

Xiii

MEXICAN SCHOOL OF PARTICLES AND FIELDS

UNIVERSITY OF SONORA & SAN CARLOS, SONORA, MEXICO.

OCTOBER
2008

2-11

The Mexican School of Particles and Fields is a biennial conference organized by the Division of Particles and Fields of the Mexican Physical Society, designed to gather specialists in different areas of high energy physics to discuss the latest developments in the field. The format of the conference will consist of morning sessions devoted to theoretical and experimental reviews and afternoon thematic sessions.

Jet Reconstruction at the LHC (Lecture 2)

Peter Loch

University of Arizona

Tucson, Arizona, USA

(loch@physics.arizona.edu)



 THE UNIVERSITY
OF ARIZONA.

Arizona's First University.

UAPhysics
THE UNIVERSITY OF ARIZONA
College of Science

Remember?

Jets are bundles of particles with correlated kinematics

↓ ~25% photons, rest charged and neutral hadrons

Calorimeters are detectors of choice for jets

↓ Generate signals from neutral and charged particles

Calorimeter response to particles depends on particle type

↓ More signal from electrons/photons than from pions of the same energy (non-compensation)

↓ Larger intrinsic fluctuations for hadrons

↓ Direct proportionality of electron signal to incoming energy

↓ Hadron response is energy dependent

↓ Hadronic showers are less compact and larger

Jet response is a convolution of the jet particle content with particle response

↓ Typically higher than hadron response

Calorimeter signal definition affects image of jet in detector

↓ Towers and clusters

How To Deal With Non-Compensation



Can we get the hadronic shower branch signals up to a signal corresponding to the electromagnetic signal?

- ↓ Lower fluctuations
- ↓ Direct proportionality of energy and signal



One approach: cell signal weighting in highly granular calorimeter

- ↓ Small signal densities in a calorimeter cell indicate hadronic deposit and should receive an additional correction (weight)
 - 👤 Pioneered by CDHS (1977) and developed by H1 (1992)
- ↓ High signal densities indicate electromagnetic signals and don't need additional corrections
 - 👤 Dense, compact showers from electrons/photons



But how can we determine these weights?

- ↓ It's mostly a matter of context: are we trying to determine them for single particles (clusters) or jets
- ↓ ATLAS works with both approaches



Jet Calibration With Cell Weights



Statistical approach

- Weights are determined by resolution minimalization fits with calorimeter jets

$$\langle E_{calo}^{jet} \rangle = \left\langle \sum_{cells} w(\rho_{cell}, \vec{X}_{cell}) \cdot E_{cells} \right\rangle = \langle E_{true}^{jet} \rangle$$

- Truth reference typically corresponding simulated particle jet
- Weights will include primary electromagnetic component of jet
 - Do we really want this? They should only correct hadronic signals!
- Weights compensate all signal inefficiencies, not only e/h
 - Dead material corrections, leakage
 - Low level of factorization!
- Weights need to be determined for all jet finders and jet finder configurations
 - Need to find the jet first in uncalibrated signals
 - Then apply correct weight for given jet finder and configuration
 - Huge task!



Hadronic Calibration



Why not calibrating calorimeter signals first?

- ⚡ No jet context
- ⚡ But need other context for cell signal weighting normalization
→ topological cell cluster
 - 👤 Energy blobs follow shower shape somewhat



Cluster based hadronic calibration

- ⚡ Advantages to jet context: can use cluster shape to parametrise cell weights
 - 👤 Measure compactness of signal cluster by cluster
 - 👤 Shape and size variables are easily reconstructed for each cluster
 - 🌐 E.g. 2nd geometrical moments

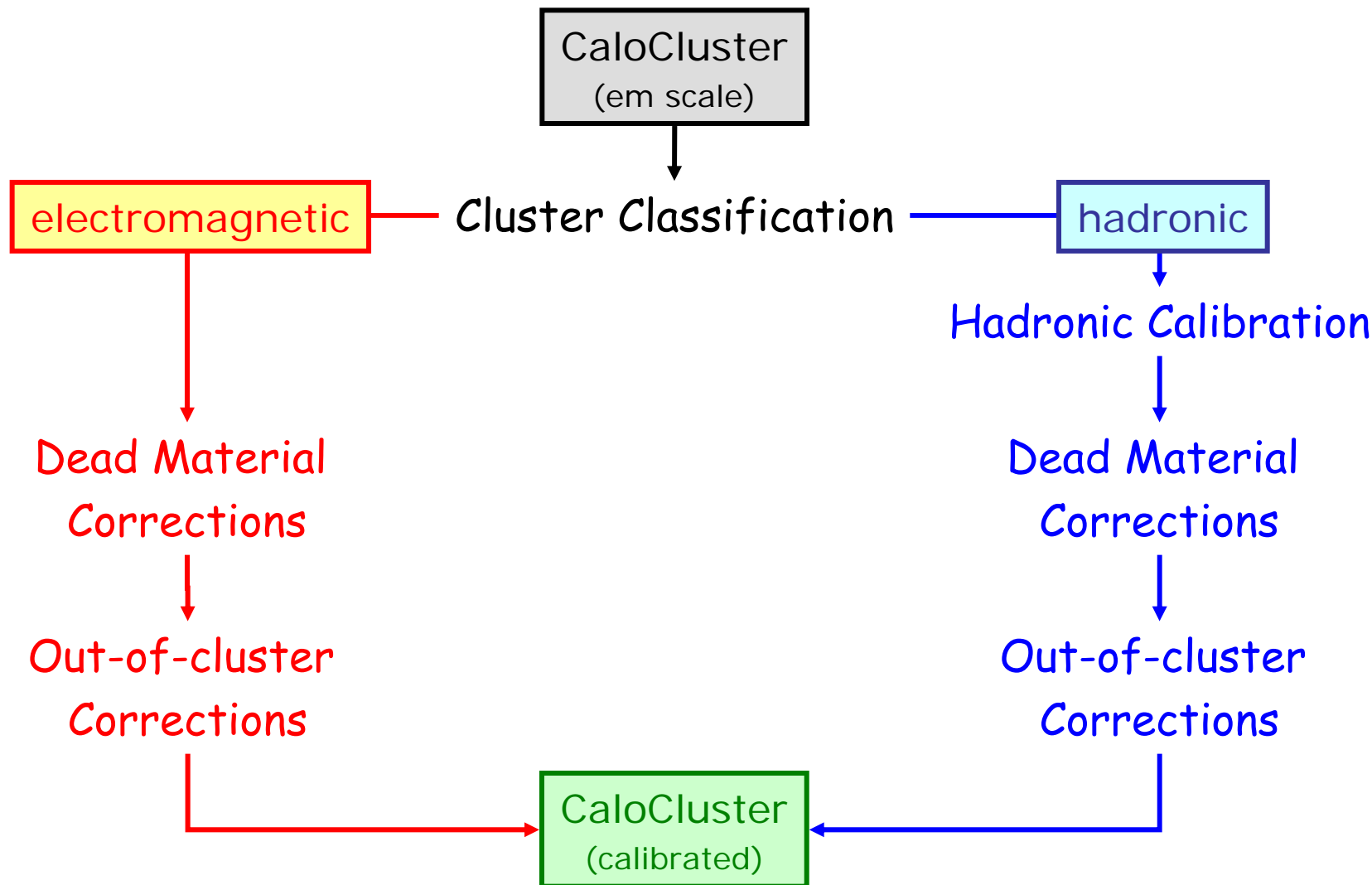


Allows factorization

- ⚡ Deal with e/h at detector level (not jet level)
- ⚡ Correct for local dead material at cluster level already
- ⚡ But need to apply further jet context corrections for particles lost in magnetic field and dead material losses not correlated with cluster signals



Local Hadronic Calibration

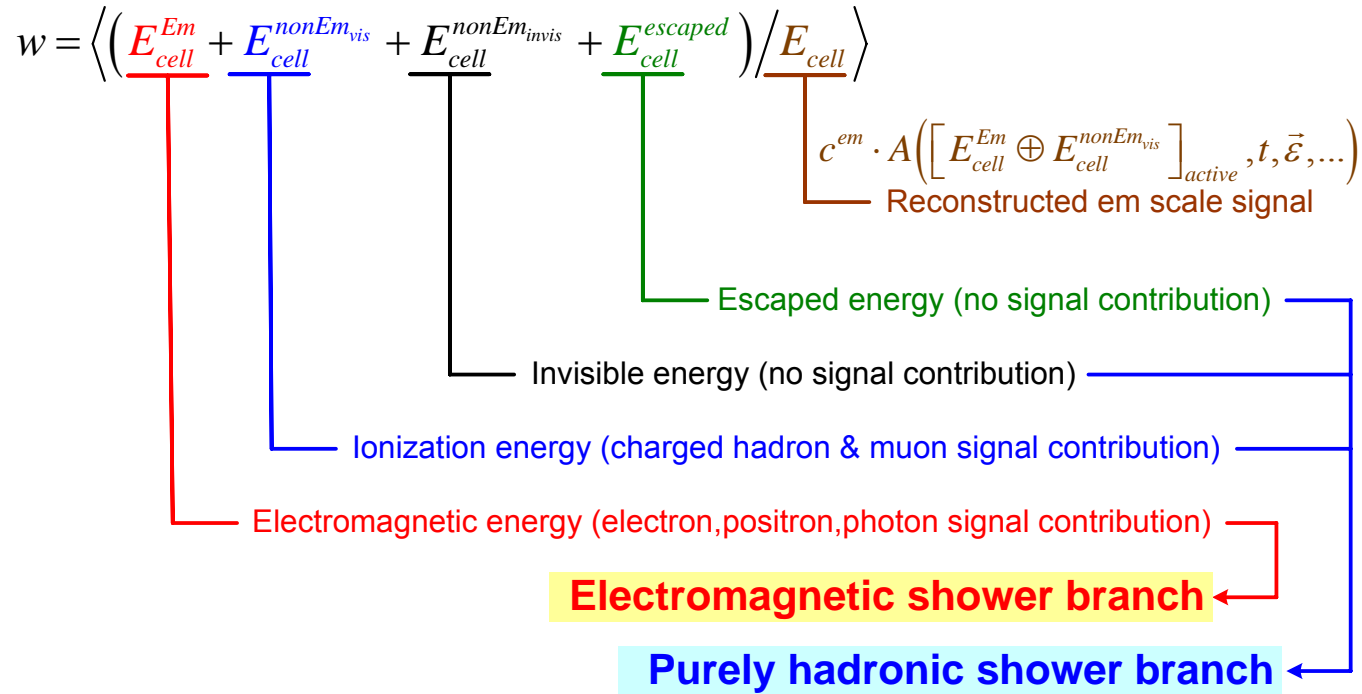


Cell Signal Weighting In ATLAS



Weights are extracted from simulations

↓ Signals and deposited energies from single charged pion MC



Weights are calculated as function of cluster & cell variables

↓ Signal environment used as additional indicator of hadronic character



Dead Material Energy Losses



Tracking device in front of a calorimeter

And maybe even a cryostat wall



Try to use the cluster signal to correct for the losses

In front

Inbetween

Central ATLAS!

Between

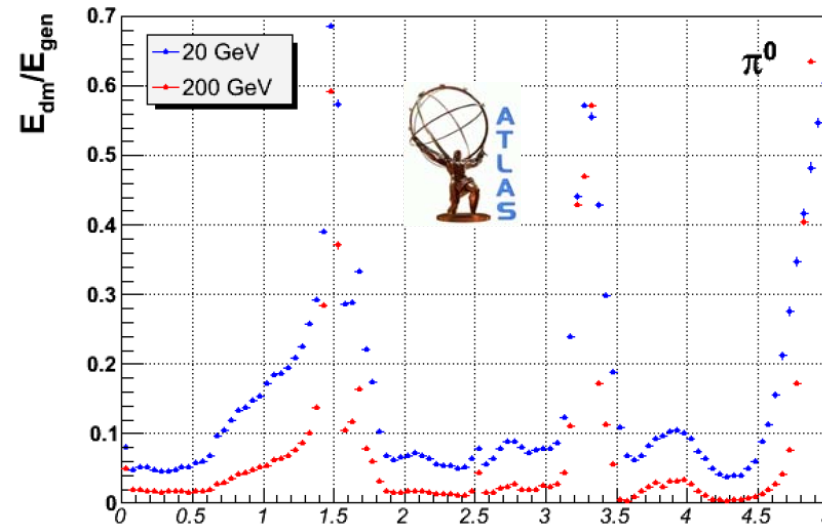
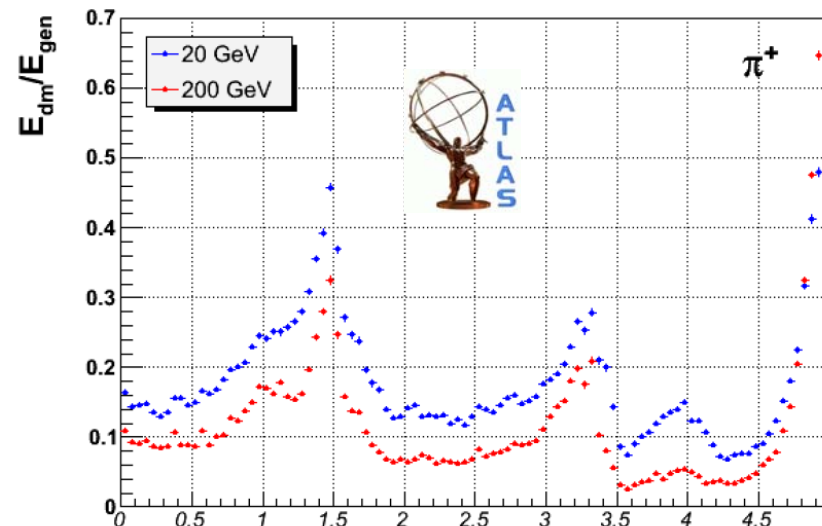
Endcap and forward cracks



Approach:

Correlate cluster signal with nearby dead material energy loss in simulations

Do this for each dead material region separately



η



Local Hadronic Calibration Summary



Attempt to calibrate hadronic calorimeter signals in smallest possible signal context

- ↓ Topological clustering implements noise suppression with least bias signal feature extraction
- ↓ No bias towards a certain physics analysis
- ↓ Good common signal base for all hadronic final state objects
 - 👤 Jets, missing Et, taus



Factorization of cluster calibration

- ↓ Cluster classification largely avoids application of hadronic calibration to electromagnetic signal objects
 - 👤 Low energy regime challenging
- ↓ Signal weights for hadronic calibration are functions of cluster and cell parameters and variables
 - 👤 Cluster energy and direction
 - 👤 Cell signal density and location (sampling layer)
- ↓ Dead material and out of cluster corrections are independently applied



Overview



Lecture 1 (Saturday, October 4th, 2008, 12:30-13:30): Signals from particle jets

- Experimentalist view on jets
- Brief review of the basics of calorimetric energy measurement
- Jet response of a non-compensating calorimeter
- Calorimeter signal reconstruction: cells, towers, clusters



Lecture 2 (Sunday, October 5th, 2008, 12:30-13:30): Jet algorithms and reconstruction

- Physics environment for jet reconstruction at LHC
- Jet algorithms and reconstruction guidelines
- Jet calibration strategies
- Jet Reconstruction Performance



Lecture 3 (Sunday, October 5th, 2008, 17:00-18:00): Refinement of jet reconstruction at LHC

- Refined calibration using other detectors
- Tagging jets from pile-up
- The origin of jets: masses and shapes
- AOB



Jet Reconstruction Environment At LHC

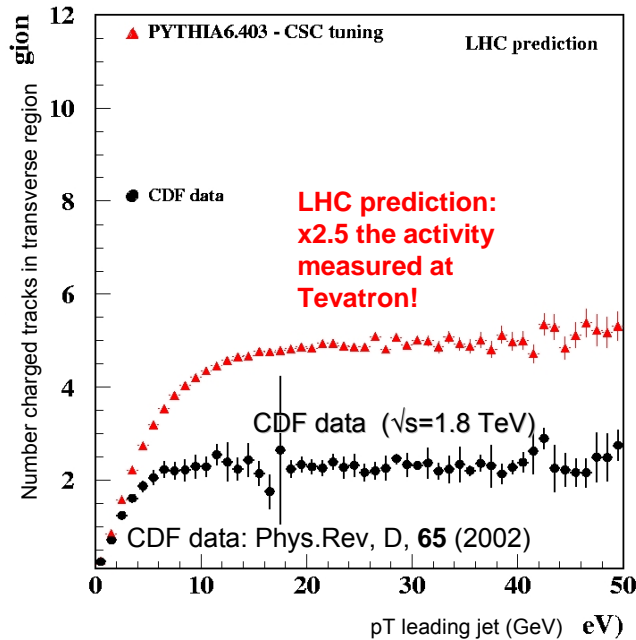


Physics Environments @ LHC

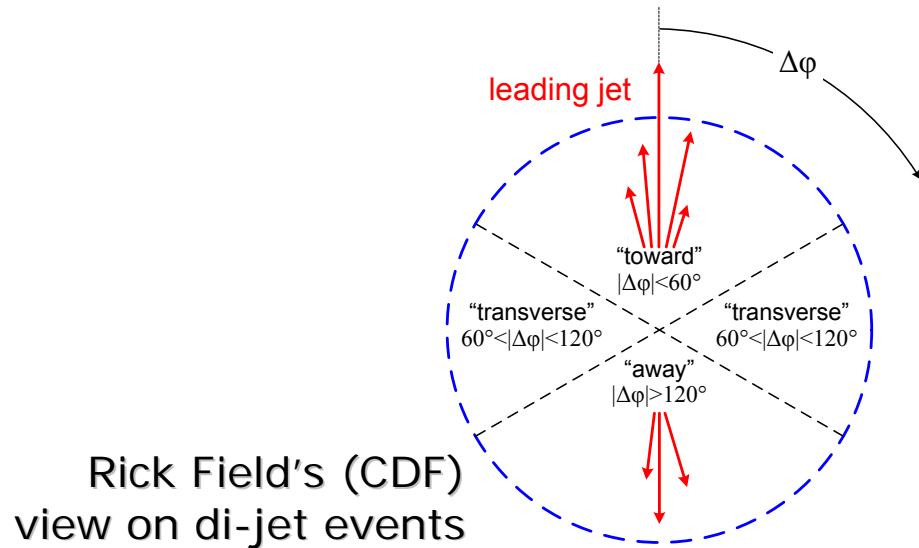
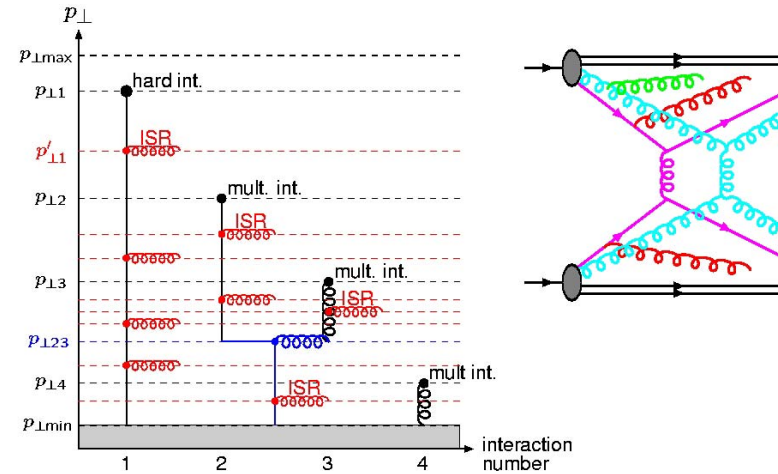


