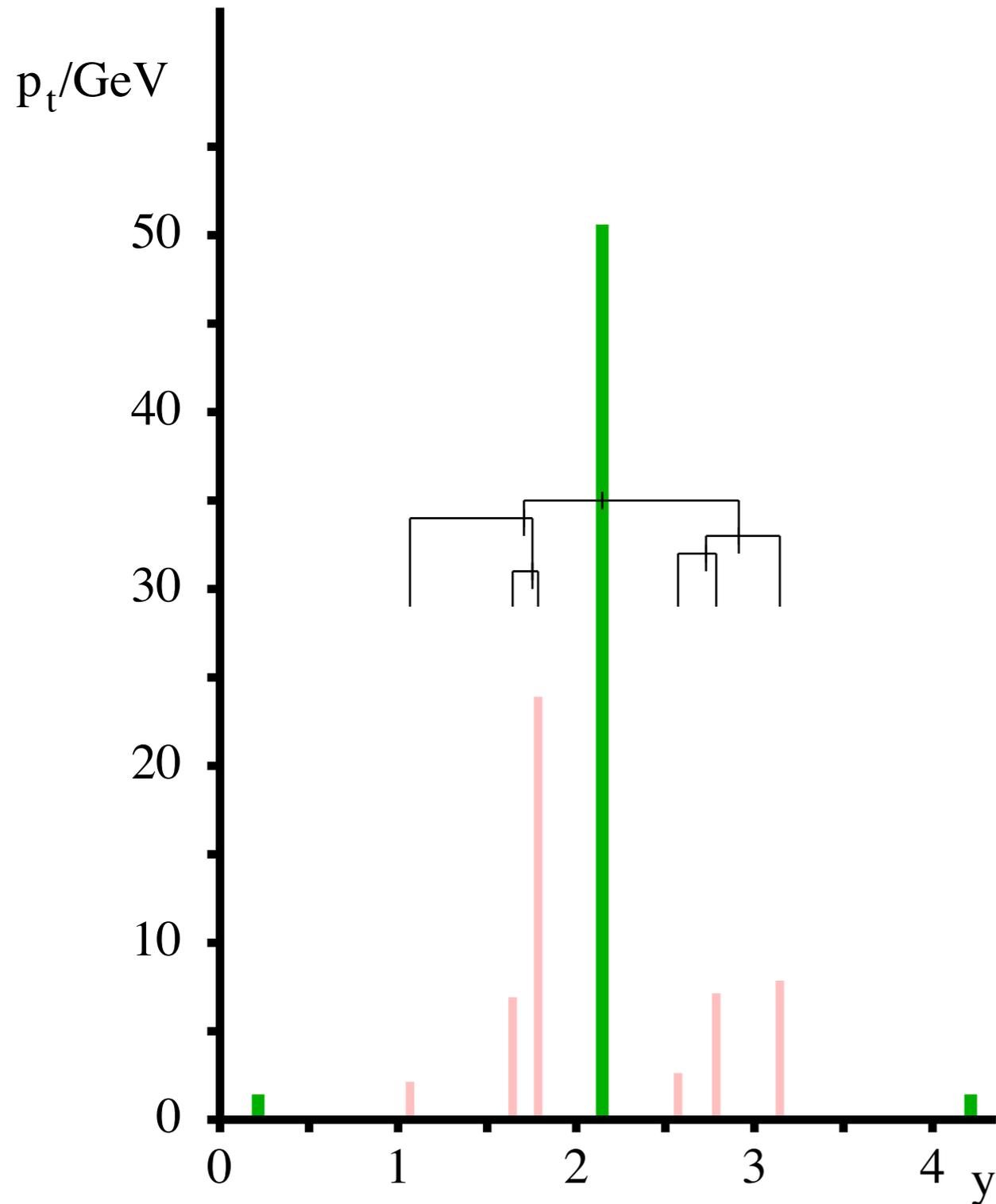
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How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

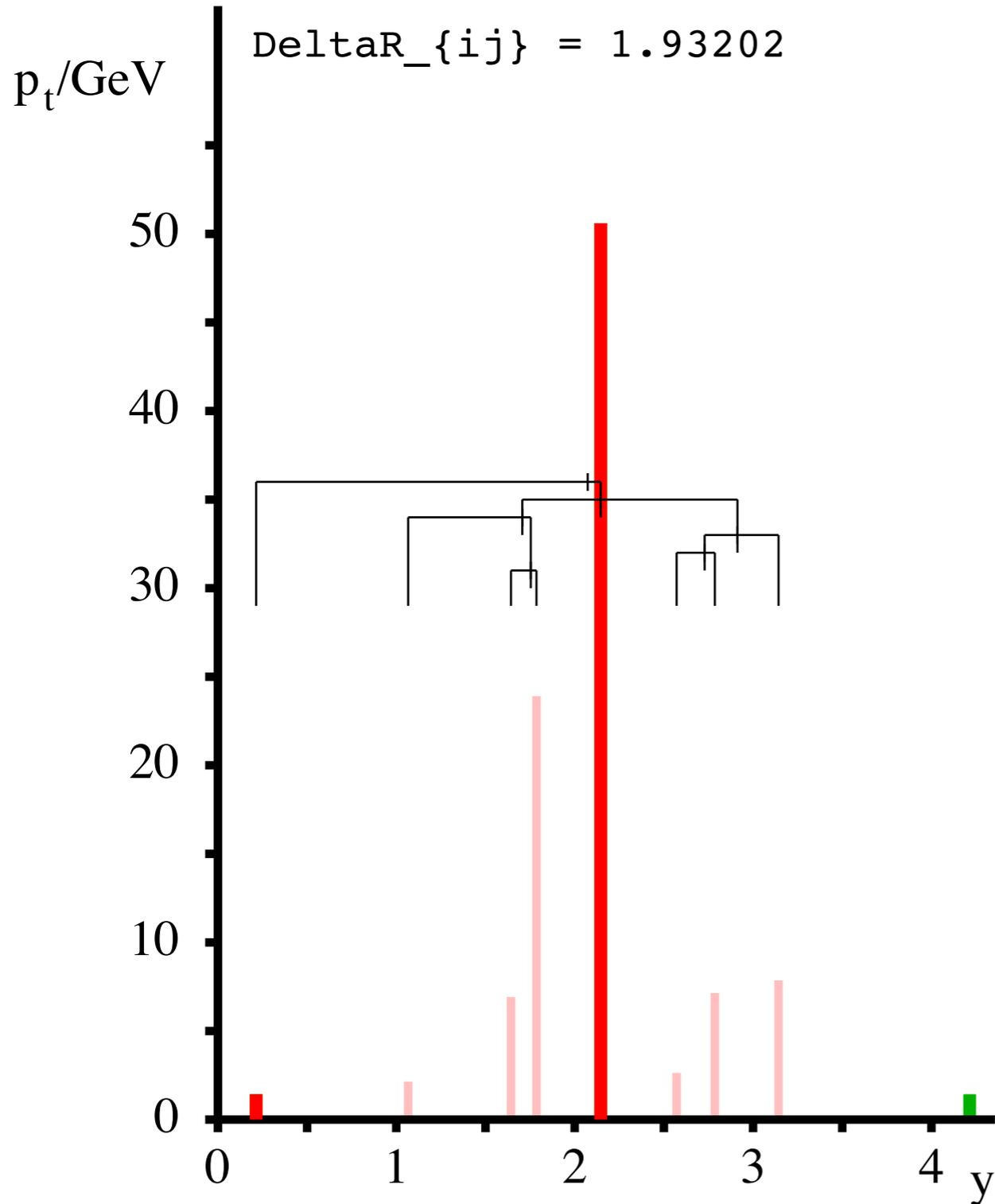
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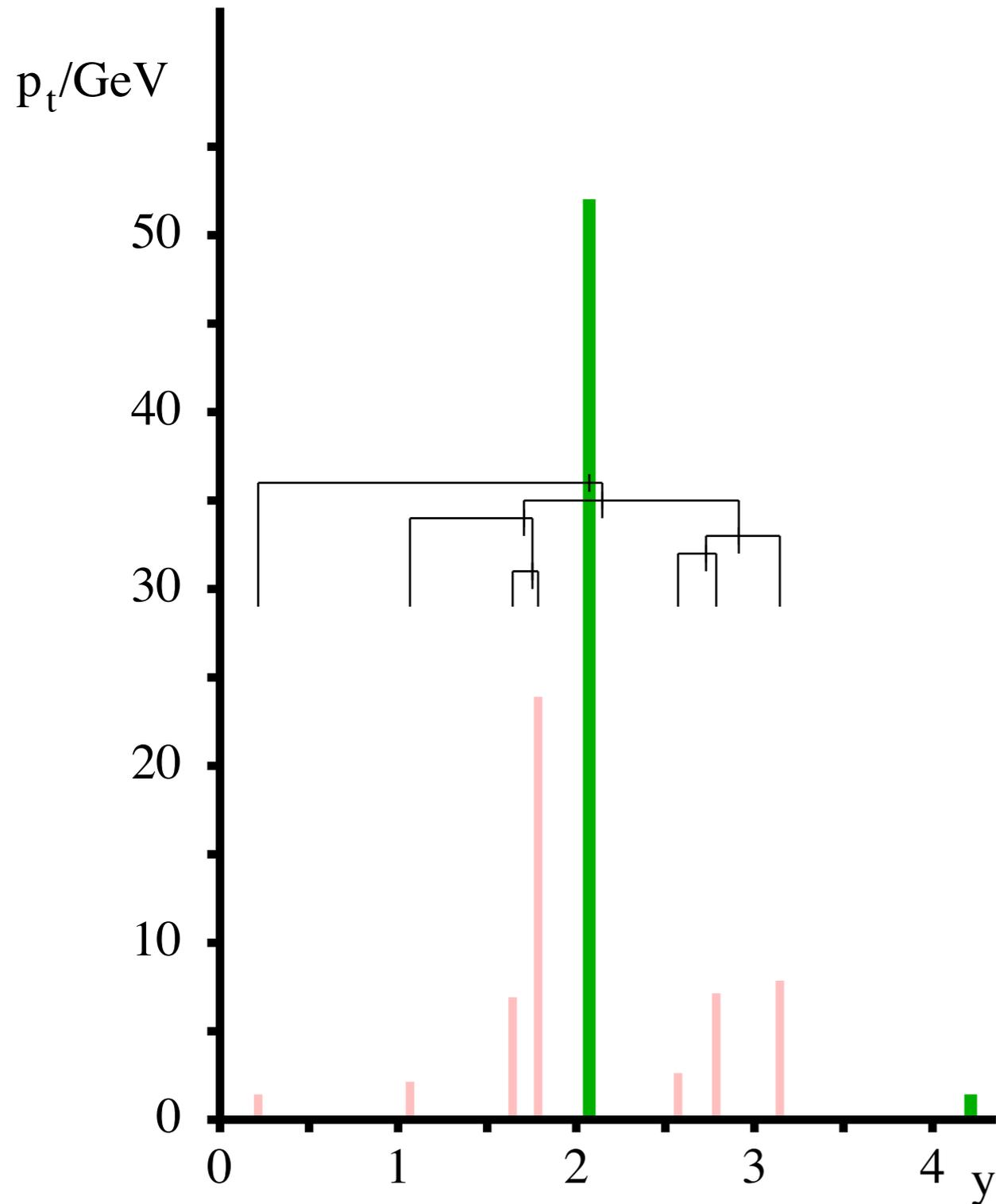
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How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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

