

Jet clustering tools

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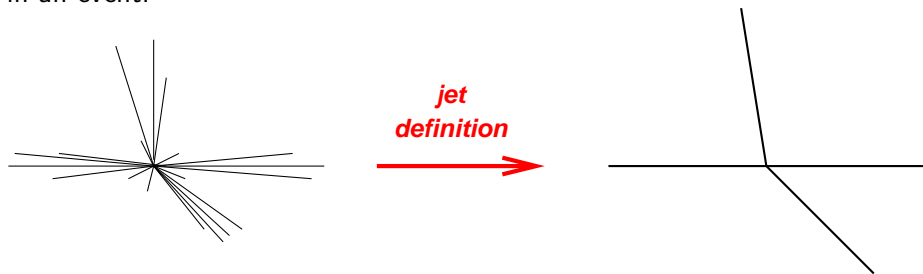
MC4LHC readiness workshop

31 March 2010

CERN

Concentrating mostly on FastJet, developed with
Matteo Cacciari and Gregory Soyez

A jet definition is a systematic procedure that **projects away the multiparticle dynamics**, so as to leave a simple picture of what happened in an event:

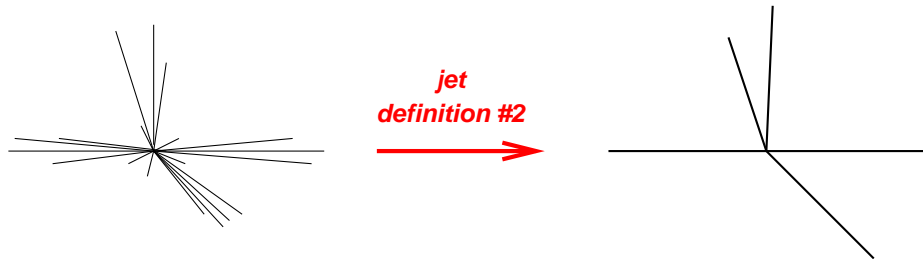


Jets are *as close as we can get to a physical single hard quark or gluon*: with good definitions their properties (multiplicity, energies, [flavour]) are

- ▶ finite at any order of perturbation theory
- ▶ insensitive to the parton \rightarrow hadron transition

NB: finiteness \longleftrightarrow set of jets depends on jet def.

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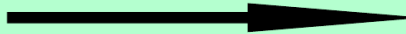
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jet definition

 $\{P_i\}$

particles,
4-momenta,
calorimeter towers, ...

jet algorithm

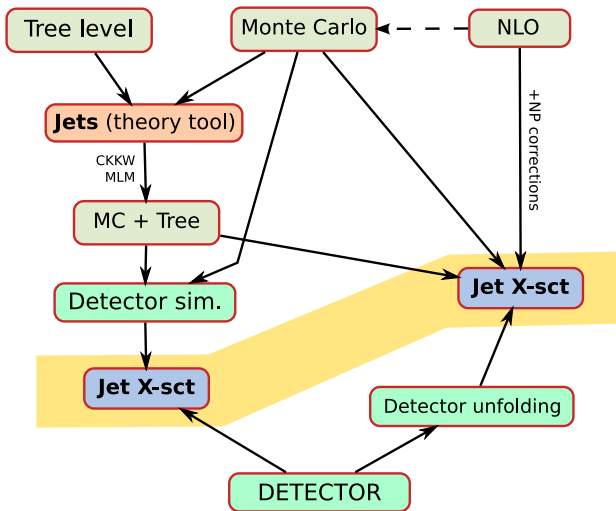
 $\{j_k\}$

jets

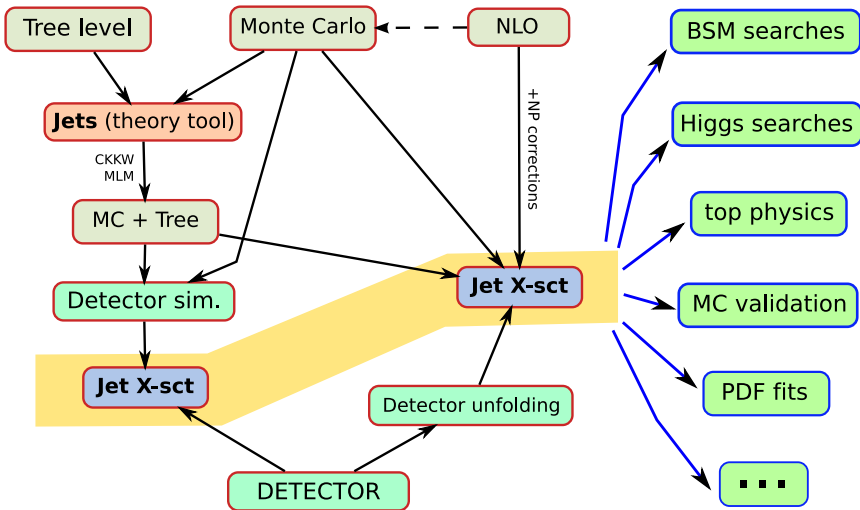
+ parameters (usually at least the radius R)

+ recombination scheme

Reminder: running a jet definition gives a well defined physical observable,
which we can measure and, hopefully, calculate



Jet (definitions) provide central link between expt., "theory" and theory
 And jets are an input to almost all analyses



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And jets are an input to almost all analyses

In rough order of first public release:

- ▶ KTCLUS (Fortran)
- ▶ ARCLUS (Fortran)
- ▶ PxCone (Fortran)
- ▶ KTJet (C++)
- ▶ Optimal Jet Finder [OJF] (Fortran)
- ▶ FastJet + plugins (C++)
- ▶ CDF MidPoint and JetClu codes (C++)
- ▶ SpartyJet (C++)
- ▶ FFTJet (C++)

Also: jet finders in non-jet tools: simple cone jet finders in Pythia, Isajet, Alpgen, PGS, AcerDet, ...; k_t and seedless/midpoint cones in MCFM, NLOJet++, etc.

FastJet

<http://fastjet.fr/>

Jan 2006 (FJ 1.0):

- ▶ Fast implementation of pp k_t algorithm Cacciari & GPS
 N^2 and $N \ln N$ timings for clustering N particles v. N^3 with earlier codes
 $N \ln N$ strategy relies on external package CGAL

Oct 2006 (FJ 2.0):

- ▶ Implementation of Cambridge/Aachen algorithm
including coding of Chan's Closest Pair algorithm
- ▶ Introduction of jet areas and background estimation/subtraction
- ▶ New interface for long-term stability

Apr 2007 (FJ 2.1):

- ▶ Plugin mechanism giving common interface to external jet finders
- ▶ Inclusion of plugins that wrap CDF (JetClu, Midpoint) code and PxCone
- ▶ Inclusion of SIScone as a plugin

Jan 2008 (FJ 2.3):

Soyez joined development team

- ▶ Added the anti- k_t algorithm (fast, native implementation)
- ▶ Added “passive” and “Voronoi” areas
- ▶ Switched to autotools for compilation/installation
- ▶ Better access to information for subset studies
- ▶ Basic Fortran wrapper

April 2009 (FJ 2.4):