Physics environment different from Tevatron

- ↓ Increased underlying event activity (more phase space)
 - 👤 Already at lowest (initial) luminosities $\sim 10^{31}-10^{32} \text{ cm}^{-2} \text{ s}^{-1}$
- ↓ Additional activity from pile-up
 - 👤 Proportional to instantaneous luminosity



Interleaved Multiple Interactions



Rick Field's (CDF) view on di-jet events

Pile-Up Events



Multiple collisions of other protons in bunch crossings

- ↓ Effect of high luminosity at LHC
- ↓ Independent of (triggered) hard scattering process
- ↓ Generates additional particles and particle jets



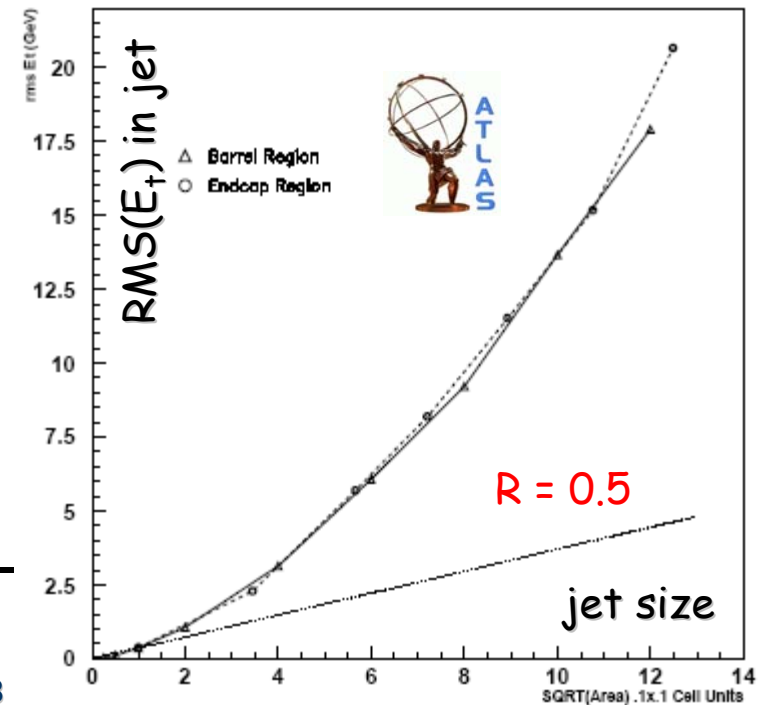
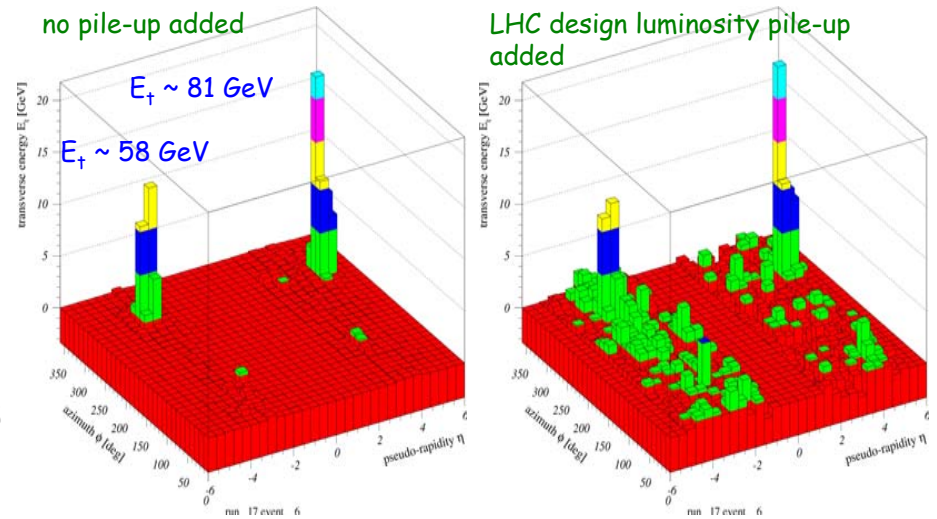
Generates additional noise in detectors

- ↓ High bunch crossing rate creates signal history from pile-up in detectors
 - 👤 Sensitivity up to 25 bunch crossings for "slow" detectors (calorimeters)
- ↓ Average contribution cancels due to chosen signal shaping



Pile-up noise is not gaussian

- ↓ Coherent effects due to physics source



Final State At LHC



Final state can make jet reconstruction more challenging

↓ “Low activity” signatures like QCD di-jets (2- \rightarrow 2 process + radiation)

👤 Mostly gluons, more quarks at high p_T

↓ Busy final states in SUSY

👤 Many leptons

👤 Many (quark) jets

👤 Higher likelihood of signal overlap



Source of jet

↓ Hadronic W decays in $t\bar{t}b\bar{a}$ production

👤 W color-disconnected to rest of collision

👤 Quark jets

↓ Prompt photon + jet(s)

👤 Mostly quark jet



Physics Environments @ LHC



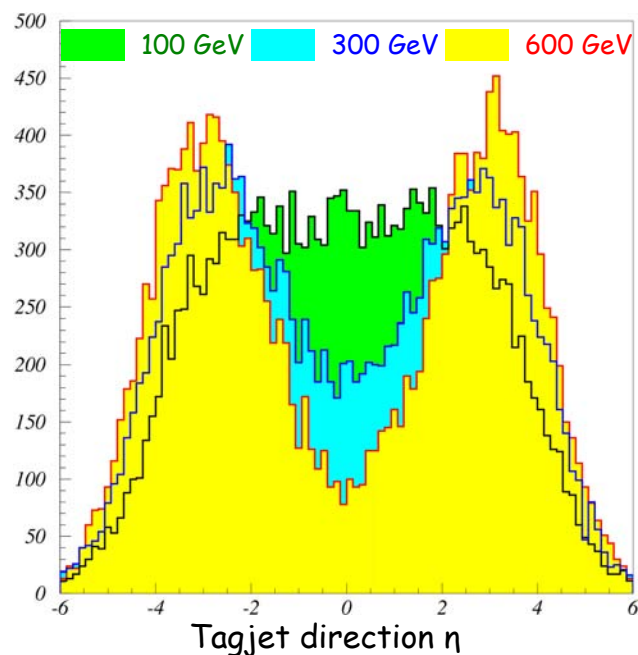
Discovery physics at LHC

↓ Expect extreme busy final states

- 👤 Lots of leptons, missing Et and jets O(10) in SUSY
- 👤 Many x 10 jets in black hole production
- 👤 Good spatial resolution power to find the jets
- 👤 Good energy resolution for reliable missing Et calculation

↓ Need large rapidity coverage

- 👤 Tag vector boson fusion production of Higgs and exotics
 - 🌐 WW, WZ, ZZ with associated quark jets
- 👤 These “tag jets” often go forward
- 👤 Jets are uncorrelated with each other, but balance the central system (Higgs)



Direction of tag jets in Higgs VBF production for $m_H = 100, 300, 600$ GeV

this is a very old plot!





Finding Jets

Jet Definition



Jet finding algorithm and its configuration

- ⤵ Seeded or seedless cone and its parameters
 - Cone size, seed threshold
 - Recombination algorithm
- ⤵ Recursive recombination algorithms
 - Distance parameter
 - Recombination algorithm



Signal or constituent definition

- ⤵ Calorimeter towers or clusters
- ⤵ Reconstructed tracks
- ⤵ Generated particles
- ⤵ Generated partons

“Snowmass”

$$E_T^{jet} = \sum E_{T,i}$$

$$\eta_{jet} = \frac{1}{E_T^{jet}} \sum E_{T,i} \cdot \eta_i$$

$$\phi_{jet} = \frac{1}{E_T^{jet}} \sum E_{T,i} \cdot \phi_i$$

4-momentum

$$(E_{jet}, \vec{p}_{jet}) = \left(\sum E_i, \sum \vec{p}_i \right)$$



Theoretical Requirements



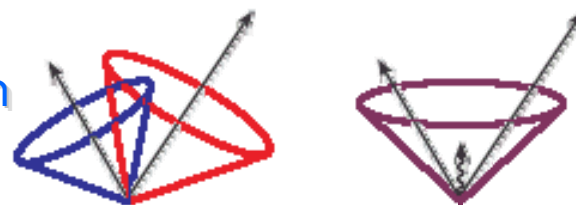
Infrared safety

- Additional soft particles should not affect jet reconstruction
- See extra slides from Gavin Salam



Collinear safety

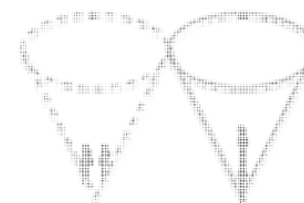
- Split energies (one instead of two particles) should not change the jet
- See extra slides from Gavin Salam



infrared sensitivity
(soft gluon radiation merges jets)



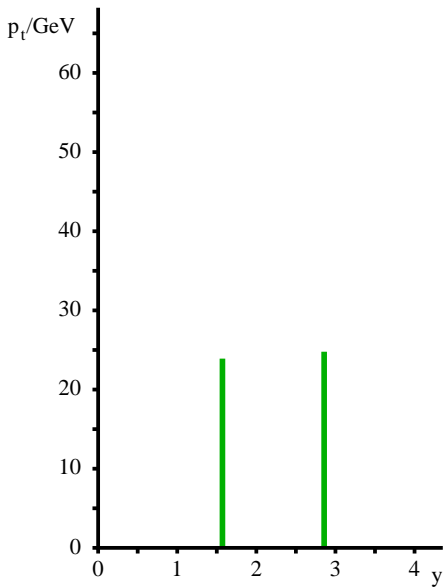
collinear sensitivity (1)
(sensitive to E_\perp , ordering of seeds)



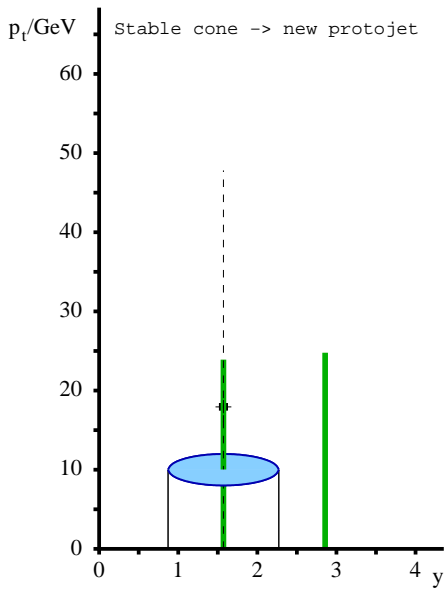
collinear sensitivity (2)
(signal split into two towers below threshold)



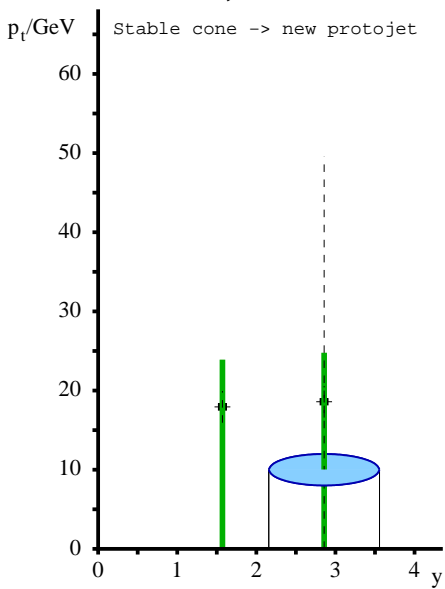
(Animation courtesy of Gavin Salam)



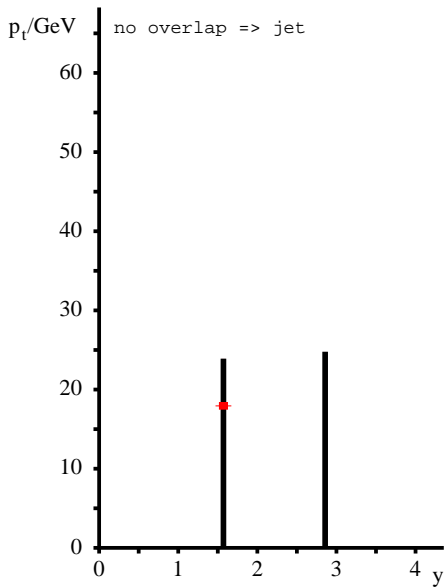
(Animation courtesy of Gavin Salam)



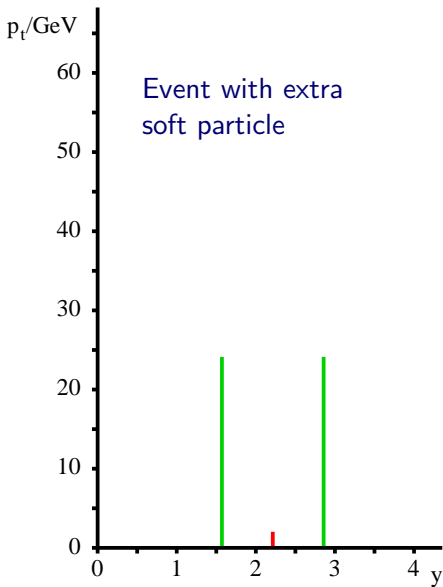
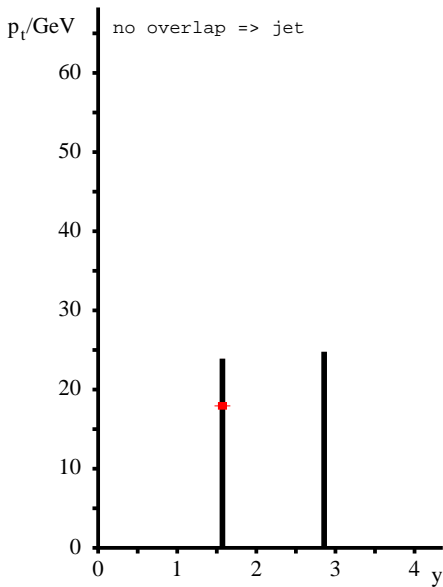
(Animation courtesy of Gavin Salam)



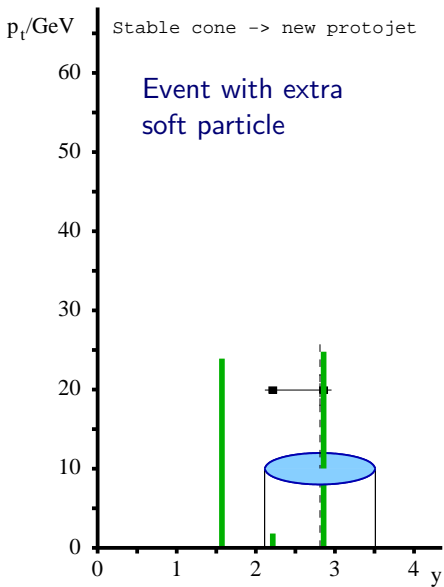
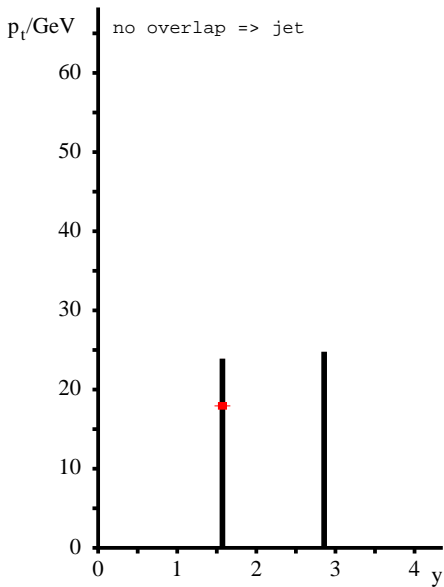
(Animation courtesy of Gavin Salam)



