

Towards Jetography

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NIKHEF Theory Group Seminar
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Based on work with

Jon Butterworth, Matteo Cacciari, Mrinal Dasgupta, Adam Davison,
Lorenzo Magnea, Juan Rojo, Mathieu Rubin & Gregory Soyez

quark

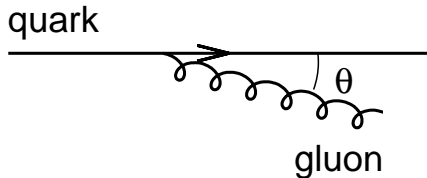


Gluon emission:

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

At low scales:

$$\alpha_s \rightarrow 1$$

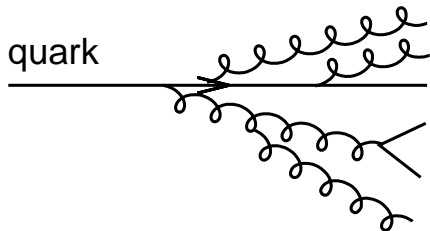


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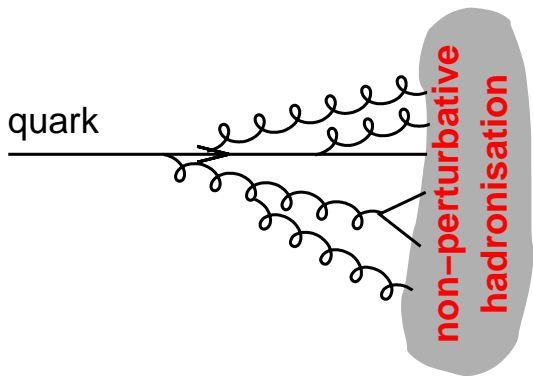


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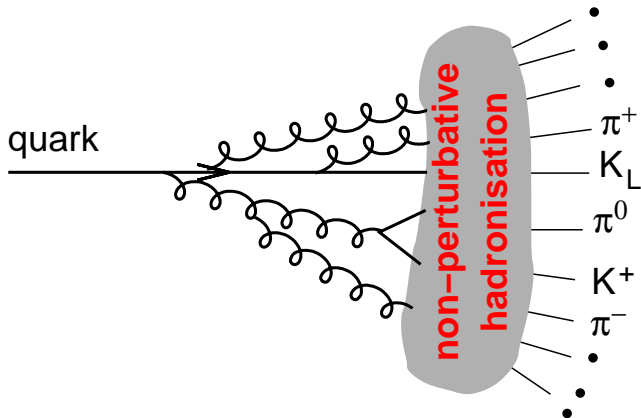


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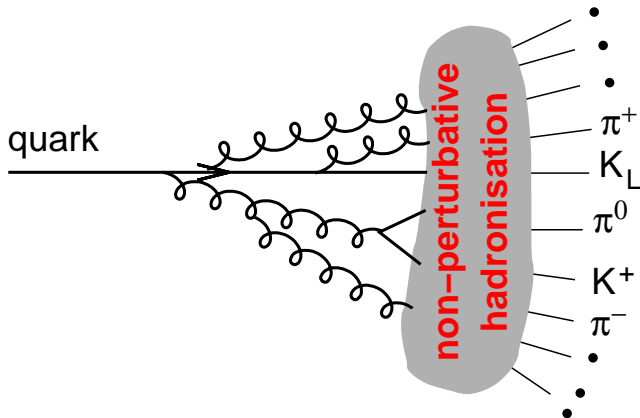


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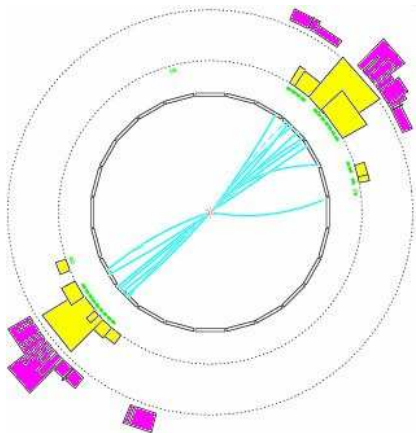
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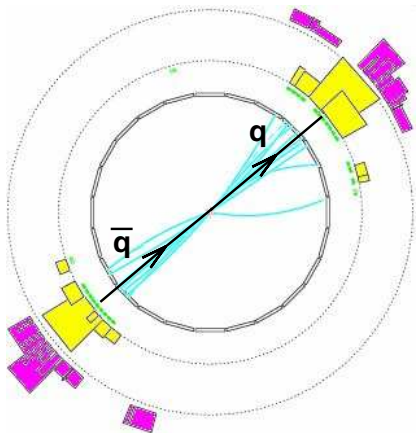
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This is a jet