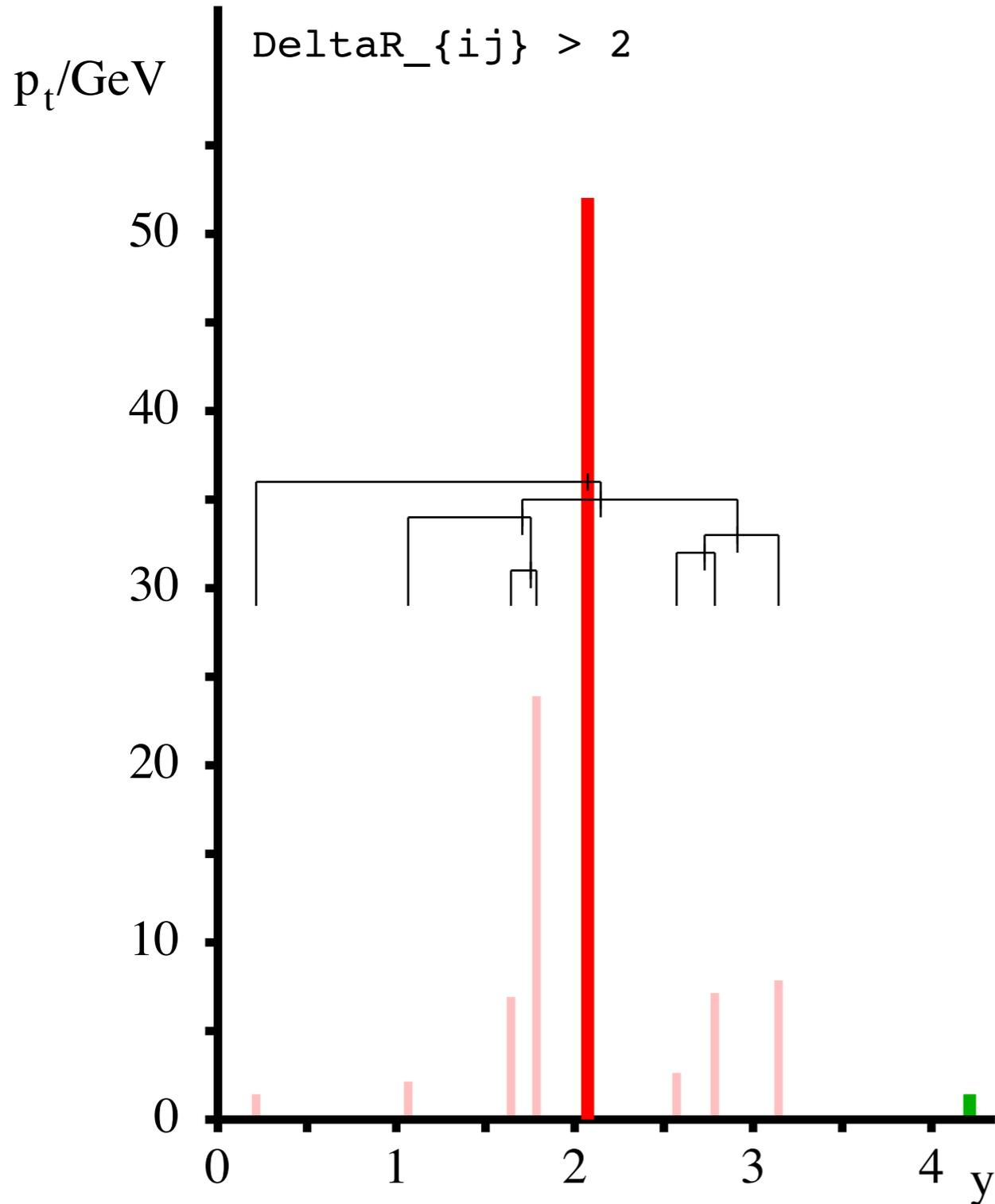
## Cambridge/Aachen algorithm



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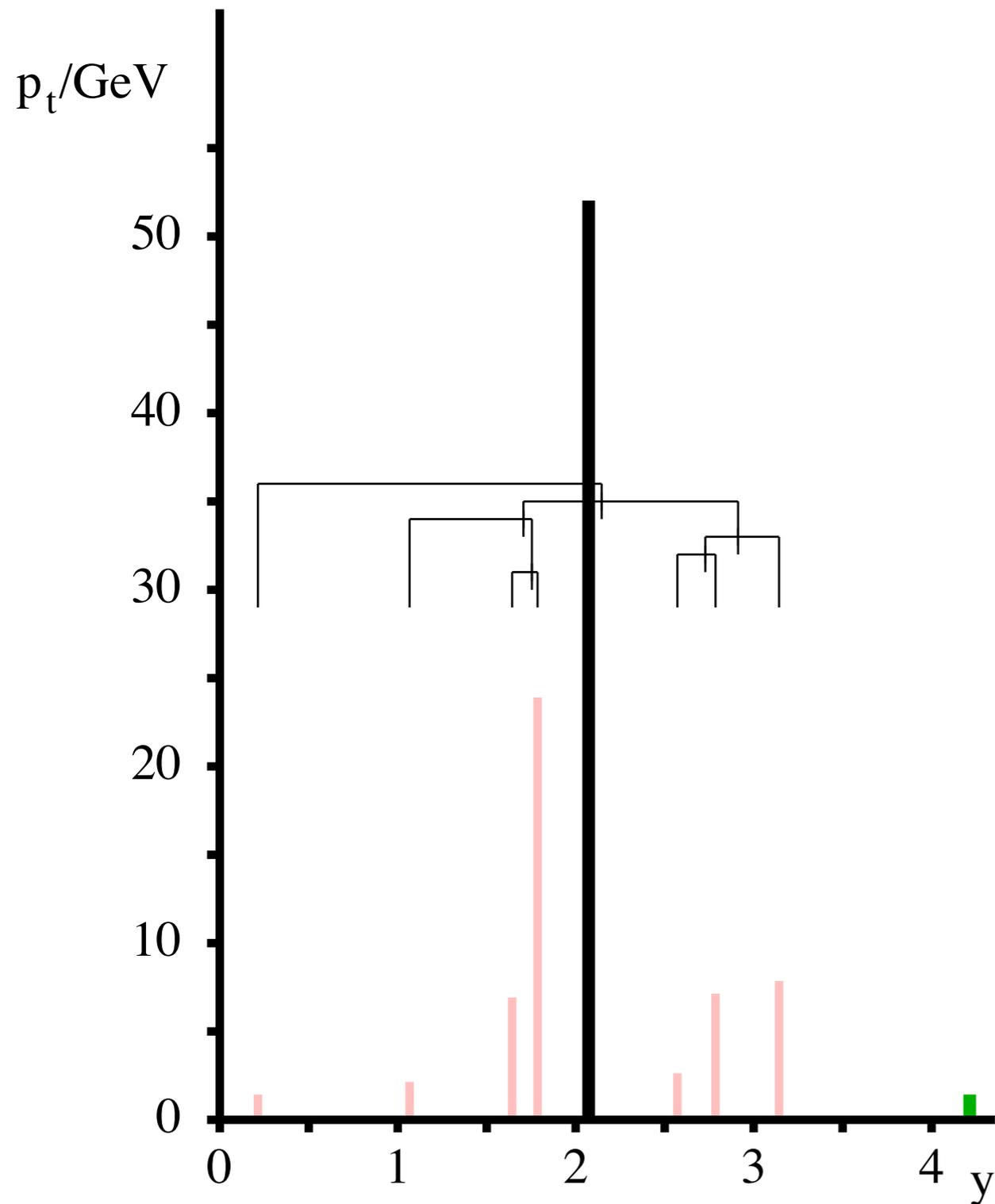
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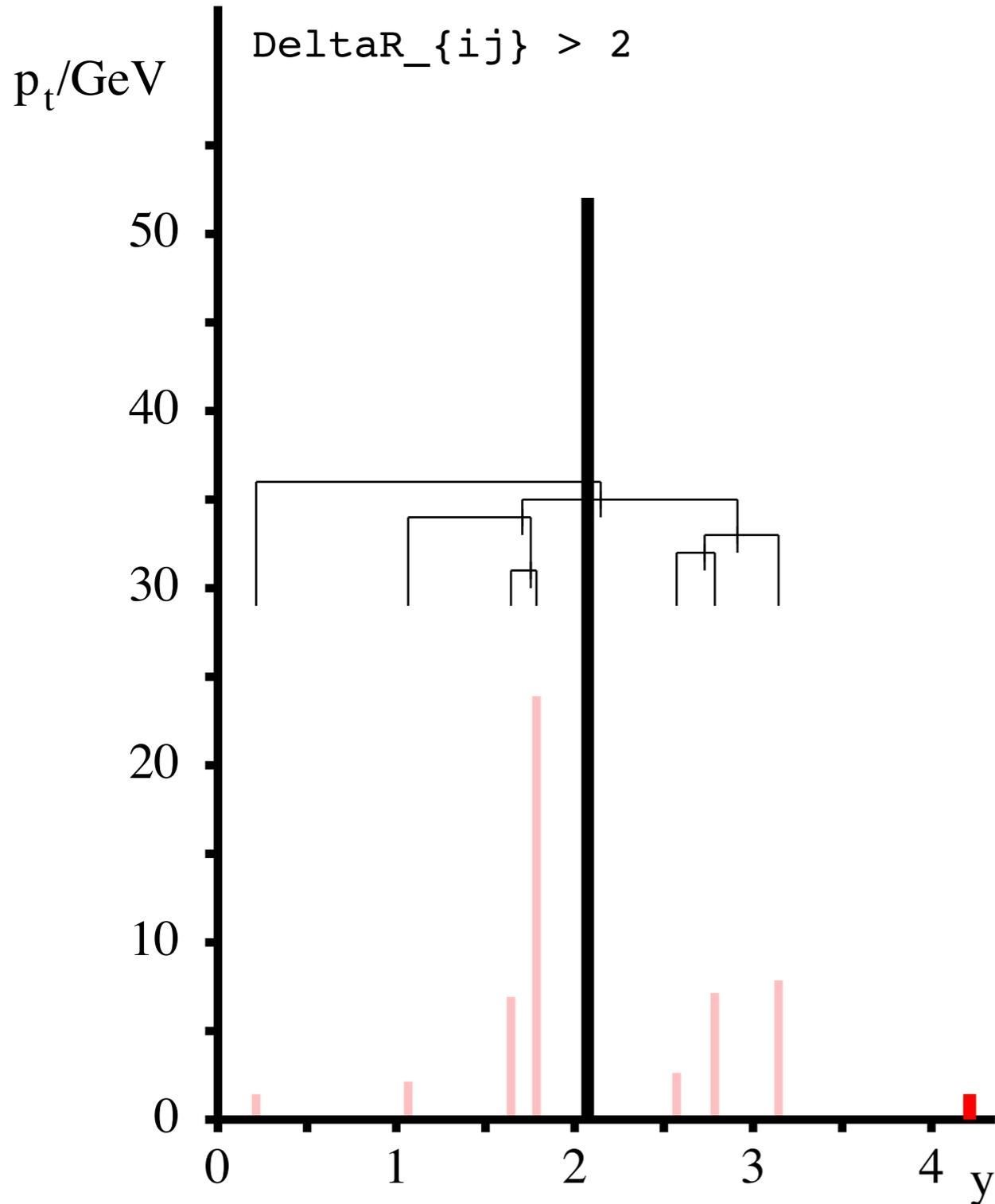
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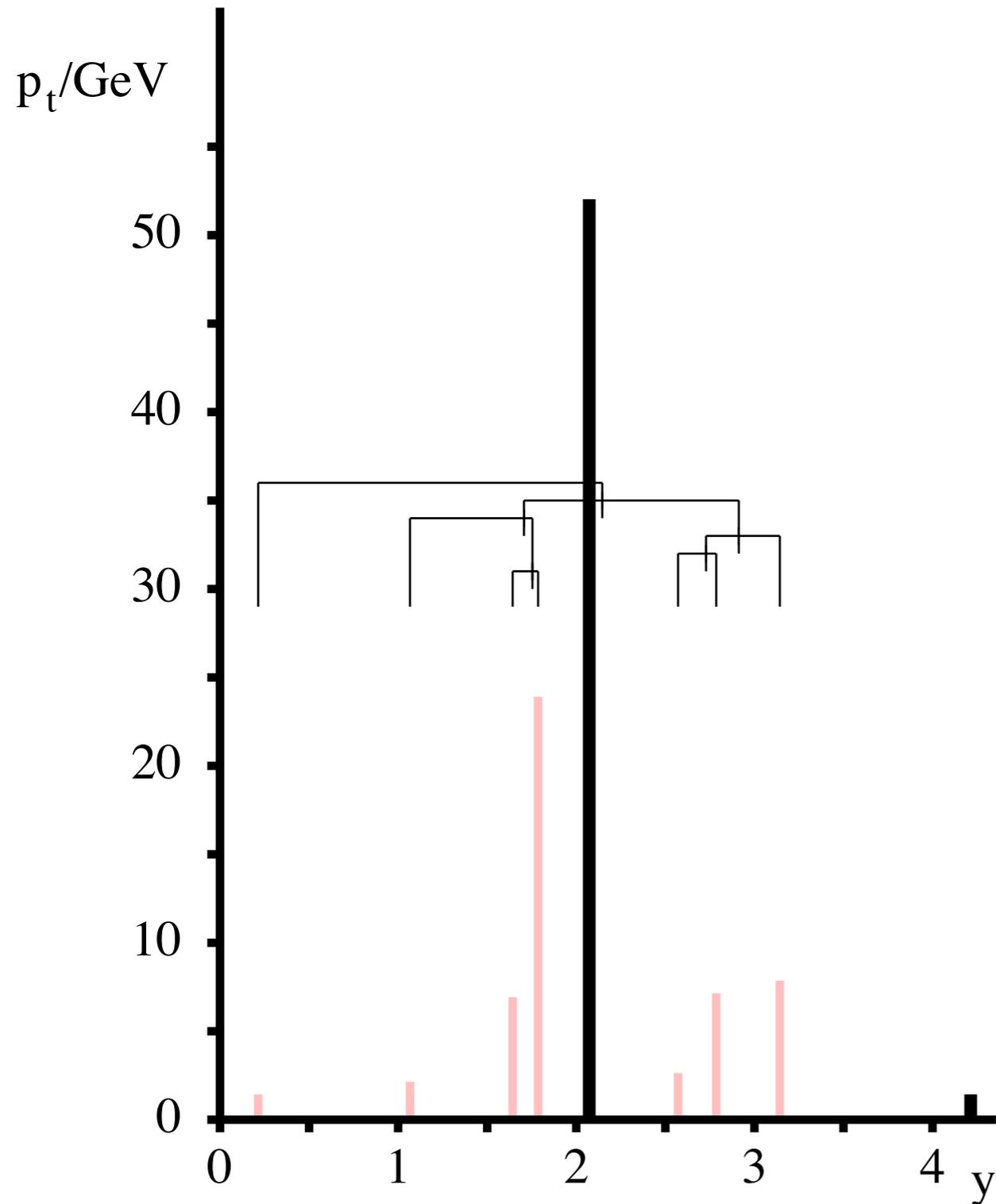
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## Cambridge/Aachen algorithm



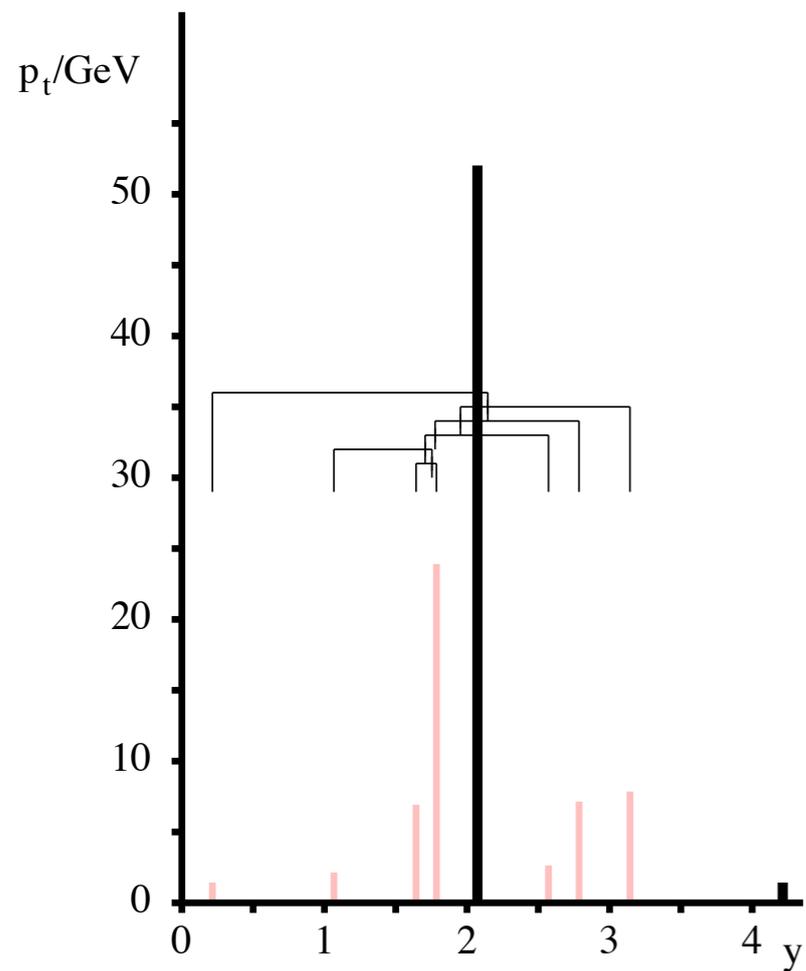
How well can an algorithm identify the “blobs” of energy inside a jet that come from different partons?