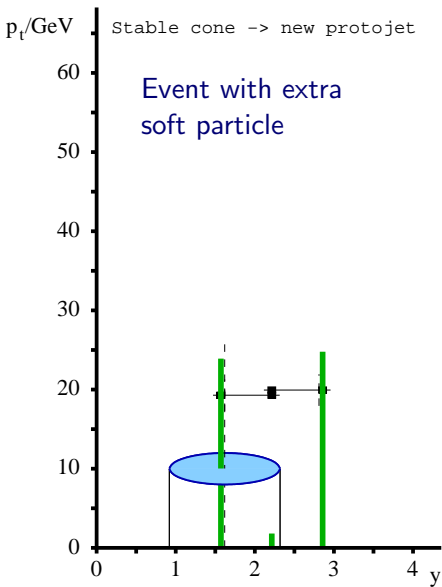
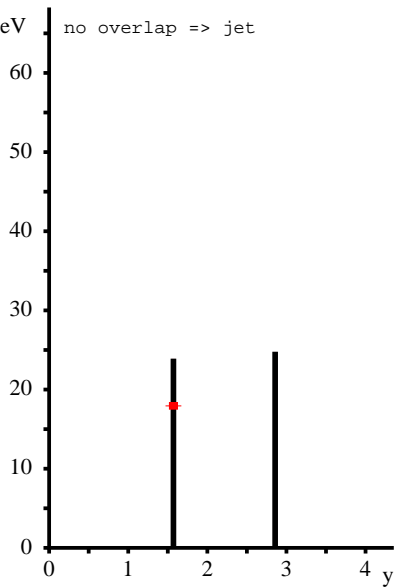
(Animation courtesy of Gavin Salam)



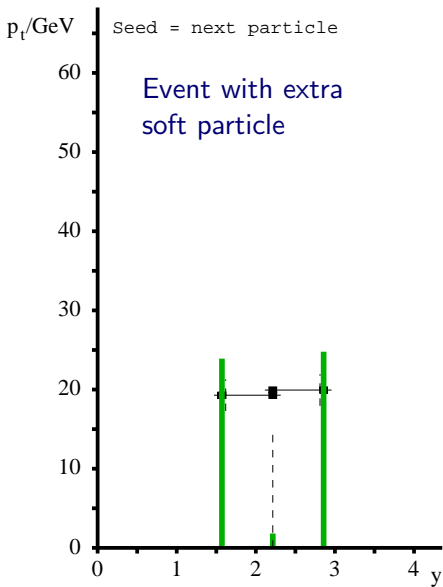
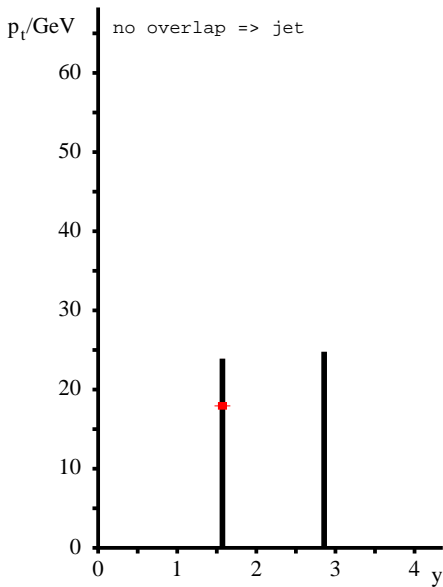
(Animation courtesy of Gavin Salam)



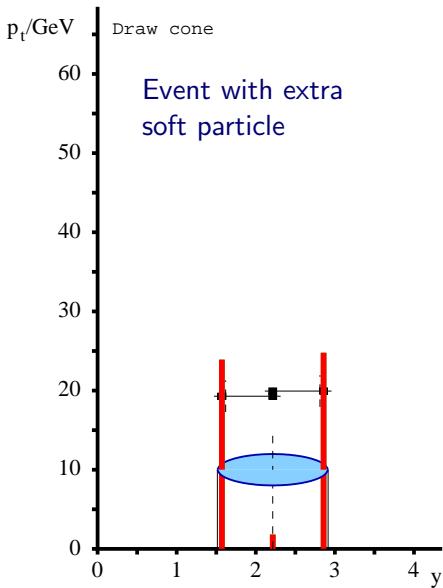
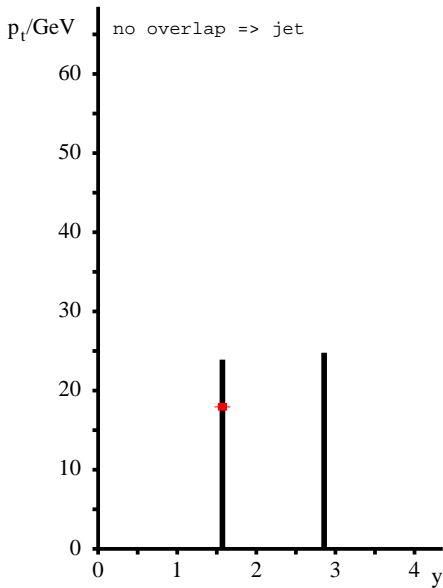
(Animation courtesy of Gavin Salam)



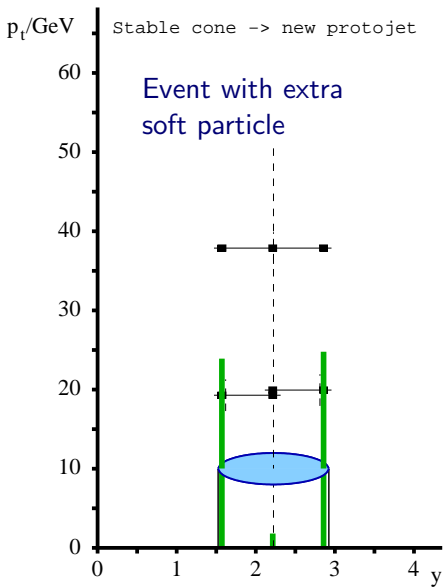
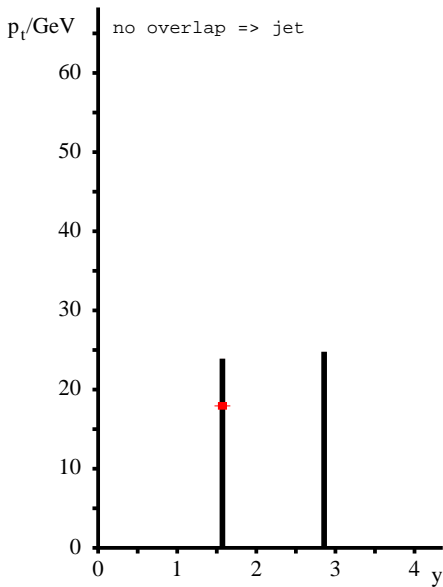
(Animation courtesy of Gavin Salam)



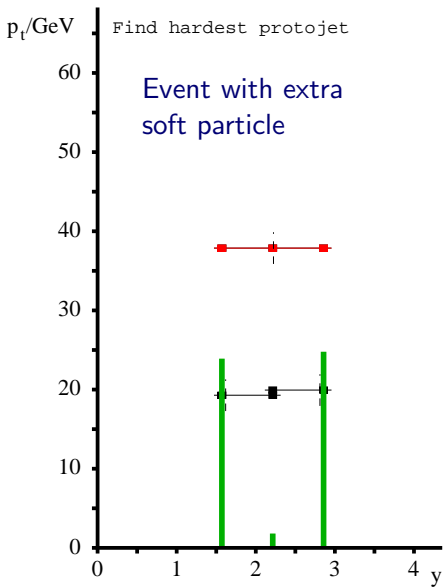
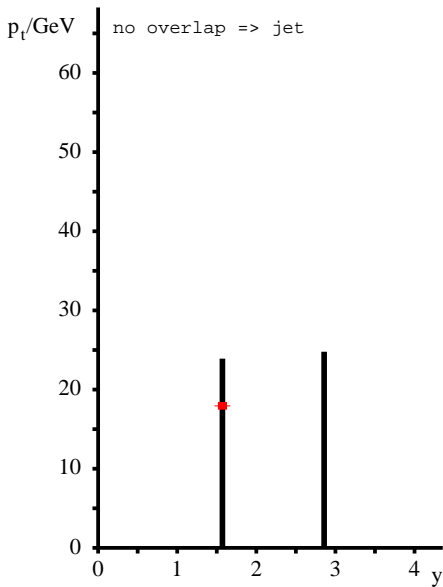
(Animation courtesy of Gavin Salam)



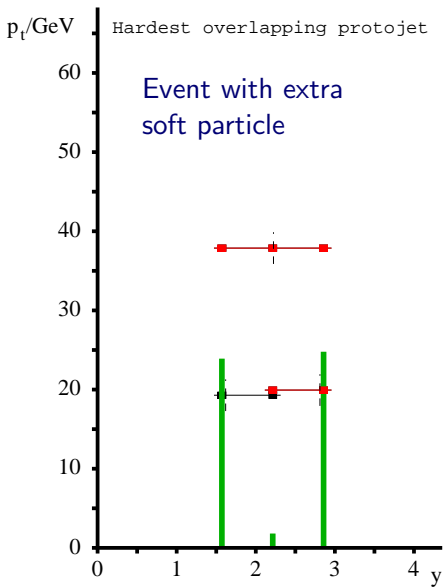
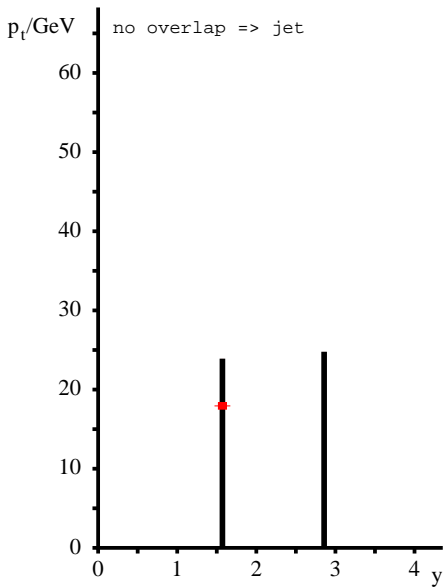
(Animation courtesy of Gavin Salam)



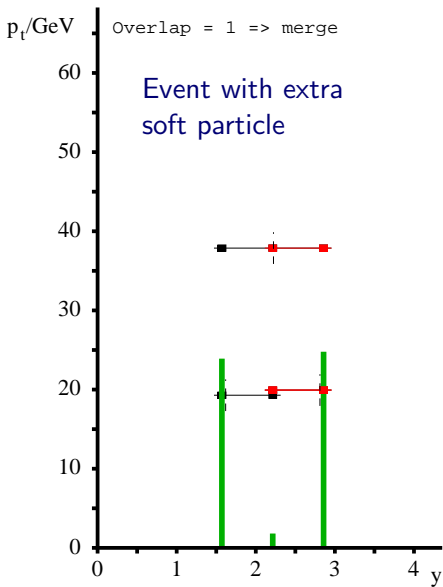
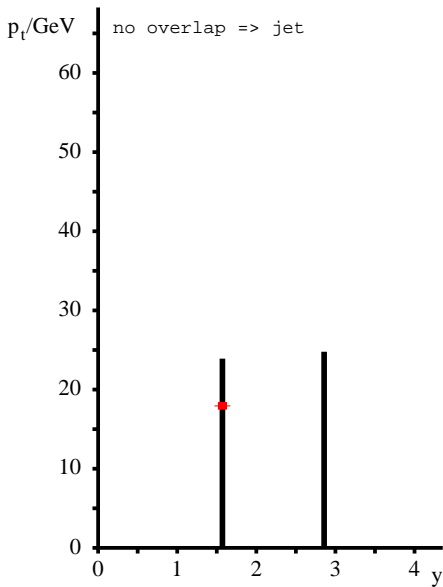
(Animation courtesy of Gavin Salam)



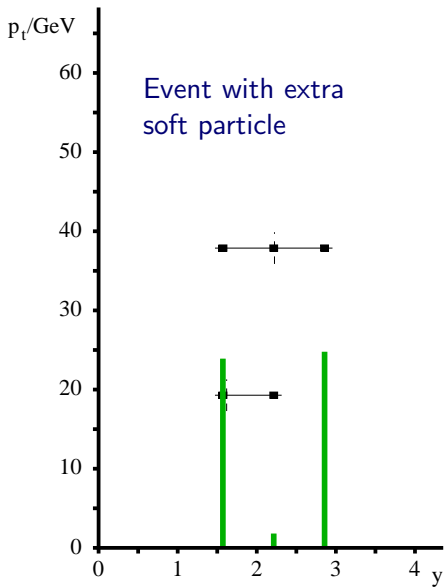
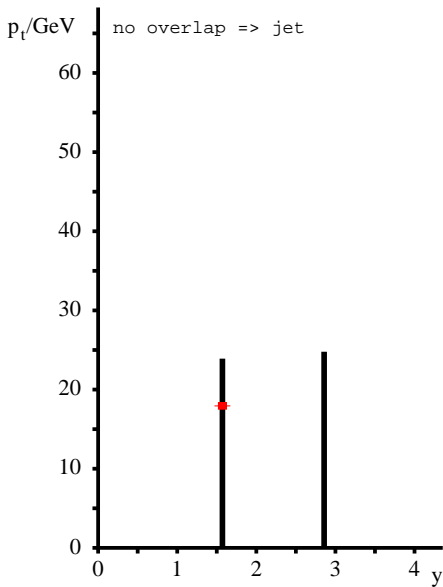
(Animation courtesy of Gavin Salam)



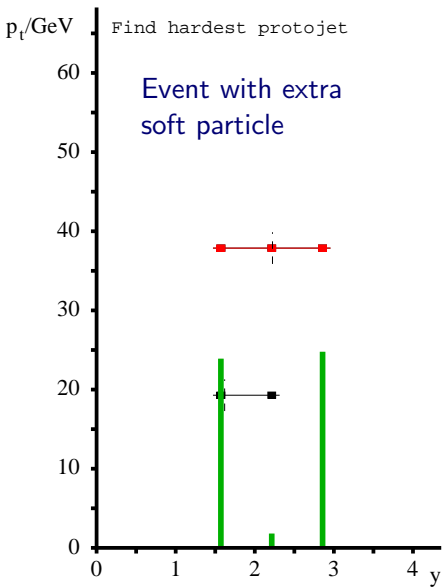
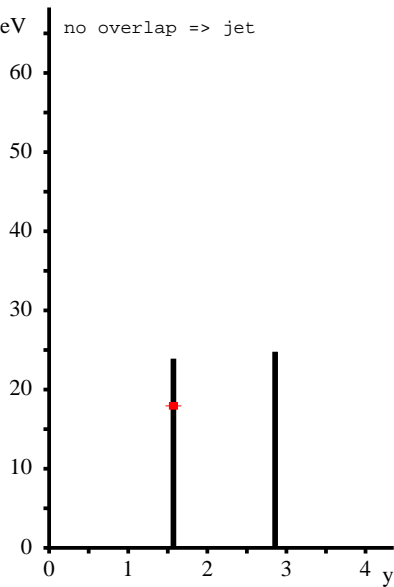
(Animation courtesy of Gavin Salam)



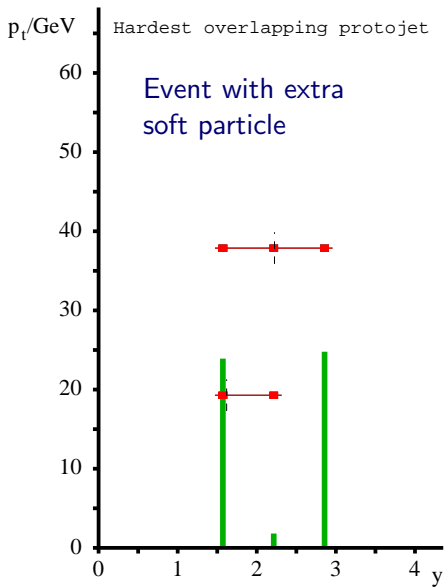
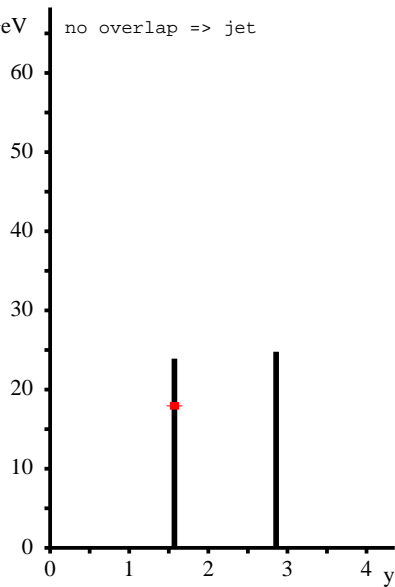
(Animation courtesy of Gavin Salam)



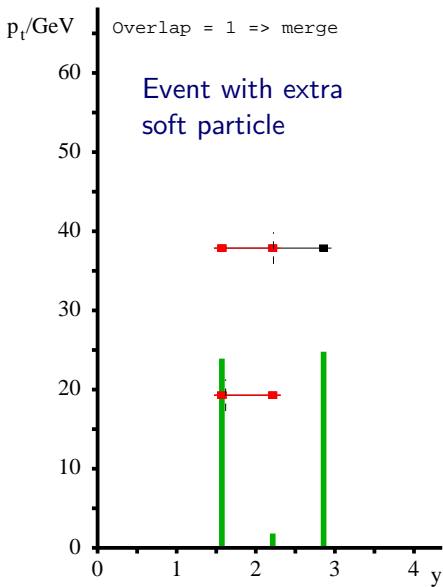
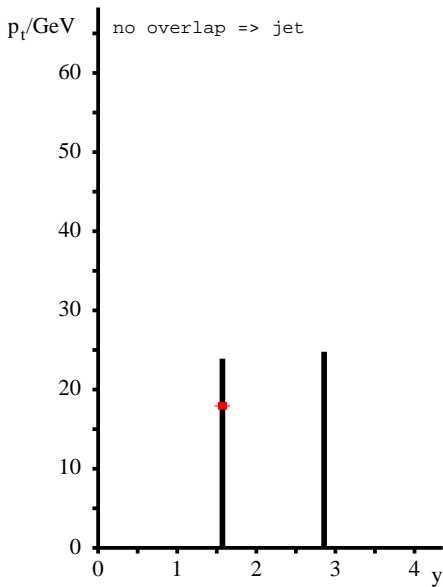
(Animation courtesy of Gavin Salam)