- ▶ Added plugins for DØRunII Cone, ATLAS cone, CMS cone, TrackJet
DØ and Trackjet code contributed by Sonnenschein
ATLAS code taken from SpartyJet
- ▶ Added gen- $k_t + e^+e^-$ algorithms (k_t , Cambridge, Jade, e^+e^- anti- k_t)
- ▶ Framework for handling user-supplied clustering distances (NNH)

Native implementations:

- ▶ longitudinally invariant kt
- ▶ (inclusive) Cambridge/Aachen
- ▶ anti-kt
- ▶ gen-kt
- ▶ e^+e^- kt and gen-kt

Plugins (distributed with FastJet)

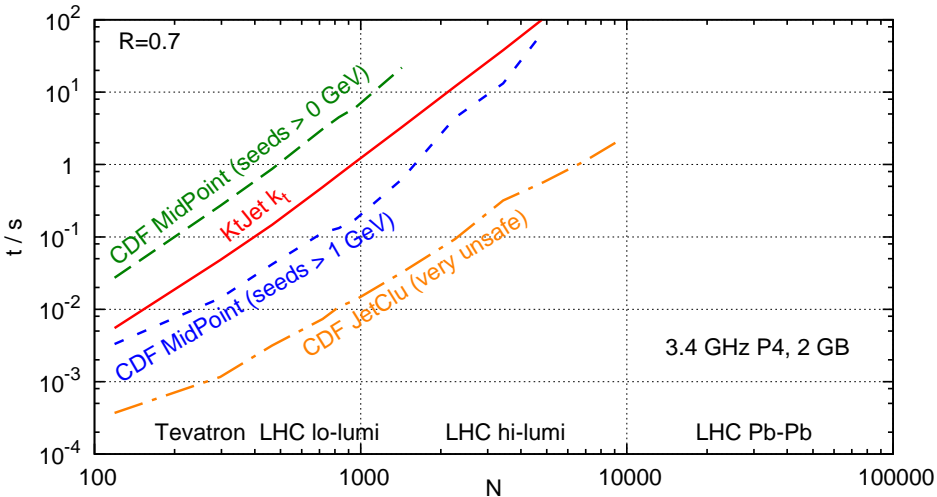
- ▶ SISCone
- ▶ CDF MidPoint [IR₃₊₁ unsafe]
- ▶ CDF JetClu [IR₂₊₁ unsafe]
- ▶ D0 Run II Cone [IR₃₊₁ unsafe]
- ▶ ATLAS Cone algorithm [IR₂₊₁ unsafe]
- ▶ CMS Cone algorithm [Coll₃₊₁ unsafe]
- ▶ TrackJet [Coll₃₊₁ unsafe]
- ▶ PxCone (fortran 77) [IR₃₊₁ unsafe]
- ▶ e^+e^- (spherical) SISCone
- ▶ e^+e^- JADE algorithm
- ▶ e^+e^- Cambridge algorithm

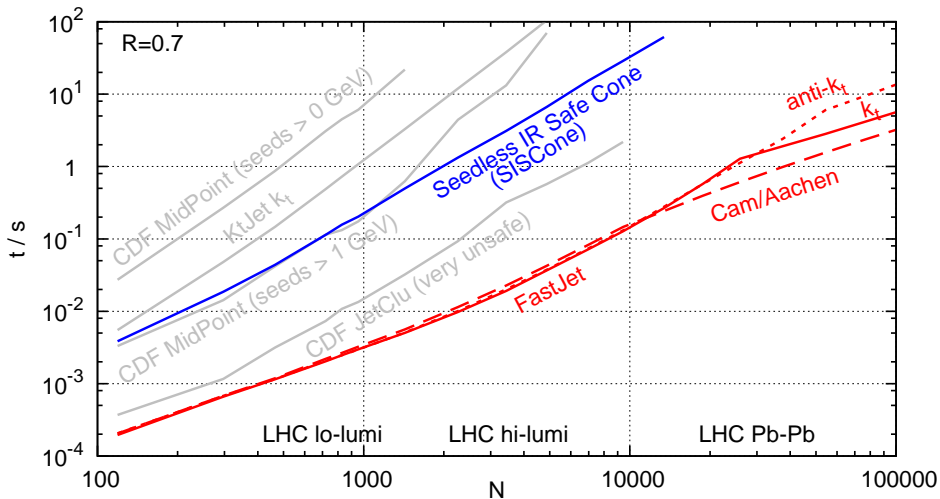
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Plugins (distributed with FastJet)

- ▶ SIScone
- ▶ CDF MidPoint [IR_{3+1} unsafe]
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Soft stuff clusters with nearest neighbour

$$k_t: d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 \longrightarrow \text{anti-}k_t: d_{ij} = \frac{\Delta R_{ij}^2}{\max(k_{ti}^2, k_{tj}^2)}$$

Hard stuff clusters with nearest neighbour
 Privilege collinear divergence over soft divergence
 Cacciari, GPS & Soyez '08

