

Jets

Lecture I

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▶ Lecture 1: jet clustering

- ▶ *Basic requirements and properties*
- ▶ *Jet clustering algorithms*
 - ▶ *jet areas and pileup subtraction*

▶ Lecture 2: jet unclustering

- ▶ *Jet grooming: taggers and filters*
 - ▶ *Jet substructure*
 - ▶ *Generalised pileup subtraction*

Code snippets detailing FastJet use for some specific cases will be included in the slides. Of course one can obtain equivalent results in a (slightly) different way, e.g. using SpartyJet (it wraps FastJet and include additional tools, i.e. ROOT interface), dedicated software from experiments, which also tend to wrap FastJet with they own interface, or the MadGraph or Delphes interfaces

These lectures include many ideas and contributions by Gavin Salam and Gregory Soyez

Some bibliography

- ▶ Les Houches 2007 proceedings, [arXiv:0803.0678](#)
- ▶ Gavin Salam, 'Towards Jetography', [arXiv:0906.1833](#)
- ▶ Proceedings of BOOST 2010 and 2011:
[arXiv:1012.5412](#) and [arXiv:1201.0008](#)
- ▶ Gavin Salam's lectures at CERN Academic Training (March 2011):
<http://indico.cern.ch/event/115078>
- ▶ Fastjet user manual, MC, G.P. Salam and G. Soyez, [arXiv:1111.6097](#).
Also available from fastjet.fr

Gluon 'discovery'

1979:

Three-jet events observed by TASSO, JADE, MARK J and PLUTO at PETRA in e^+e^- collisions at 27.4 GeV

Interpretation:
large angle emission of a hard gluon

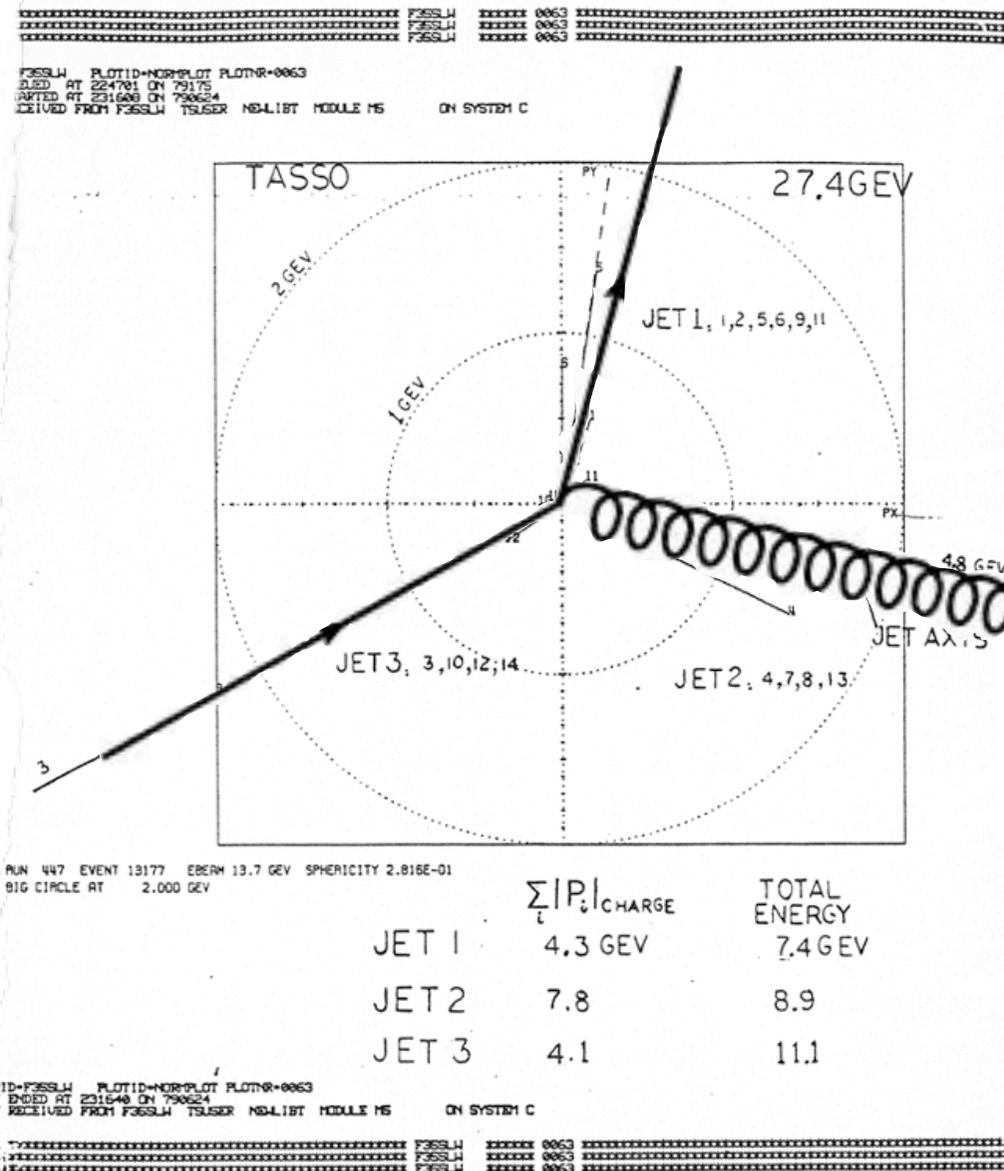


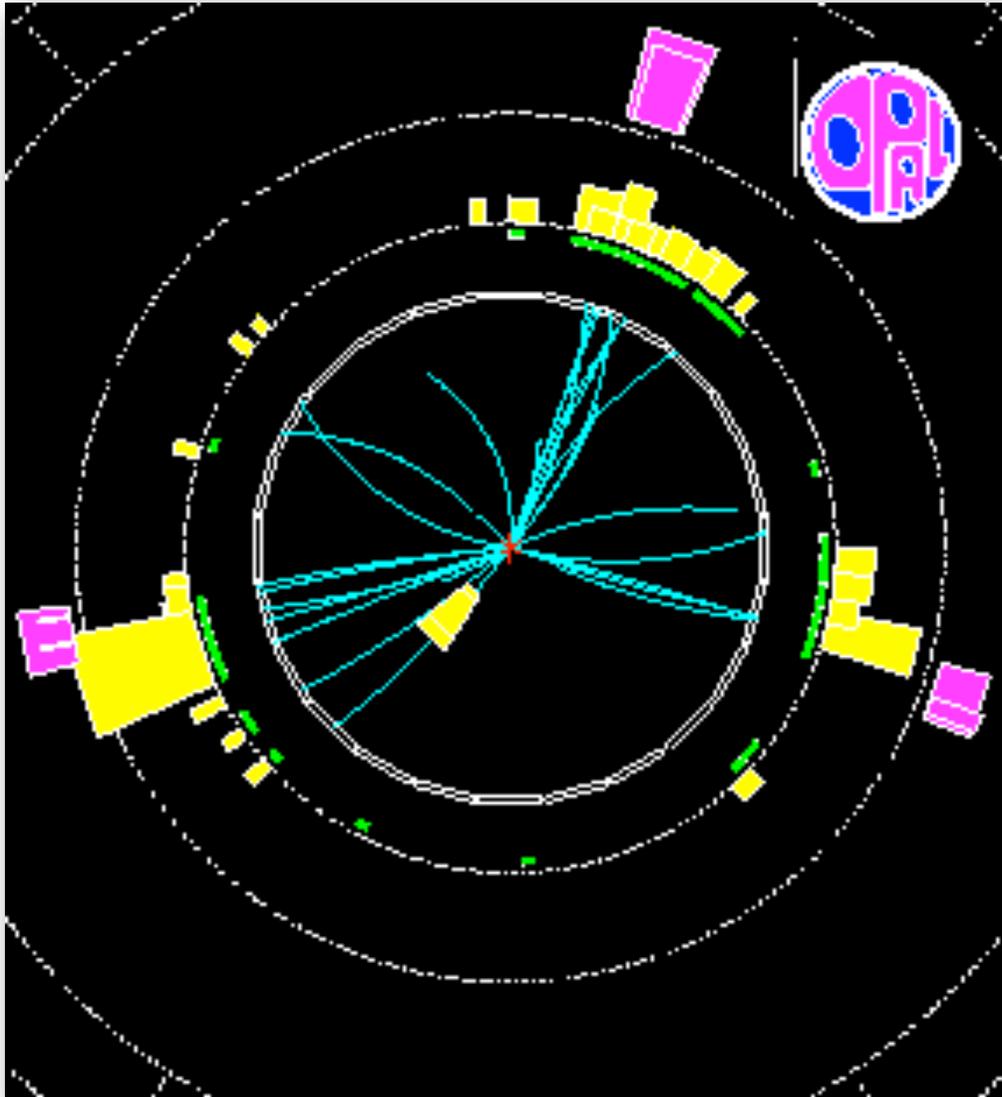
FIGURE 3

From PETRA to LEP

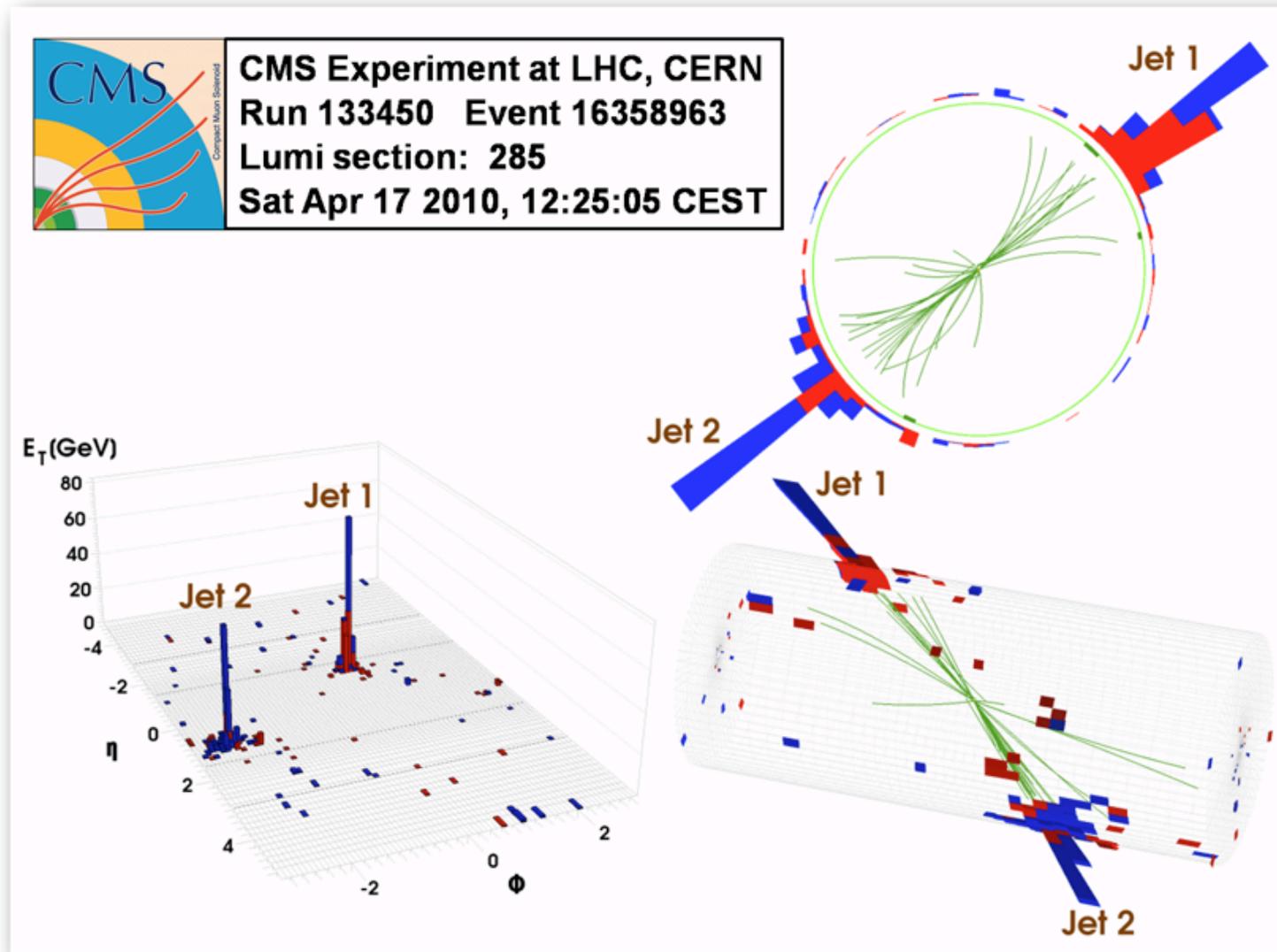
A **jet** is something that happens in high energy events: **a collimated bunch of hadrons flying roughly in the same direction**