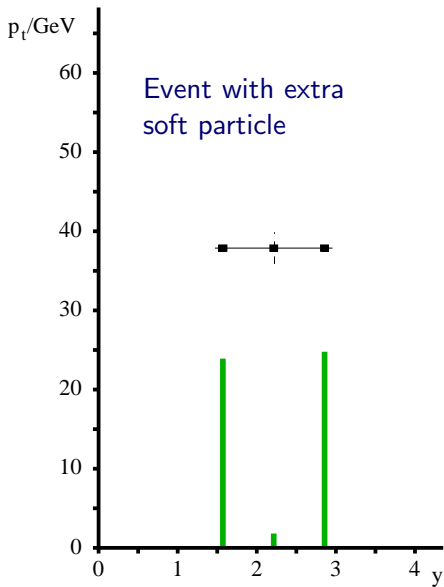
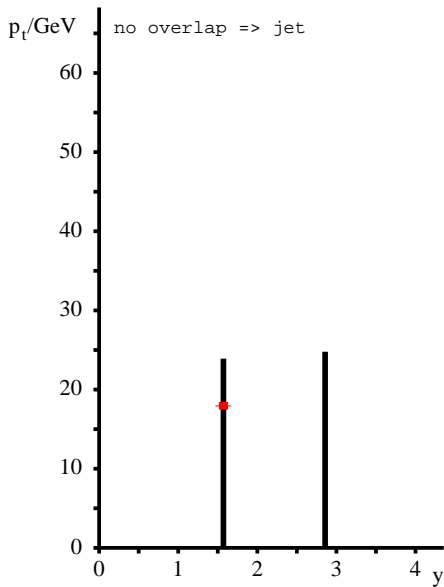
(Animation courtesy of Gavin Salam)



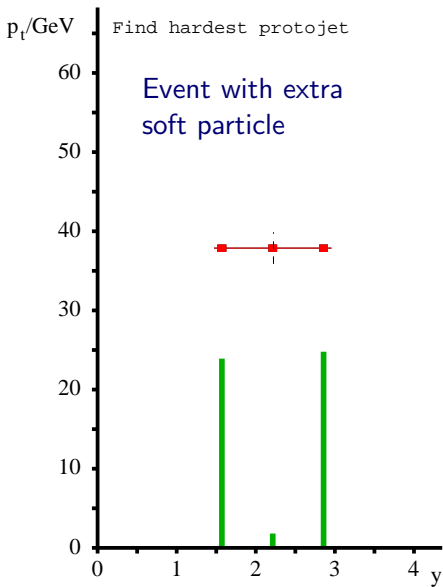
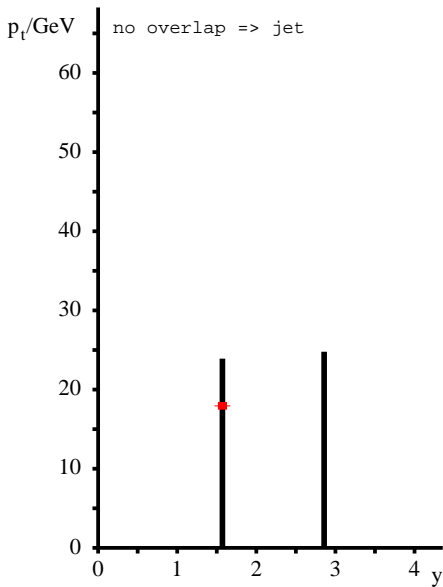
(Animation courtesy of Gavin Salam)



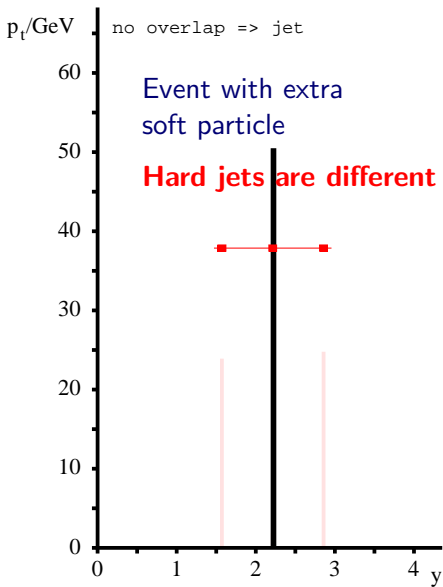
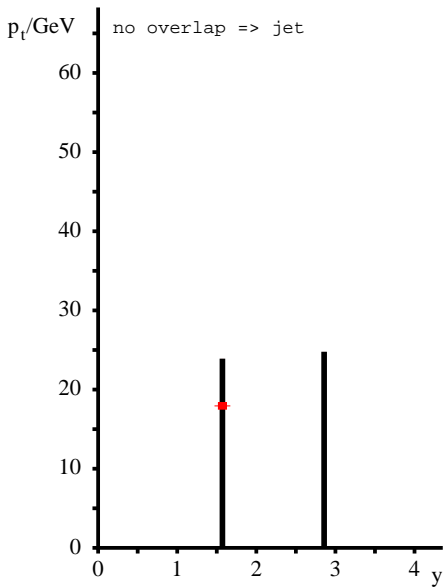
(Animation courtesy of Gavin Salam)



(Animation courtesy of Gavin Salam)



(Animation courtesy of Gavin Salam)



Theoretical Requirements



Infrared safety

Additional soft particles should not affect jet reconstruction

See extra slides from Gavin Salam



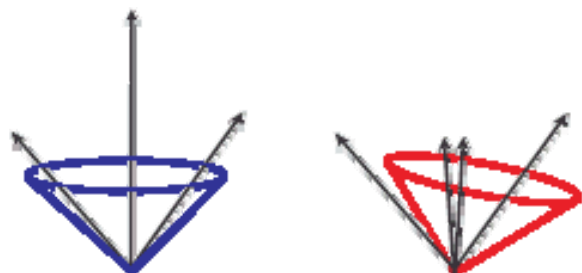
infrared sensitivity
(soft gluon radiation merges jets)



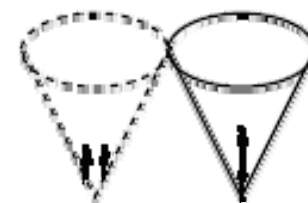
Collinear safety

Split energies (one instead of two particles) should not change the jet

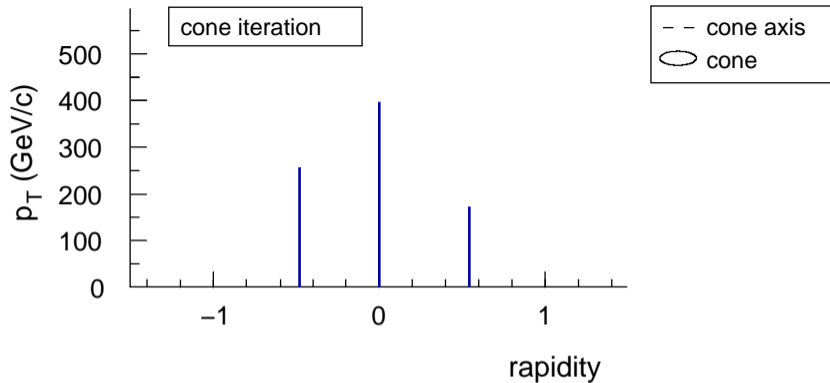
See extra slides from Gavin Salam



collinear sensitivity (1)
(sensitive to E_{\perp} ordering of seeds)

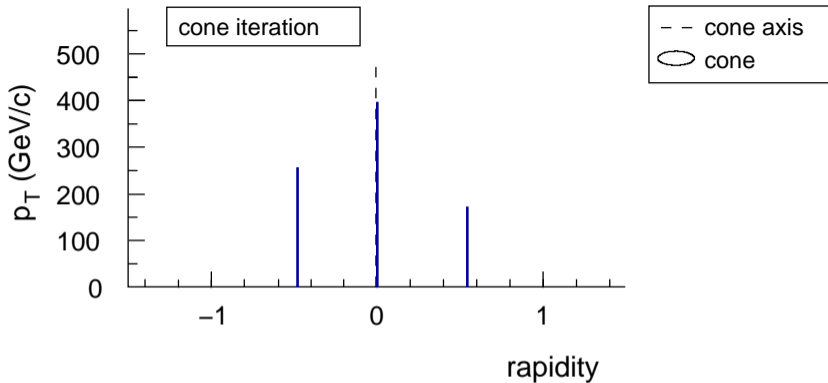


collinear sensitivity (2)
(signal split into two towers below threshold)



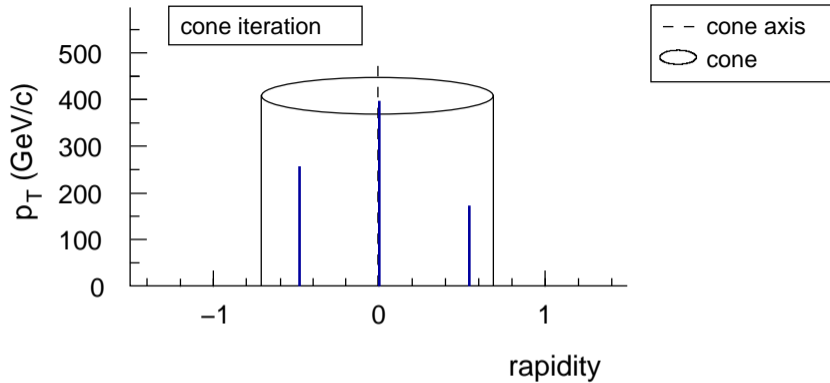
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



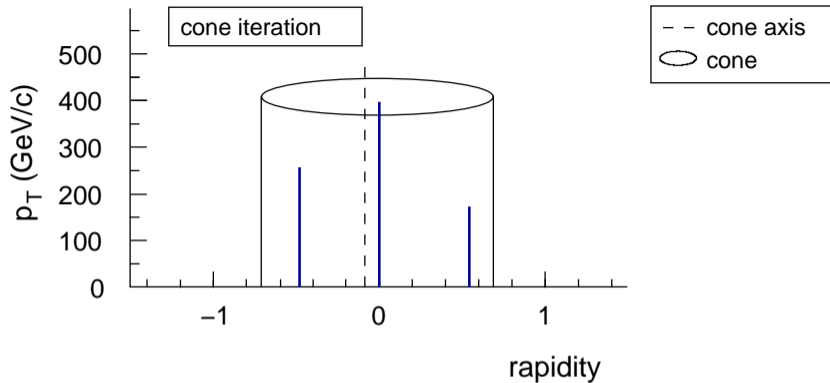
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



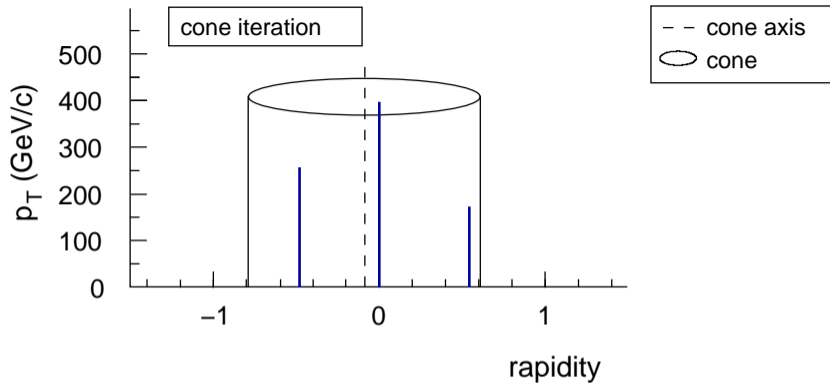
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



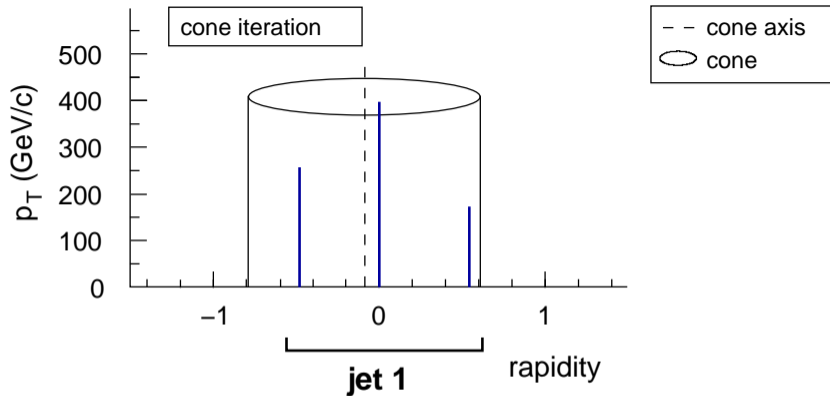
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



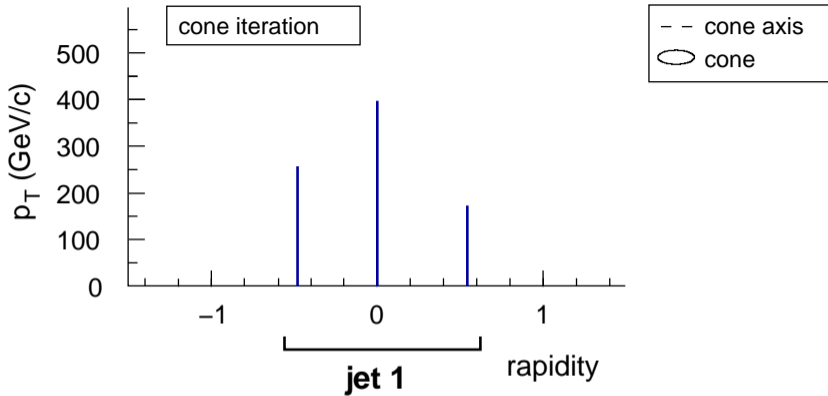
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



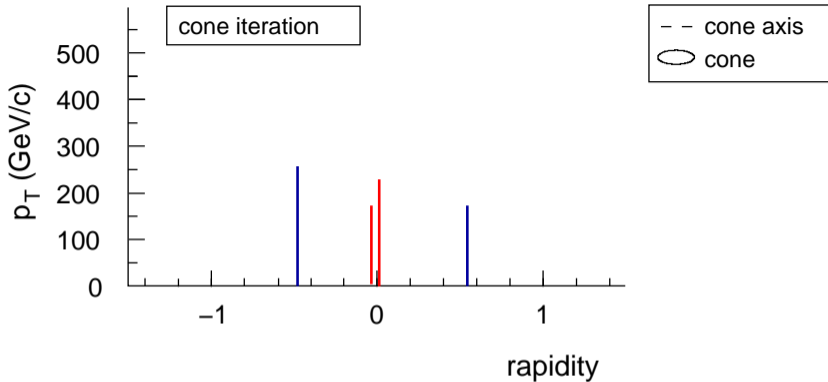
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



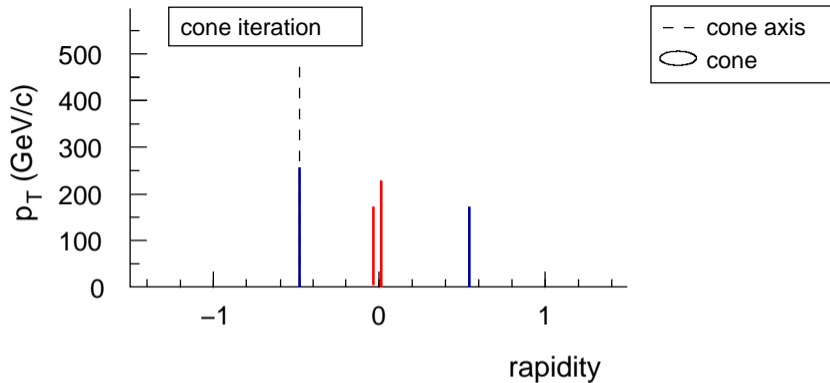
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



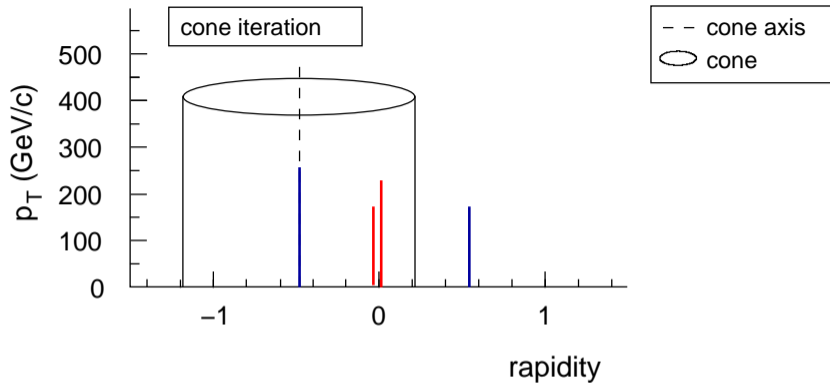
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



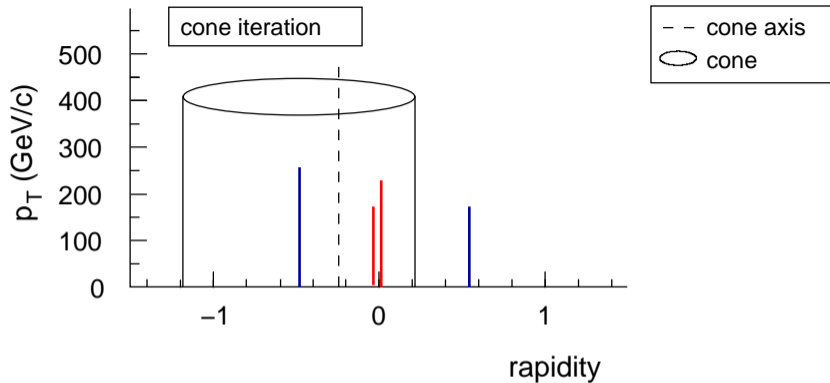
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