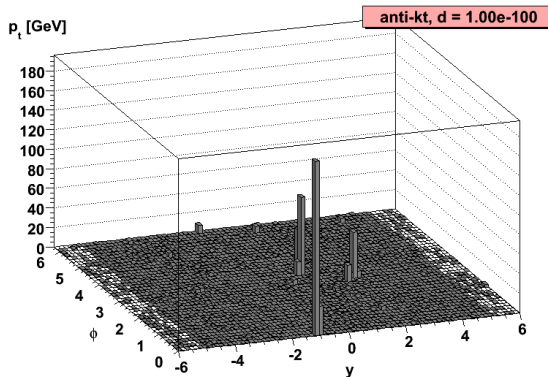
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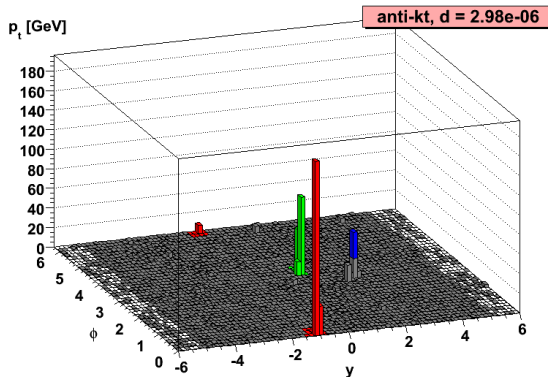
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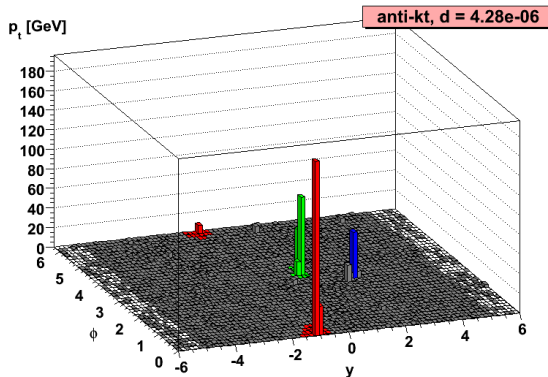
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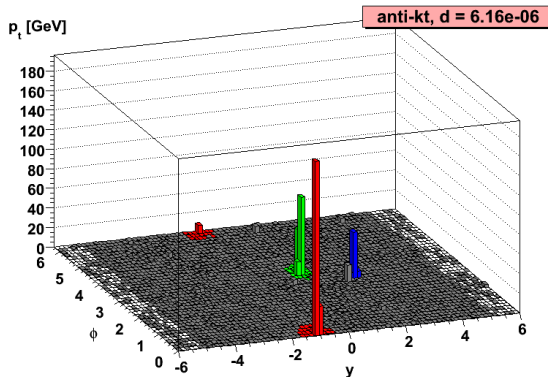
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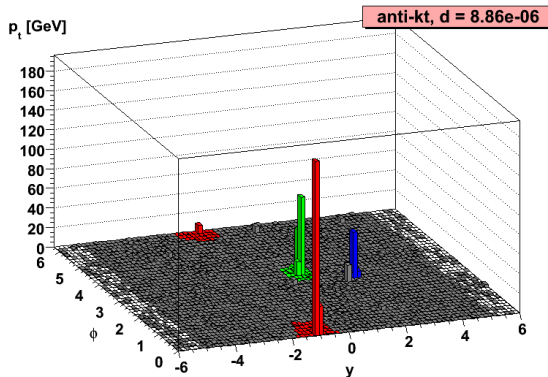
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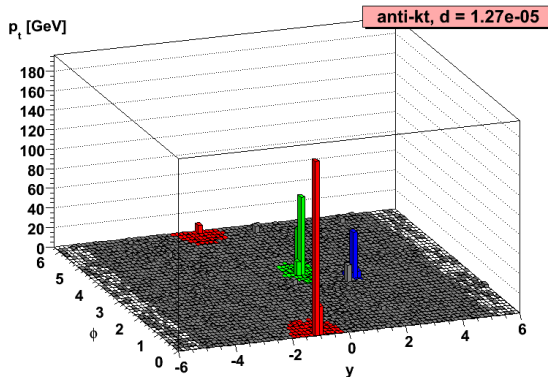
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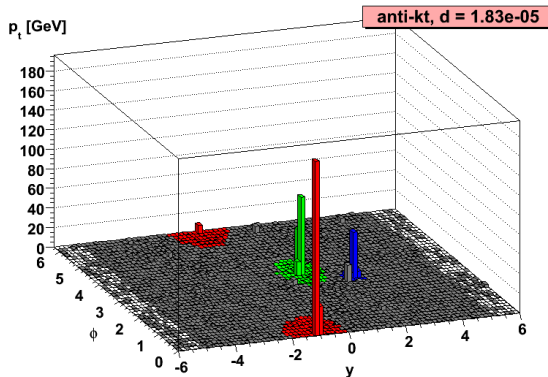
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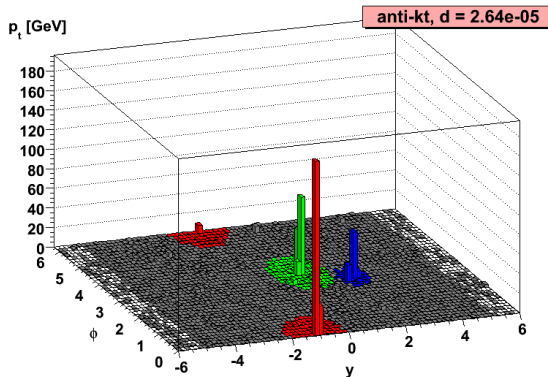
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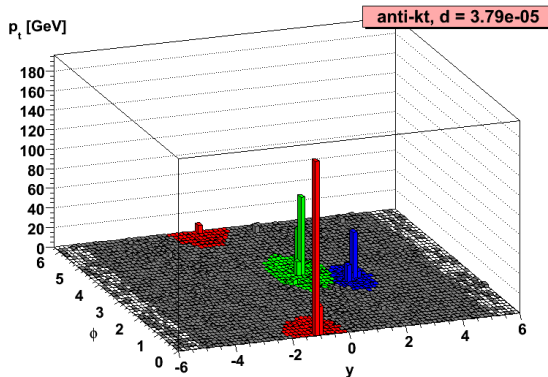
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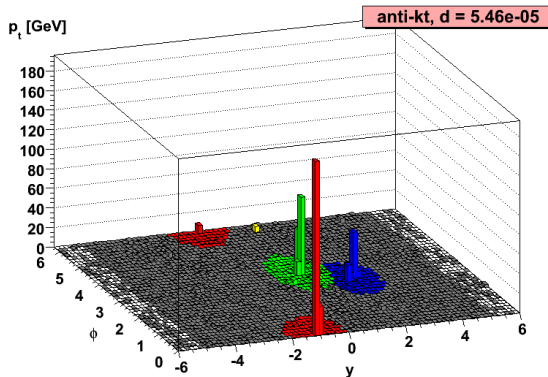
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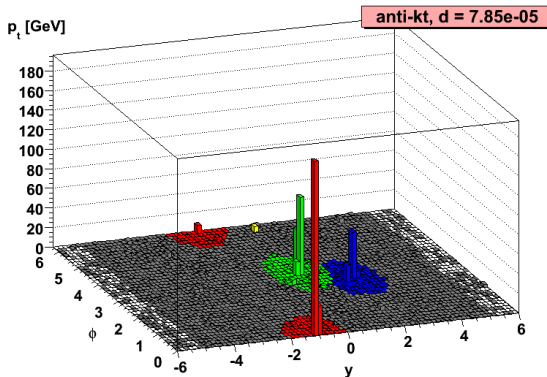
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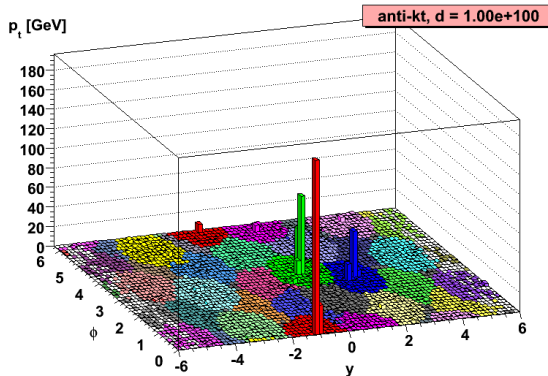
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anti- k_t gives
cone-like jets
without using stable
cones

```
#include "fastjet/ClusterSequence.hh"
using namespace fastjet;
using namespace std;

int main () {
    // choose a jet definition
    double R = 0.7;
    JetDefinition jet_def(kt_algorithm, R);

    vector<PseudoJet> particles;
    // build event with 2 particles:   px py pz   E
    particles.push_back( PseudoJet( 100.0, 0, 0, 100.0) );
    particles.push_back( PseudoJet(-100.0, 0, 0, 100.0) );

    // run the clustering, extract the jets
    ClusterSequence cs(particles, jet_def);
    vector<PseudoJet> jets = cs.inclusive_jets();
}
```

```
#include "fastjet/ClusterSequence.hh"
using namespace fastjet;
using namespace std;

int main () {
    // choose a jet definition
    double R = 0.7, f = 0.75;
    JetDefinition jet_def = new SISconePlugin(R, f);

    vector<PseudoJet> particles;
    // build event with 2 particles:   px py pz   E
    particles.push_back( PseudoJet( 100.0, 0, 0, 100.0) );
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```

Jets “ecosystem”

Individuals

- ▶ Anyone needing simple jet finding need stable, simple interface
- ▶ People playing with new jet ideas need flexible interface