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

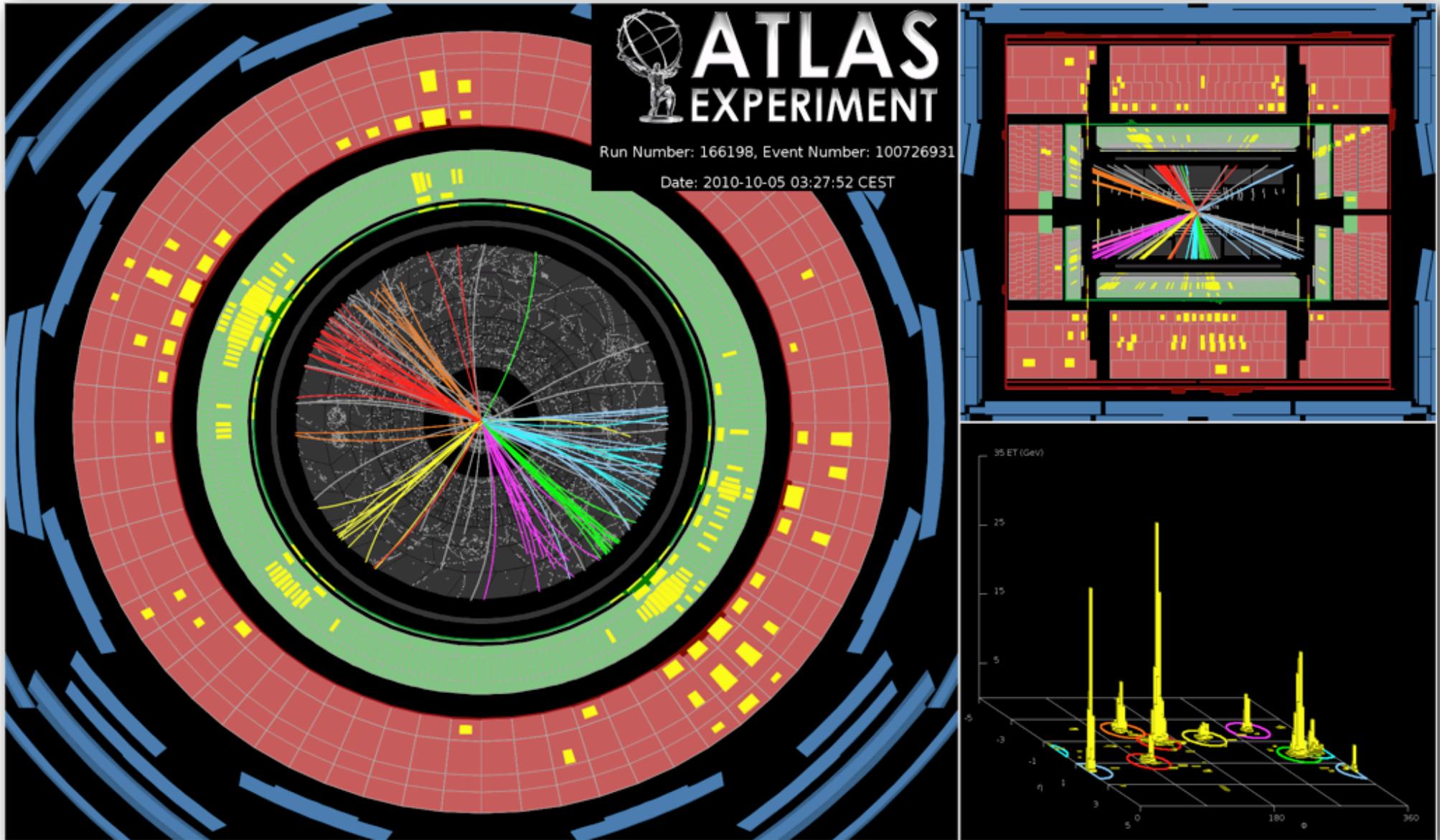
The classification of particles into jets is best done using a **clustering algorithm**



A few decades after PETRA and LEP, the event displays got prettier, but jets are still pretty much the same



Dijet event from CMS

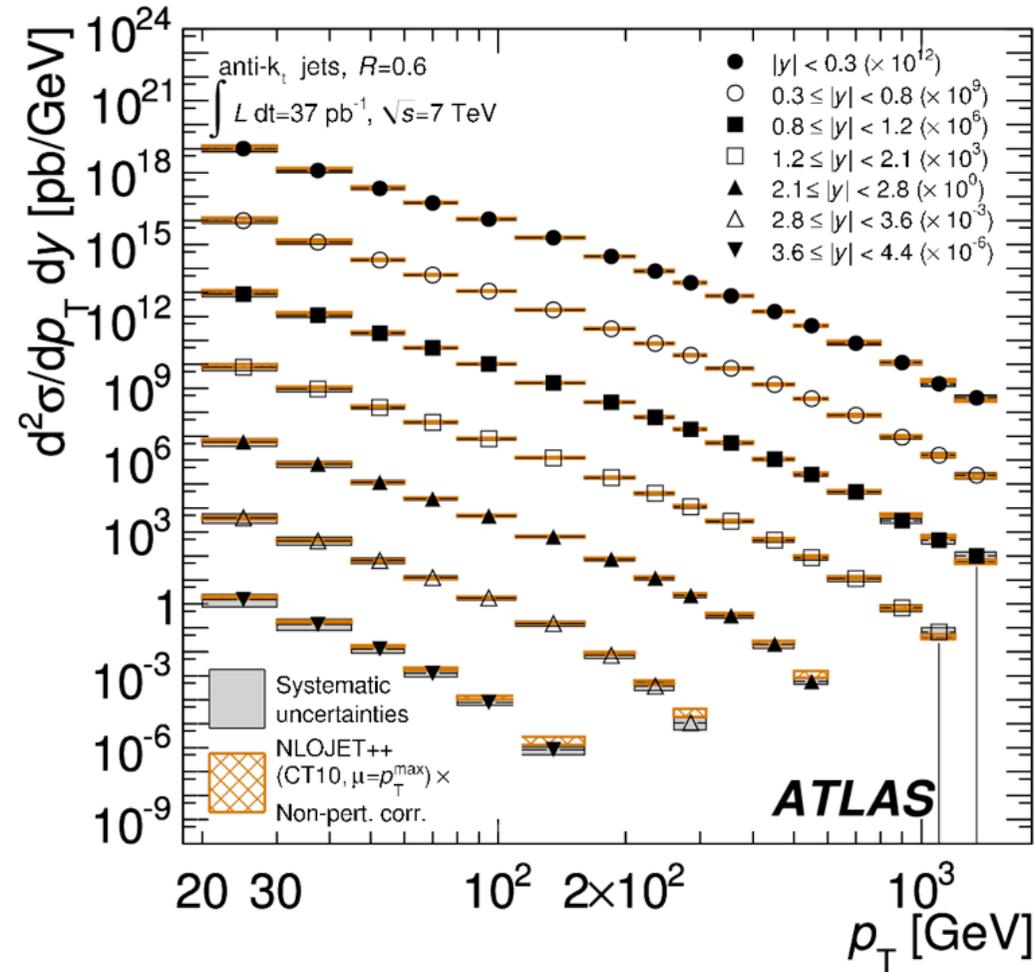
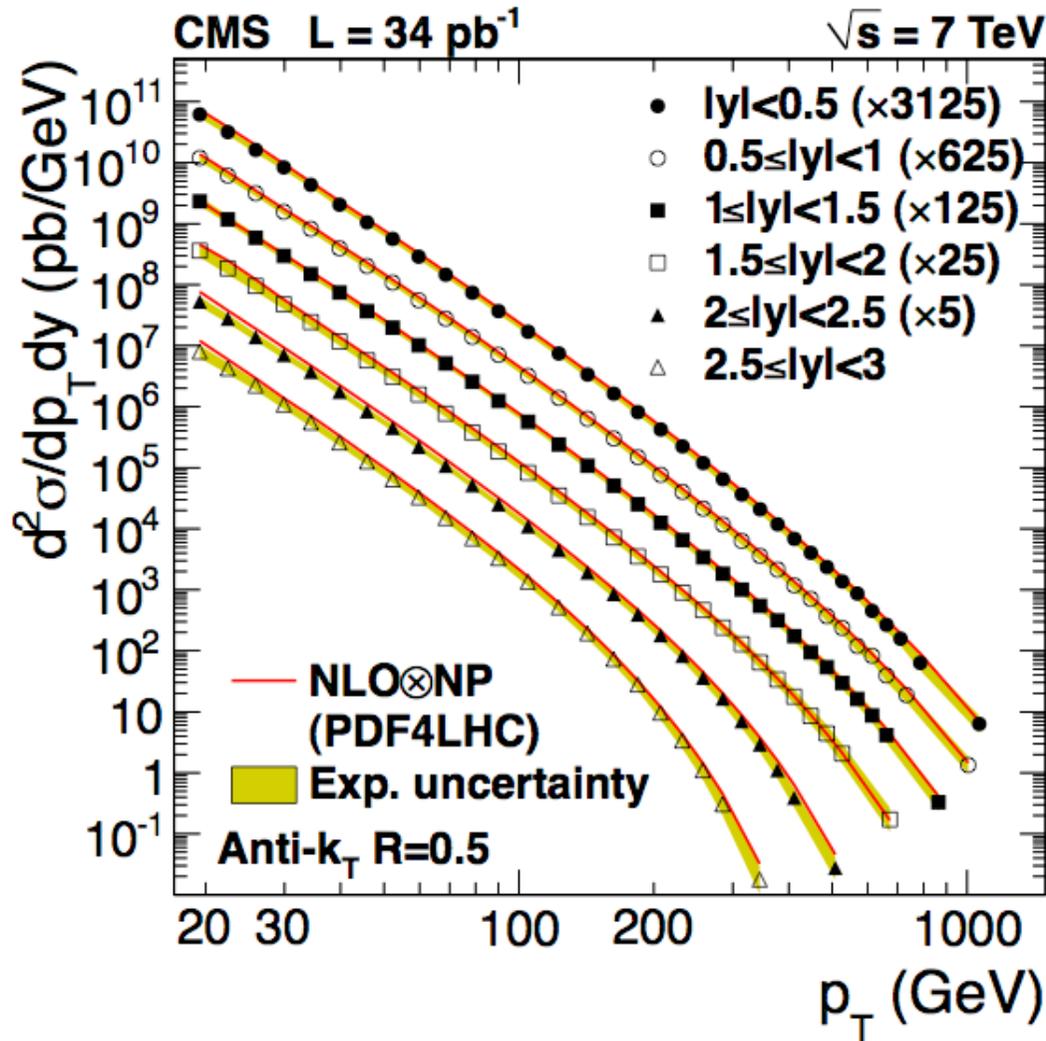


8(!) jets event from ATLAS

Example of a jet observable

arXiv:1106.0208

arXiv:1112.6297

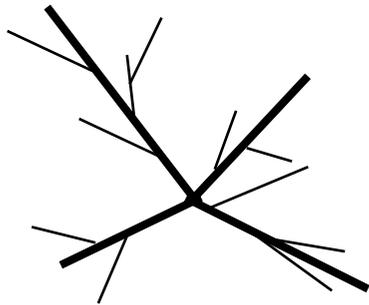


Jets extensively measured in hadronic collisions.