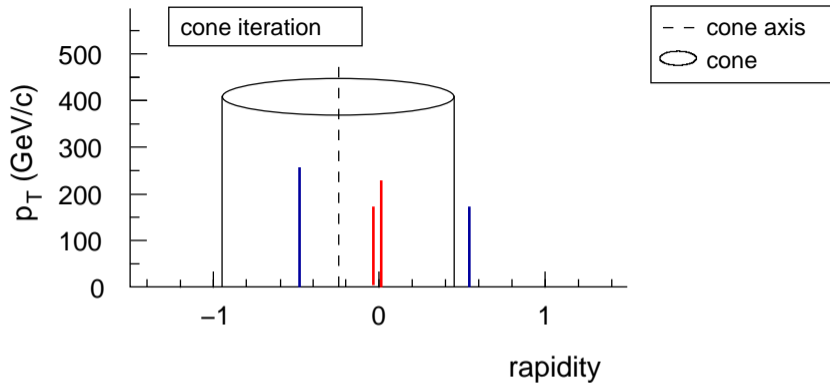
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



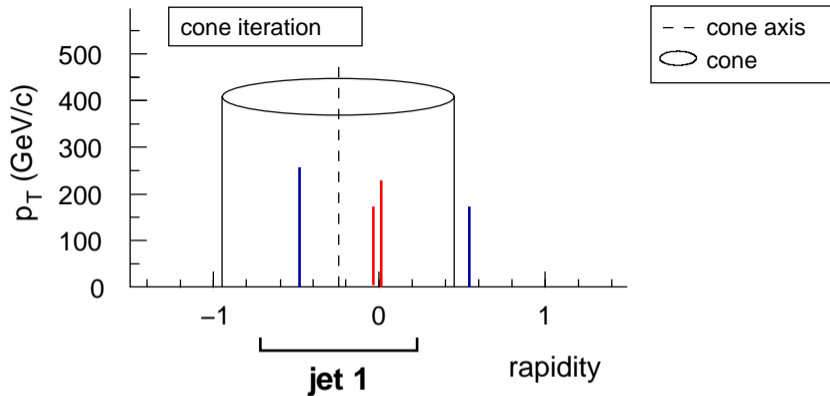
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



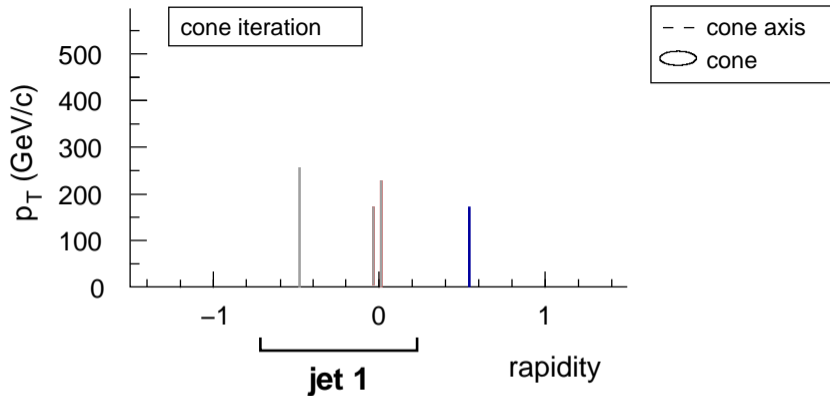
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



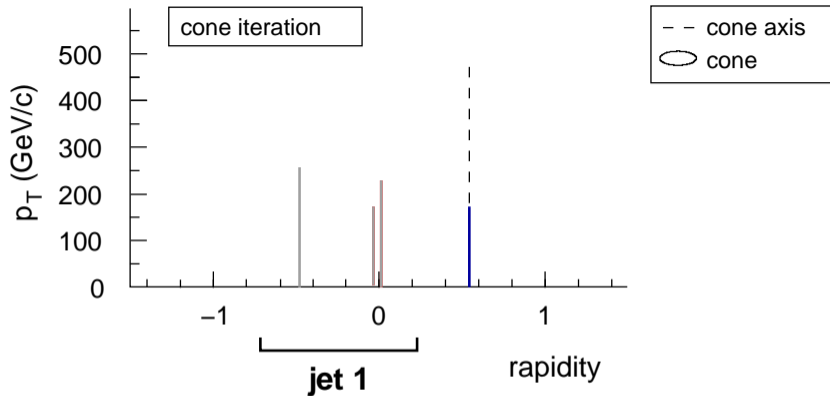
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



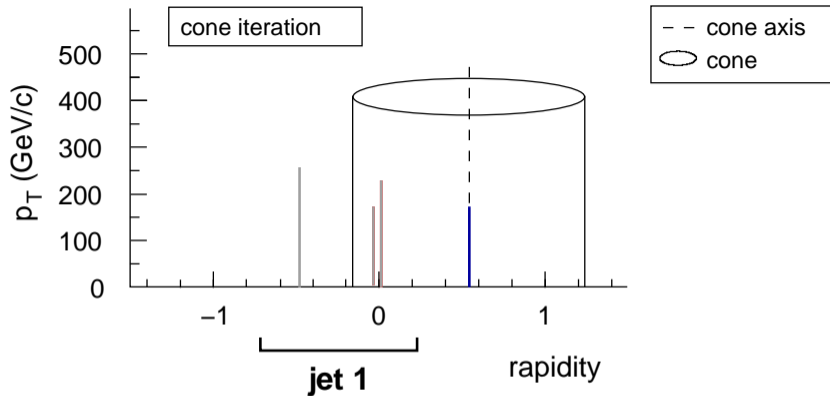
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



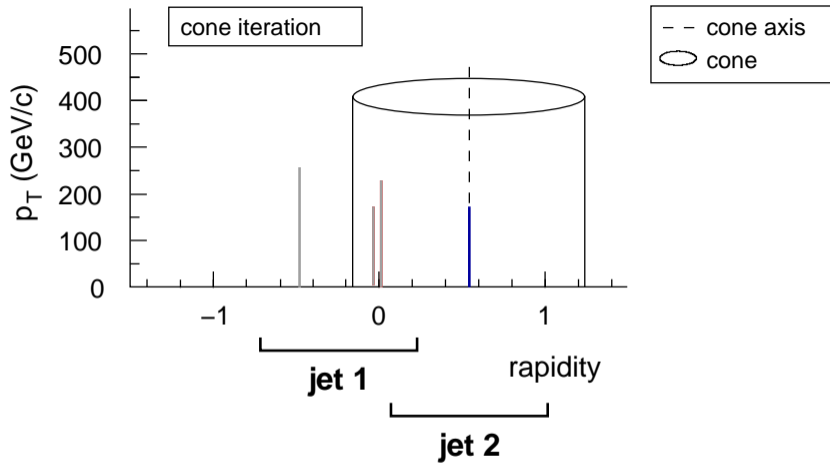
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



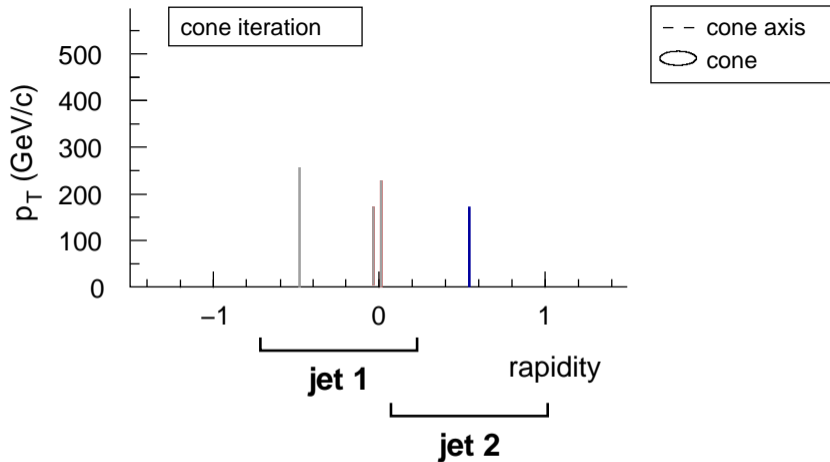
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \Rightarrow perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



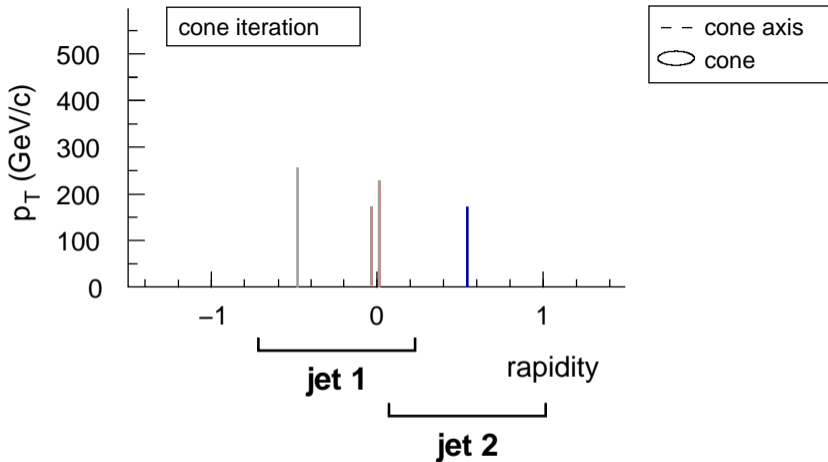
Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \implies perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \implies perturbative calculations give ∞

(Animation courtesy of Gavin Salam)



Collinear splitting can modify the hard jets: ICPR algorithms are collinear unsafe \implies perturbative calculations give ∞

(Animation courtesy of Gavin Salam)

Experimental Requirements for Jet Finders



Detector technology independence

- ↓ Minimal contributions to spatial and energy resolution
- ↓ Insignificant effects of detector environment
 - 👤 Noise
 - 👤 Dead material
 - 👤 Cracks
- ↓ Easy to calibrate
 - 👤 Well...



Environment independence

- ↓ Stability with changing luminosity
- ↓ Identify all physically interesting jets from energetic partons in perturbative QCD (pQCD)
- ↓ High reconstruction efficiency



Implementation

- ↓ Fully specified
 - 👤 All selections and other configurations known
- ↓ Efficient use of computing sources



Popular Jet Algorithms in ATLAS



Seeded cone

- ↓ Place cone with radius R around seed
 - 👤 $p_T > 1 \text{ GeV}$
- ↓ Collect all particles in cone
- ↓ Re-calculate energy and direction of cone
 - 👤 4-momentum recombination
- ↓ Find more particle in new cone
- ↓ Stop until no more particles to be found
 - 👤 Stable solution
 - 👤 Particles can be shared between jets
- ↓ Is not infrared safe
 - 👤 Needs split & merge (50% threshold)
- ↓ May miss significant energy
 - 👤 Dark jets



Recursive recombination (kT)

- ↓ Calculate for all particles i and pairs ij :

$$d_{ij} = \min(p_{t,i}^2, p_{t,j}^2) \frac{\Delta R_{ij}^2}{D^2}$$

$$= \min(p_{t,i}^2, p_{t,j}^2) \frac{\Delta y_{ij}^2 + \Delta \phi_{ij}^2}{D^2}$$

$$d_i = p_{t,i}^2$$
- ↓ Find minimum d_{min} from all d_i, d_{ij}
- ↓ If d_{min} is a d_i , call i a jet and remove it from the list
- ↓ Else combine i and j into a jet
 - 👤 4-momentum recombination
- ↓ Calculate new combinations
 - 👤 Stop when all particles declared jets
 - 👤 Each particle is part of one jet only (exclusive assignment)
- ↓ Infrared safe



kT Clustering Visualization (Gavin Salam)



k_t alg.: Find smallest of

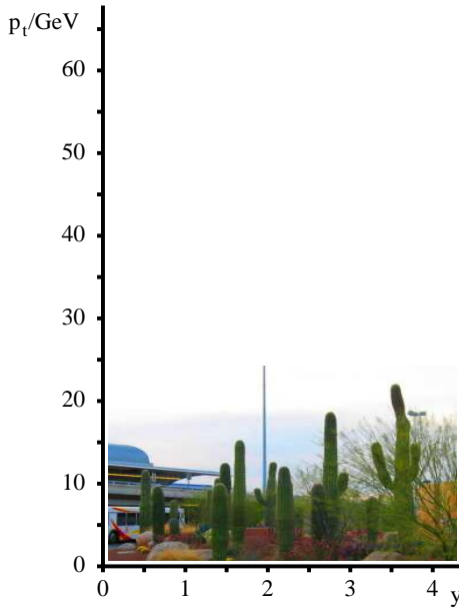
$$d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 / R^2, \quad d_{iB} = k_{ti}^2$$

If d_{ij} recombine; if d_{iB} , i is a jet
Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



(Animation courtesy of Gavin Salam)



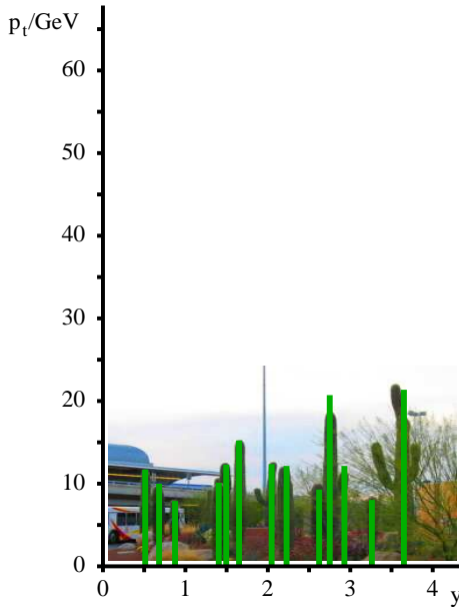
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)



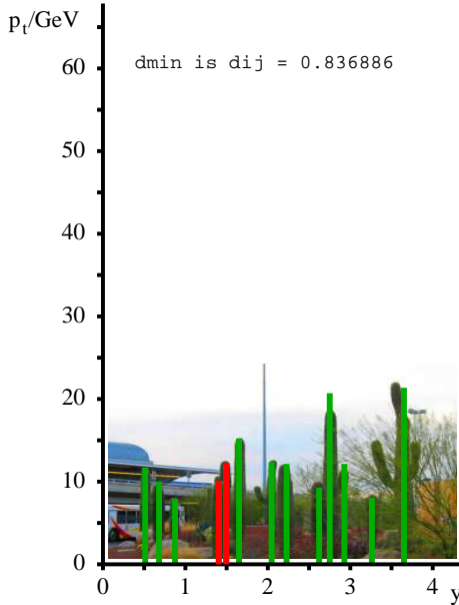
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)



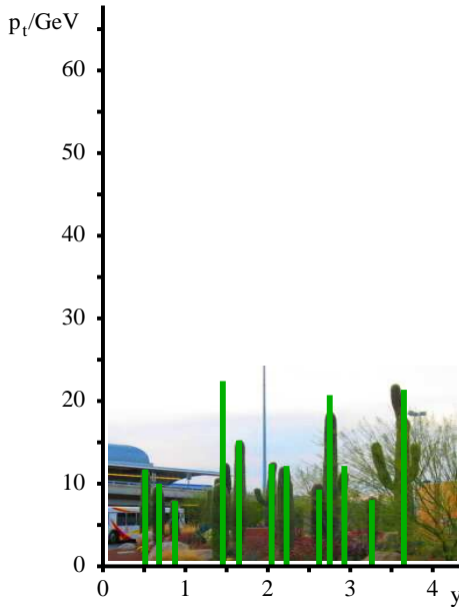
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)



k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

p_t/GeV

d_{\min} is $d_{ij} = 1.42534$

60

50

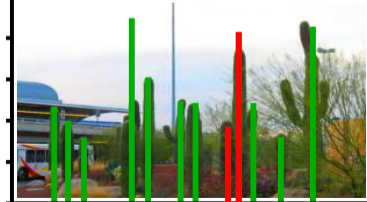
40

30

20

10

0



0

1

2

3

4

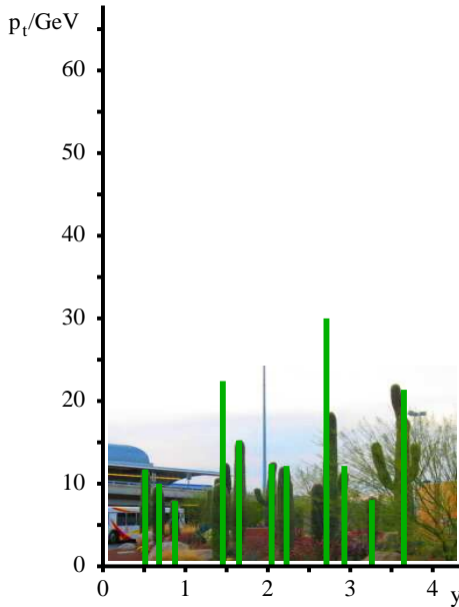
y (Animation courtesy of Gavin Salam)

k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

p_t/GeV

d_{\min} is $d_{ij} = 2.34324$

60

50

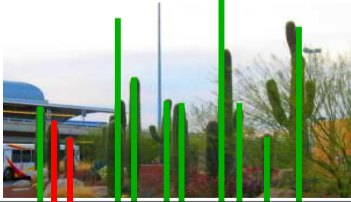
40

30

20

10

0



0

1

2

3

4

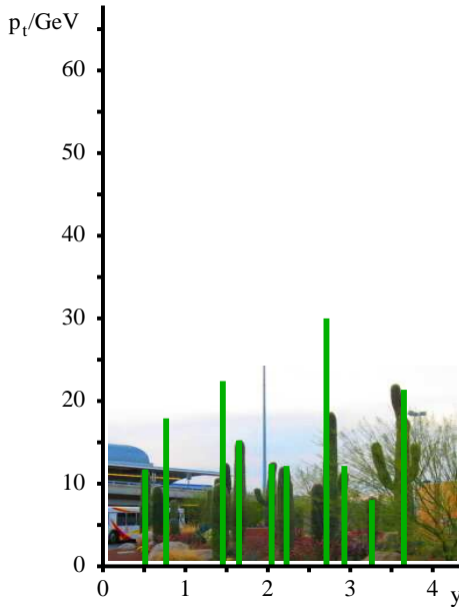
y (Animation courtesy of Gavin Salam)

k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



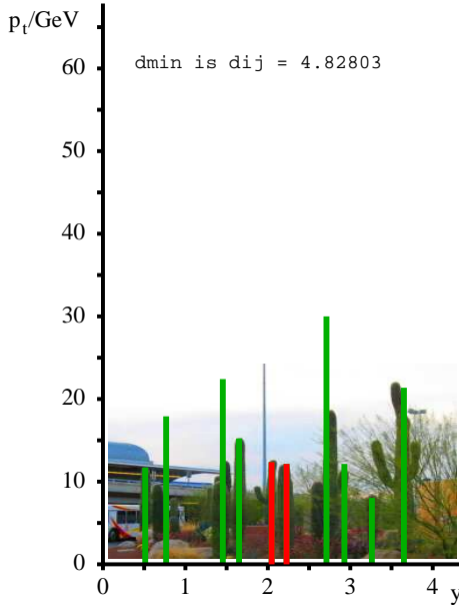
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)