- ▶ Theorists who still like Fortran

Community-wide projects

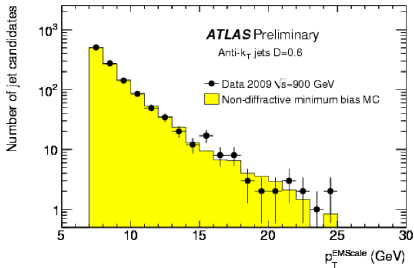
- ▶ Rivet One of the drivers for inclusion of “legacy” jet algorithms
- ▶ Delphes detector simulation

Experiments

- ▶ The four main LHC experiments all use FastJet
- ▶ ATLAS and CMS have chosen anti- k_t as the first jet alg. to calibrate
- ▶ ATLAS uses FastJet in the high-level trigger It had better not crash!

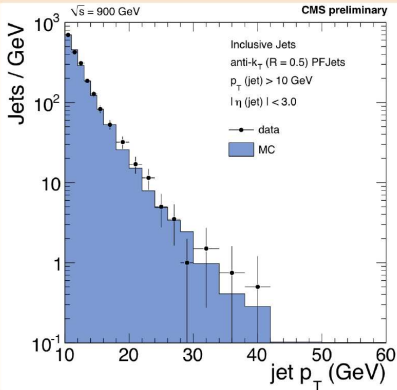
ATLAS

Jets from AntiK $_T$ D=0.6 algorithm
 $|\eta| < 2.6$ and $p_T > 7 \text{ GeV}$



CMS

Particle Flow



External plugins for FastJet:

- ▶ Variable R plugin
- ▶ Pruning plugin
- ▶ Trimming plugin

(not included in release)

Krohn, Thaler & Wang '09

Ellis, Vermillion & Walsh '09

Krohn, Thaler & Wang '10

Algorithms not naturally “pluggable” into FastJet:

FastJet designed for algorithms for which each particle ends up in at most 1 jet. Not all algorithms fit this picture:

- ▶ ARCLUS ($3 \rightarrow 2$ clustering)
- ▶ OJF (a particle has weighted assignment to multiple jets)
- ▶ FFTJet, in its “fuzzy” mode (weighted assignment)

SpartyJet [↗](#)

Delsart, Geerlings, Huston & Martin '06-

- ▶ Provides root interface to FastJet, including PyRoot access
- ▶ Provides visualisation tools
- ▶ Also has a number of native implementations of jet algs

FastJet Tools page [↗](#)

- ▶ A range of boosted-particle finders (Higgs, top, etc.)
Our own, links to other people's, and our implementations of other people's
- ▶ Background (UE/pileup) estimation and subtraction tools
Already in FJ, more flexible versions in the works
- ▶ Filtering
cleanup of UE/pileup noise to improve resolution
[Butterworth, Davison, Rubin & GPS '08]
[“trimming” is closely related]

Physics Roadmap:

Questions include

- a) Developing (analytical) understanding of different uses of jets
- b) Designing better analyses as a result

What follows is an illustration

What R is best for an isolated jet?

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

PT radiation:

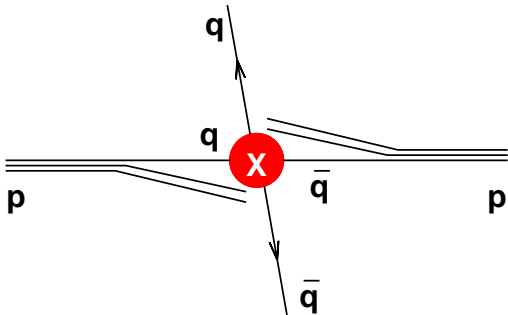
$$q : \langle \Delta p_t \rangle \simeq \frac{\alpha_s C_F}{\pi} p_t \ln R$$

Hadronisation:

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

Underlying event:

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



Minimise fluctuations in p_t

Use crude approximation:

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

in small- R limit (!)
 cf. Dasgupta, Magnea & GPS '07

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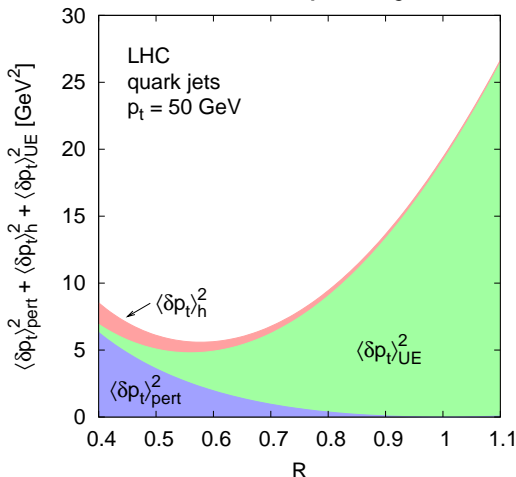
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50 GeV quark jet



in small- R limit (!)

cf. Dasgupta, Magnea & GPS '07

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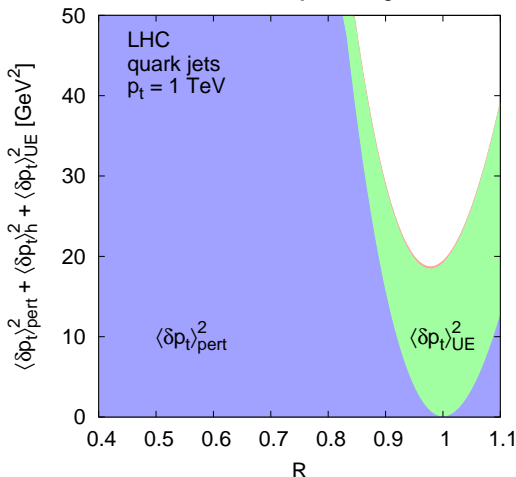
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1 TeV quark jet



in small- R limit (!)

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What R is best for an isolated jet?

PT radiation:

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

Hadronization:

$q :$

At high p_t , perturbative effects dominate over non-perturbative $\rightarrow R_{best} \sim 1$.

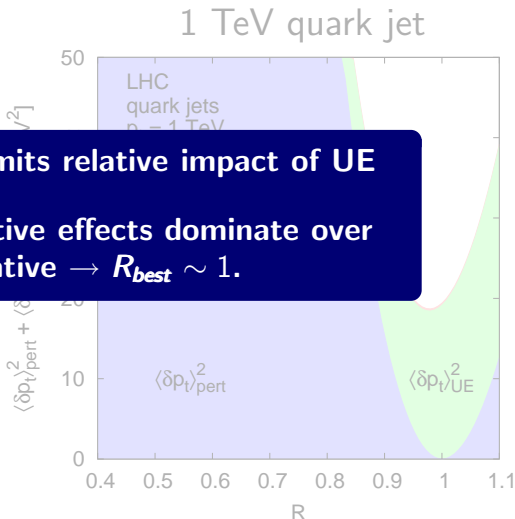
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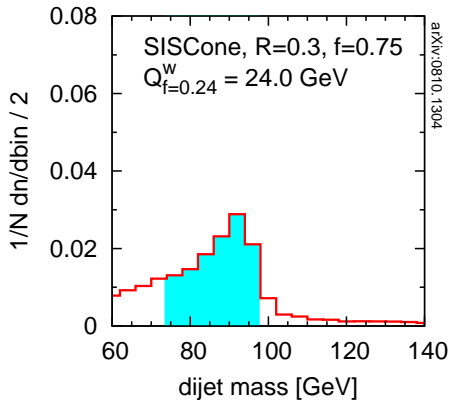


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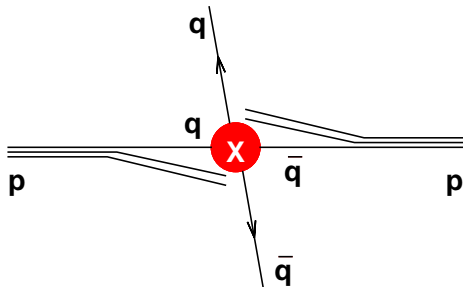
cf. Dasgupta, Magnea & GPS '07

$R = 0.3$

$qq, M = 100 \text{ GeV}$

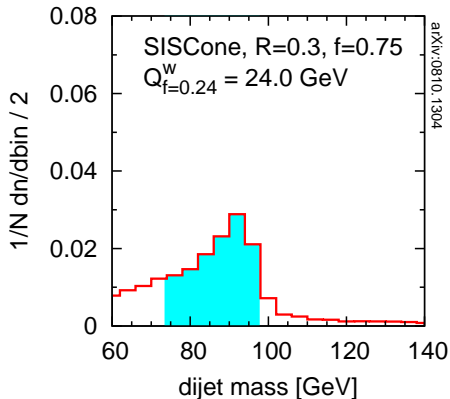


Resonance X \rightarrow dijets

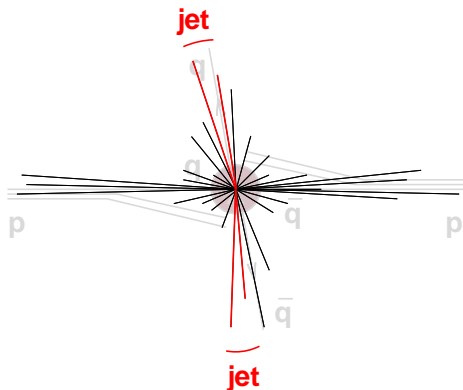


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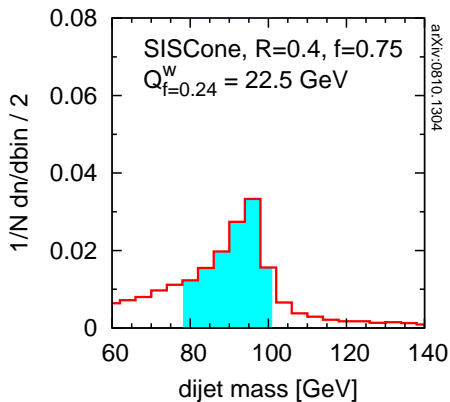


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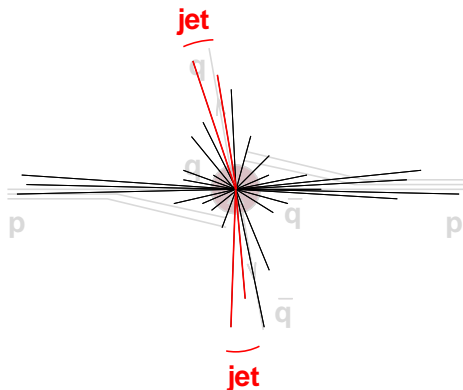


$R = 0.4$

qq, $M = 100$ GeV

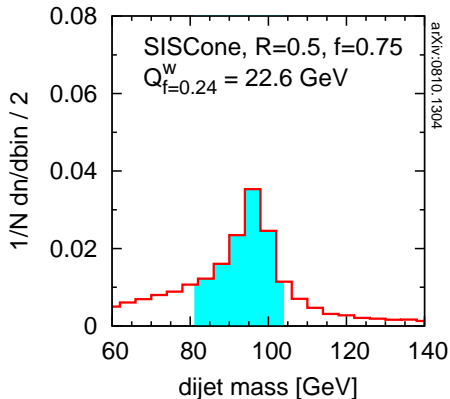


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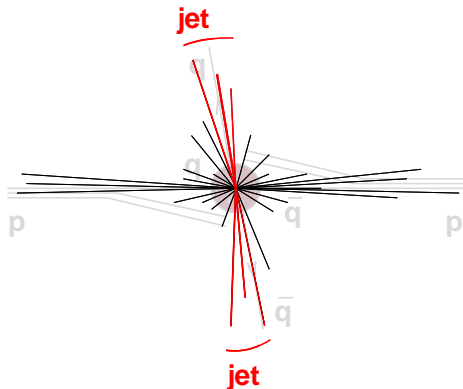


$R = 0.5$

qq, $M = 100$ GeV

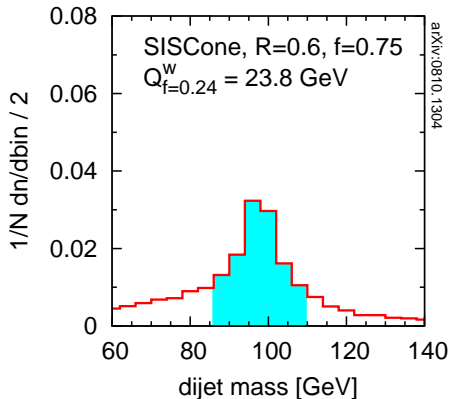


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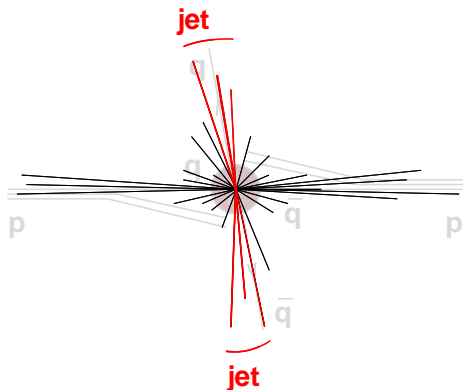


$R = 0.6$

qq, $M = 100$ GeV

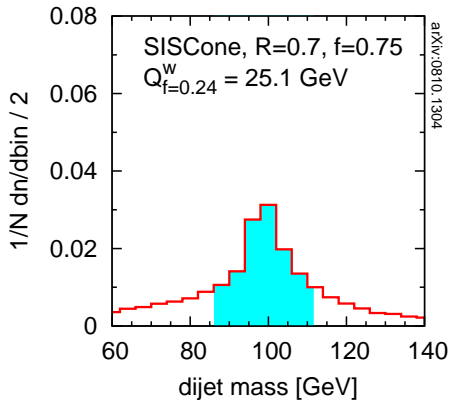


Resonance X \rightarrow dijets

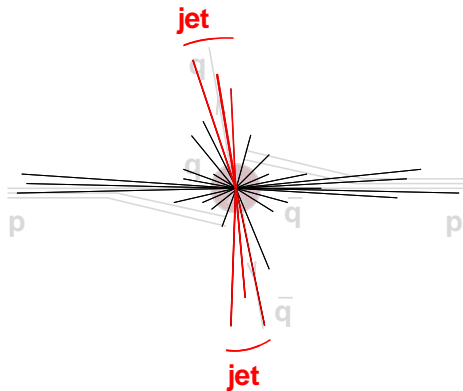


$R = 0.7$

qq, $M = 100$ GeV

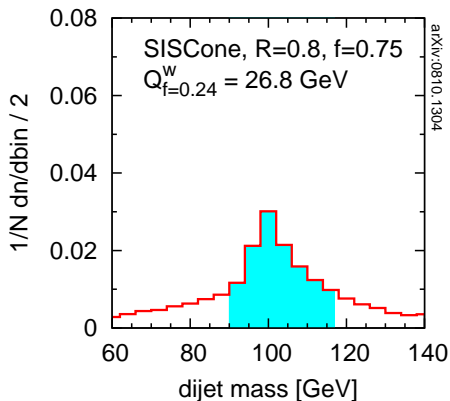


Resonance X \rightarrow dijets

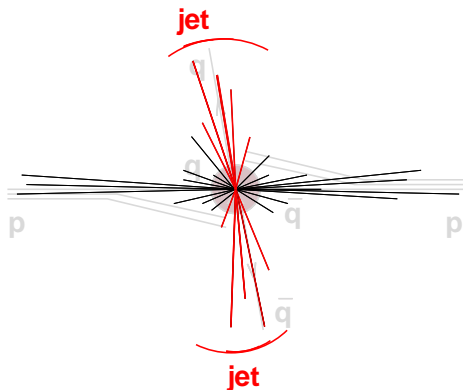


$R = 0.8$

qq, $M = 100$ GeV

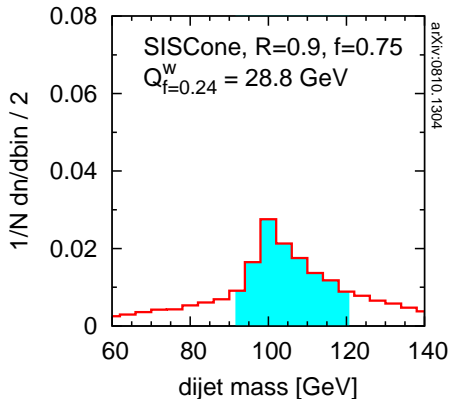


Resonance X \rightarrow dijets

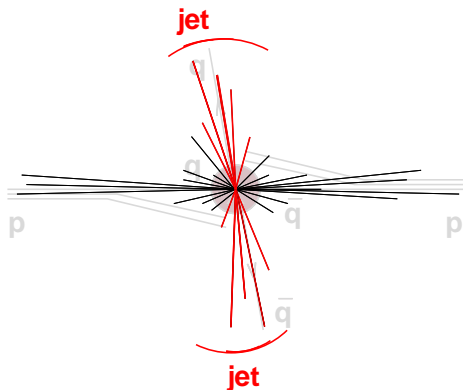


$R = 0.9$