Very good agreement with pQCD predictions over 10 orders of magnitude

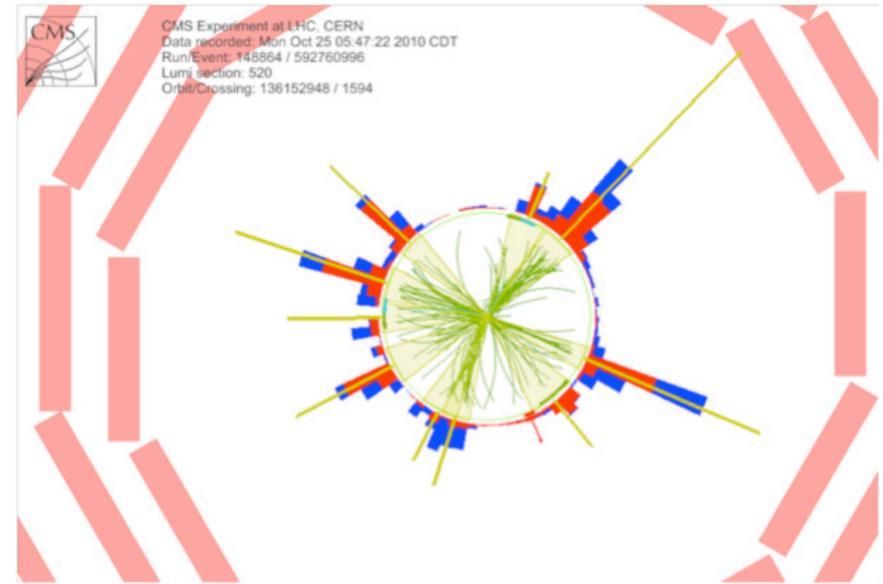
Taming reality

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

Clustering goes well beyond particle physics.

It is used in many fields (genetics, data mining, information retrieval, ...) as a form of **unsupervised learning**: it is meant to **group similar objects together**, and therefore suggest a classification

▶ What does “similar” mean?

- ▶ We need to define a dissimilarity function between two objects.

Almost infinite freedom of choice, as long as

$D(A,B) = D(B,A)$	(symmetry)
$D(A,B) = 0$ iff $A=B$	(self-similarity)
$D(A,B) \geq 0$	(positivity)
$D(A,B) \leq D(A,C) + D(B,C)$	(triangle inequality)

▶ How do we group the objects together?

- ▶ This is specified by the choice of a clustering algorithm
 - ▶ *partitional*
 - ▶ *hierarchical*

Clustering algorithms

Partitional: construct partitions of the set of objects and evaluate them by some criterion.

X Number of clusters must be provided in advance

✓ Easy to implement, fast

Example: **k-means**. Choose K centroids at random, assign objects to them, recalculate centroids iterate until clusters are stable (i.e. they are all closest to the centroid of the cluster they belong too)

Hierarchical: create a hierarchical decomposition of the set of objects by some criterion

X Computationally heavy

✓ No need to specify number of clusters in advance

✓ 'In depth' view of structure (hierarchy).

Can decide a posteriori on a criterion for number of clusters.

Example: **hierarchical agglomerative clustering**. Combine objects iteratively starting from those with the smallest distance (i.e. largest similarity). Stop when a single cluster is left.

- ▶ While we could take almost any clustering algorithm and, with a reasonable distance, use it to construct jets, i.e. clusters of hadrons, the result may not be particularly useful.

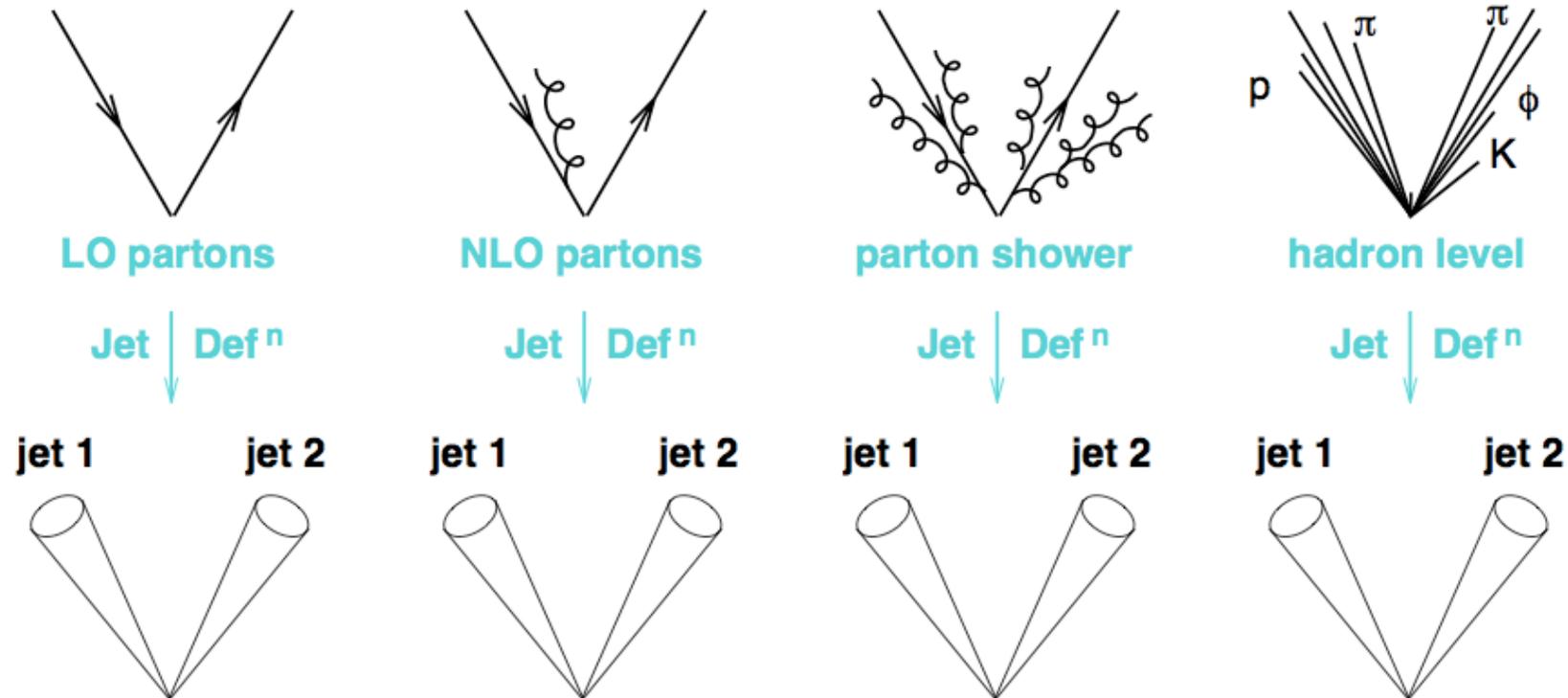
We must also be guided by physics, so that

- ▶ *the procedure leads to calculable results* → *infrared and collinear safety*
- ▶ *the results are robust with respect to dynamics that we cannot calculate in detail* → *resiliency to hadronisation effects*

This puts strong constraints on the distances and algorithms that we can use

Jets as proxies

A good jet definition should be resilient to QCD effects



NB. 'Resiliency' does not mean 'total insensitivity'
A 'hadron jet' is **not** a parton

Most definitions will give very similar results (especially for inclusive observables), but it is important to be aware of potential differences, and not to compare apples with oranges.

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**:

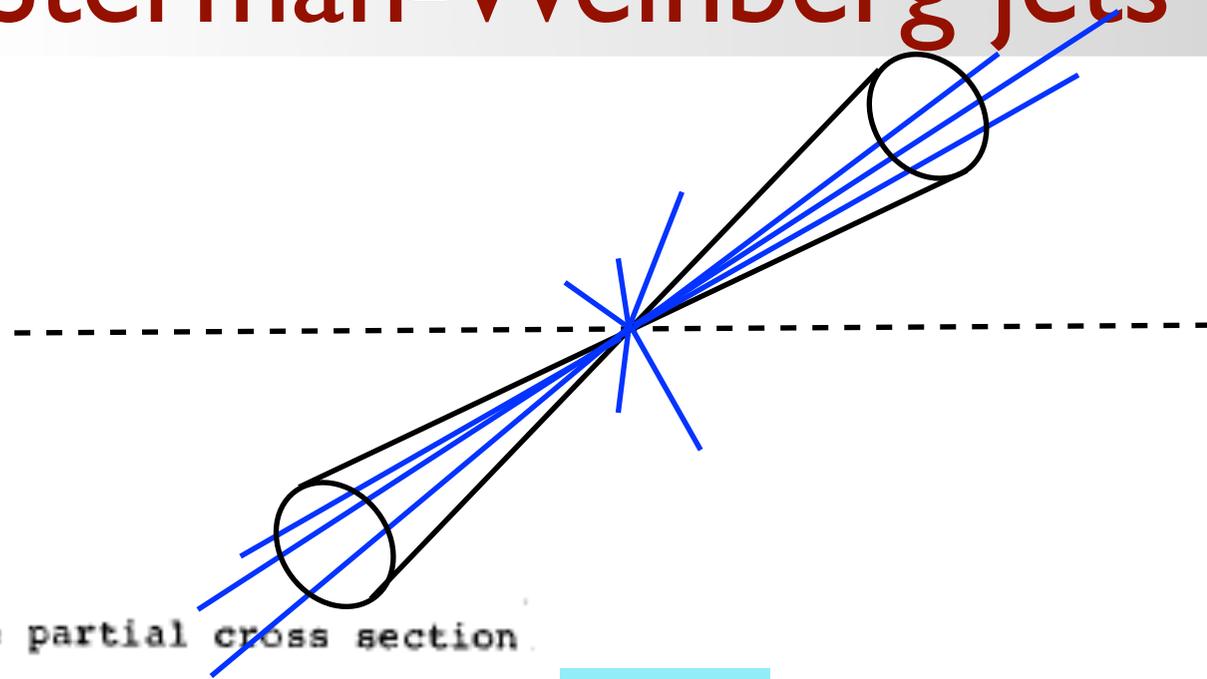
$$O(X; p_1, \dots, p_n, p_{n+1} \rightarrow 0) \rightarrow O(X; p_1, \dots, p_n)$$

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

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):



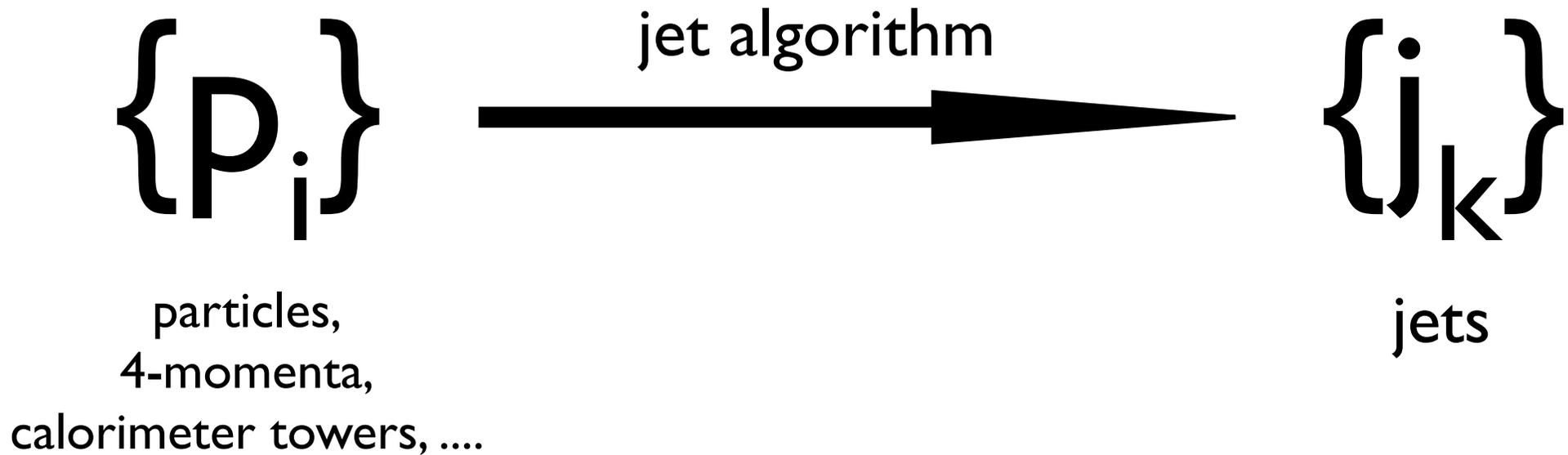
To study jets, we consider the partial cross section $\sigma(E, \theta, \Omega, \epsilon, \delta)$ for e^+e^- hadron production events, in which all but a fraction $\epsilon \ll 1$ of the total e^+e^- energy E is emitted within some pair of oppositely directed cones of half-angle $\delta \ll 1$, lying 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

Jet 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
(more about this later on)

A jet algorithm
+
its parameters (e.g. R)
+
a recombination scheme
=
a **Jet Definition**

“Jet [definitions] are legal contracts between theorists and experimentalists”

-- MJ Tannenbaum

Jet definition in FastJet

```
/// define a jet definition  
JetDefinition jet_def(JetAlgorithm jet_algorithm,  
                      double R,  
                      RecombinationScheme rec_sch = E_scheme);
```

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 ,

→ hierarchical clustering

▶ 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.....