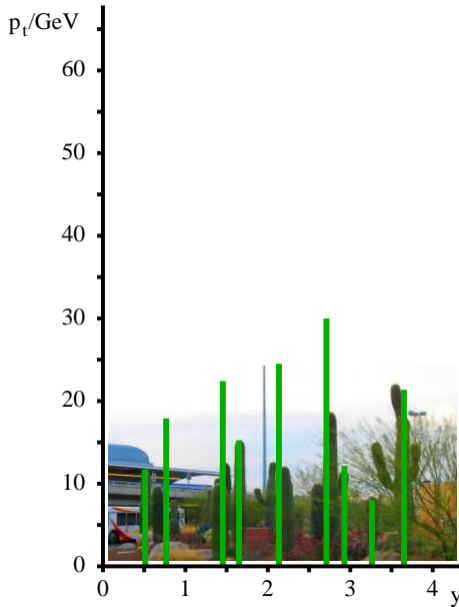


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



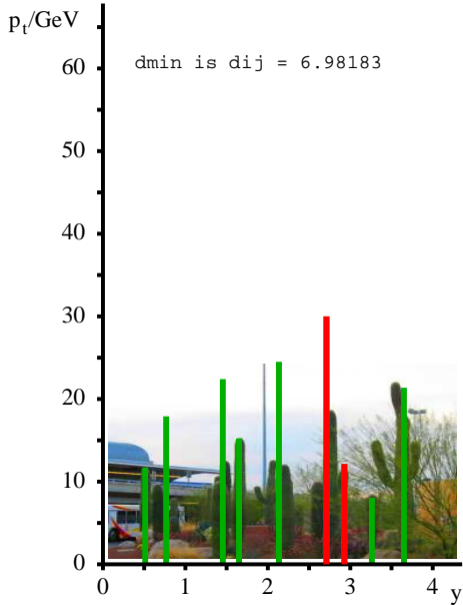
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)

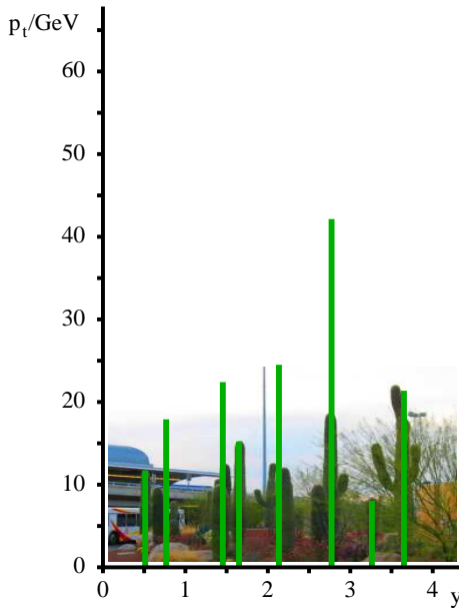


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



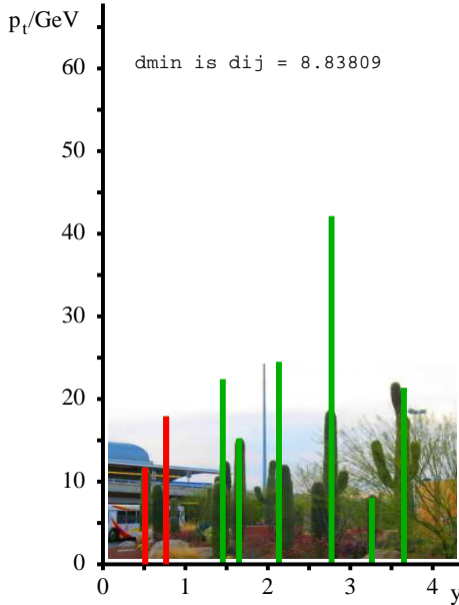
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

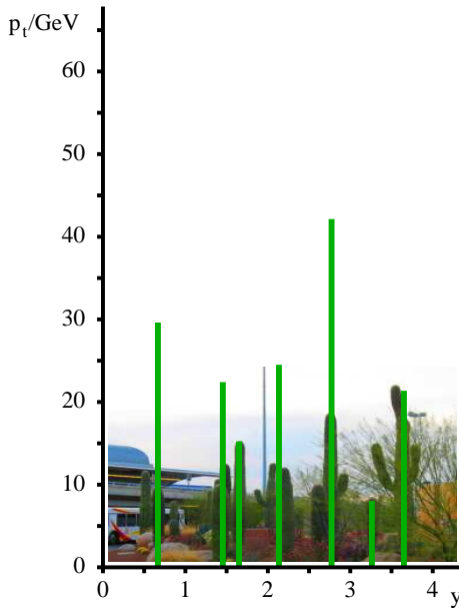


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



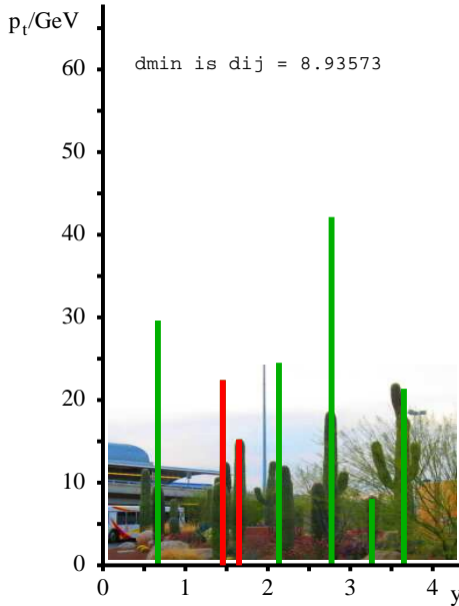
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)

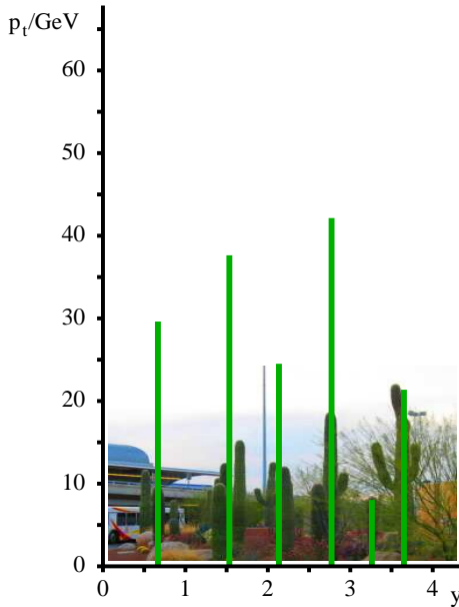


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

p_t/GeV

dmin is dij = 9.75598

60

50

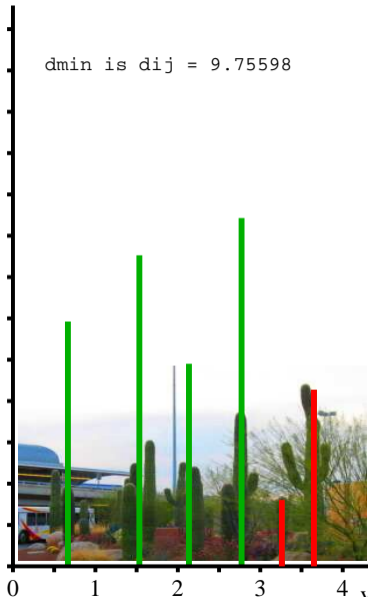
40

30

20

10

0



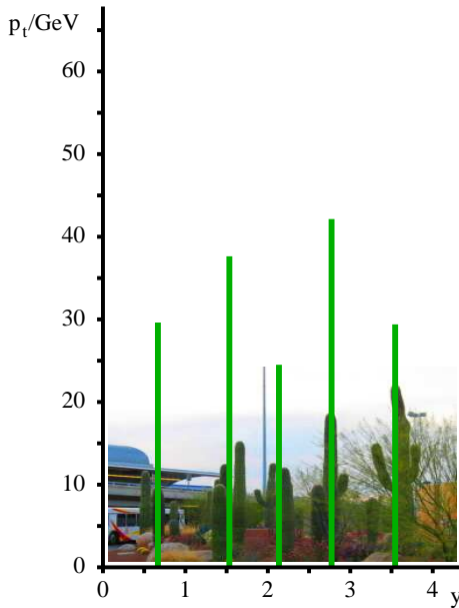
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)



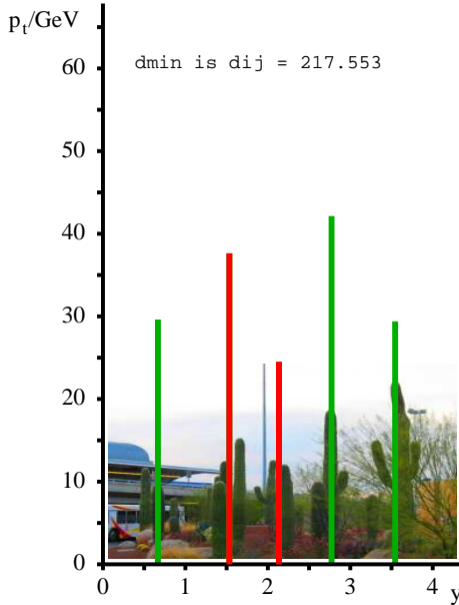
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

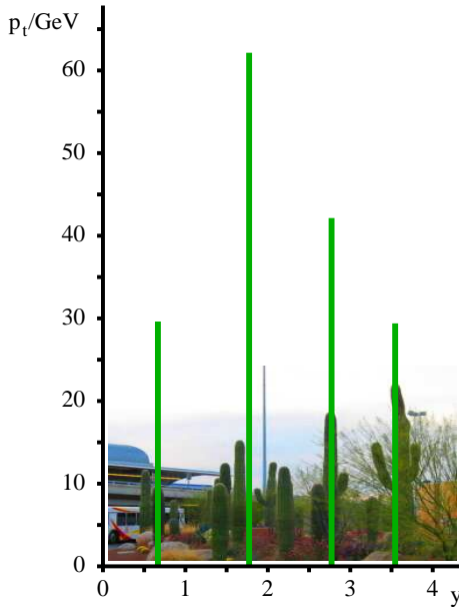


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



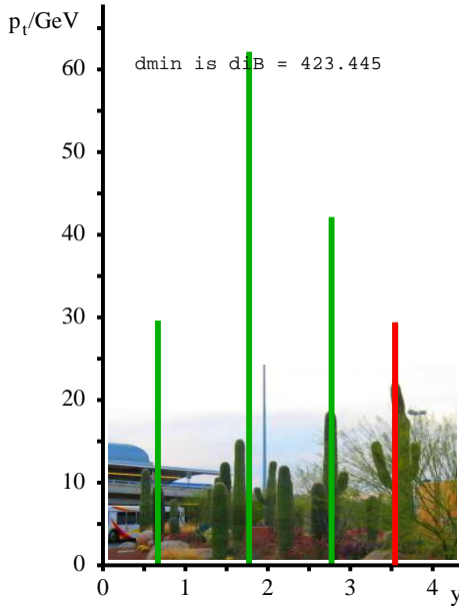
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

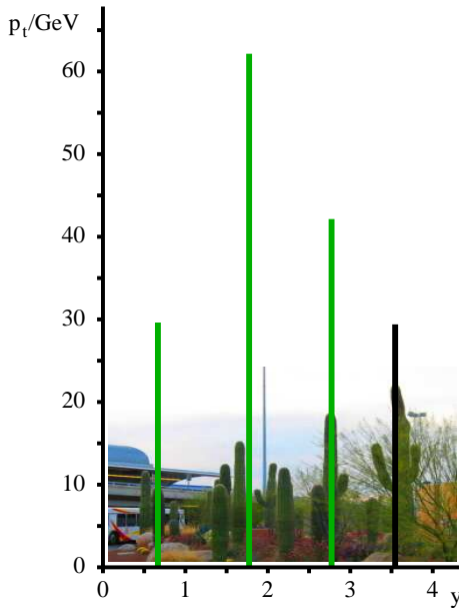


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



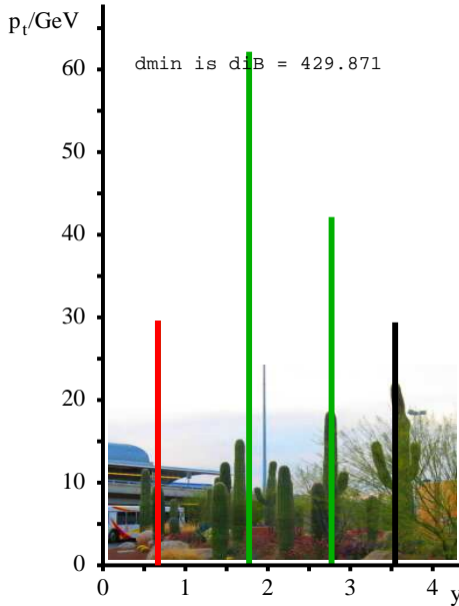
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)



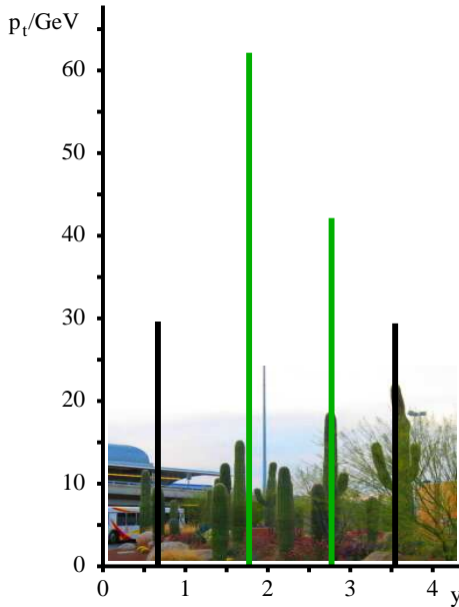
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)



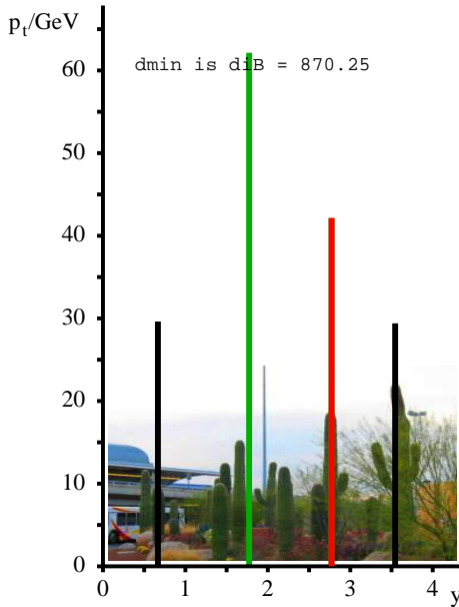
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)

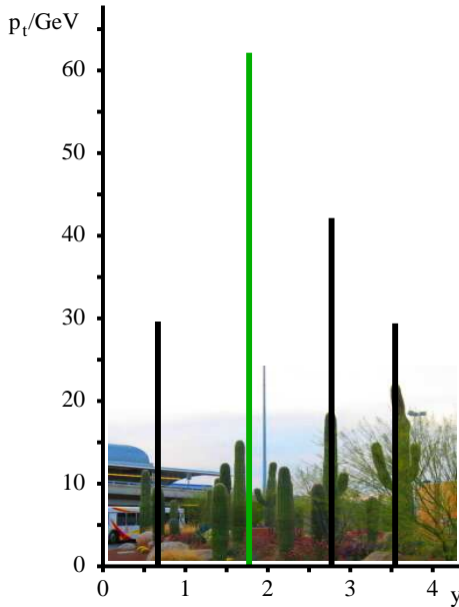


k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers



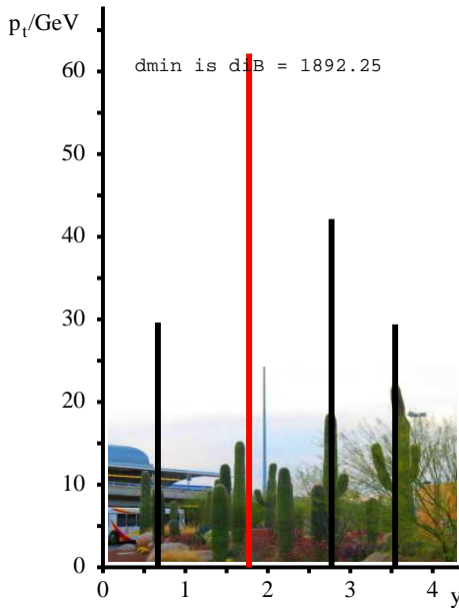
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)



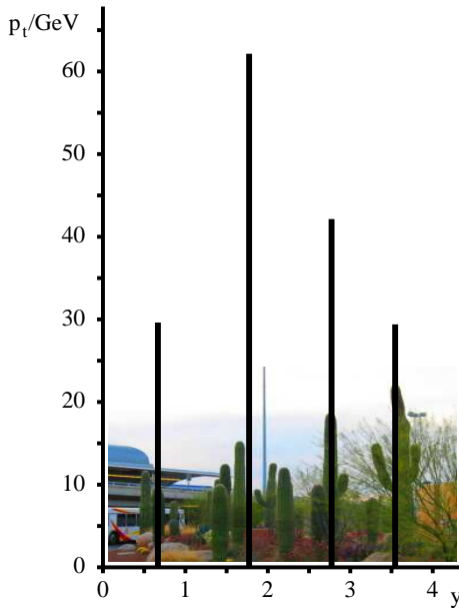
k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

(Animation courtesy of Gavin Salam)



k_t alg.: Find smallest of

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

If d_{ij} recombine; if d_{iB} , i is a jet
 Example clustering with k_t algorithm, $R = 0.7$

ϕ assumed 0 for all towers

y (Animation courtesy of Gavin Salam)

Jet Calibration



Jet Response



Jets are bundles of particles

- ⚡ ~25% initial photons
- ⚡ Hadronic particles include mostly pions, kaons, protons and their anti-particles
- 👤 Different response!