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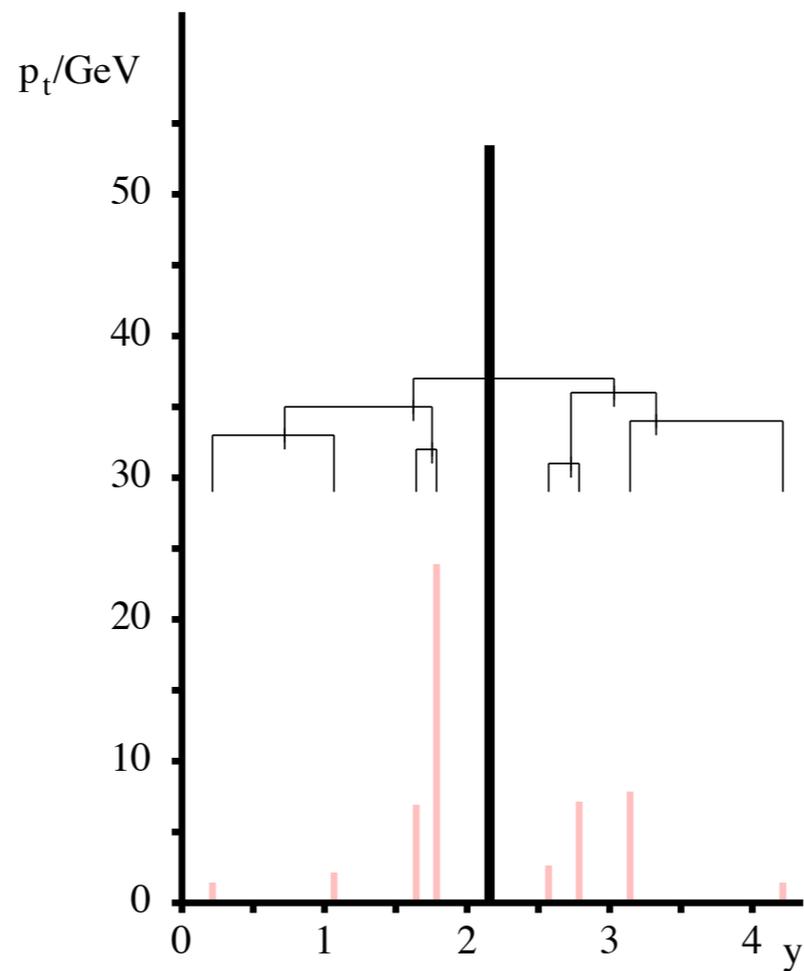
The interesting substructure is buried inside the clustering sequence — **it's less contaminated by soft junk, but needs to be pulled out with special techniques**

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

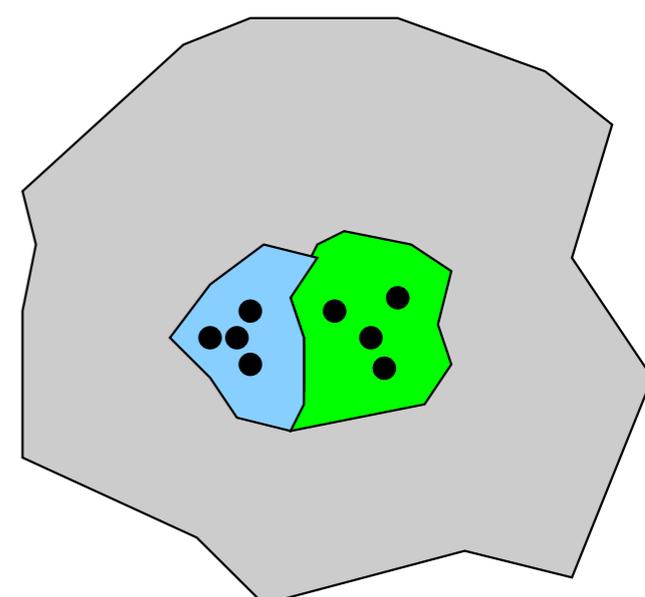
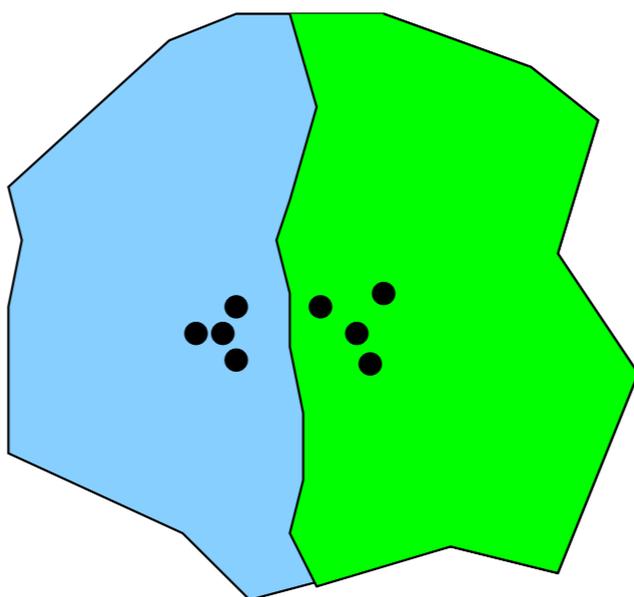
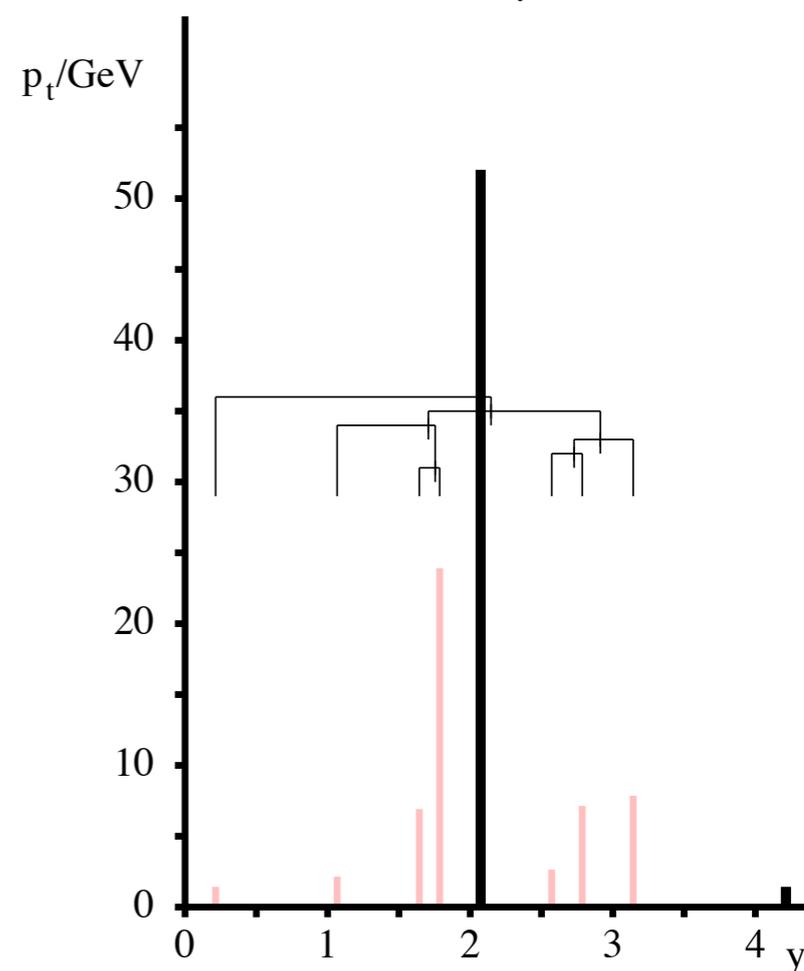
### anti- $k_t$ algorithm



### $k_t$ algorithm



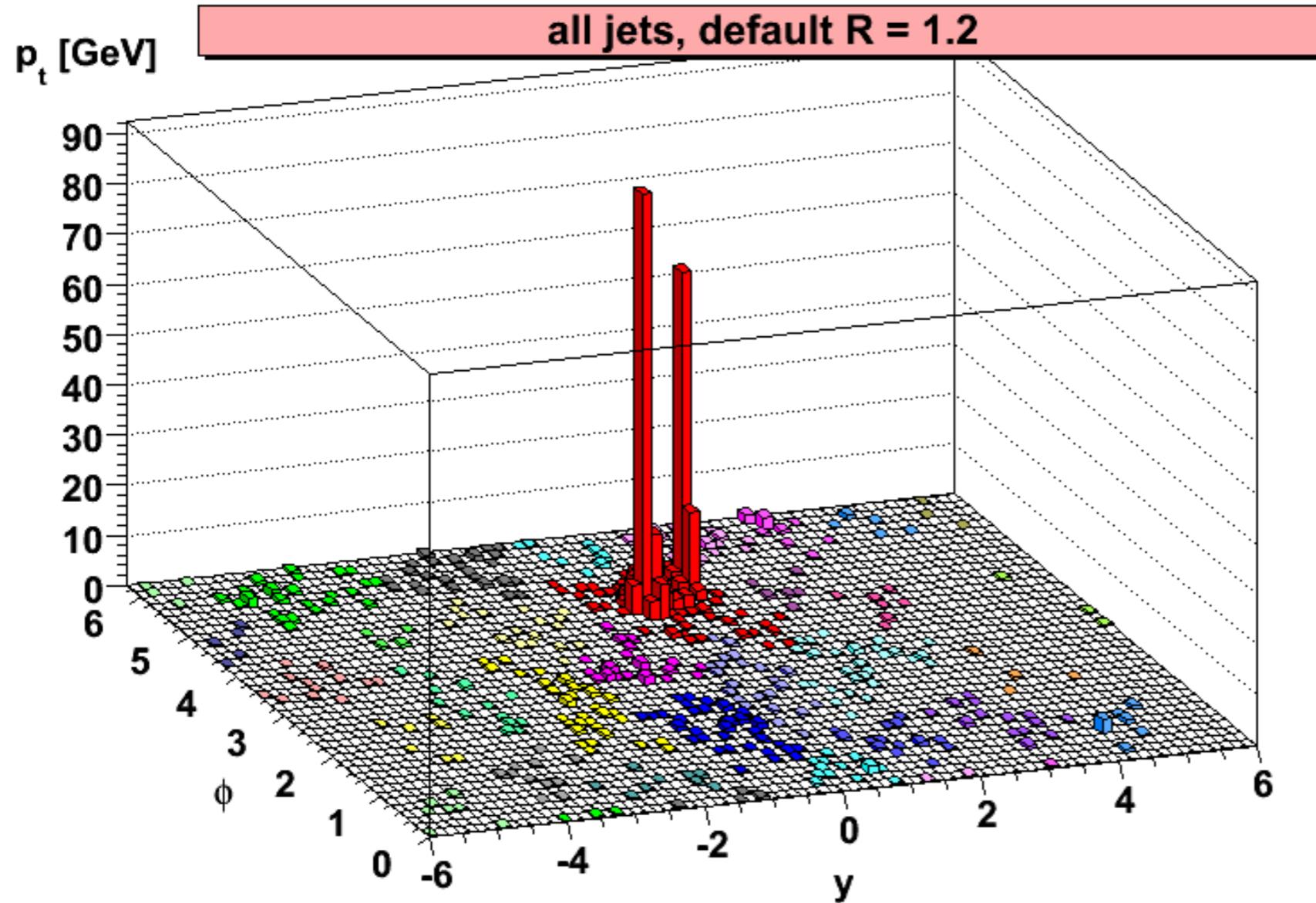
### Cambridge/Aachen



# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

SIGNAL

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



Zbb BACKGROUND

Cluster event, C/A, R=1.2

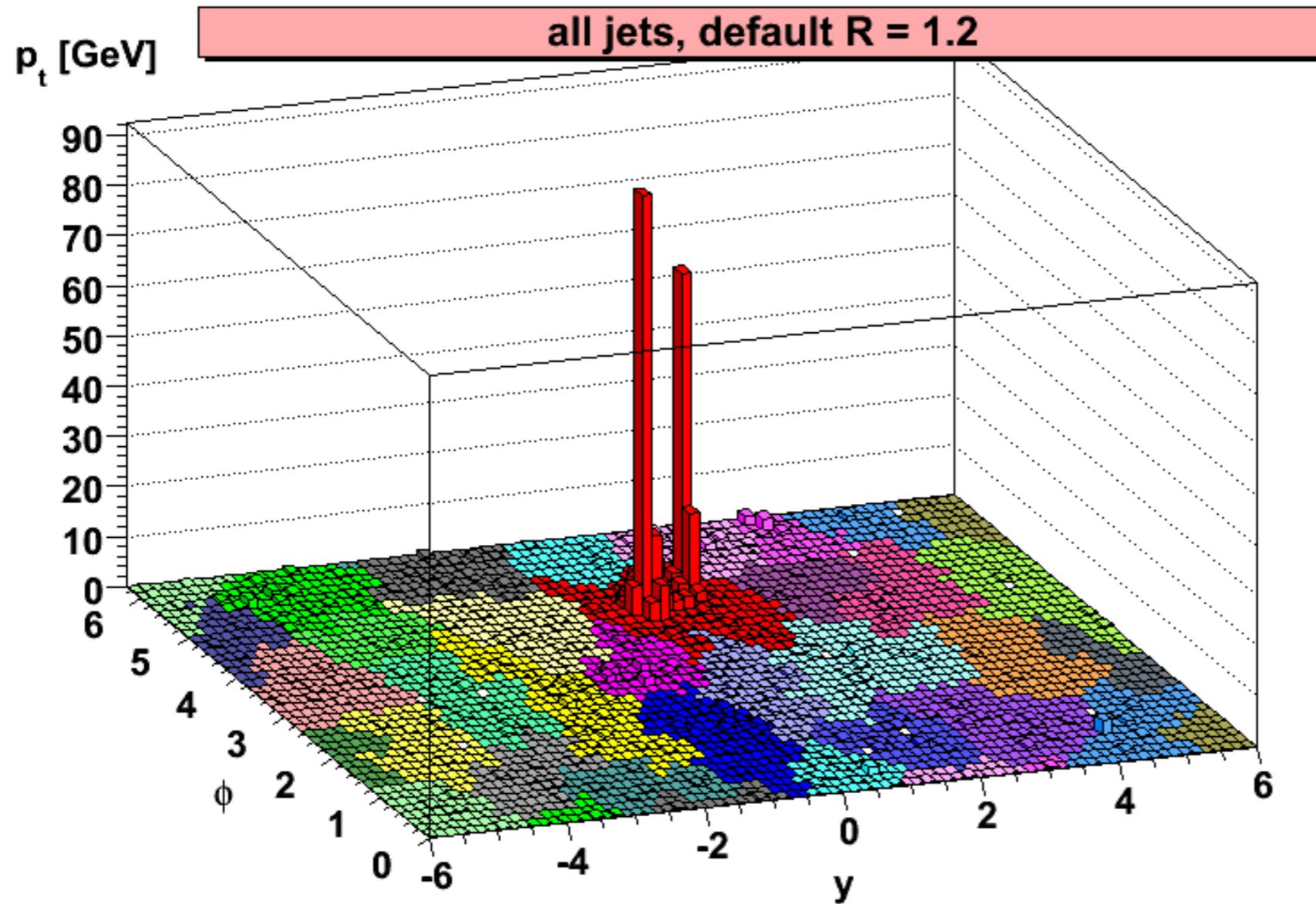
Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

SIGNAL

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



Zbb BACKGROUND

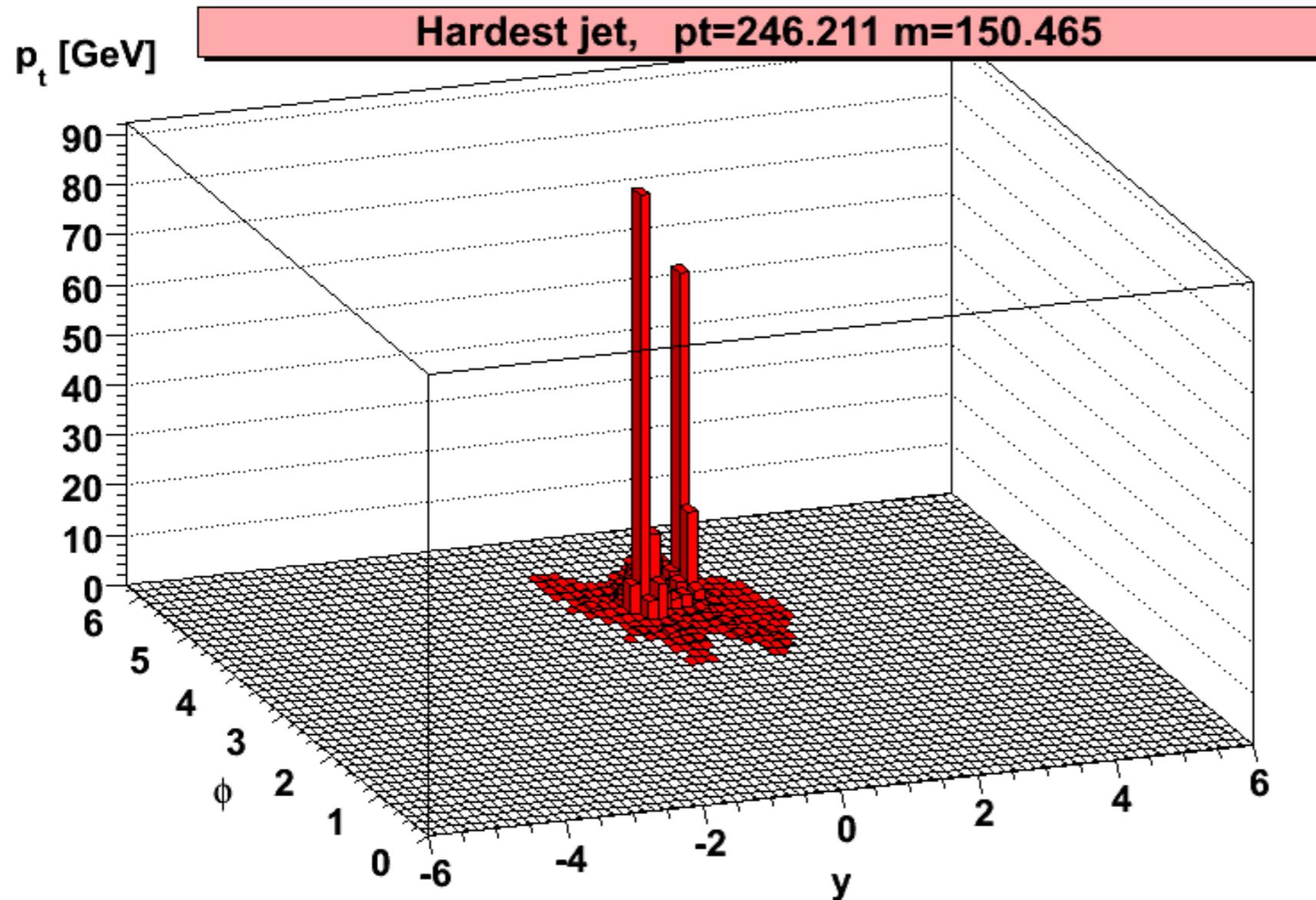
Fill it in,  $\rightarrow$  show jets more clearly

Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}, @14\text{ TeV}, m_H = 115\text{ GeV}$$

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3

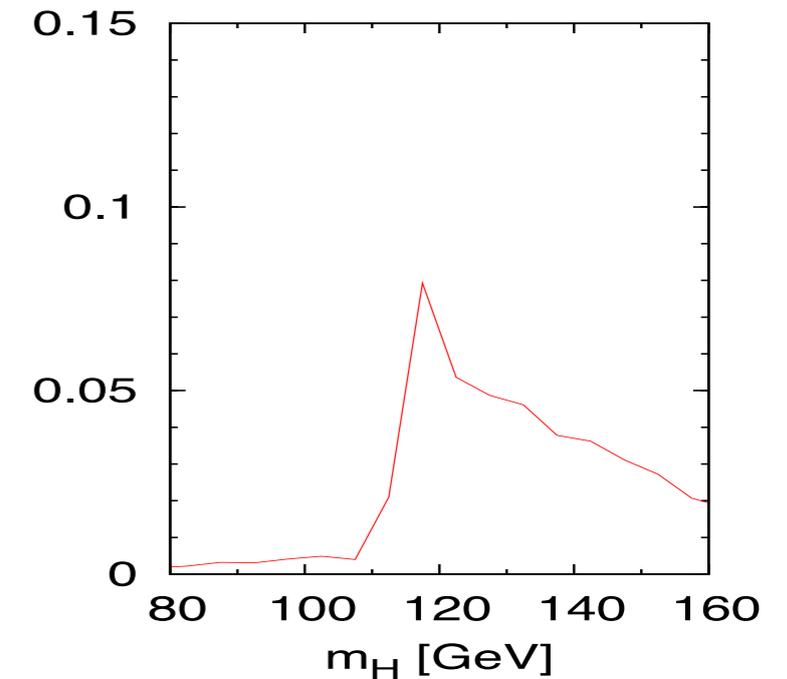


Consider hardest jet,  $m = 150$  GeV

Butterworth, Davison, Rubin & GPS '08

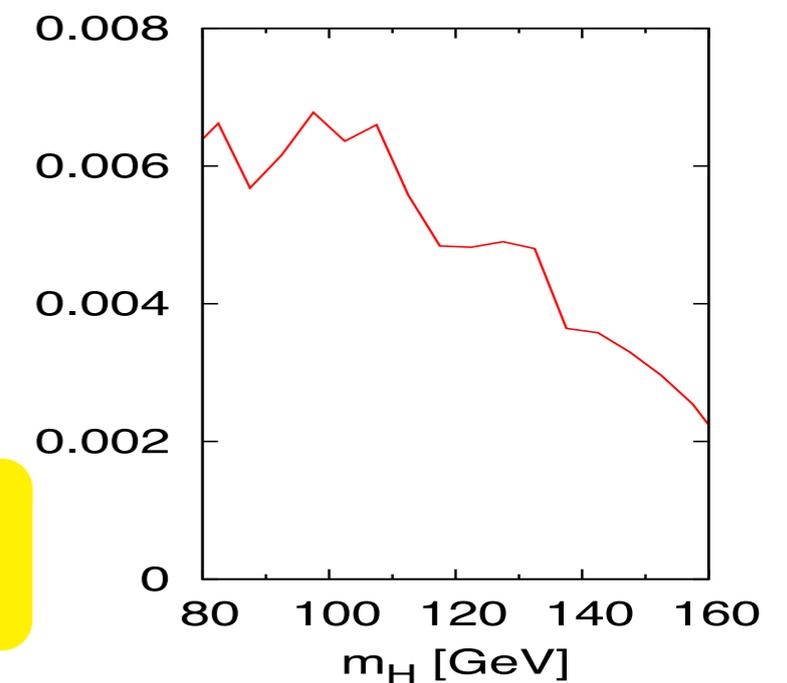
SIGNAL

$200 < p_{tZ} < 250$  GeV



Zbb BACKGROUND

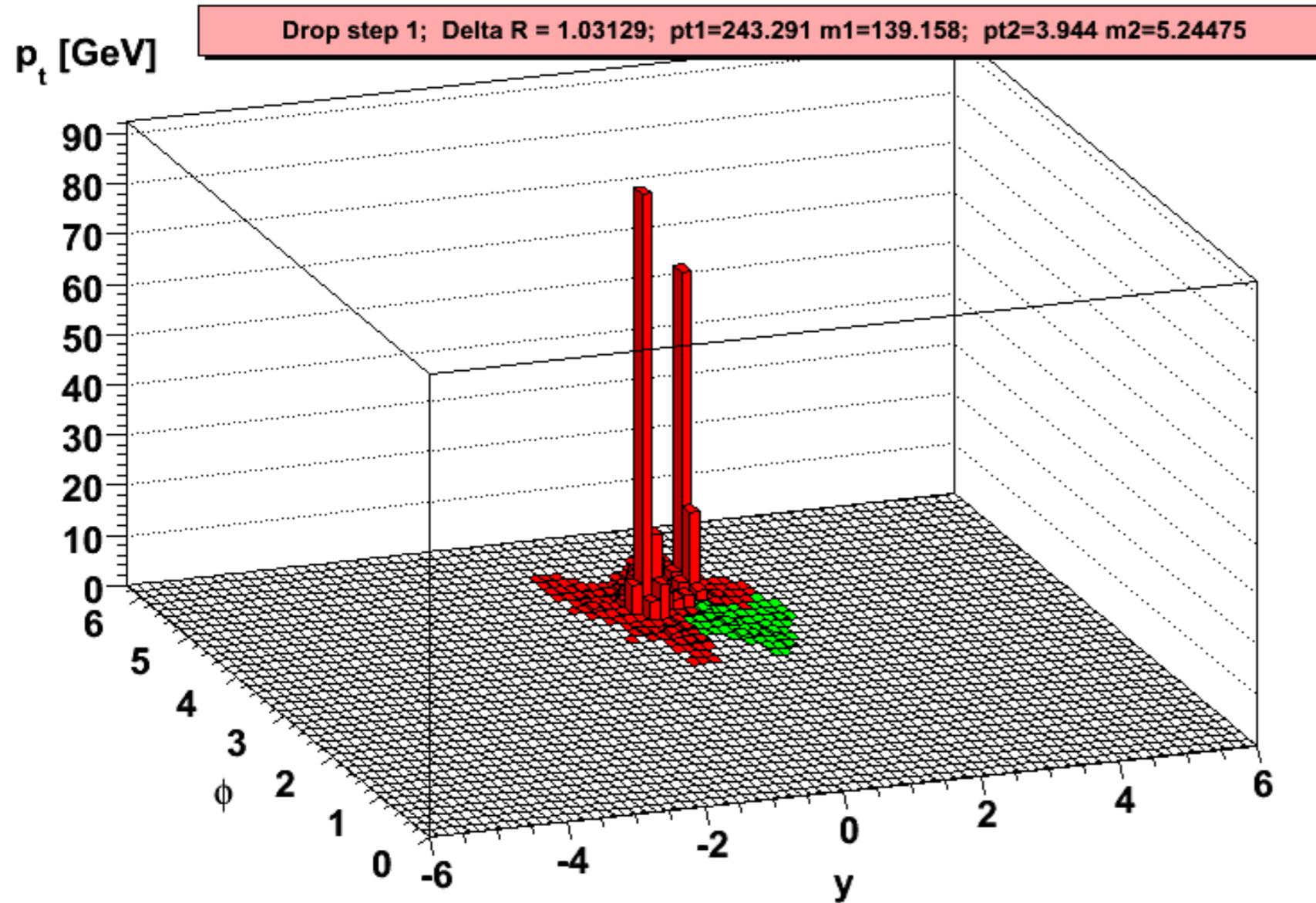
$200 < p_{tZ} < 250$  GeV



arbitrary norm.

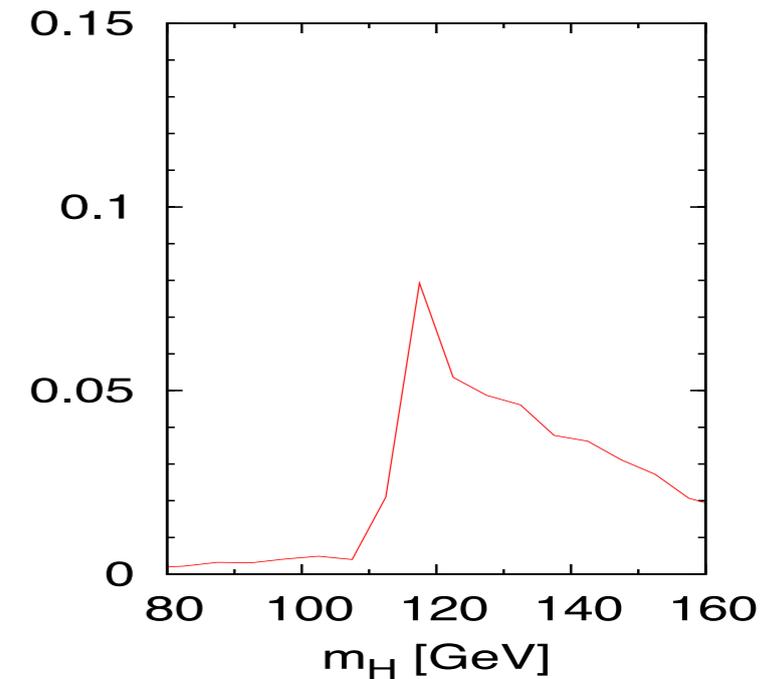
# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



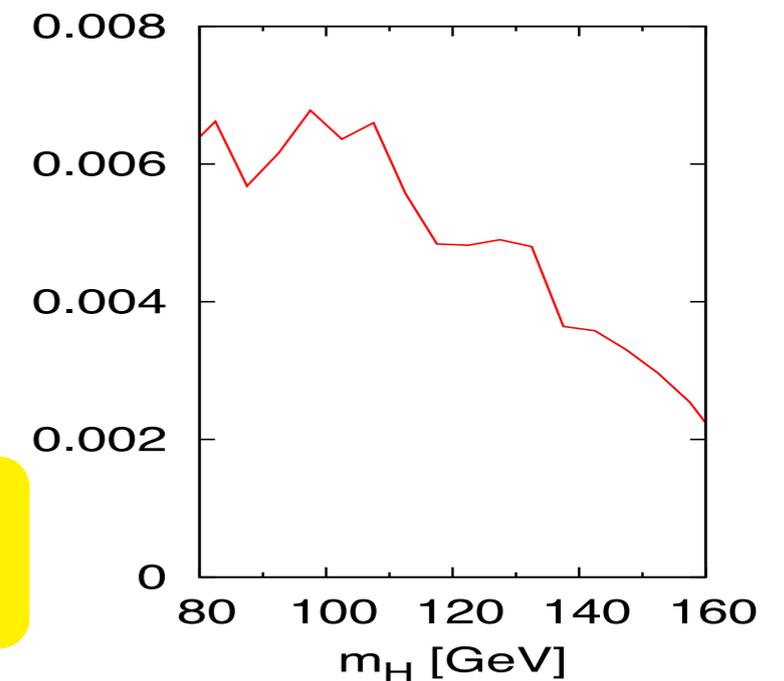
## SIGNAL

$200 < p_{tZ} < 250$  GeV



## Zbb BACKGROUND

$200 < p_{tZ} < 250$  GeV



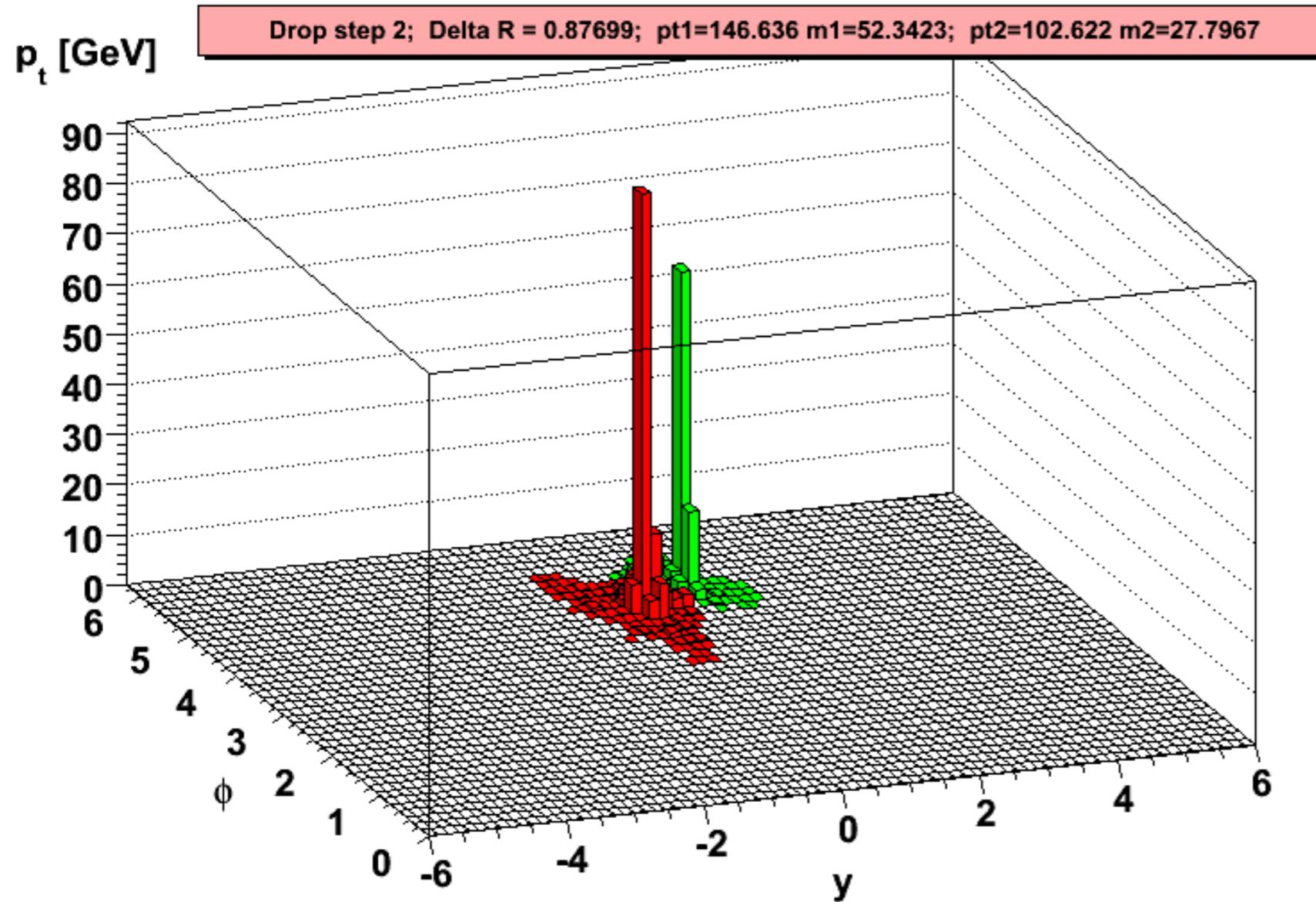
split:  $m = 150$  GeV,  $\frac{\max(m_1, m_2)}{m} = 0.92 \rightarrow$  repeat

Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

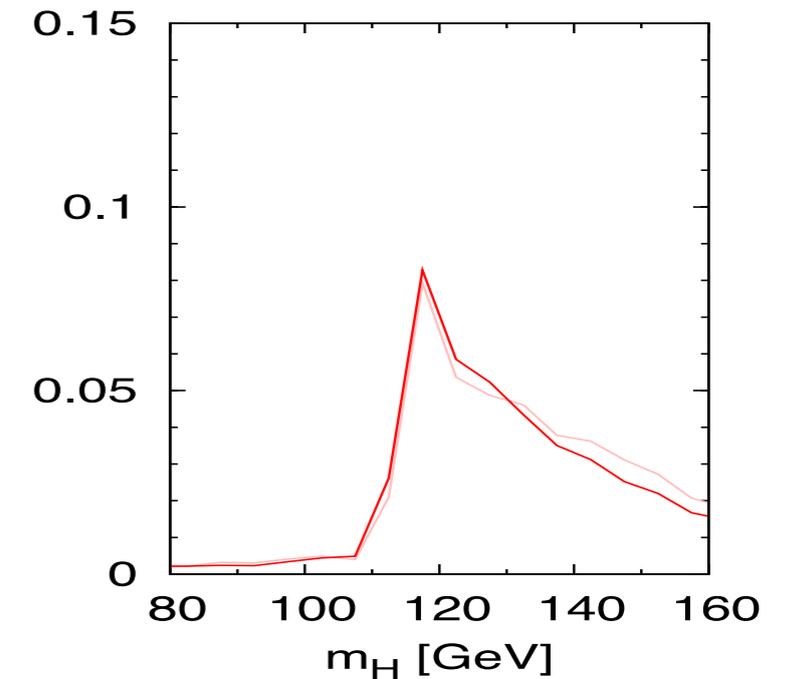
# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



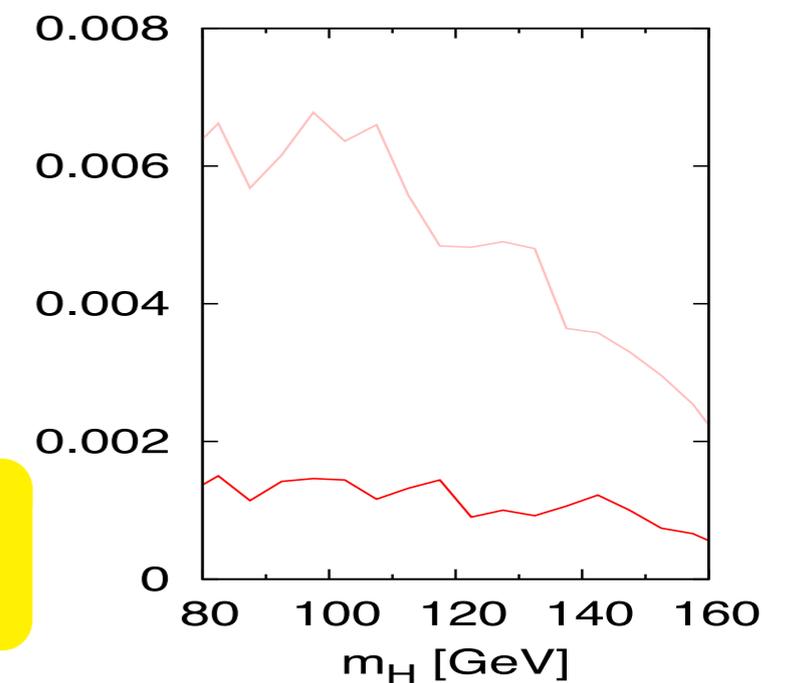
## SIGNAL

$200 < p_{tZ} < 250$  GeV



## Zbb BACKGROUND

$200 < p_{tZ} < 250$  GeV



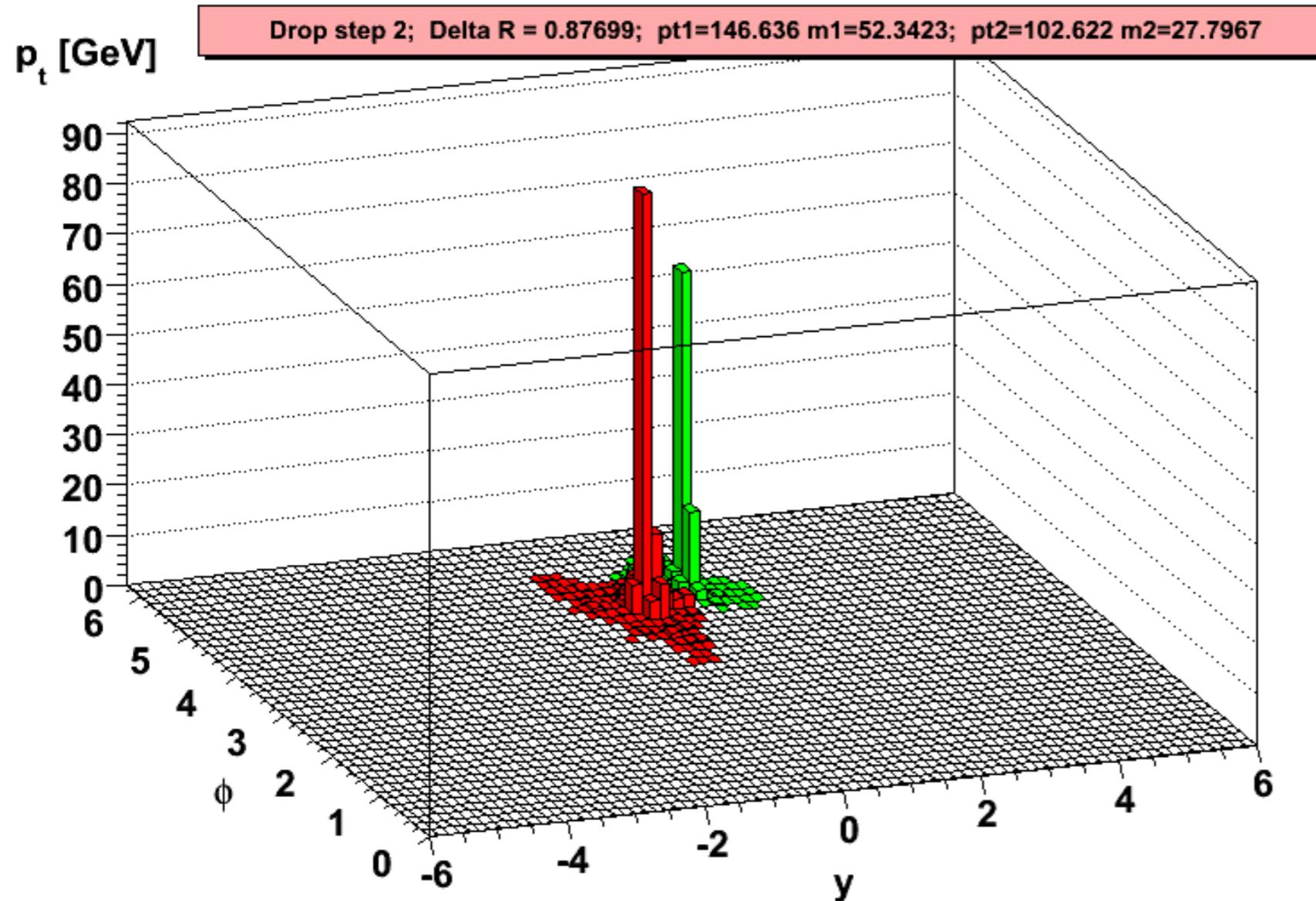
split:  $m = 139$  GeV,  $\frac{\max(m_1, m_2)}{m} = 0.37 \rightarrow$  mass drop

Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

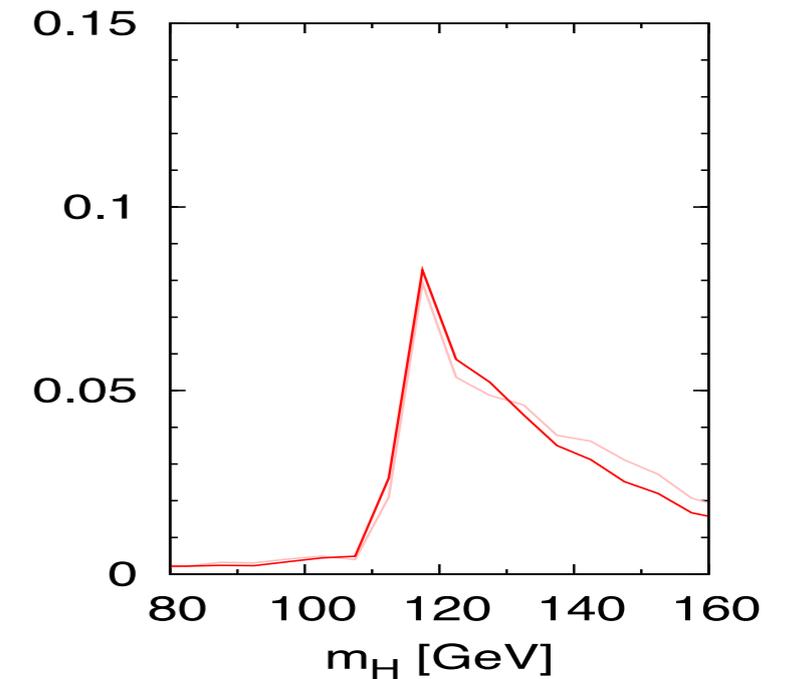
# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



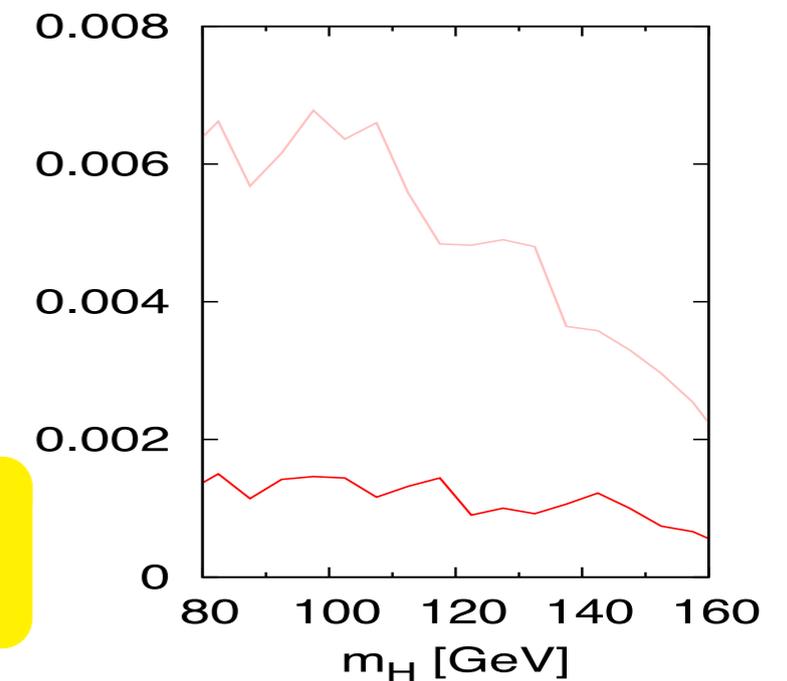
## SIGNAL

$200 < p_{tZ} < 250$  GeV



## Zbb BACKGROUND

$200 < p_{tZ} < 250$  GeV



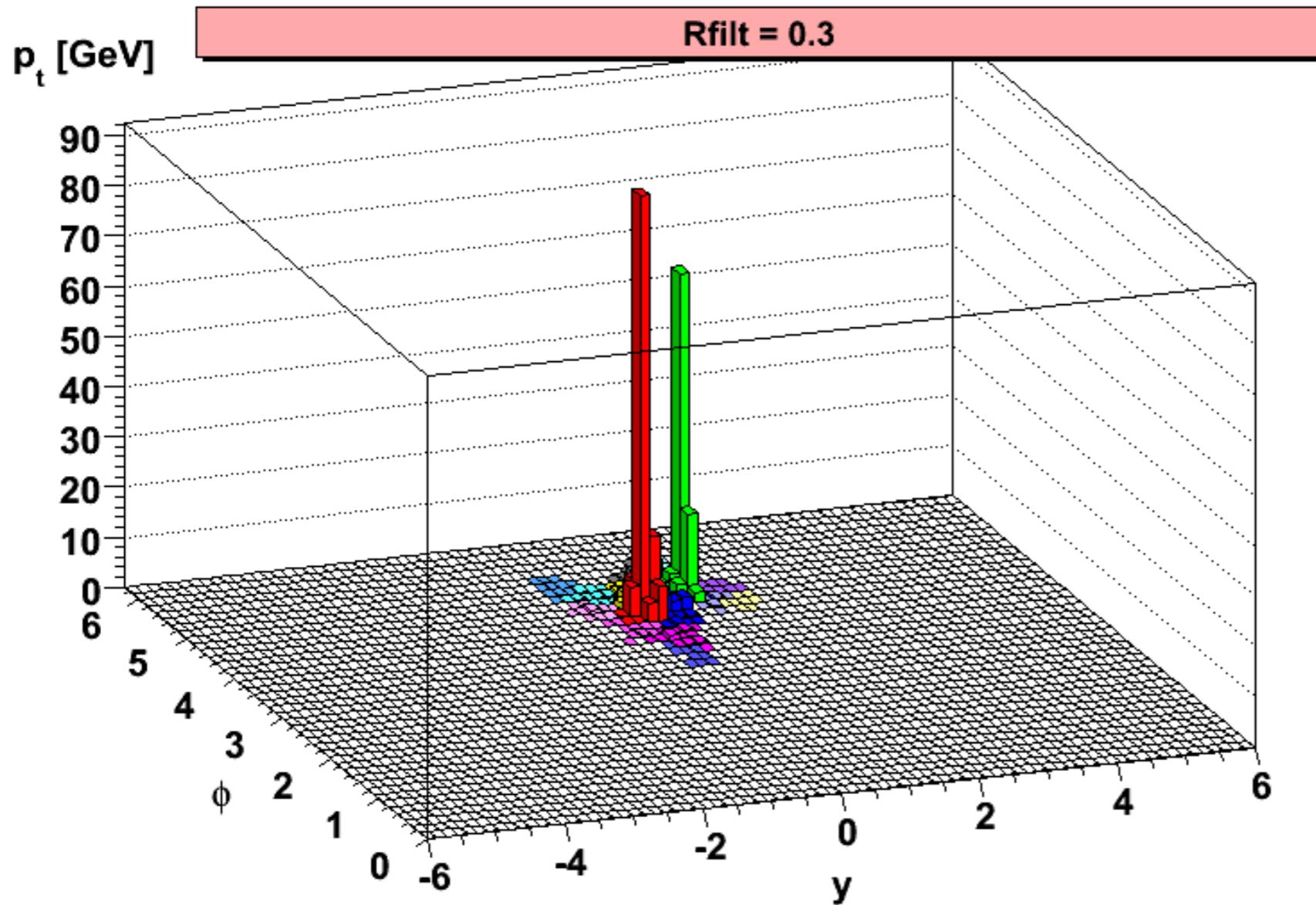
check:  $y_{12} \simeq \frac{p_{t2}}{p_{t1}} \simeq 0.7 \rightarrow \text{OK} + 2 \text{ } b\text{-tags (anti-QCD)}$

Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3

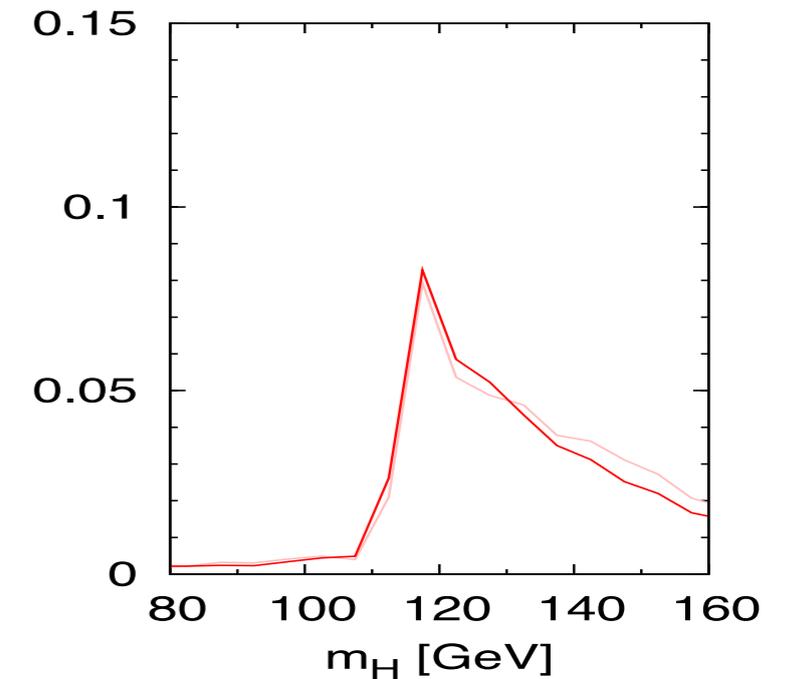


$R_{filt} = 0.3$

Butterworth, Davison, Rubin & GPS '08

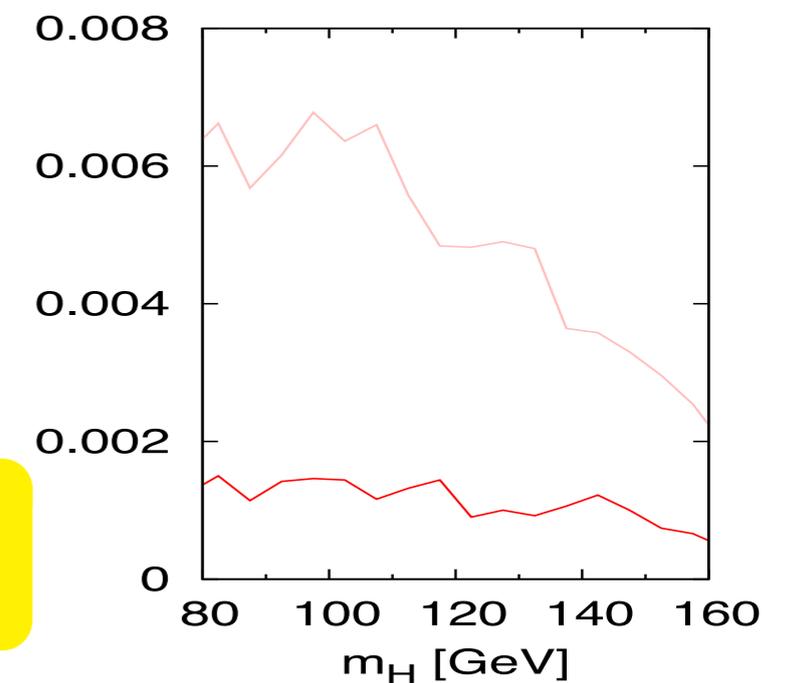
## SIGNAL

$200 < p_{tZ} < 250$  GeV



## Zbb BACKGROUND

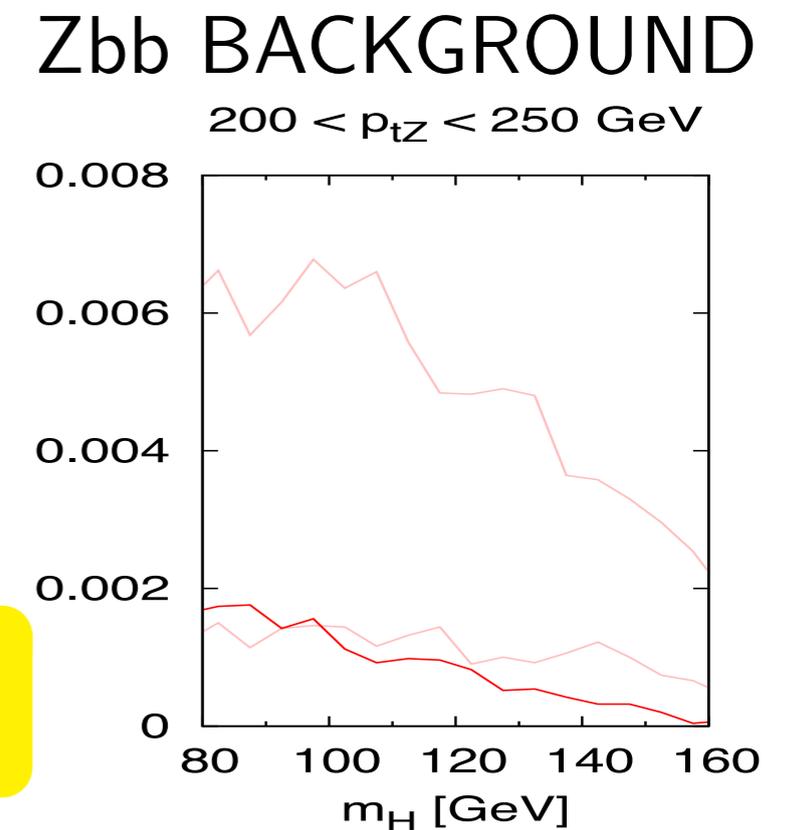
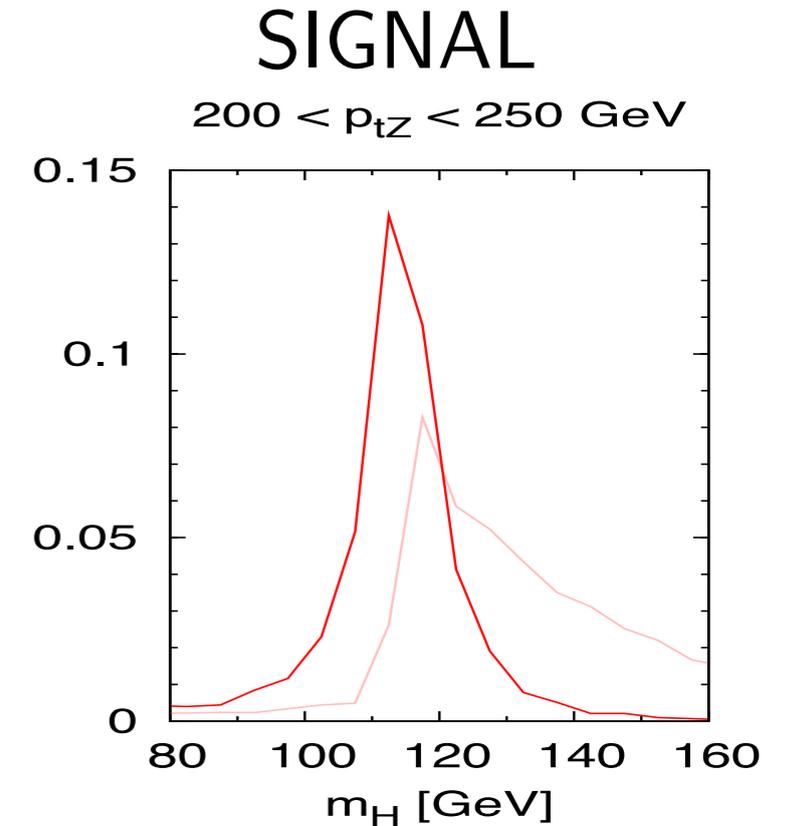
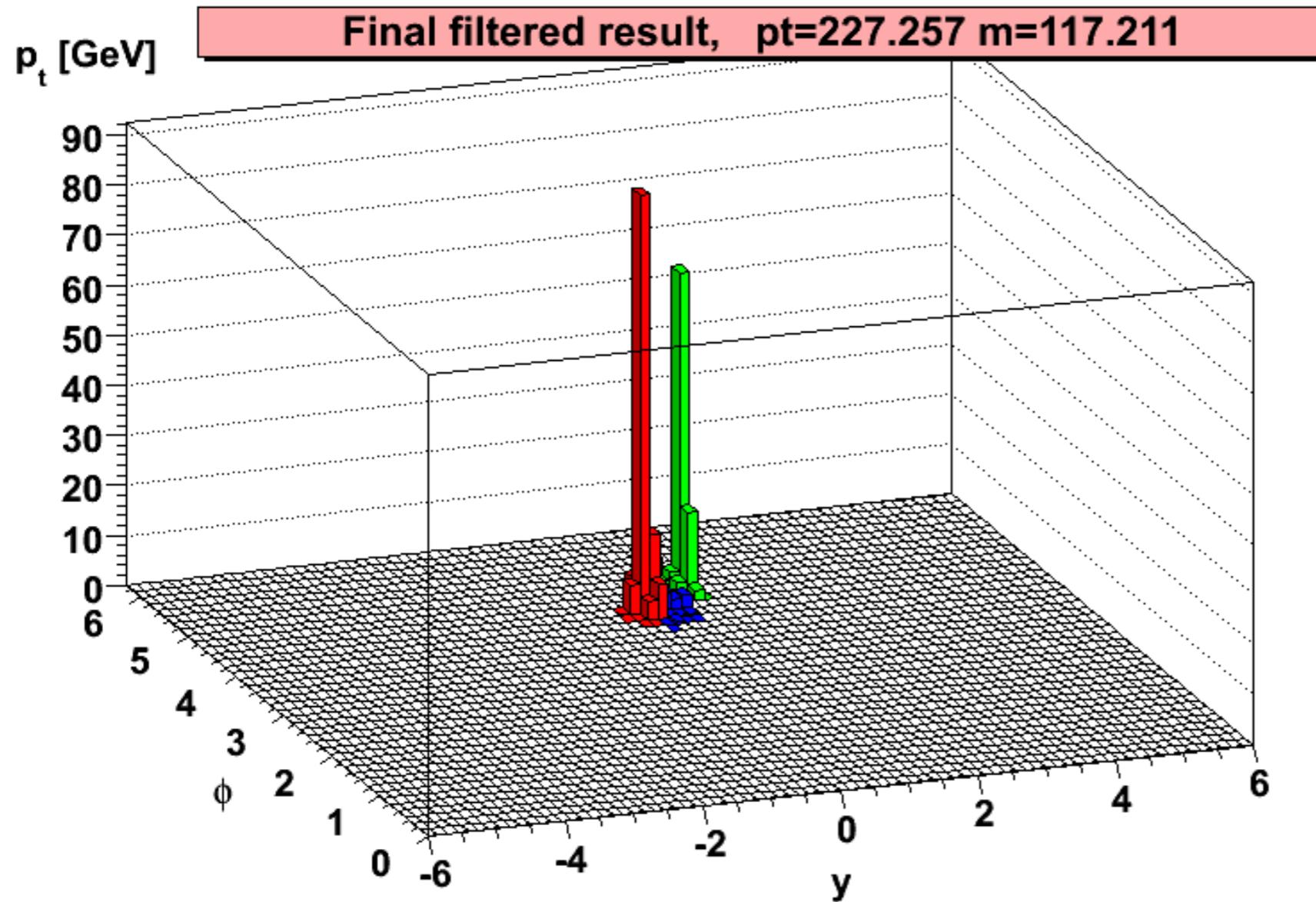
$200 < p_{tZ} < 250$  GeV



arbitrary norm.