→ partitional clustering

Finding stable cones

In partitional-type algorithms (i.e. cones), one wishes to find the **stable configurations**:

axis of cones coincides with sum of 4-momenta of the particles it contains

The 'safe' way of doing so is to test **all possible combinations** of N objects

Unfortunately, this takes $N2^N$ operations:
the time taken is the age of the universe for only 100 objects

An approximate way out is to use seeds (e.g. à la k-means)
However, the final result can depend on the choice of the seeds and,
such jet algorithms usually turn out to be **IRC unsafe**

Seedless IRC-safe Cone (SC-SM): SISCon

Salam, Soyez, arXiv:0704:0292

Seeds are a problem:
they lead to finding only some of the stable cones

Obvious solution:
find ALL stable cones, testing all possible combinations of N particles

Unfortunately, this takes $N2^N$ operations:
the age of the universe for only 100 particles

Way out: a geometrical solution → SISCon

The first (and only?) IRC-safe cone algorithm for hadronic collisions

SISCon is guaranteed to find ALL the stable cones

IRC safety in real life

Strictly speaking, one needs IRC safety not so much to find jets, but to be able to calculate them in pQCD

If you are not interested in theory/data comparisons, you may think of doing well enough with an IRC-unsafe jet algorithm

However

- ▶ Detectors may split/merge collinear particles, and be poorly understood for soft ones
- ▶ High luminosity (or heavy ions collisions) add a lot of soft particles to hard event

IRC safety provides resiliency to such effects
(plus, at some point in the future you may wish to compare your measurement to a calculation)

Recombination algorithms

These are hierarchical clustering algorithms

- ▶ First introduced in e^+e^- collisions in the '80s
- ▶ Typically they work by calculating a '**distance**' between particles, and then recombine them pairwise according to a given order, until some condition is met (e.g. no particles are left, or the distance crosses a given threshold)

IRC safety can usually be seen to be trivially guaranteed

JADE algorithm

distance:

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

- ▶ 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

Problem of this particular algorithm:

two soft particles emitted at large angle get easily recombined into a single jet:
counterintuitive and perturbatively troublesome

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

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

Identical to JADE,
but with 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

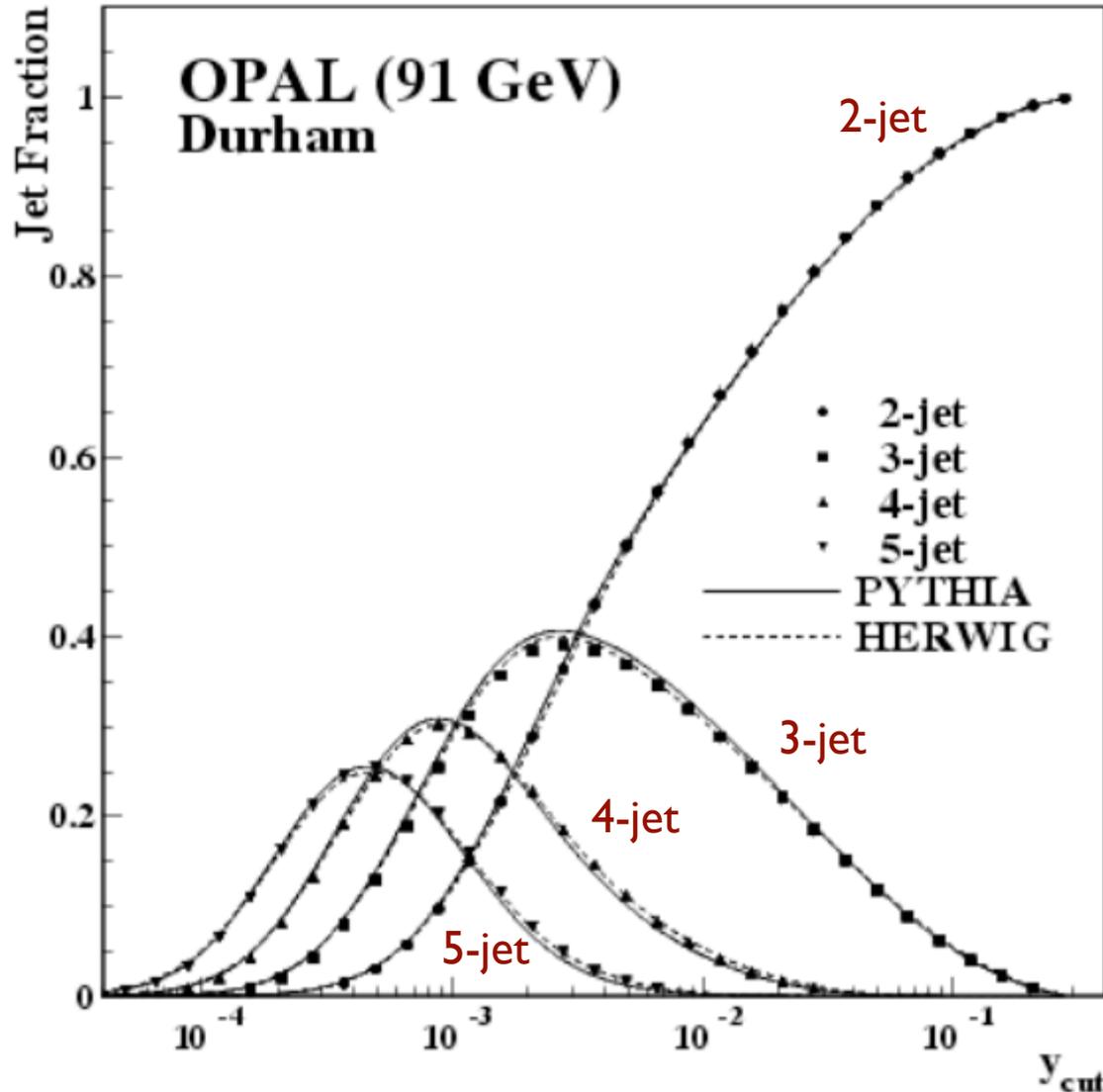
The use of the $\min()$ avoids the problem of recombination of back-to-back particles present in JADE: a soft and a hard particle close in angle are 'closer' than two soft ones at large angle

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 k_t algorithm inverts the QCD branching sequence (the pair which is recombined first is the one with the largest probability to have branched)

$e^+e^- k_t$ (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

k_t algorithm in hadron collisions

(Inclusive and longitudinally invariant version)

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

- ▶ Calculate the distances between the particles: \mathbf{d}_{ij}
 - ▶ Calculate the beam distances: \mathbf{d}_{iB}
 - ▶ Combine particles with **smallest distance** d_{ij} or, if d_{iB} is smallest, call it a jet
 - ▶ Find again smallest distance and repeat procedure until no particles are left (this stopping criterion leads to the **inclusive** version of the k_t algorithm)
-
- ▶ Given N particles this is, naively, an $O(N^3)$ algorithm: calculate N^2 distances, repeat for all N iterations. 1 second to cluster 1000 particles: too slow for practical use.
 - ▶ An $O(N^2)$ implementation (the 'FastJet algorithm') exists: 1ms for 1000 particles. Can even use it in the trigger.
 - ▶ The same algorithm can also be implemented with only $O(N \ln N)$ complexity.

The k_t algorithm and its siblings

One can generalise the k_t distance measure:

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

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 safe algorithms

k_t	<p>SR</p> $d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 / R^2$ <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> $d_{ij} = \min(k_{ti}^{-2}, k_{tj}^{-2}) \Delta R_{ij}^2 / R^2$ <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)

Jet clustering in FastJet

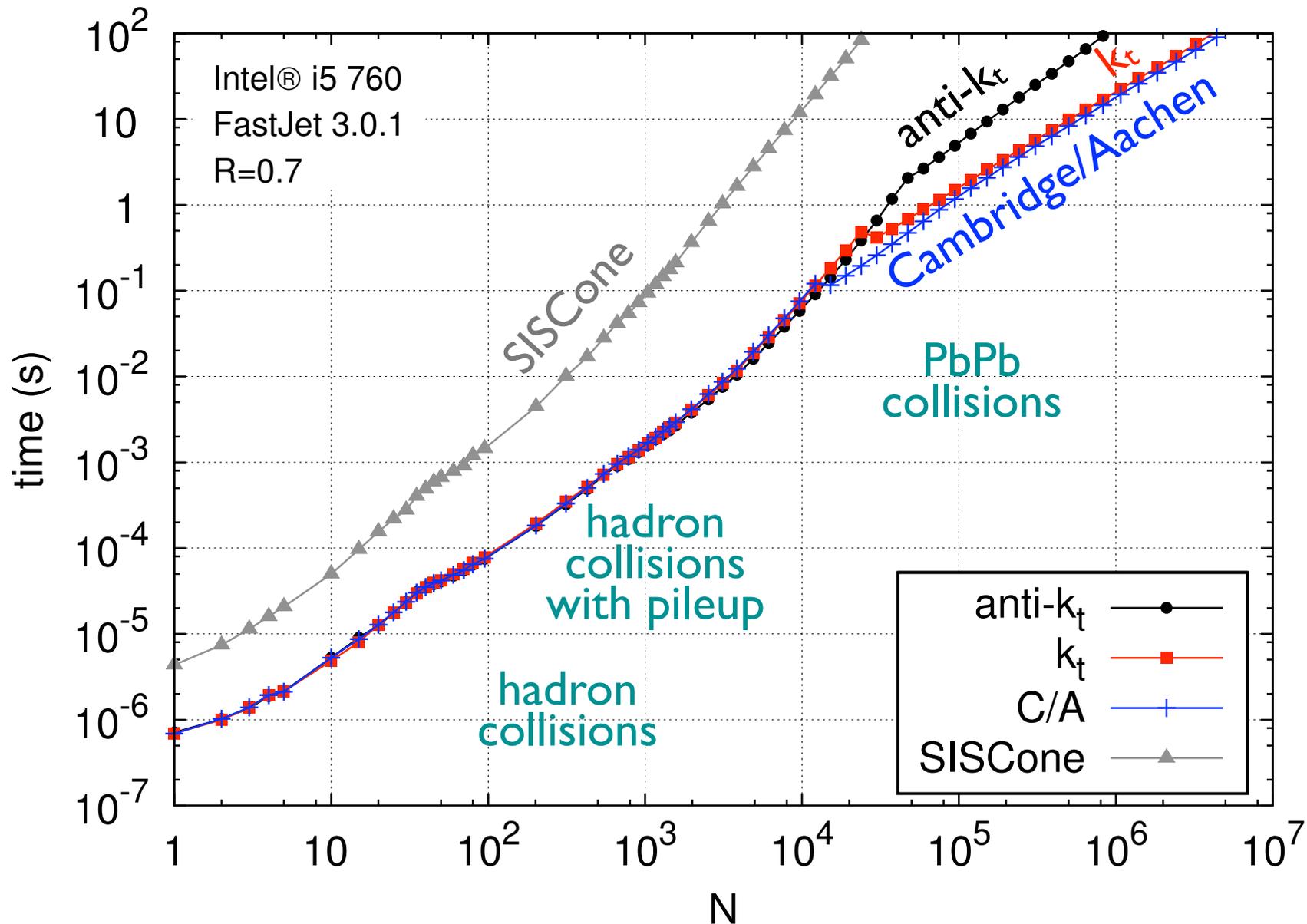
```
/// define a jet definition
JetDefinition jet_def(JetAlgorithm jet_algorithm,
                      double R,
                      RecombinationScheme rec_sch = E_scheme);
```

jet_algorithm can be any one of the four IRC safe algorithms, or also most of the old IRC-unsafe ones, for legacy purposes

```
/// create a ClusterSequence, extract the jets
ClusterSequence cs(input_particles, jet_def);
vector<PseudoJet> jets = sorted_by_pt(cs.inclusive(jets));
...
// pt of hardest jet
double pt_hardest = jets[0].pt();
...
// constituents of hardest jet
vector<PseudoJet> constits = jets[0].constituents();
```