Jet response in non-compensating calorimeters

- ⚡ Jet signal depends on fragmentation/particle content
- 👤 Significant jet-to-jet response variations due to more or less photons

$$E_{true}^{jet} = E_{dep}^{calo} + E_{mag}^{loss} + E_{low}^{loss} + E_{leak}^{loss} + E_{out}^{loss} - E_{UE\otimes PU}^{gain} - E_{env}^{gain}$$

E_{dep}^{calo} energy deposited in the calorimeter within signal definition

E_{mag}^{loss} charged particle energy lost in solenoid field

E_{low}^{loss} particle energy lost in dead material

E_{leak}^{loss} energy lost due to longitudinal leakage

E_{out}^{loss} energy lost due to jet algorithm/calorimeter signal definition

$E_{UE\otimes PU}^{gain}$ energy added by underlying event and/or pile-up

E_{env}^{gain} energy added by response from other nearby particles/jets

↖ only source of signal!



Jet Calibration Strategies



Preliminaries

- ↓ There is no universal model for jet calibration
 - 👤 Immediate consequence from the fact that there is no universal jet finder (or jet finder configuration) appropriate for all physics reconstruction/analysis
 - 👤 Calibration approach depends on much more than detector signal characteristics and signal definitions
 - 👤 But there two general strategies
- ↓ Publications often refer to jets corrected to parton level
 - 👤 Ill-defined concept in pp (c.f. Gavin Salam)
 - 👤 Maybe more useful in e^+e^- or deep inelastic scattering (?)
- ↓ At LHC jets are foremost calibrated to the particle (hadron) level
 - 👤 First aim to reconstruct the energy carried by particles into the detector (calorimeter, of course)
 - 👤 Needs detailed and most accurate detector signal simulations for testbeams and physics processes
 - 👤 Link to interaction physics needs full modeling of collision processes
 - 👤 Need all particles, not only hard scatter fragments (don't really know which those are anyway)
- ↓ Factorize jet calibration as much as possible
 - 👤 Better control of systematics
 - 👤 Can even use hadron testbeams to a point (see later)



Most of all: every experiment needs its own model in the end!



Jet Calibration Strategies Overview



Two models:

↓ Model I: Calibration in jet context

👤 First find jet, then calibrate, then correct if needed

↓ Model II: Calibration in cluster context

👤 Calibrate calorimeter signals, then find jet, then correct (likely needed)

🌐 Local hadronic calibration plugs in here!



Best calibration likely a combination of both models

↓ Need to keep track of systematics!



Full Calibration in Jet Context



Find the jet using basic (electromagnetic) energy scale signals in the calorimeter

- ↓ Assumes that all elementary signal corrections (electronics etc.) are taken care of
- ↓ Relative mis-calibration between input to jet finder can O(30%) or more in non-compensating calorimeters
 - 🐇 Can be a problem especially for KT!
- ↓ Best for compensating calorimeters, as basic energy scale is ~hadronic scale



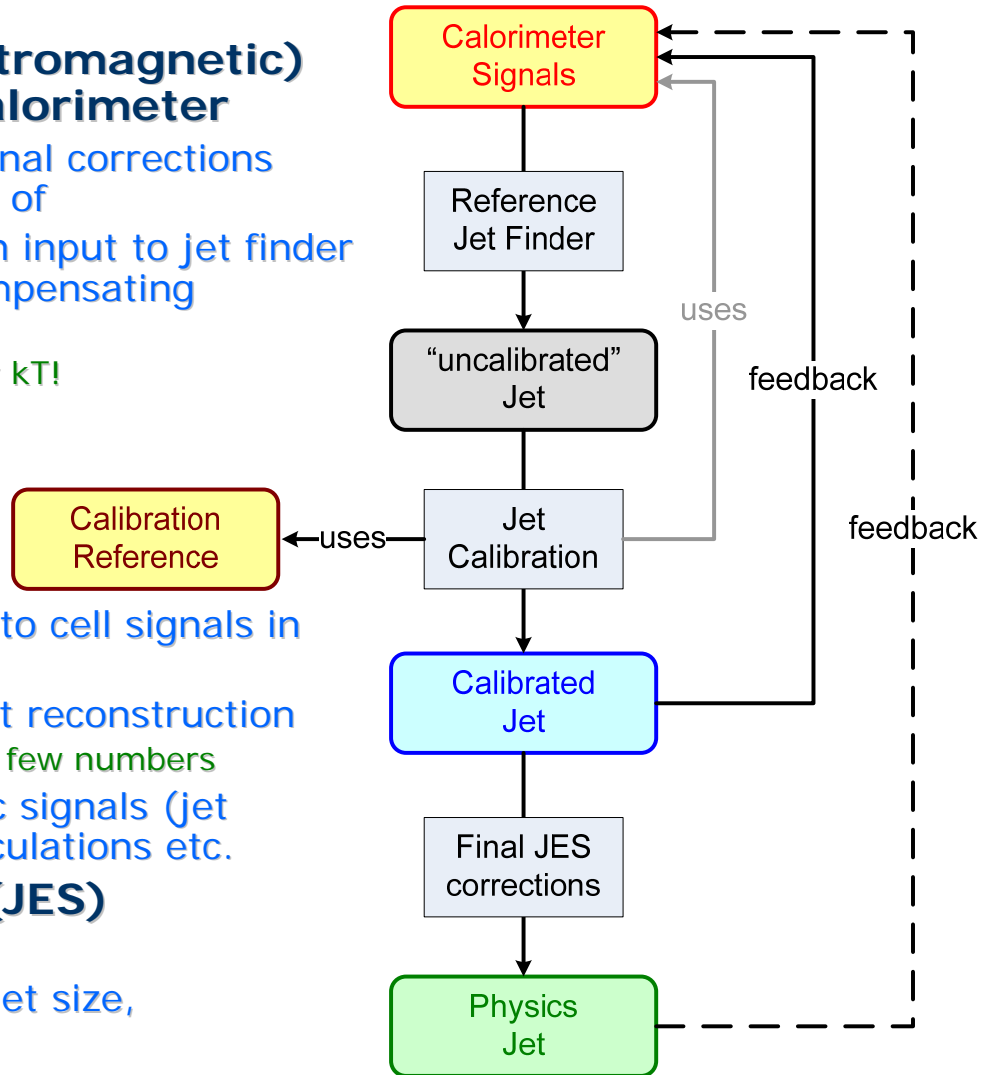
Then calibrate it

- ↓ Complex signal weights applied to cell signals in jet (default)
- ↓ Lower level of factorization of jet reconstruction
 - 🐇 Many corrections absorbed in a few numbers
- ↓ Feedback of calibrations to basic signals (jet constituents) for missing ET calculations etc.



Apply final Jet Energy Scale (JES) corrections

- ↓ Correct for different algorithm, jet size, calorimeter signal definition



Cell Calibration Functions From Jets



Sample of fully simulated QCD di-jet events from $p_T > 17$ GeV/c to ~ 4000 TeV/c

- Match reconstructed calorimeter jet with close-by particle jet
- Both jets reconstructed with Seeded Cone $R=0.7$
- Match only successful if only one jet close by
- Calorimeter jets are based on tower signals in a grid of $\Delta\eta \times \Delta\eta = 0.1 \times 0.1$



Determine cell signal weights (H1-style)

- De-compose matched calorimeter jet into individual cell signals
- Determine cell signal weights in resolution optimization fit using truth particle jet energy as normalization
 - Weights are function of cell location and cell signal density
 - Dense signals – em, less dense signals hadronic
- Re-calculate jet four-momentum using cell weights
 - Jet energy and direction change



Determine additional correction functions

- Different jet size, cluster jets, different jet algorithm (kT)



"H1" Style Cell Signal Weighting in ATLAS



Fit constraint for weights:

$$\frac{\partial}{\partial w} \sum_{j=1}^{N_{evts}} \left(\left(\sum_{i=1}^{N_{cells}} w(\rho_i, \vec{X}_i) E_i \right) - E_{truth}^{jet} \right)^2 = 0$$



Jet four-momentum calculation after fit

$$\left(E_{reco}^{jet}, \vec{p}_{reco}^{jet} \right) = \left(\sum_{i=1}^{N_{cells}} w(\rho_i, \vec{X}_i) (E_i, \vec{p}_i) \right) \times \underbrace{f(\eta_{jet}, p_t^{jet})}_{=1 \text{ for reference jets}}, \text{ with } E_i = |\vec{p}_i|$$



Cluster Context Jet Calibration



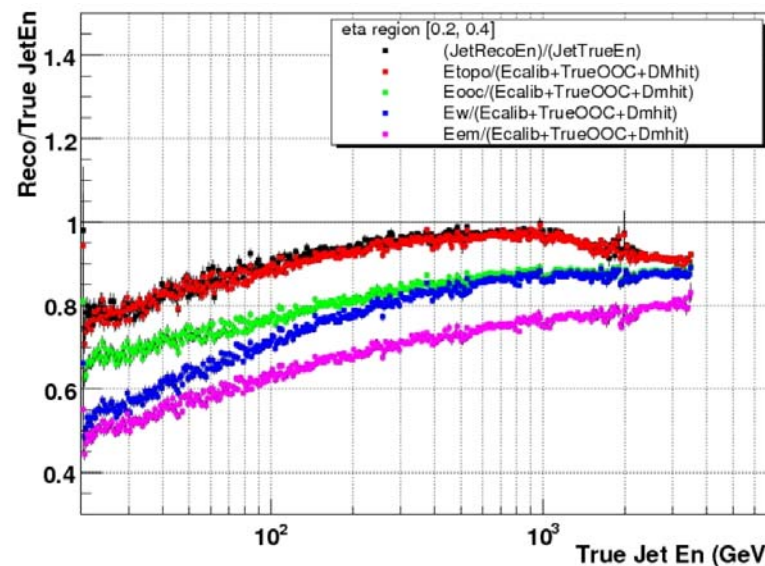
Calibrate calorimeter signals first as much as possible, then find jets

- ↓ **Detector motivated (use measured signal shapes)**
 - 👤 Applies calibration in the context of a specific calorimeter signal definition (topological clusters)
 - 👤 No jet context needed
 - 👤 Provides calibrated input to jet finding
 - 👤 Better for kT
 - ↓ **Needs final jet energy scale corrections**
 - 👤 Calibration derived from single particles
 - 👤 Feedback of final corrections for missing ET calculations etc.
 - ↓ **High level of factorization, better control of systematics (?)**
 - 👤 To be investigated



Provides hadronic calibration outside of jet context

- ↓ Previous discussion of local hadronic calibration



“Data Only” Jet Calibration

	Task	JES	Tool
1	PileUp Subtraction	$E_{bc}^{jet}(\eta_{jet}, \varphi_{jet}) = E_0^{jet}(\eta_{jet}, \varphi_{jet}) - \bar{\rho}_0^{mb}(N_{vtx}, \eta_{jet}, \varphi_{jet}) \cdot A_{\eta\varphi}^{jet}$	minbias events (determine E/Et density in pile-up as function of # vertices, next slide)
2	Relative response corrections (η, φ)	$E_{rel}^{jet} = \bar{f}(\eta_{jet}, \varphi_{jet}) \cdot E_{bc}^{jet}(\eta_{jet}, \varphi_{jet})$	di-jet pT balance (equalize jet response of calorimeter system with respect to central region in slices of φ , next slide)
3	Absolute energy scale corrections	$E_{rec}^{jet} = \bar{C}(p_{t,rel}^{jet}, \dots) \otimes E_{rel}^{jet}$	photon/jet pT balance in direct photon production (correct JES from pT balance with photon, as function of jet pT etc.)

Conceptional Features & Limitations

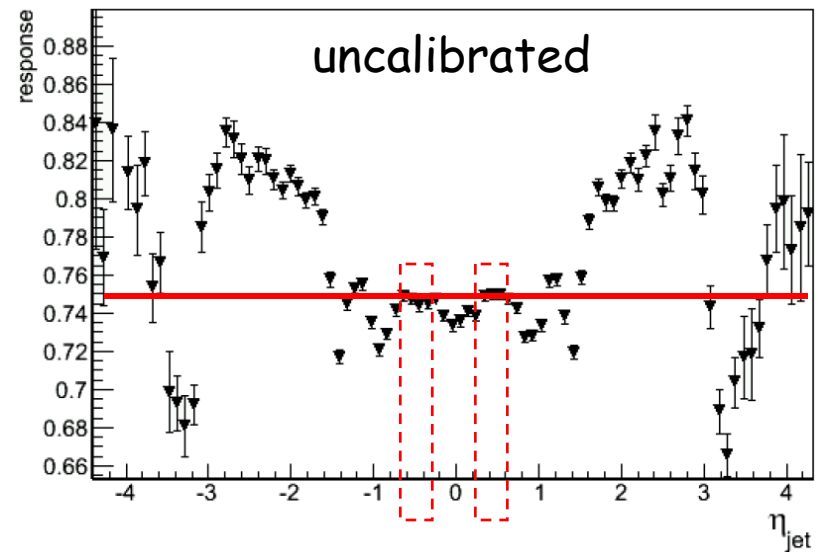
Data only approach from three different events

- ↓ MB for PileUp corrections
 - 👤 Underlying event is folded in
- ↓ Di-jets for response equalization in pseudorapidity and azimuth
 - 👤 Use central jet as reference
 - 👤 Back-to-back topologies needed
- ↓ Prompt photons for absolute JES corrections
 - 👤 Precision limit by photon reconstruction quality and di-jet background
 - 👤 Also theoretical uncertainties (ISR,FSR) important for very high precision
- ↓ Control of systematics seems to be possible
 - 👤 "independent" calibration and correction sources

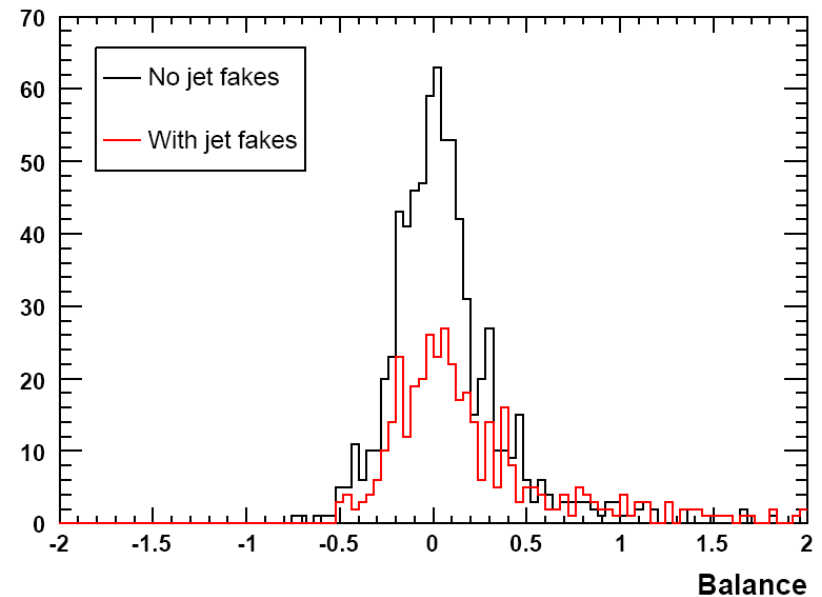
More is needed care is needed...

- ↓ E.g., mostly gluon jets di-jets versus mostly quark jets in prompt photon production
- ↓ Absolute correction is to interaction (parton) level, not particle level!