qq, $M = 100$ GeV

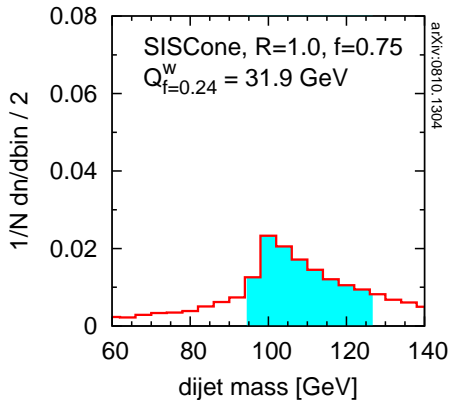


Resonance X \rightarrow dijets

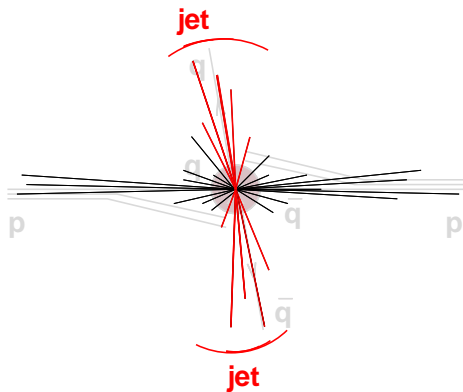


$R = 1.0$

qq, $M = 100$ GeV

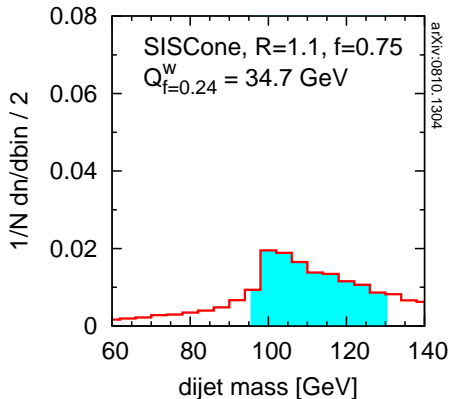


Resonance X \rightarrow dijets

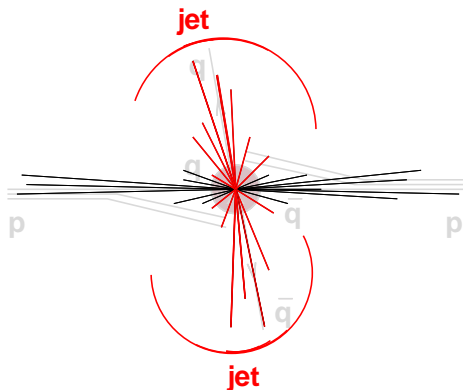


$R = 1.1$

qq, $M = 100$ GeV

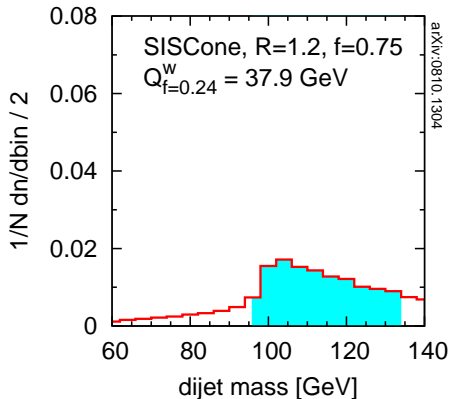


Resonance X \rightarrow dijets

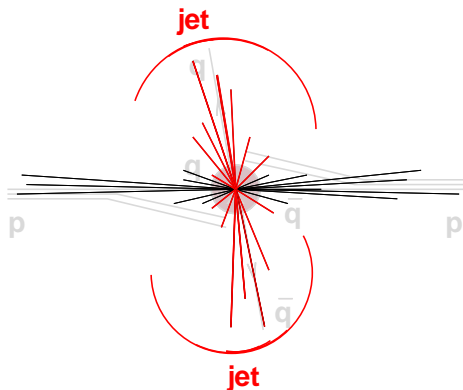


$R = 1.2$

qq, $M = 100$ GeV

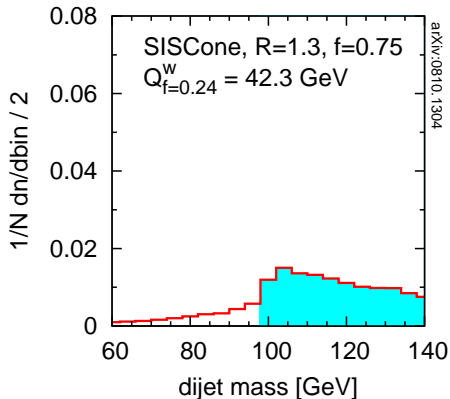


Resonance X \rightarrow dijets

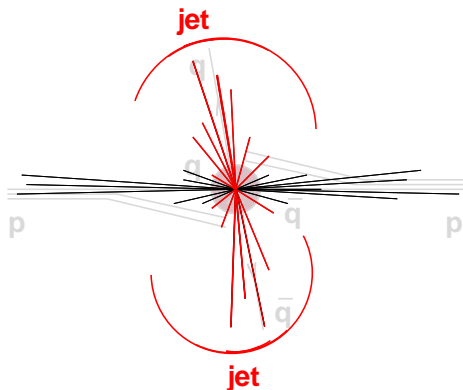


$R = 1.3$

qq, $M = 100$ GeV

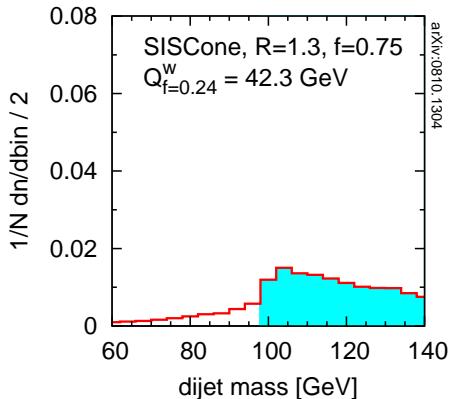


Resonance X \rightarrow dijets

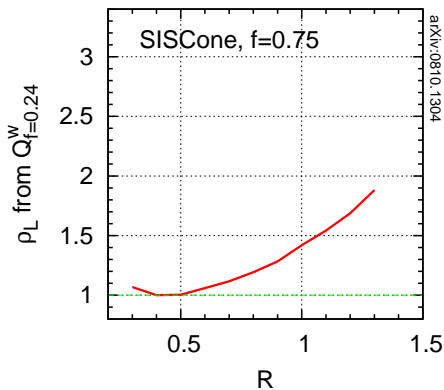


$R = 1.3$

qq, $M = 100$ GeV



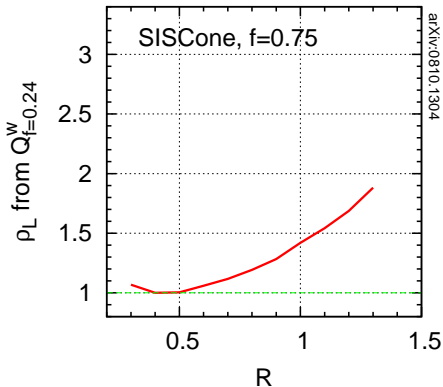
qq, $M = 100$ GeV



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

$m_{qq} = 100 \text{ GeV}$

$qq, M = 100 \text{ GeV}$



Best R is at minimum of curve

- Best R depends strongly on mass of system
- Increases with mass, just like crude analytical prediction
- NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

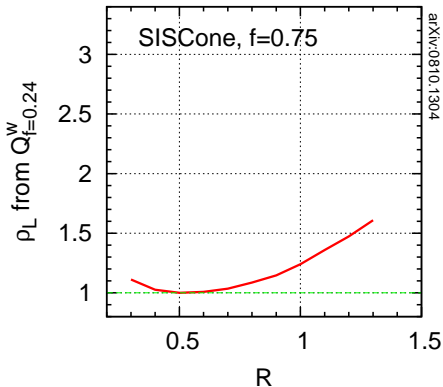
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$m_{qq} = 150 \text{ GeV}$

$qq, M = 150 \text{ GeV}$



Best R is at minimum of curve

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BUT: so far, LHC's plans involve running with fixed smallish R values

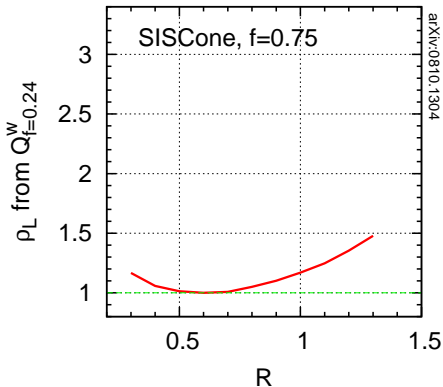
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$m_{qq} = 200 \text{ GeV}$

$qq, M = 200 \text{ GeV}$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
- NB: current analytics too crude

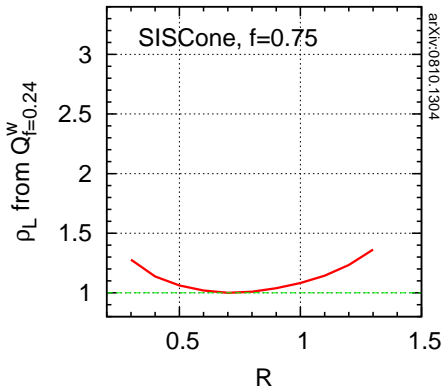
BUT: so far, LHC's plans involve running with fixed smallish R values

e.g. CMS arXiv:0807.4961

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

$$m_{qq} = 300 \text{ GeV}$$

$$qq, M = 300 \text{ GeV}$$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
- NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

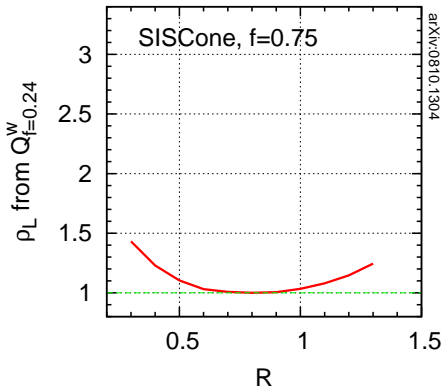
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$$m_{q\bar{q}} = 500 \text{ GeV}$$

$$q\bar{q}, M = 500 \text{ GeV}$$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
- NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