FastJet speed test

Time needed to cluster an event with N particles



Jets 'reach'

Algorithmically, a jet is simply a collection of particles

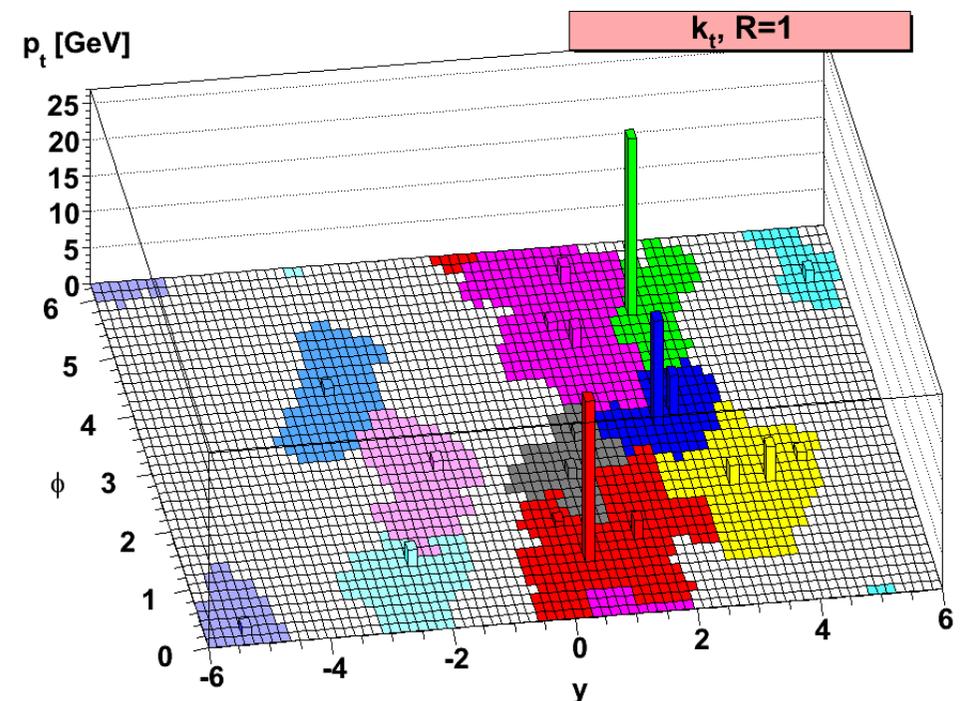
For a number of reasons, it is however useful to consider its **spatial extent**, i.e. given the position of its axis, up to where does it collect particles? What is its shape?

These details are important for a number of corrections of various origin: perturbative, non-perturbative (hadronisation), pileup, detector related, etc

Note that the intuitive picture of a jet being a cone (of radius R) is **wrong**.

This is what k_t jets can look like:

(more later about what this plot really means)



From jet 'reach' to jet areas

MC, Salam, Soyez, 0802.1188

Not one, but three **definitions** of a jet's size:

▶ **Passive area**

Place a single very soft particle (a '**ghost**') in the event, measure the extent of the region where it gets clustered within a given jet

Reach of jet for **pointlike** radiation

▶ **Active area**

Fill the events with many very soft particles ('**ghosts**'), cluster them together with the hard ones, see how many get clustered within a given jet

Reach of jet for **diffuse** radiation

▶ **Voronoi area**

Sum of areas of intersections of Voronoi cells of jet constituents with circle of radius R centred on each constituent

Coincides with passive area for k_t algorithm

(In the large number of particles limit all areas converge to the same value)

Active Area

Add **many** ghost particles in random configurations to the event.
Cluster many times. *Allow ghosts to cluster among themselves too.*
Count how many ghosts on average get clustered into a given jet J.

$$A(J | \{g_i\}) = \frac{N_g(J)}{\nu_g} \equiv A_g N_g(J)$$

Active area of jet J for a single ghosts configuration

Number of ghosts in jet J

Ghost density

Area of a single ghost

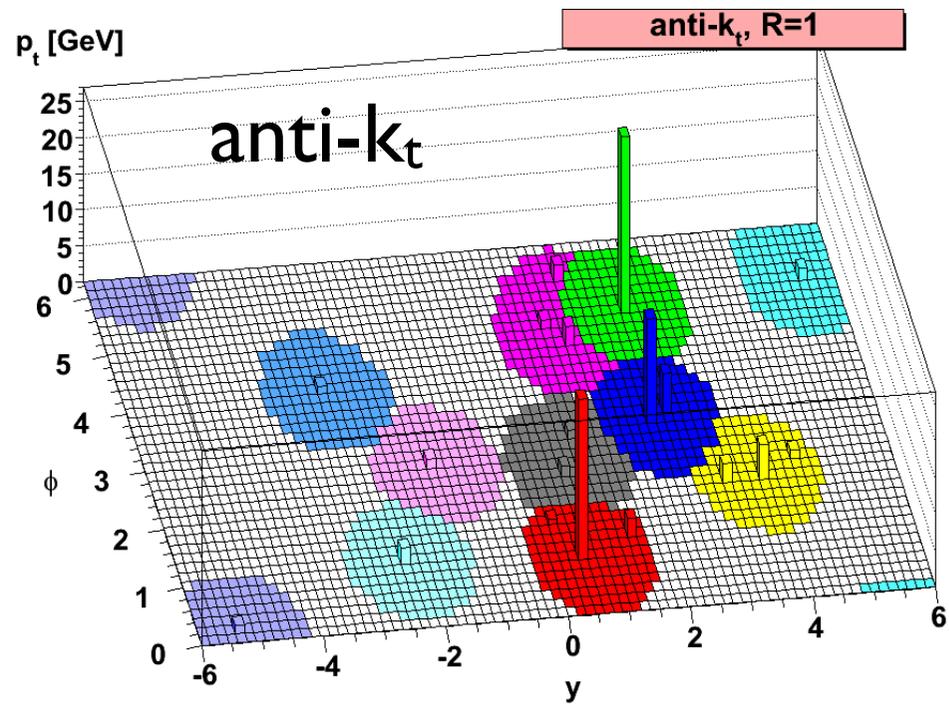
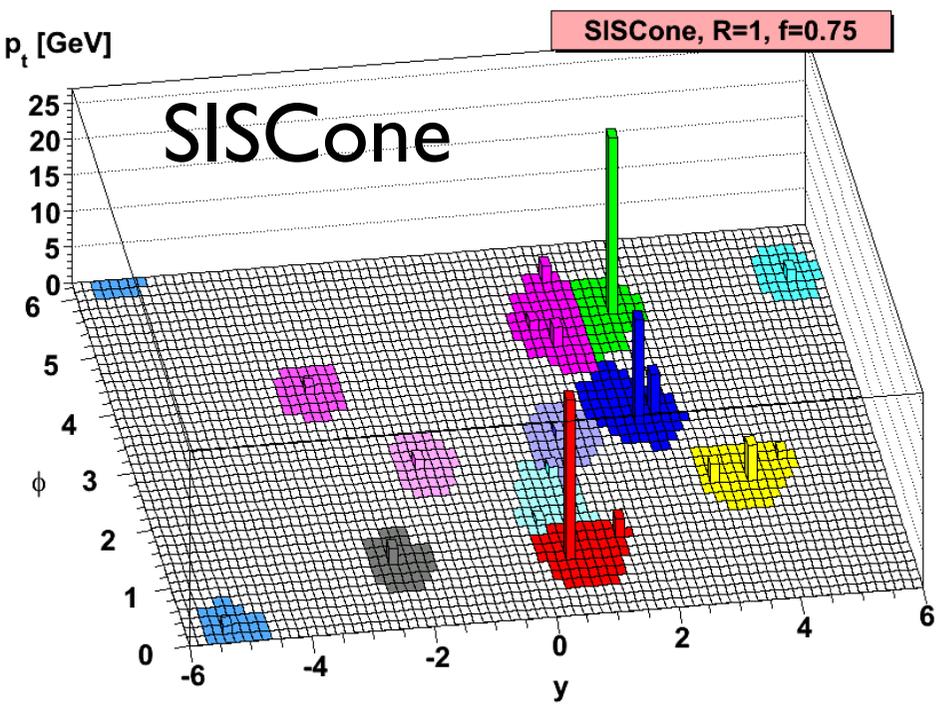
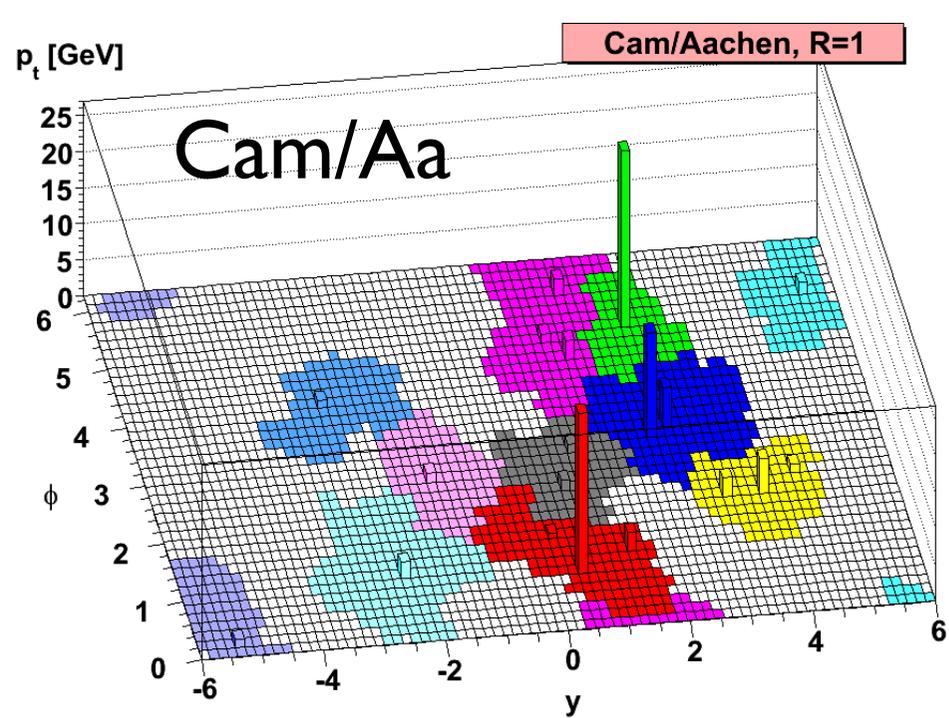
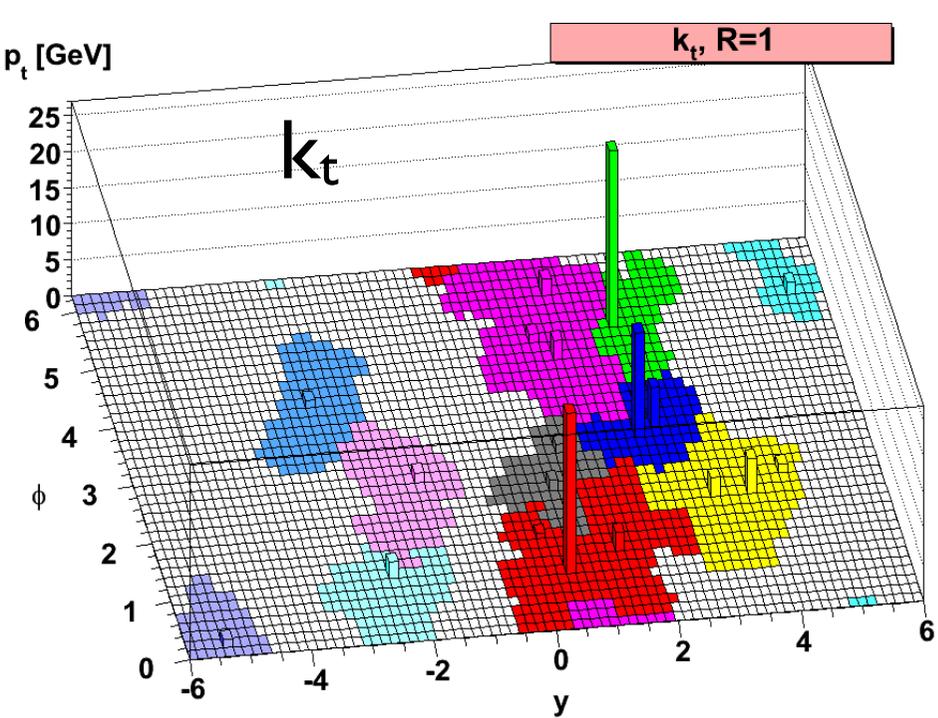
$$A(J) = \lim_{\nu_g \rightarrow \infty} \langle A(J | \{g_i\}) \rangle_g$$