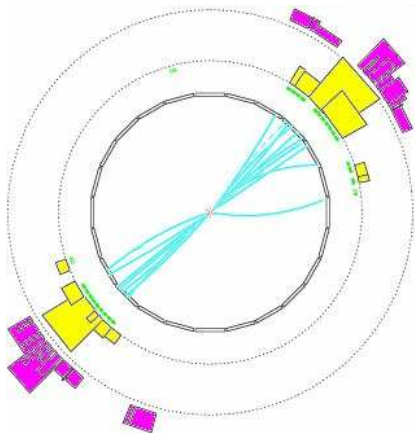
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Clearly(?) 2 jets here

How many jets do you see?
Do you really want to ask yourself
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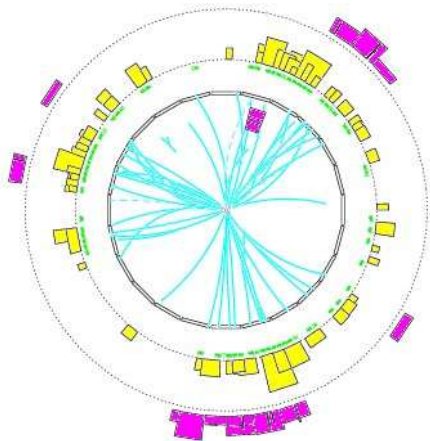


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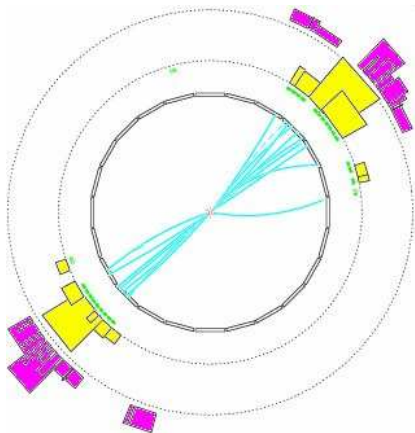


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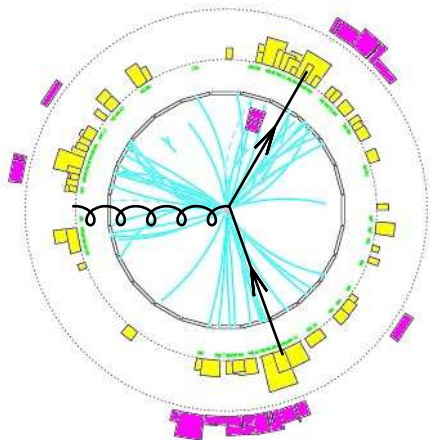


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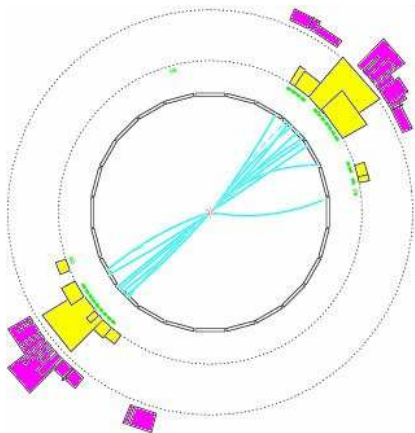
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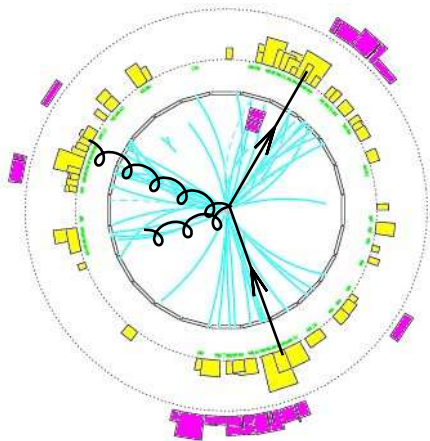
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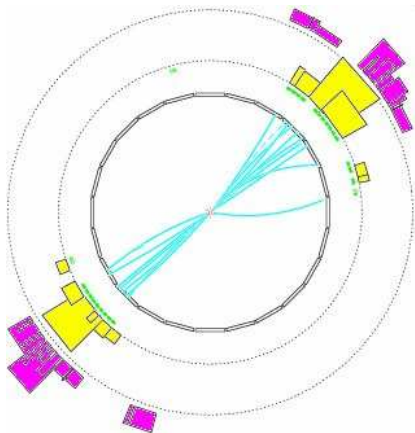
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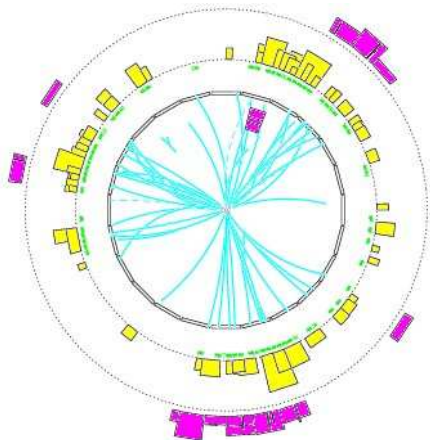
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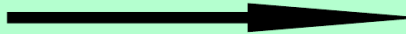
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jet definition