# $pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}$ , @14 TeV, $m_H = 115$ GeV

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



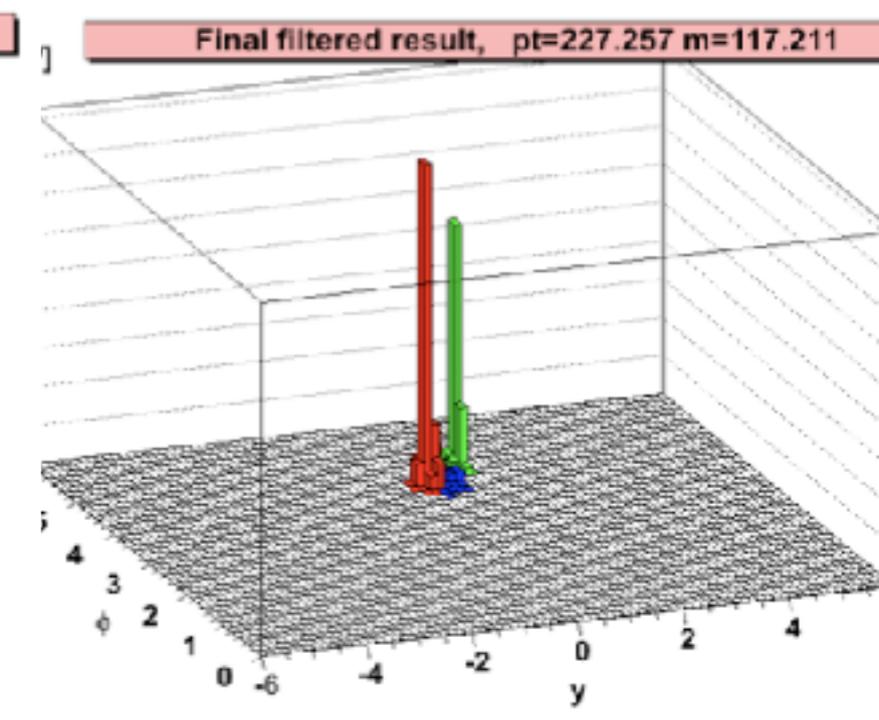
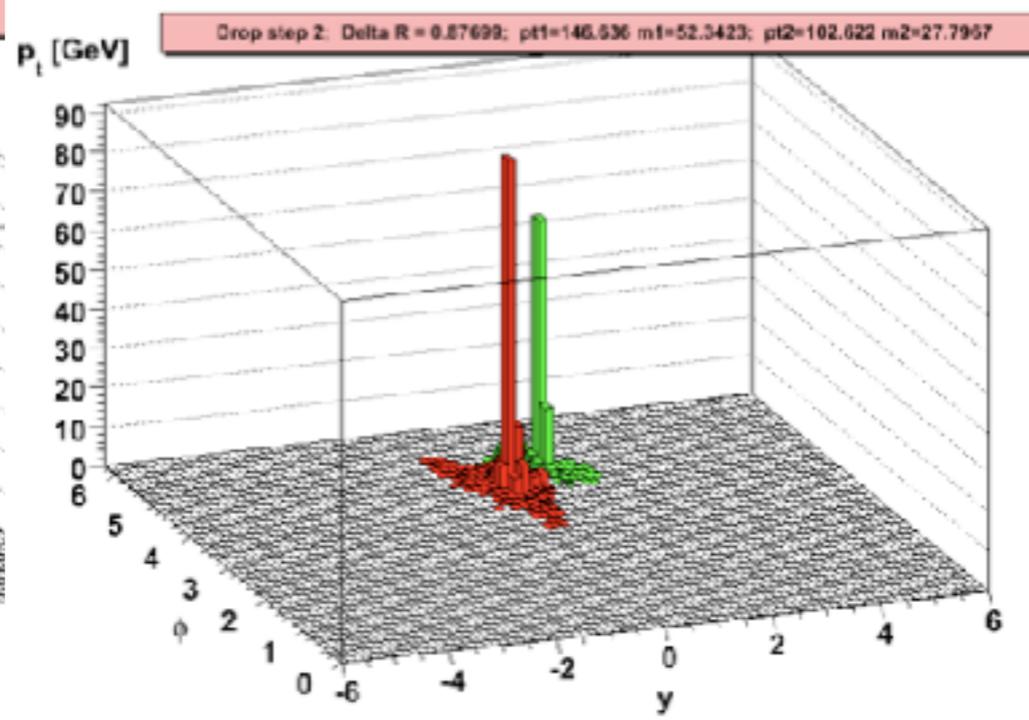
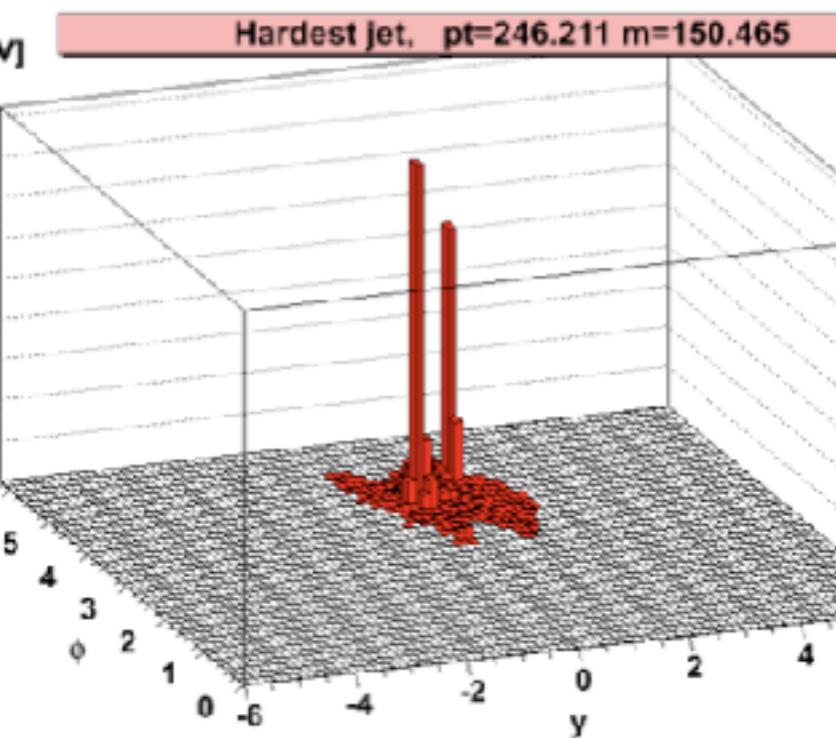
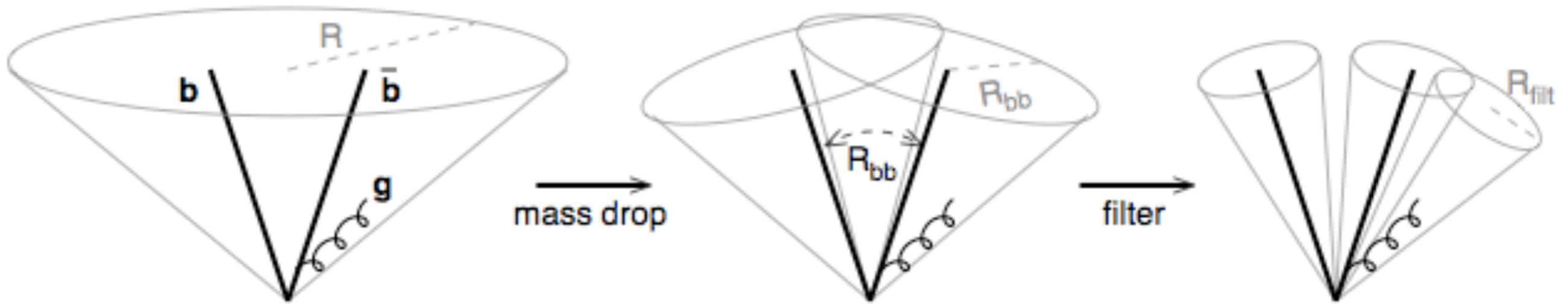
$R_{filt} = 0.3$ : take 3 hardest,  $m = 117$  GeV

Butterworth, Davison, Rubin & GPS '08

arbitrary norm.

# Boosted Higgs analysis

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



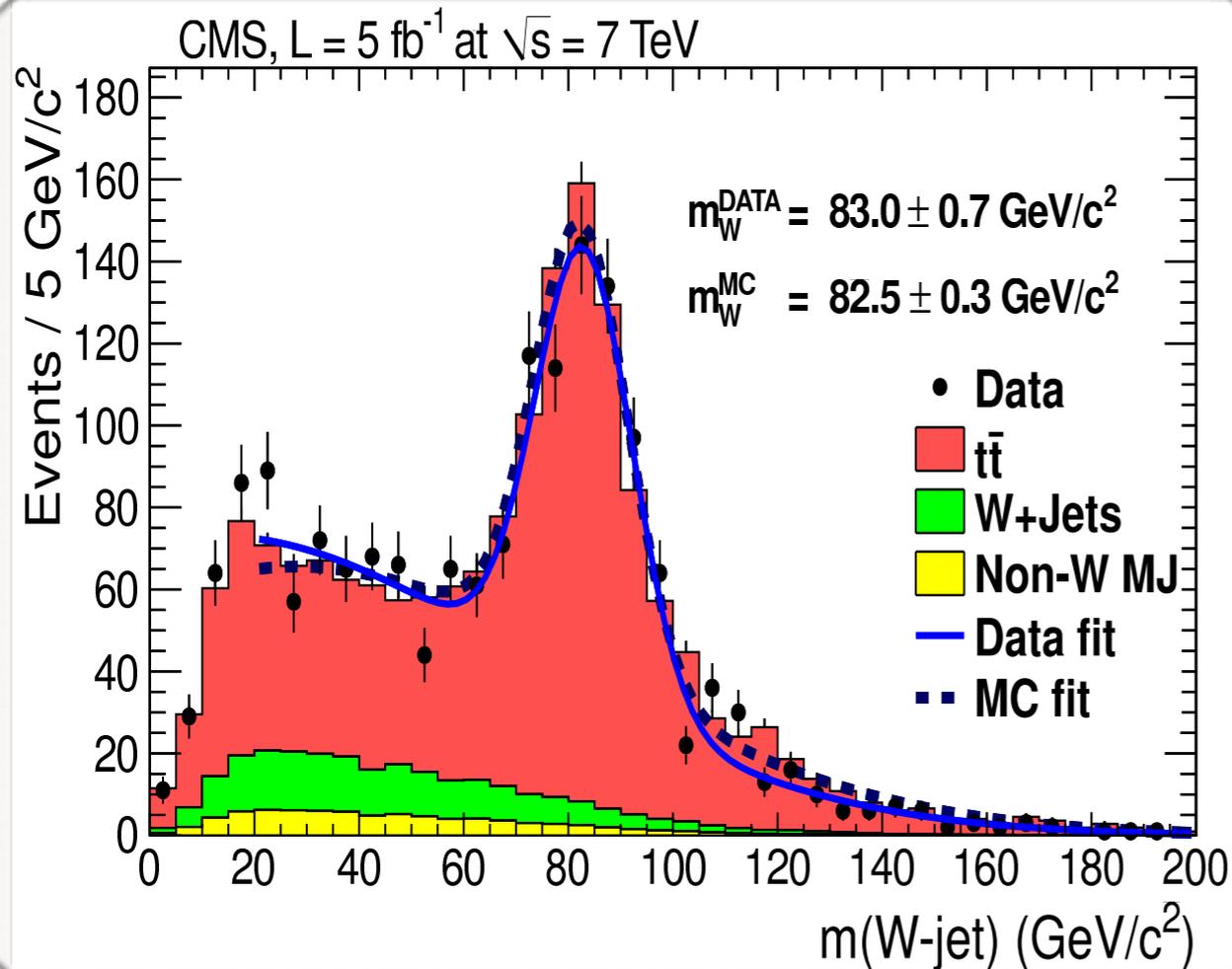
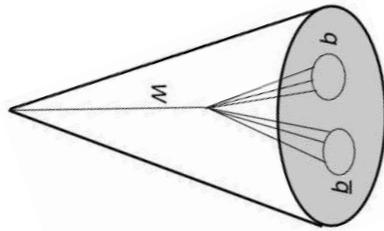
Cluster with a large  $R$

Undo the clustering into subjets, until a large mass drop is observed

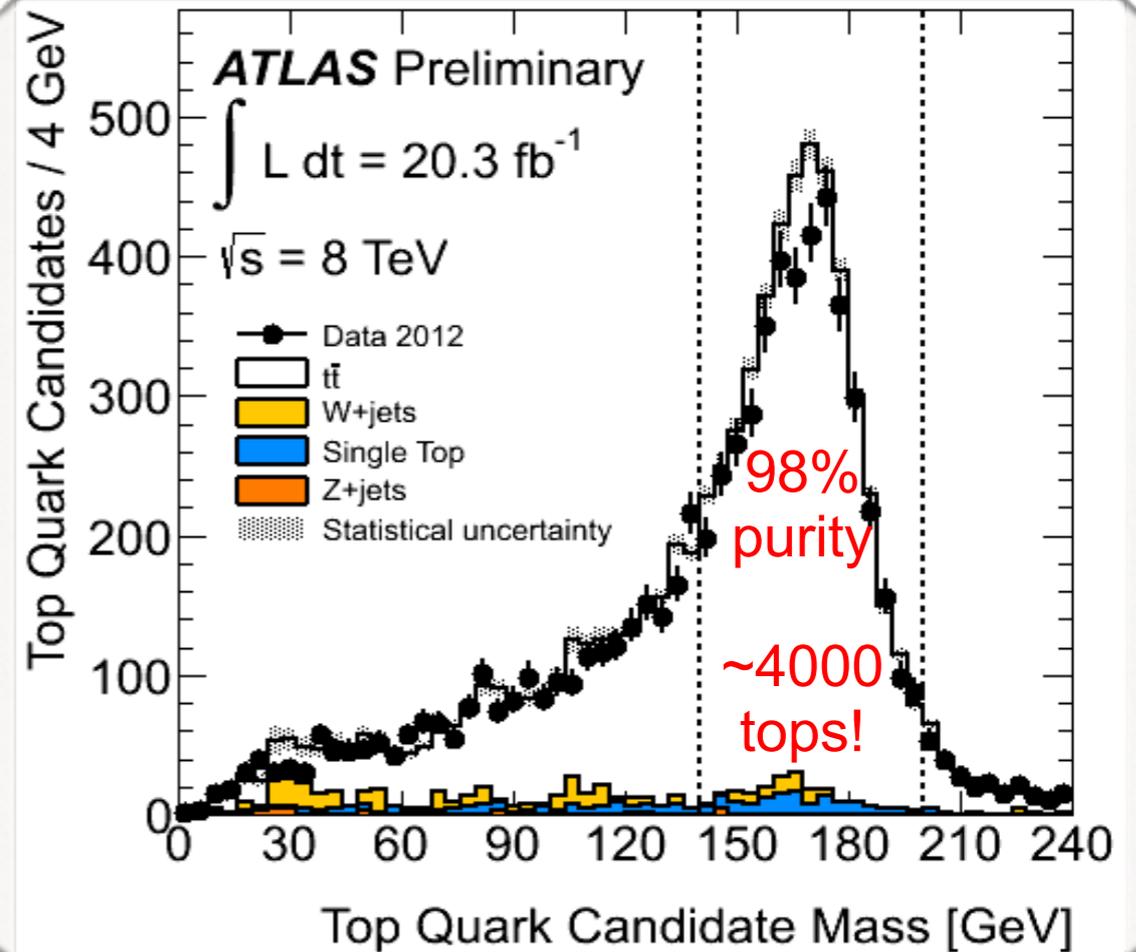
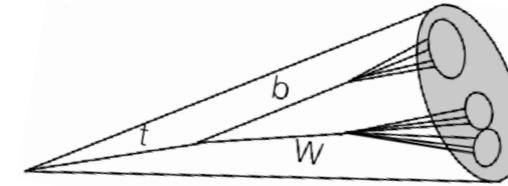
Re-cluster with smaller  $R$ , and keep only 3 hardest jets