Active area of jet J

```
/// specify where to place the ghosts, their area, how many times
/// to repeat the clustering
double rapmax = 2.5;
int nrepeat = 1;
double ghost_area = 0.01;
GhostedAreaSpec gas(rapmax, nrepeat, ghost_area);

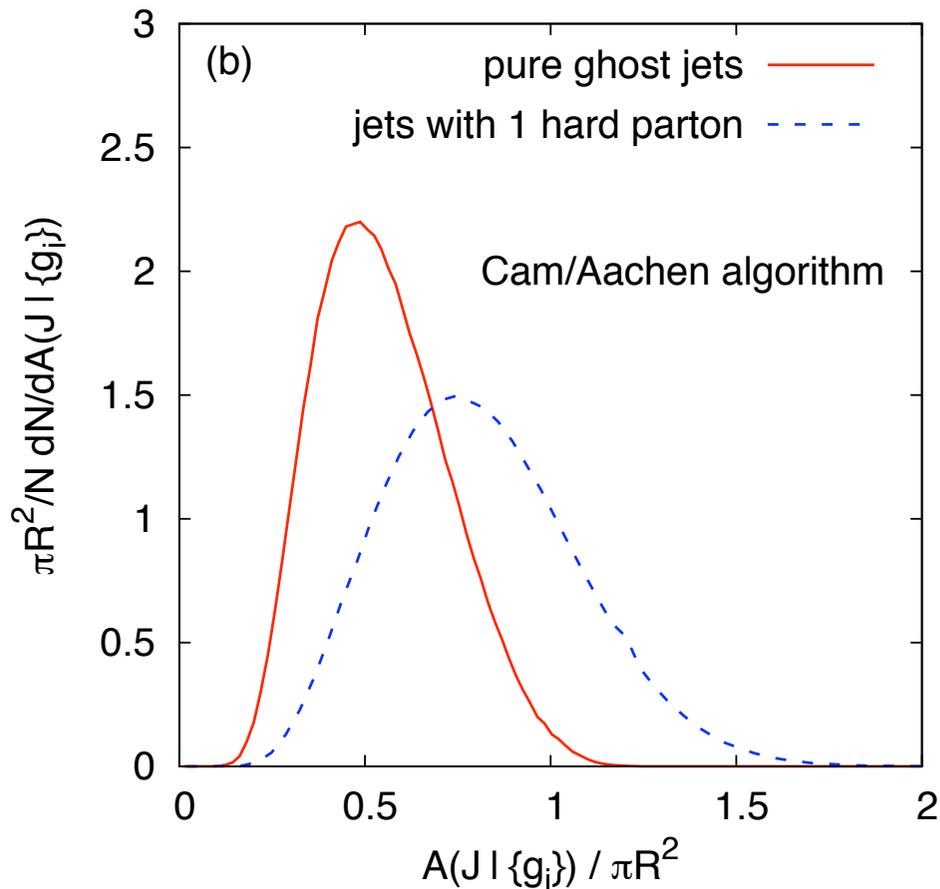
/// use this configuration to define the area
/// A sensible default for gas is provided
AreaDefinition area_def(active_area, gas);

/// construct cluster sequence with areas
ClusterSequenceArea(input_particles, jet_def, area_def);
....
jets[0].area()
```

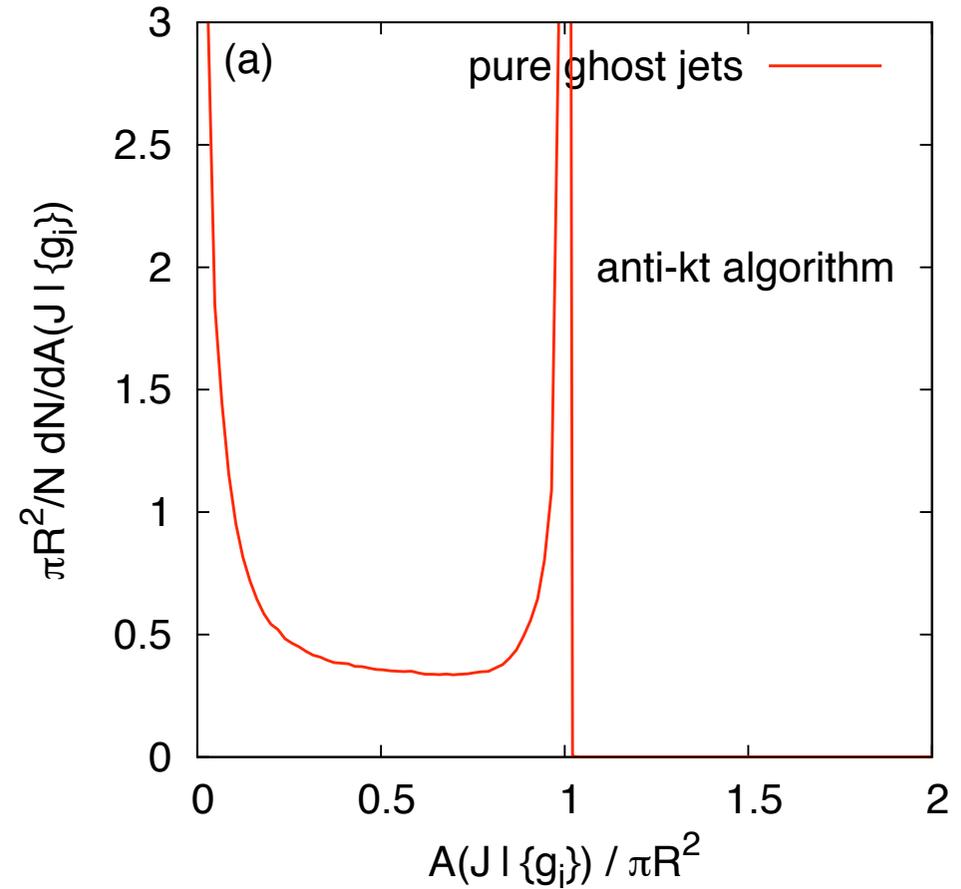


Active area distributions

k_t and Cam/AA



anti- k_t



For a roughly uniformly soft background, anti- k_t gives many small jets and many large ones (you can't fill a plane with circles!)

Jet areas physical meaning

A jet's active area expresses the susceptibility of that jet's transverse momentum to contamination from a uniform background

Consider a jet of transverse momentum p_t , made up of N_g ghosts, each with transverse momentum $p_{t,g}$.

It holds
$$\frac{\partial p_t}{\partial p_{t,g}} = \frac{\partial(N_g p_{t,g})}{\partial p_{t,g}} = N_g$$

Recalling the definition of active jet area, $A_J = A_g N_g$, we can then rewrite

$$A_J = A_g \frac{\partial p_t}{\partial p_{t,g}} = \frac{\partial p_t}{\partial \rho}$$

Transverse momentum density, $\rho = p_{t,g}/A_g$

The jet area is therefore the susceptibility of a jet's p_t to contamination, because for a generic background density ρ it will hold

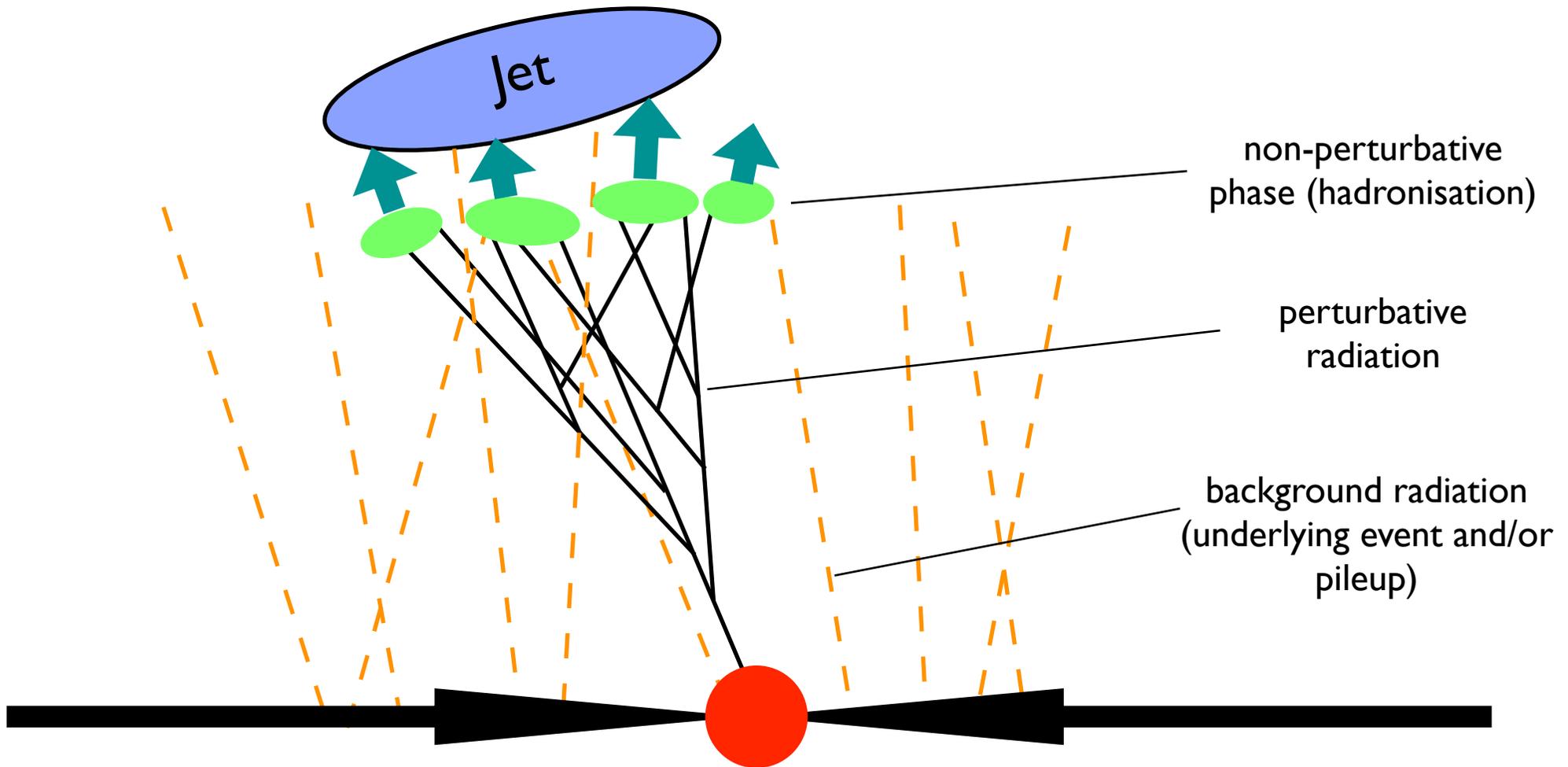
$$\Delta p_t = \rho \frac{\partial p_t}{\partial \rho} = \rho A_J$$

Susceptibility

Jet area: summary

- ▶ Jets CAN have an area, but one must define it
- ▶ The **jet (active) area expresses the susceptibility of a jet's transverse momentum to contamination from a uniform background**
 - ▶ Given a background transverse momentum density ρ , the jet's p_t will be modified by a quantity $A\rho$
- ▶ Different jet algorithms can have very different area properties:
 - ▶ Jet areas in many algorithms can fluctuate significantly from a jet to another. *Isolated hard jets in anti- k_t are an exception*
 - ▶ Jet areas can depend on a jet's p_t , driven by a (calculable) anomalous dimension that is specific to each jet algorithm. *Anti- k_t jets are again an exception: the anomalous dimension is zero*

What contributes to a jet's transverse momentum?



Effect of background

How are the hard jets modified by the background?