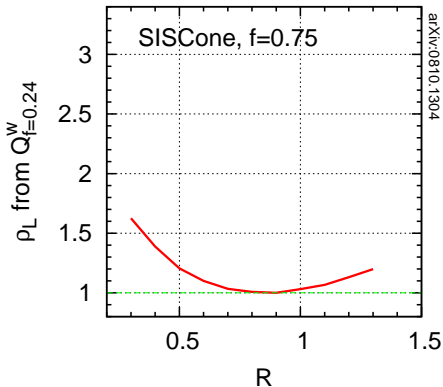
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$$m_{qq} = 700 \text{ GeV}$$

$$qq, M = 700 \text{ GeV}$$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
 - NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

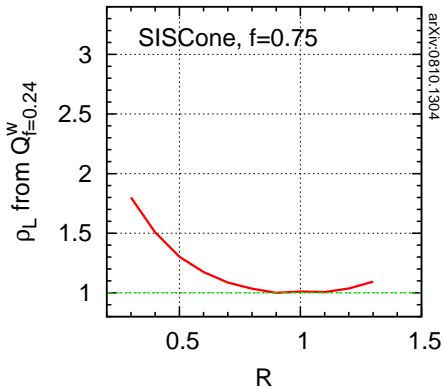
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$m_{q\bar{q}} = 1000 \text{ GeV}$

$q\bar{q}, M = 1000 \text{ GeV}$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
 - ▶ Increases with mass, just like crude analytical prediction
- NB: current analytics too crude

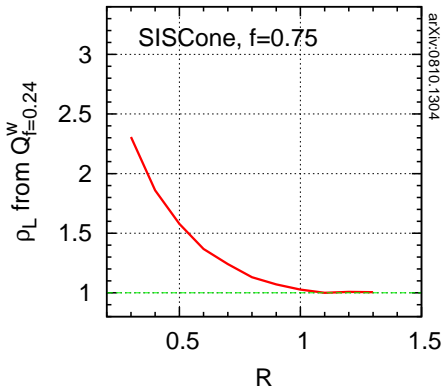
BUT: so far, LHC's plans involve running with fixed smallish R values

e.g. CMS arXiv:0807.4961

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

$m_{q\bar{q}} = 2000 \text{ GeV}$

$q\bar{q}, M = 2000 \text{ GeV}$



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
 - NB: current analytics too crude

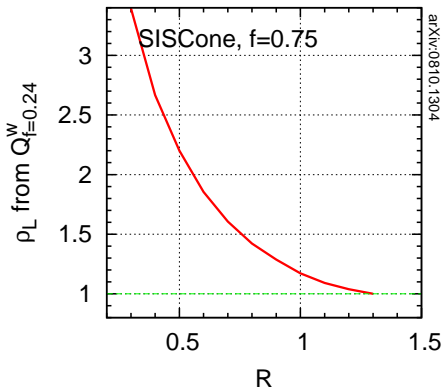
BUT: so far, LHC's plans involve running with fixed smallish R values

e.g. CMS arXiv:0807.4961

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

$m_{q\bar{q}} = 4000 \text{ GeV}$

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



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction

NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

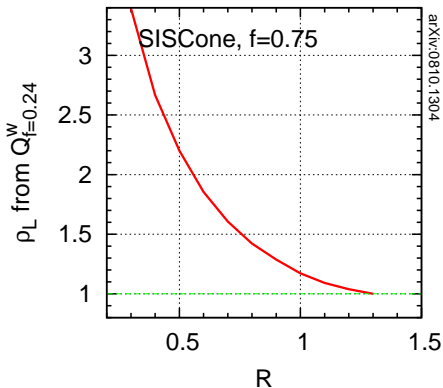
e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

$m_{q\bar{q}} = 4000 \text{ GeV}$

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



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction

NB: current analytics too crude

BUT: so far, LHC's plans involve running with fixed smallish R values

e.g. CMS arXiv:0807.4961

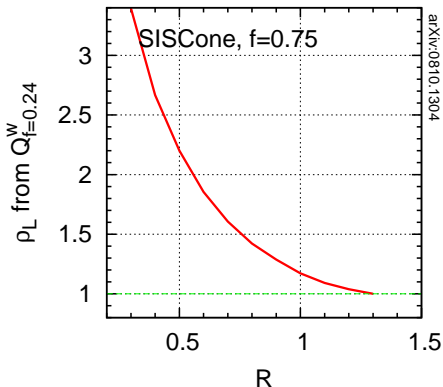
NB: 100,000 plots for various jet algorithms, narrow $q\bar{q}$ and $g\bar{g}$ resonances

from <http://quality.fastjet.fr>

Cacciari, Rojo, GPS & Soyez '08

$m_{q\bar{q}} = 4000 \text{ GeV}$

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



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
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NB: current analytics too crude

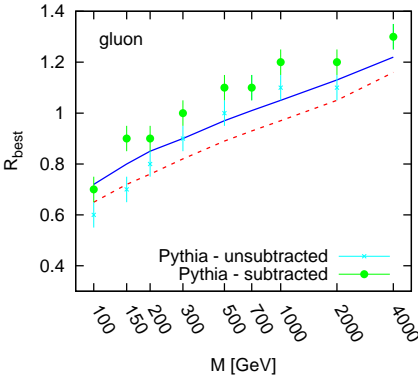
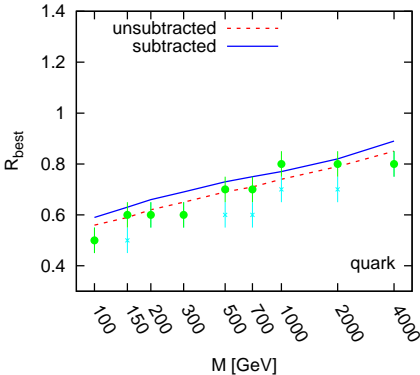
BUT: so far, LHC's plans involve running with fixed smallish R values