# Seeing W's and tops in a single jet

## W's in a single jet

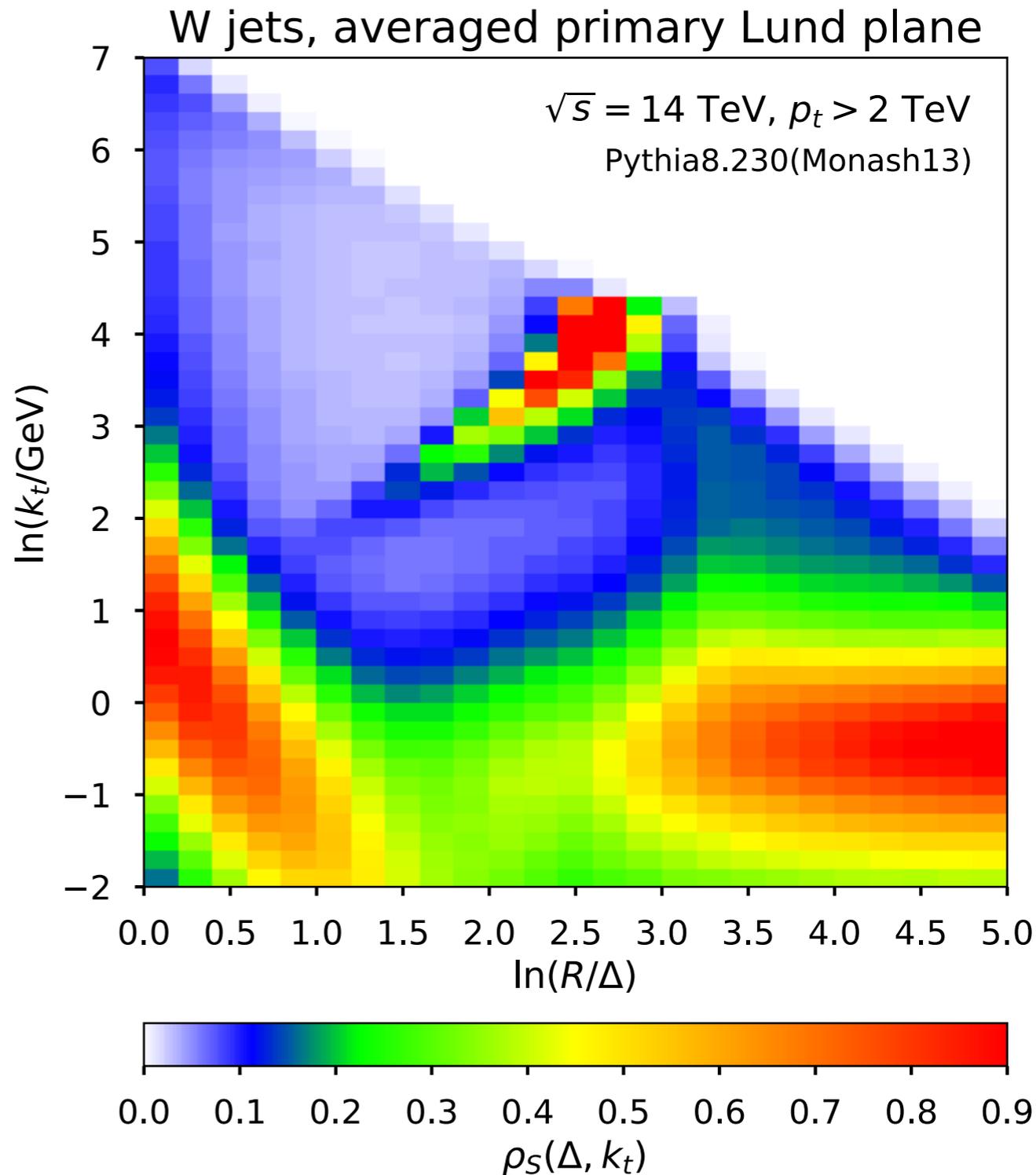
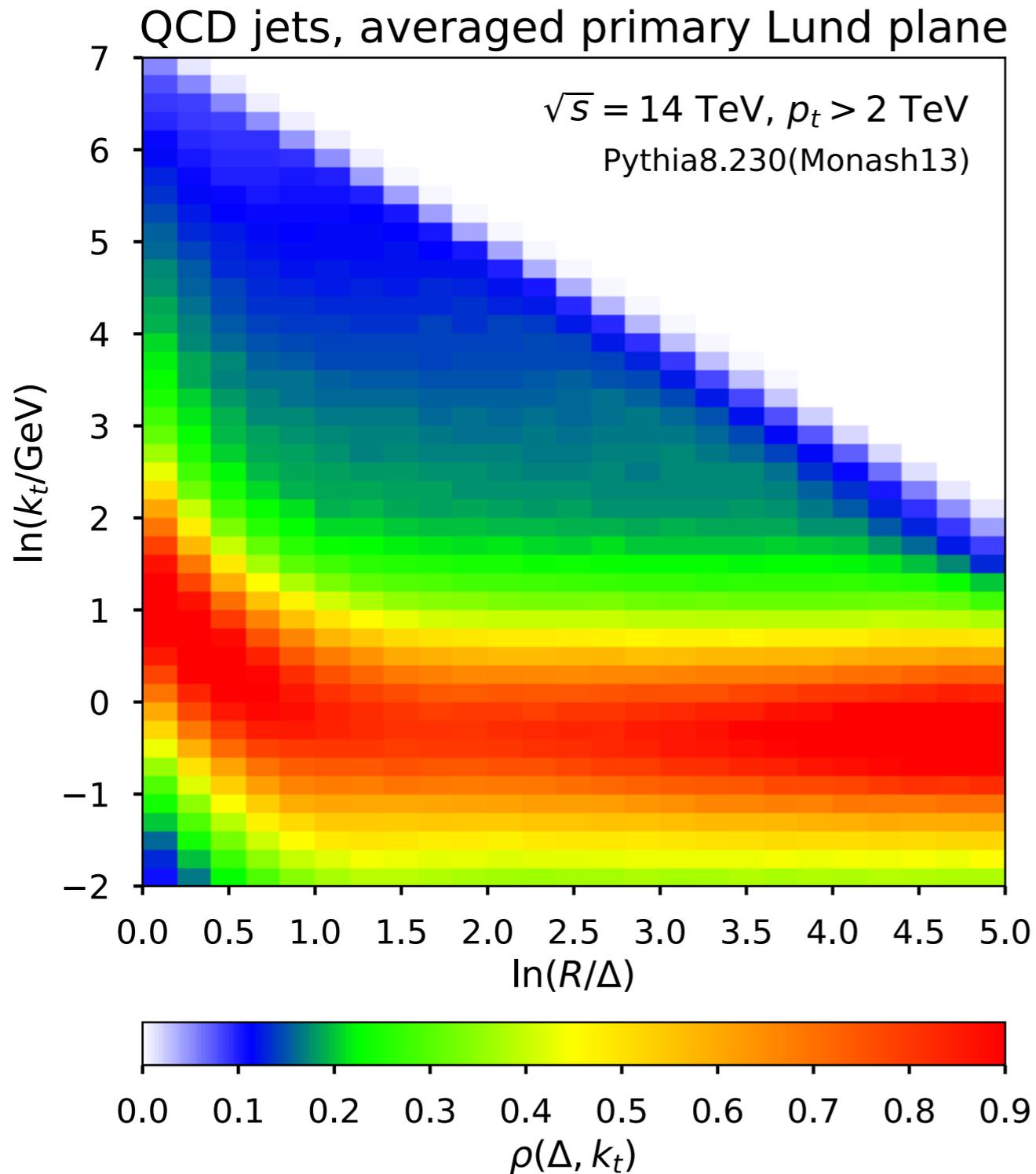


## tops in a single jet



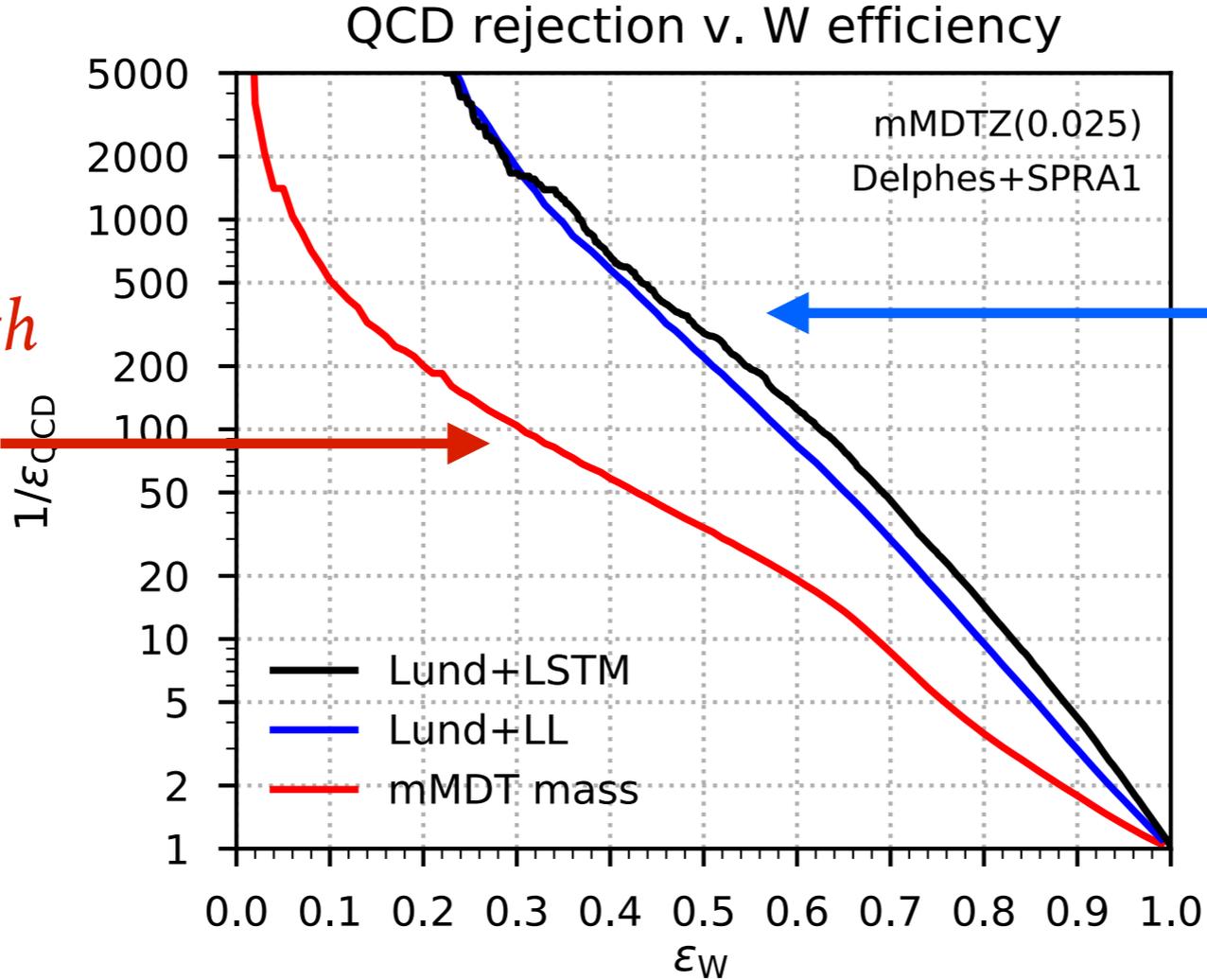
- **SoftDrop:** uses the same key ideas of C/A declustering, but with better theoretical properties and more flexibility in phasespace
- **Subjettiness / energy-energy-correlations / energy-flow polynomials / Lund Plane structure:** all try to measure the energy flow around the core  $n$ -prong structure of a jet (e.g. 2-prong for Higgs decay)
- **Machine learning:** jet substructure is one of the most dynamic playgrounds for ML, with large gains to be had in pulling out all info from jets

# intrajet energy flow for QCD jets & W jets



# using intra-jet energy flow for W tagging

*QCD rejection with just jet mass (SD/mMDT)*



*QCD rejection with use energy-flow within the jet (beyond just hard prongs)*

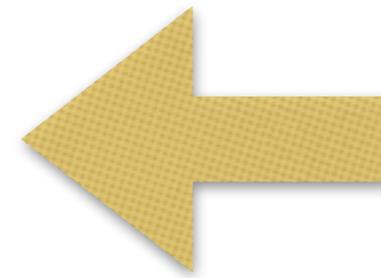
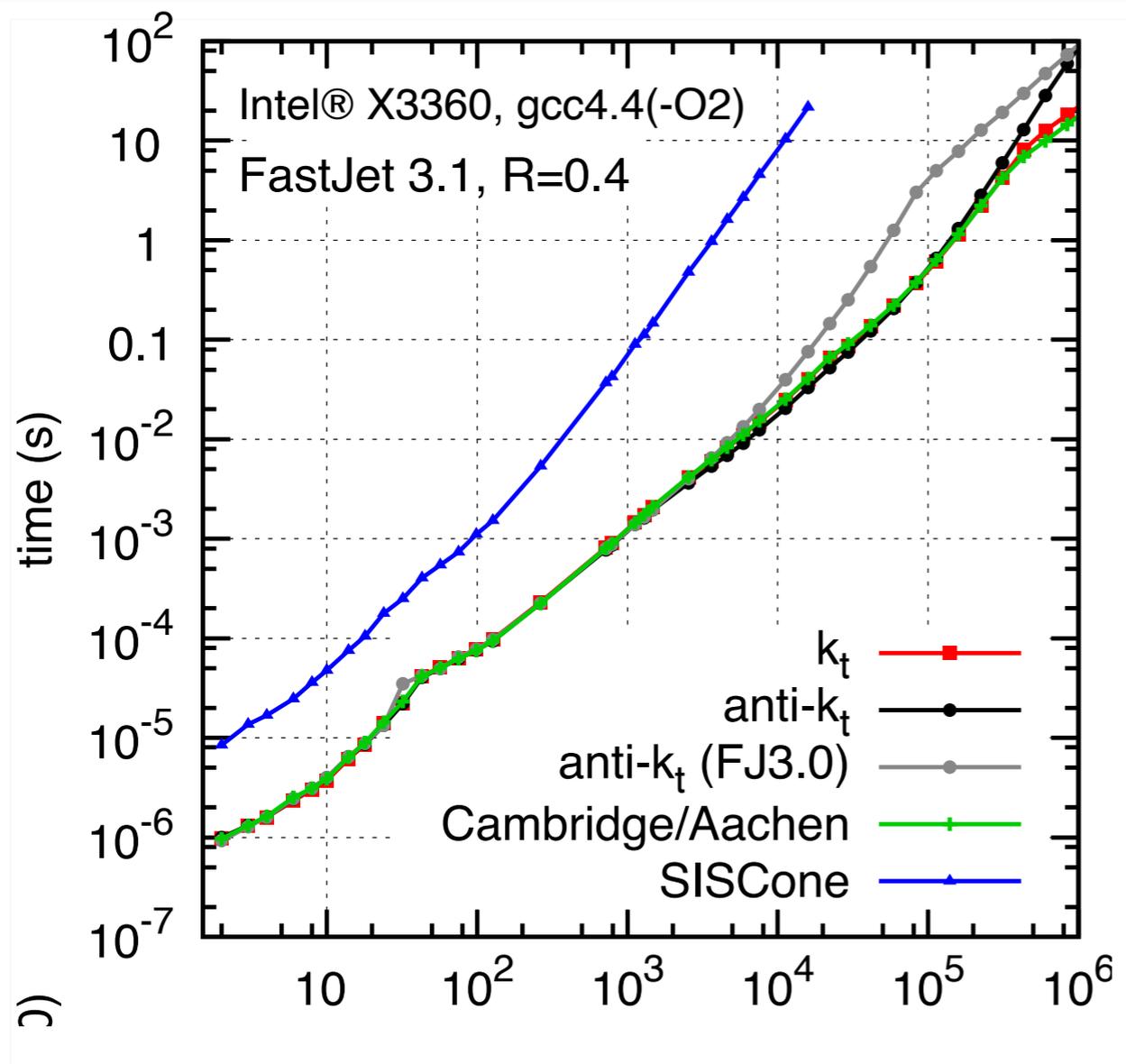
*5–10x better (and newer ML can do even better)*

Dreyer, GPS & Soyez, [arXiv:1807.04758](https://arxiv.org/abs/1807.04758)

- There are myriad approaches to jet finding
- For applications with a single moderately hard scale (e.g.  $t\bar{t}$ ), anti-kt,  $R=0.4$ , with a  $p_t$  cut of a few tens of GeV is often a good default
- For problems with multiple hard scales (e.g. highly boosted top / W / H / etc.) one needs to look at events on multiple angular scales: jet substructure

EXTRAS

# Time to cluster N particles in FastJet



Time to cluster N particles

# FJContrib packages

Version 1.043 of FastJet Contrib is distributed with the following packages

Package	Version	Release date	Information
ClusteringVetoPlugin	1.0.0	2015-05-04	README NEWS
ConstituentSubtractor	1.4.5	2020-02-23	README NEWS
EnergyCorrelator	1.3.1	2018-02-10	README NEWS
FlavorCone	1.0.0	2017-09-07	README NEWS
GenericSubtractor	1.3.1	2016-03-30	README NEWS
JetCleanser	1.0.1	2014-08-16	README NEWS
JetFFMoments	1.0.0	2013-02-07	README NEWS
JetsWithoutJets	1.0.0	2014-02-22	README NEWS
LundPlane	1.0.3	2020-02-23	README NEWS
Nsubjettiness	2.2.5	2018-06-06	README NEWS
QCDAwarePlugin	1.0.0	2015-10-08	README NEWS
RecursiveTools	2.0.0-beta2	2018-05-29	README NEWS
ScJet	1.1.0	2013-06-03	README NEWS
SoftKiller	1.0.0	2014-08-17	README NEWS
SubjetCounting	1.0.1	2013-09-03	README NEWS
ValenciaPlugin	2.0.2	2018-12-22	README NEWS
VariableR	1.2.1	2016-06-01	README NEWS