(Can be underlying event and/or pileup)

Susceptibility

(how much bkgd gets picked up)

Jet areas

Resiliency

(how much the original jet changes)

Backreaction

Hard jets and background

Modifications of the hard jet

$$\Delta p_t = \underbrace{\rho A \pm (\sigma \sqrt{A} + \sigma_\rho A + \rho \sqrt{\langle A^2 \rangle - \langle A \rangle^2})}_{\text{background}} + \underbrace{\Delta p_t^{BR}}_{\text{back-reaction}}$$

Background momentum density (per unit area)

background

back-reaction

'susceptibility'

'resiliency'

Background subtraction

Once the **background momentum density ρ** has been measured, it can be used to **correct** the transverse momentum of the hard jets:

$$p_T^{\text{hard jet, corrected}} = p_T^{\text{hard jet, raw}} - \rho \times \text{Area}_{\text{hard jet}}$$

MC, Salam, 0707.1378

If ρ is measured on an event-by-event basis, and each jet subtracted individually, this procedure will remove many fluctuations and generally improve the resolution of, say, a mass peak

$$\Delta p_t = \rho A \pm (\underbrace{\sigma \sqrt{A}}_{\text{Irreducible fluctuations:}} + \underbrace{\sigma_\rho A}_{\text{uncertainty of the subtraction}} + \underbrace{\rho \sqrt{\langle A^2 \rangle - \langle A \rangle^2}}_{\text{uncertainty of the subtraction}}) + \Delta p_t^{BR}$$

Irreducible fluctuations:
uncertainty of the subtraction

Background estimation and subtraction

```
// constructor for a background estimator
JetMedianBackgroundEstimator bge(Selector sel,
                                   JetDefinition jet_def,
                                   AreaDefinition area_def);

// an alternative background estimator
//GridMedianBackgroundEstimator bge(Selector sel, grid_step);

bge.set_particles(input_particles);
....
double rho = bge.rho(jet);    // extract rho estimation
```

```
// define a subtractor
Subtractor sub(&bge);

// apply it to a jet (or a vector of jets)
PseudoJet subtracted_jet = sub(jet);
```

The IRC safe algorithms

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

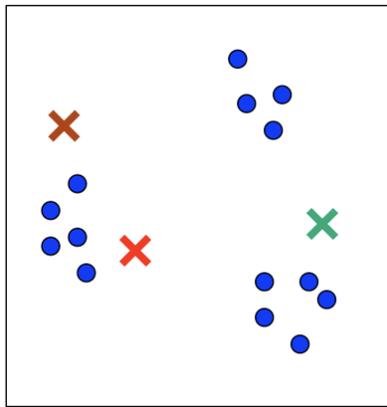
Instructions for the tutorial

1. Get yourself a copy of Fastjet at fastjet.fr and install it or, at least, get a copy of the manual if you plan to use the virtual machine
2. Get the code for the first tutorial, **01.tgz**, from the school Indico page, indico.cern.ch/conferenceDisplay.py?confId=253947, and unpack it somewhere convenient
 - 2.1. Note that you'll likely need to modify the location of the Fastjet install in the Makefile
3. Get the datafiles **pythia8-Zprime-npileupXX-nev1000.UW.gz** with $XX=\{20,50\}$ from the USB stick (NB. 50MB and 117MB files)
 - 3.1. These files contain 1000 MC events each of signal from a heavy Z' of unknown mass, produced together with an average of 20 and 50 minbias events from pileup respectively
4. Use the 01-subtraction.cc code to produce a histogram of the invariant mass of the two hardest jets in each event. **Your job is to measure the mass of the Z'**
5. In fact, what you see is not the correct mass, because the pileup has been adding p_t to each jet. **Different pileup levels give different masses.**
 - 5.1. Modify the code and use the Fastjet techniques to **subtract the pileup** from each jet, and recover the correct value for the mass.
 - 5.1.1. You may also try varying the jet algorithm and/or the jet radius

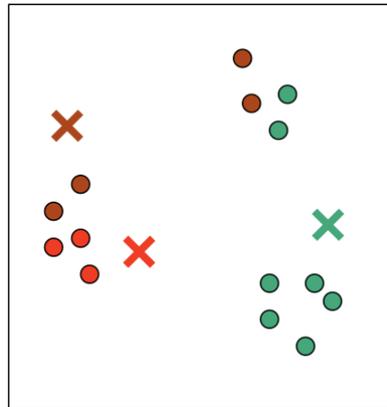
Backup slides

Example of a partitional algorithm

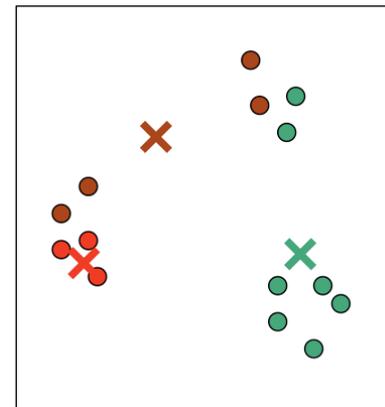
- 1) Choose K centroids at random
- 2) Assign objects to closest centroid, forming K clusters
- 3) Calculate centroid (mean of distances) of each cluster, update centroids
- 4) Check if an object in a cluster is closer to another centroid. Reallocate in case.
- 5) Repeat from step 3 until no object changes cluster anymore.



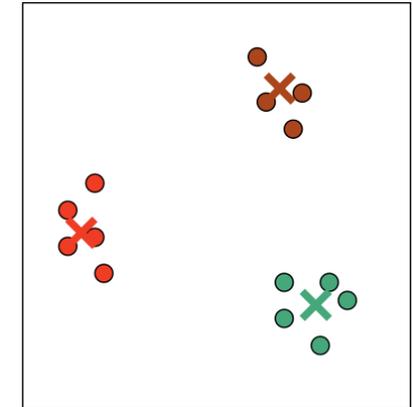
Step 1
(random centroids)



Step 2
(allocate objects)



Step 3
(move centroids)



Step 5
(end of iteration)

[Figures by E. Garrett-Mayer]

One of the main shortcomings:

result of final convergence can be highly sensitive to choice of initial seeds.

Also, the concept of 'mean distance' (to calculate the centroid) must be defined.

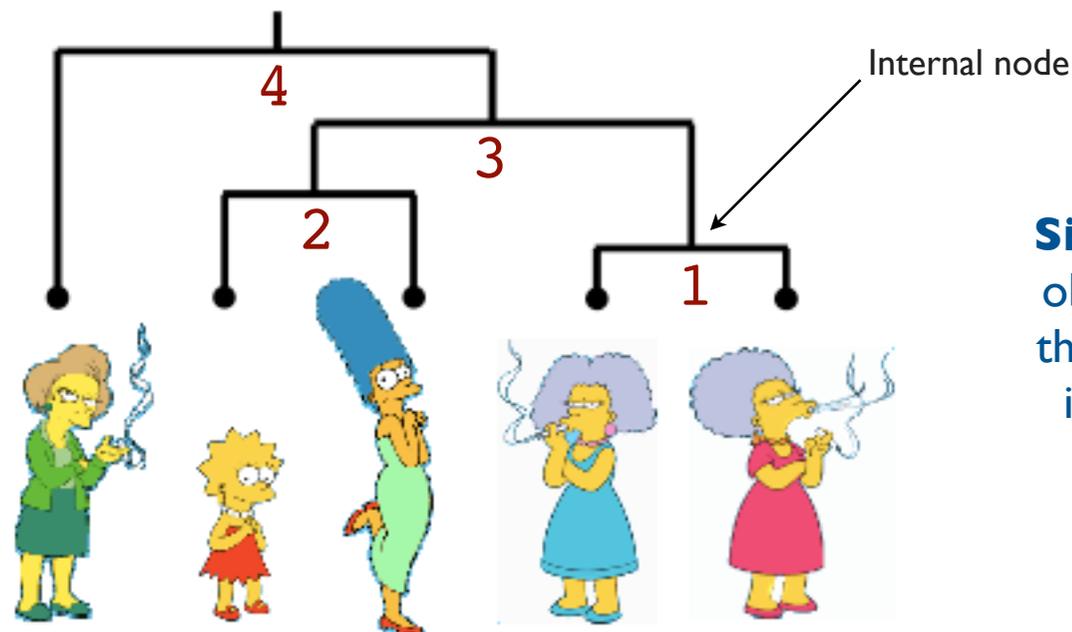
Agglomerative clustering

Example of a hierarchical algorithm

(often denoted HAC: Hierarchical Agglomerative Clustering)

- 1) Choose a dissimilarity function, calculate distance matrix between all objects
- 2) Choose a linkage method
- 3) **Cluster** two objects with **smallest** dissimilarity
- 4) Update the distance matrix
- 5) Repeat from step 3 until a single cluster is left
- 6) Look at the resulting hierarchy, and decide what 'best' number of clusters is

Dendrogram



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

Order of clustering is 1,2,3,4

Cancellation of singularities in total cross section

$$\sigma_{tot} = \int_n \overset{\text{Born}}{|M_n^B|^2} d\Phi_n + \int_n \overset{\text{Virtual}}{|M_n^V|^2} d\Phi_n + \int_{n+1} \overset{\text{Real}}{|M_{n+1}^R|^2} d\Phi_{n+1}$$

CANCELLATION

A generic observable

$$\frac{d\sigma}{dX} = \int_n |M_n^B|^2 O(X; p_1, \dots, p_n) d\Phi_n + \int_n |M_n^V|^2 O(X; p_1, \dots, p_n) d\Phi_n + \int_{n+1} |M_{n+1}^R|^2 O(X; p_1, \dots, p_n, p_{n+1}) d\Phi_{n+1}$$

In order to ensure the same cancellation existing in σ_{tot} , the definition of the observable $O(X; p_1, \dots, p_{n+1})$ **must not affect** the soft/collinear limit of the real emission term $|M_{n+1}^R|^2$, because it is there that the real/virtual cancellation takes place

Areas as a dynamical jet property

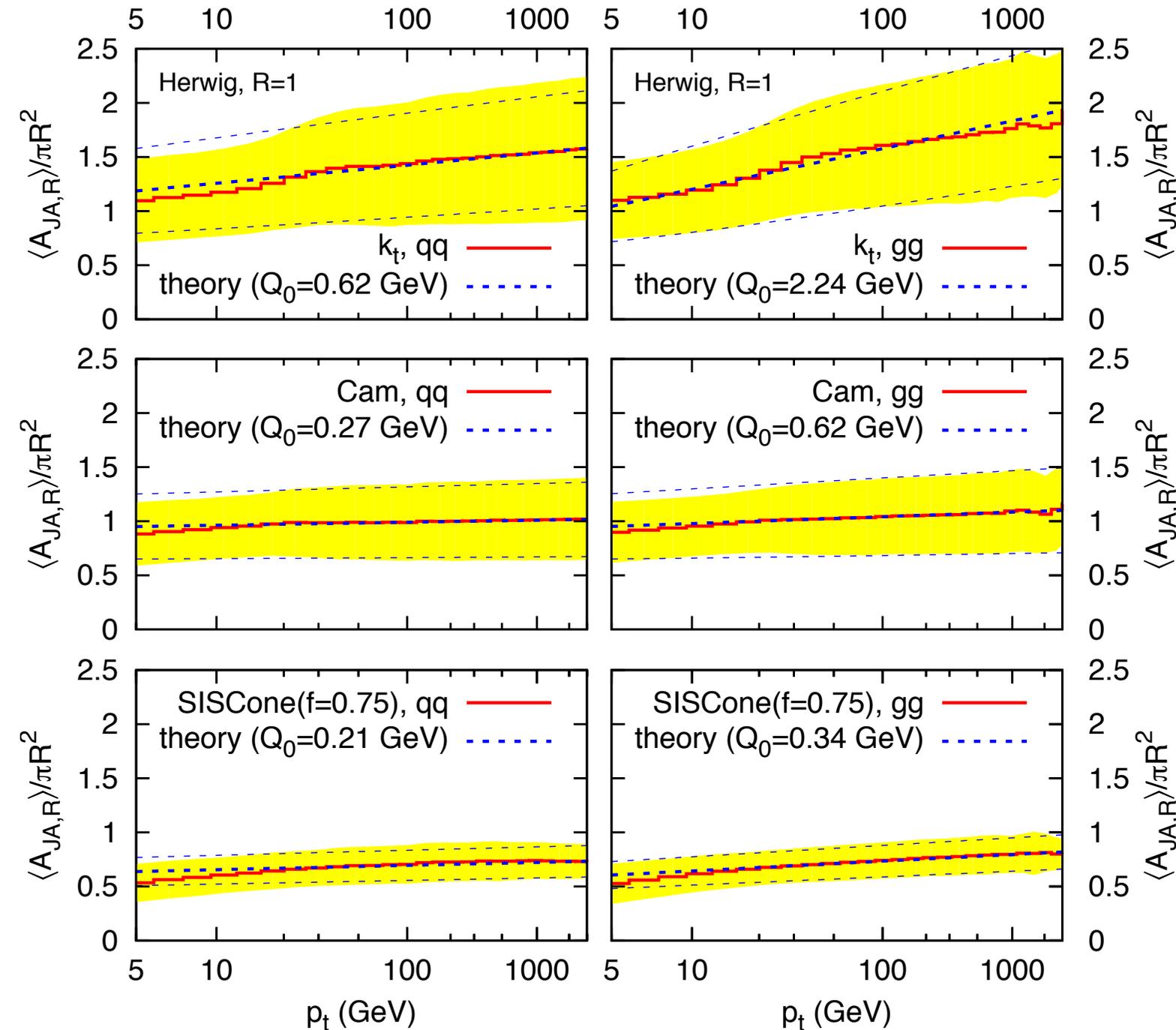
The average area of a jet can change with its p_t :