e.g. CMS arXiv:0807.4961

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

Cacciari, Rojo, GPS & Soyez '08

Medium-term aim: have ability for FastJet to suggest near-optimal parameter choices for different classes of analysis. Can be based on MC study, or analytical calculations:



Soyez, preliminary (LHC 14 TeV)

- ▶ Search strategies with jets in complex events
 - ▶ Boosted objects [Several groups working]
 - ▶ Non-boosted objects
- ▶ Further work on noise reduction (UE/pileup removal)
- ▶ Further understanding of UE/pileup characterisation
 - e.g. Cacciari, GPS & Sapeta '10
- ▶ Jets in heavy-ion collisions

Software Roadmap

Long-term maintainability

Facilitation of “advanced” usage

The most frequently used core code (N2Tiled strategy) was written in the space of a couple of days in 2005.

Not quite spaghetti code (C-style macaroni code?)

It could do with a cleanup

An important part of maintainability is validation:

- ▶ We currently check compilation and clustering results for 1000 MC events for all algorithms every night on several systems.
We would like to switch to 10k events
- ▶ Other aspects of FastJet (e.g. jet areas) not yet subject to automatic validation, but that should probably change.

We aim to maintain backwards compatibility for extended periods of time (allow 2-3 years from “deprecation” to “removal” of any feature).

Apparently “small” user-interface additions. E.g. from

```
vector<PseudoJet> constituents = cluster_sequence->constituents(jet)
```

to

```
vector<PseudoJet> constituents = jet.constituents();
```

Has memory management implications. But can help significantly with advanced usage.

One step on the way to a simple “tools” interface.

Goal: easy and centralized access to helpful utilities

Conclusions

Main public jet-clustering package currently is FastJet

- ▶ Code is quite stable
- ▶ Provides access to a lot more than just native FastJet algorithms

Future evolution

- ▶ Physics-driven: how can we make better use of jets?
- ▶ Code should provide facilities to make this easy in practice

User feedback is welcome!

- ▶ It has driven inclusion of “legacy” plugins
- ▶ It can help shape future evolution of code
- ▶ E.g. should there be built-in access from Python, PyRoot?