D. Schouten, M. Vetterli,
ATL-PHYS-INT-2007-011



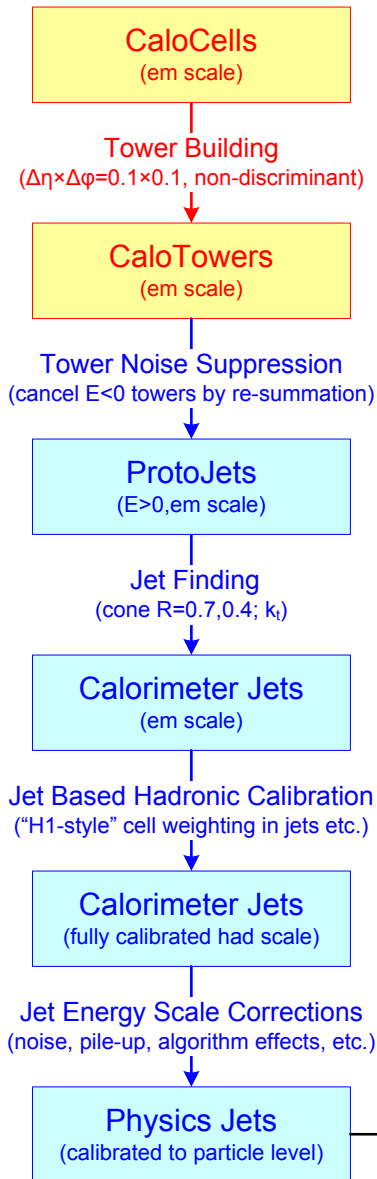
M. Hurwitz, FDR-1 User
Meeting April 1, 2008



Jet Reconstruction



Tower Jets in ATLAS



Sum up electromagnetic scale calorimeter cell signals into towers

- 🌱 Fixed grid of $\Delta\eta \times \Delta\phi = 0.1 \times 0.1$
- 🌱 Non-discriminatory, no cell suppression
- 🌱 Works well with pointing readout geometries
- 👤 Larger cells split their signal between towers according to the overlap area fraction

Tower noise suppression

- 🌱 Some towers have net negative signals
- 🌱 Apply "nearest neighbour tower recombination"
- 👤 Combine negative signal tower(s) with nearby positive signal towers until sum of signals > 0
- 👤 Remove towers with no nearby neighbours
- 🌱 Towers are "massless" pseudo-particles

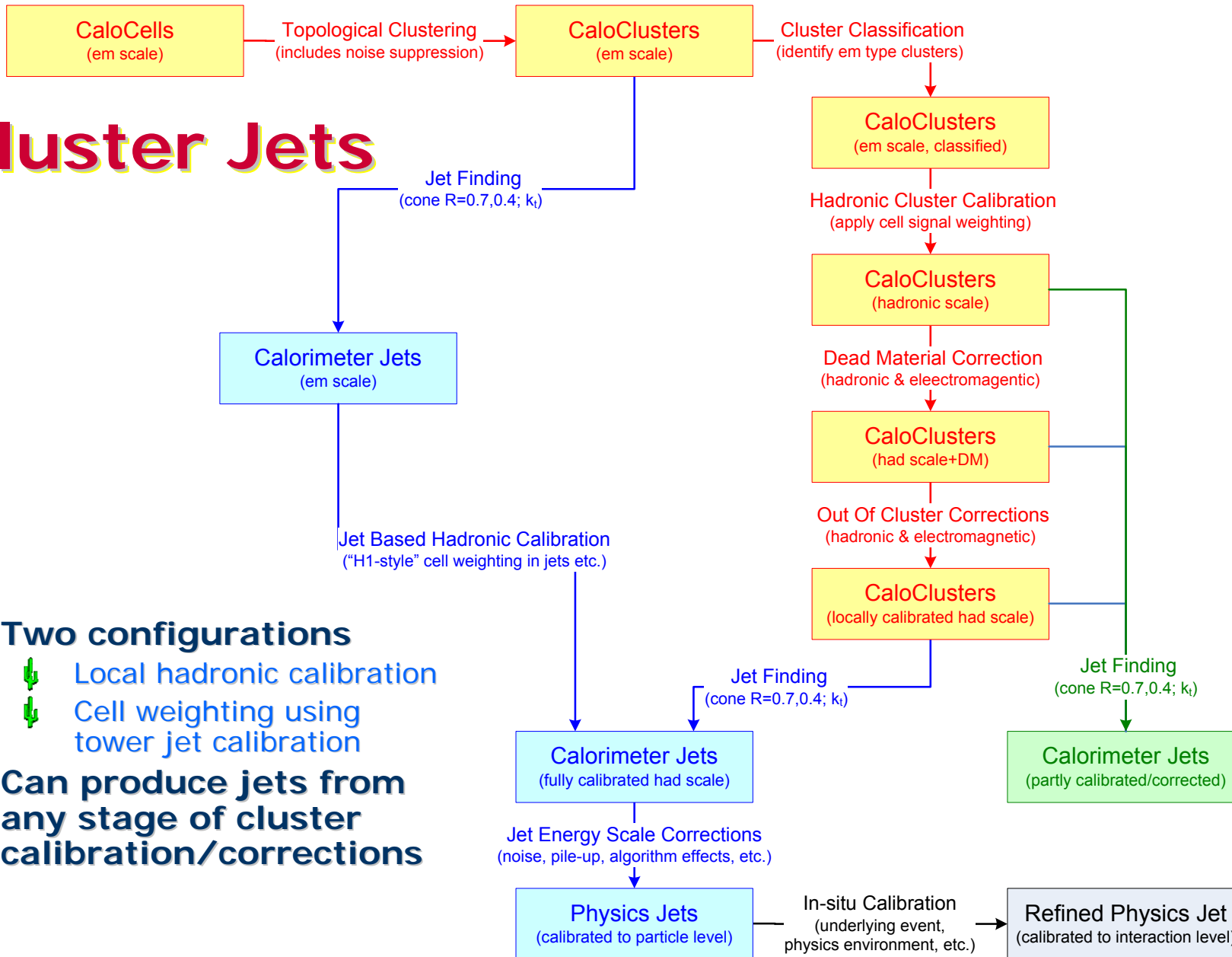
Find jets

- 🌱 Note: towers have signal on electromagnetic energy scale

Calibrate jets

- 🌱 Retrieve calorimeter cell signals in jet
- 🌱 Apply signal weighting functions to these signals
- 🌱 Recalculate jet kinematics using these cell signals
- 👤 Note: there are cells with negative signals!
- 🌱 Apply final corrections

Cluster Jets



Two configurations

- ↓ Local hadronic calibration
- ↓ Cell weighting using tower jet calibration

Can produce jets from any stage of cluster calibration/corrections



Deviations Signal Linearity

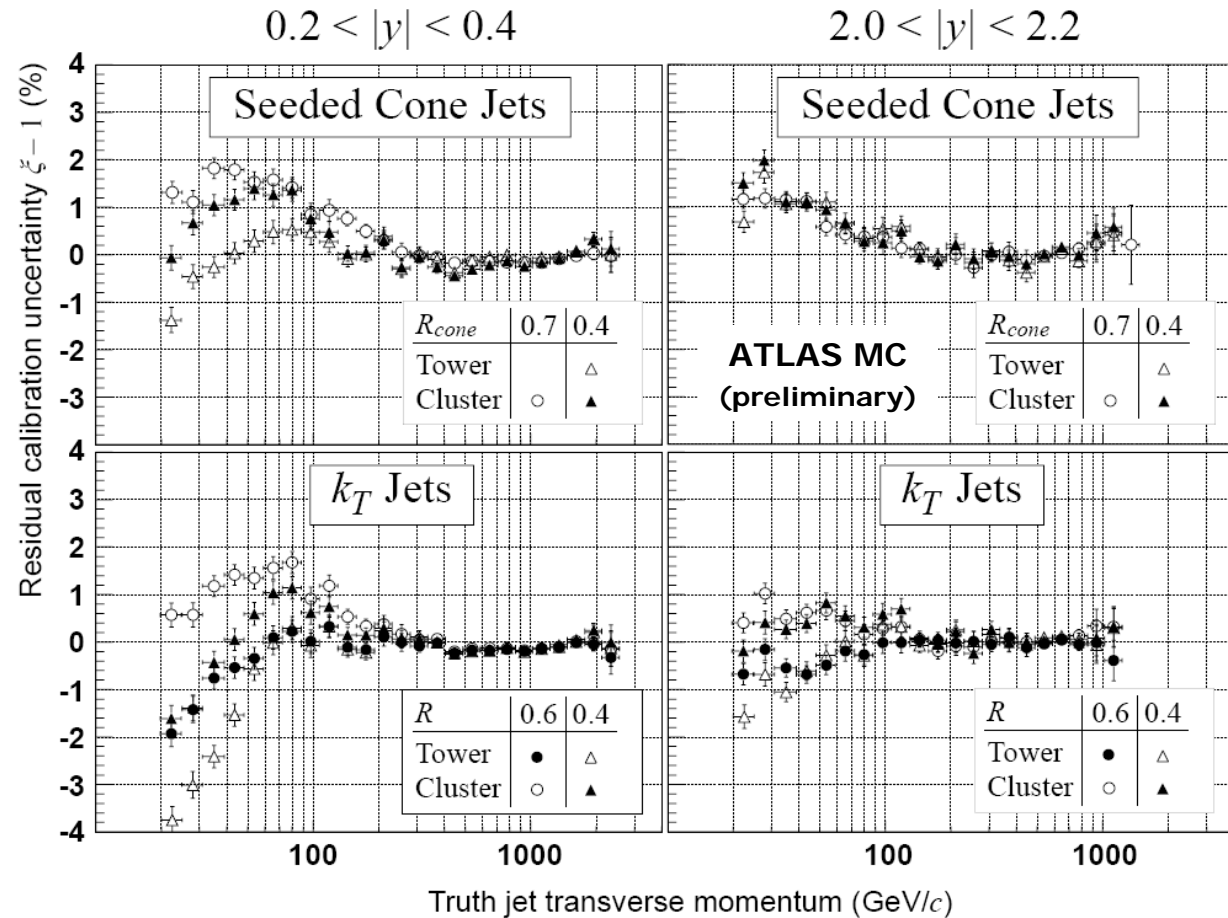


Estimated effect of a distorted detector

$\xi =$

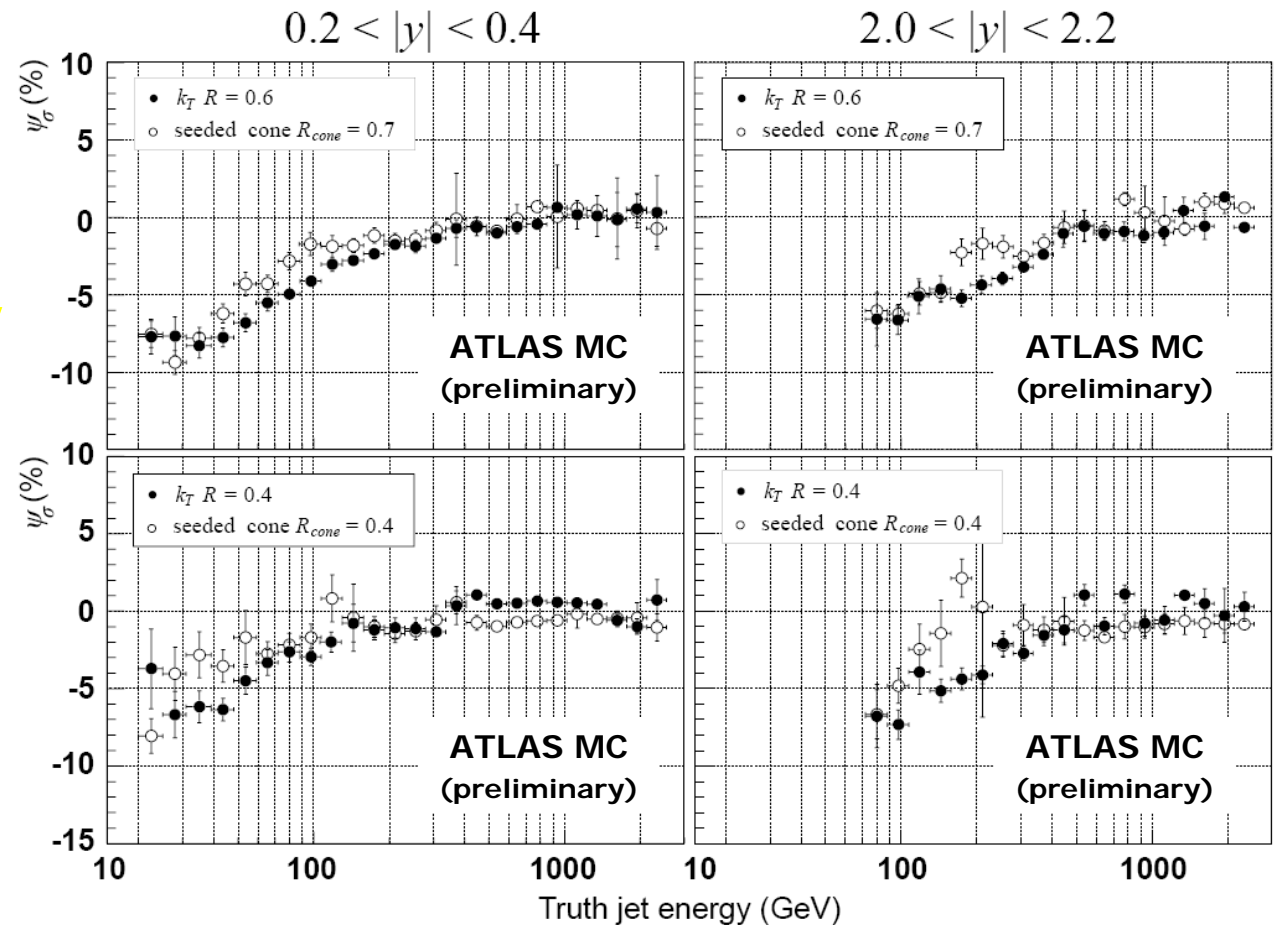
$$\frac{\left(E_{jet}^{calo} / E_{jet}^{truth} \right)_{alt}}{\left(E_{jet}^{calo} / E_{jet}^{truth} \right)_{ref}}$$

Effect of detector distortion depends on jet size, calo signal choice, and kinematic domain:
 ~ 2% for cone jets, up to ~4% for central (narrow) kT jets!



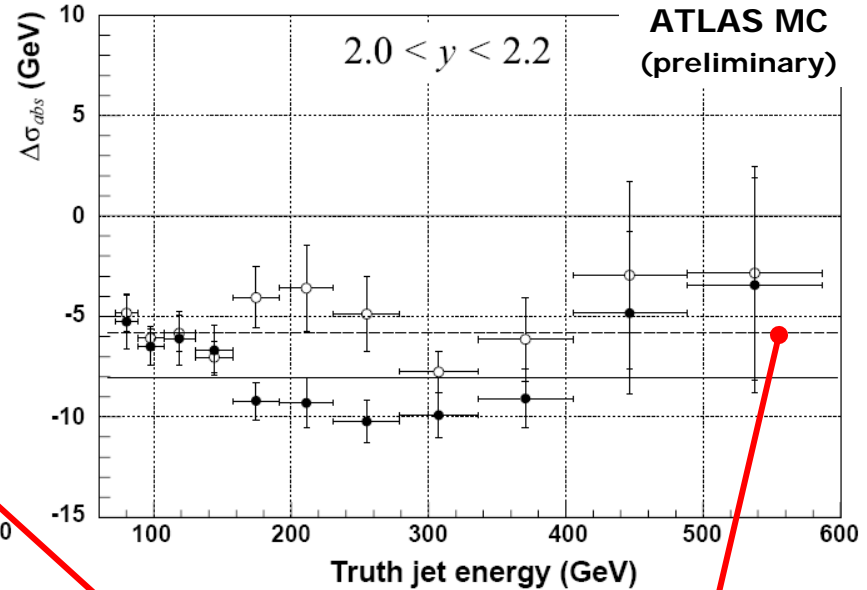
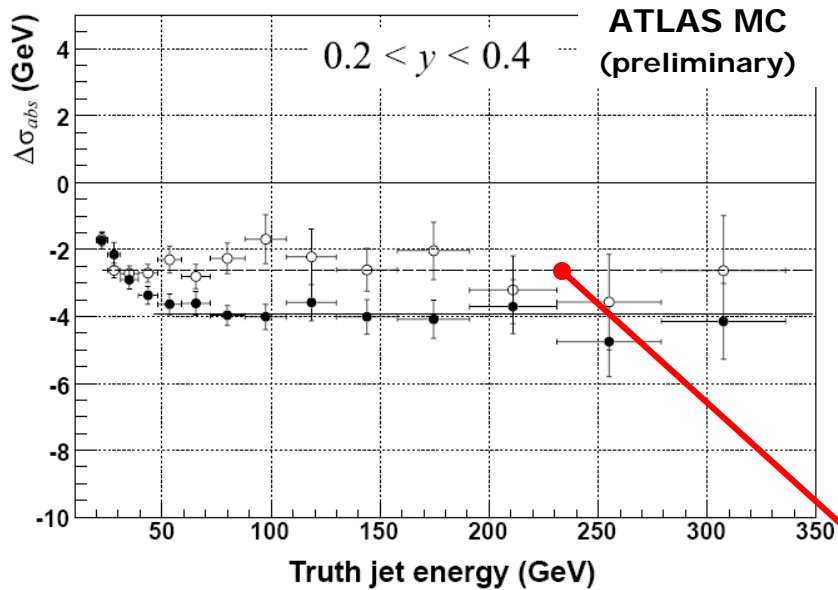
Effect of calorimeter signal choice on jet energy resolution

Cluster jets have better resolution at low energies even for non-optimal cell weights! -> noise?



$$\psi_\sigma = \begin{cases} \sqrt{\Delta\sigma_{rel}} & \Delta\sigma_{rel} > 0 \\ -\sqrt{-\Delta\sigma_{rel}} & \Delta\sigma_{rel} < 0 \end{cases}, \Delta\sigma_{rel} = \left(\frac{\sigma(E_{cluster})}{E_{cluster}} \right)^2 - \left(\frac{\sigma(E_{tower})}{E_{tower}} \right)^2$$

Energy Resolution Differences



open symbols: cone, filled symbols: kT

$$\Delta\sigma_{abs} = \psi_{\sigma} \cdot E_{jet}$$

Clear indication of noise suppression by TopoClusters!