$$\langle \Delta A \rangle = \mathbf{D} \frac{C_1}{\pi b_0} \ln \frac{\alpha_s(Q_0)}{\alpha_s(Rp_{t1})}$$

	k_t	Cam/Aa	SISCone	anti- k_t
D	0.52	0.08	0.12	0

Again, only **anti- k_t** has a typical area that does **not** increase with p_t

Jet areas scaling violations



Averages and dispersions evolution from Monte Carlo simulations (dijet events at LHC) in good agreement with simple LL calculations

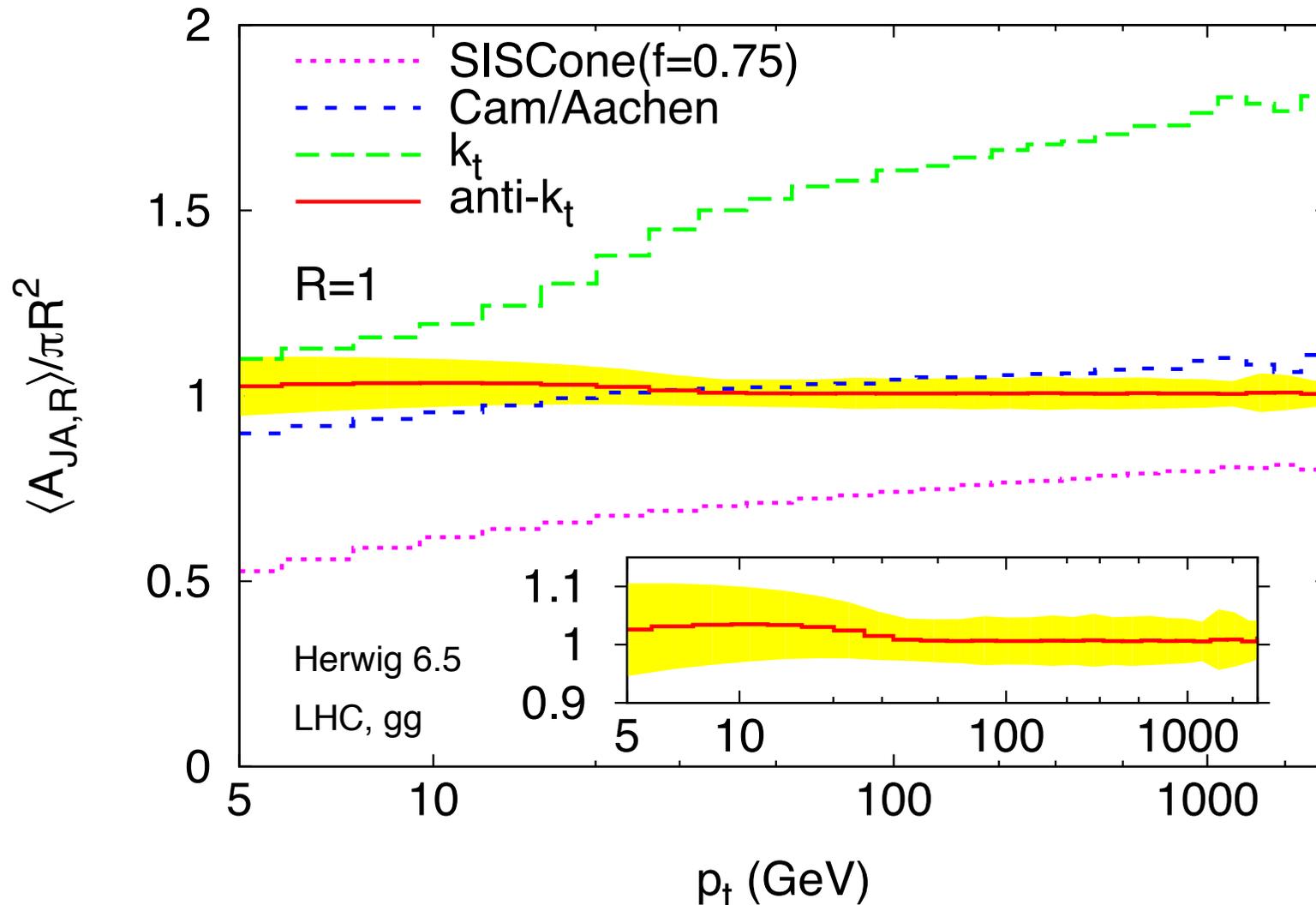
Area scaling violations are a legitimate observable.

(Though they might not be the best place where to measure α_s ...)

Jet areas scaling violations

MC, Salam, Soyez, arXiv:0802.1189

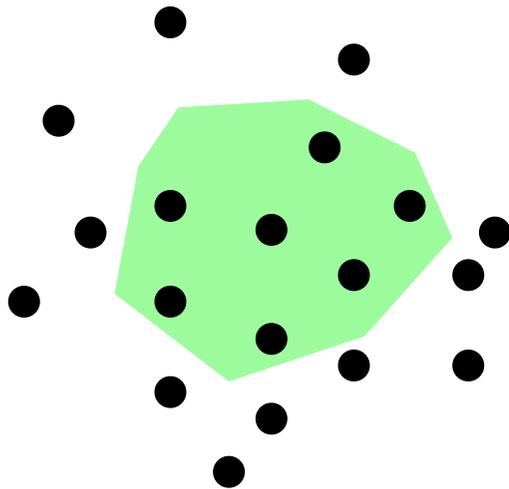
Check anti- k_t behaviour: **scaling violations** indeed **absent**, as predicted



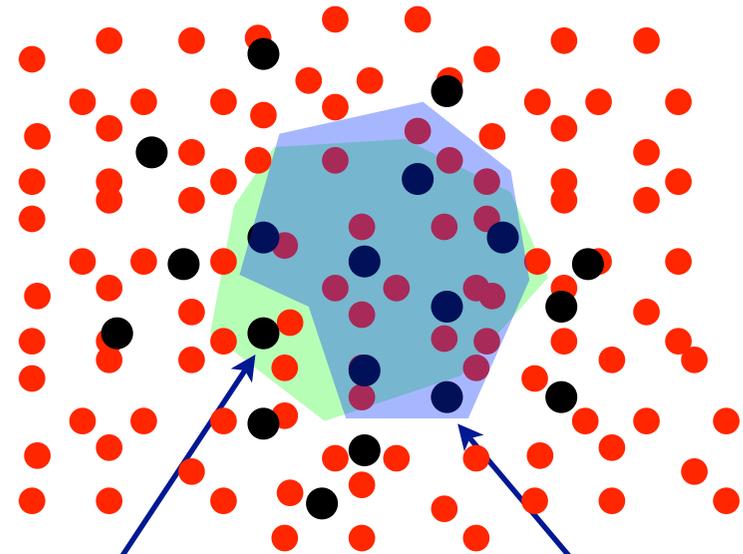
Resiliency: backreaction

“How (much) a jet changes when immersed in a background”

Without
background



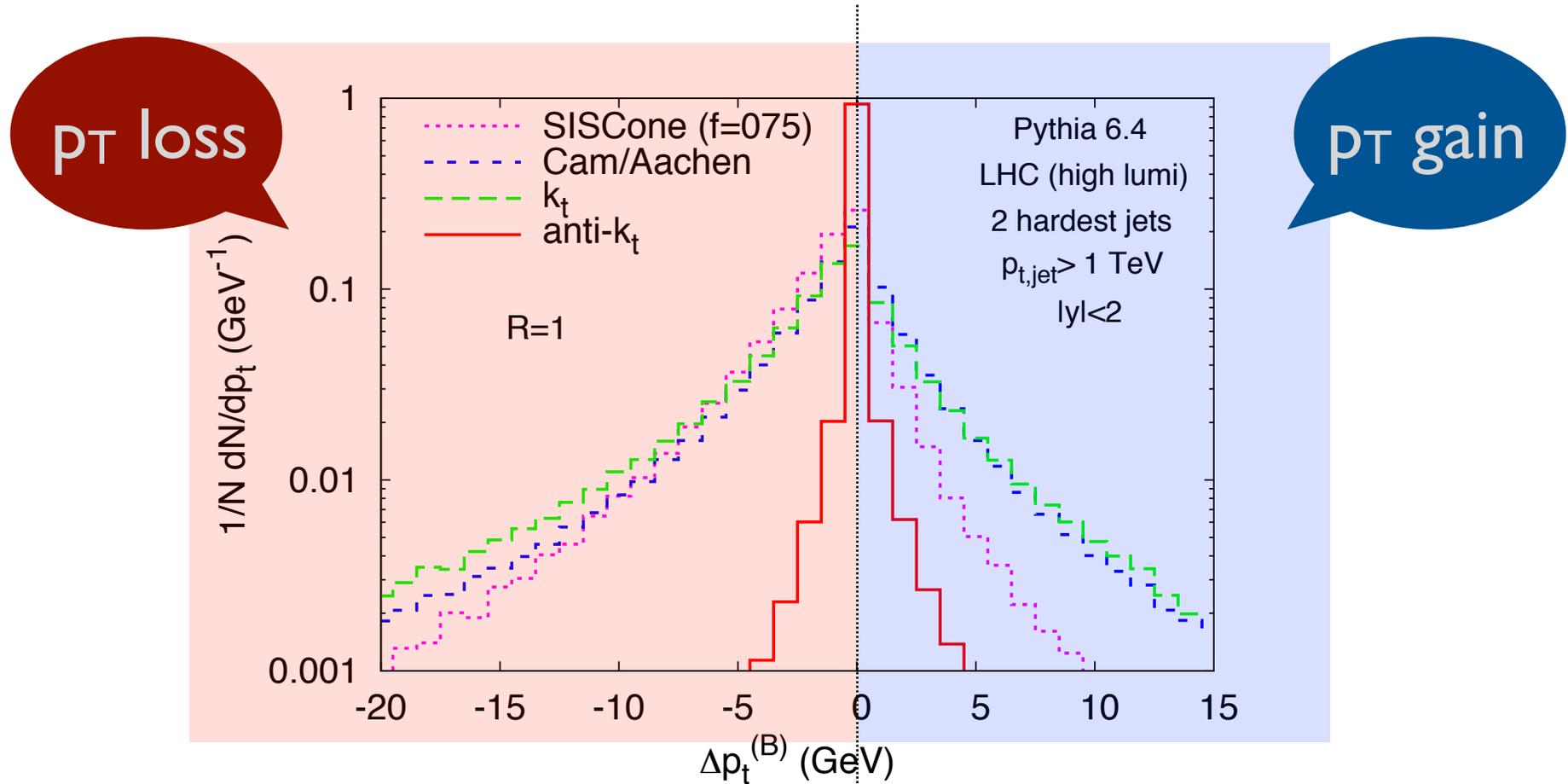
With
background



Backreaction **loss**

Backreaction **gain**

Resiliency: backreaction



Anti- k_t jets are much more resilient to changes from background immersion

(NB. Backreaction is a minimal issue in pp background and at large p_t .
Can be much more important in Heavy Ion collisions)