 $\{P_i\}$

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

jet algorithm

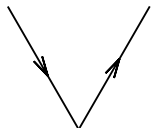
 $\{j_k\}$

jets

+ parameters (usually at least the radius R)

+ recombination scheme

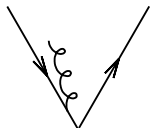
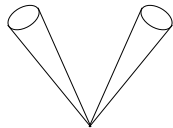
Reminder: running a jet definition gives a well defined physical observable,
which we can measure and, hopefully, calculate



LO partons

Jet ↓ Defⁿ

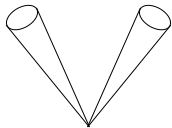
jet 1 jet 2



NLO partons

Jet ↓ Defⁿ

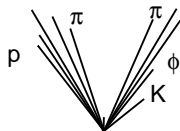
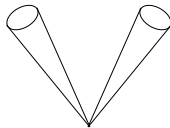
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parton shower

Jet ↓ Defⁿ

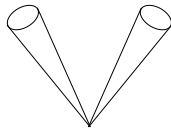
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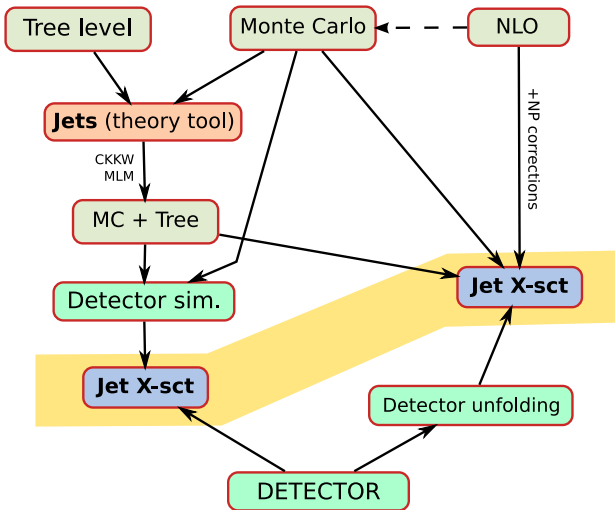
hadron level

Jet ↓ Defⁿ

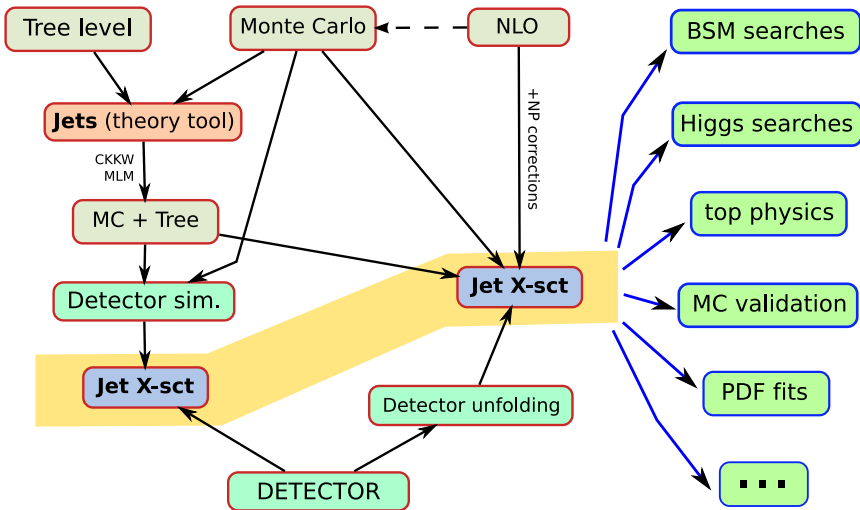
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Projection to jets should be resilient to QCD effects



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

What jet algorithms are out there?

sequential recombination (k_t)
& cone type

Sequential recombination

k_t , Jade, Cam/Aachen, ...

Bottom-up:

Cluster 'closest' particles repeatedly until few left \rightarrow jets.

Works because of mapping:

closeness \Leftrightarrow QCD divergence

Loved by e^+e^- , ep and theorists

Cone

UA1, JetClu, Midpoint, ...

Top-down:

Find coarse regions of energy flow (cones), and call them jets.

Works because *QCD only modifies energy flow on small scales*

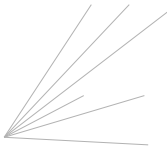
Loved by pp and few(er) theorists

k_t algorithm

Catani, Dokshitzer, Olsson, Seymour, Turnock, Webber '91-'93
Ellis, Soper '93

- ▶ Find smallest of all $d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 / R^2$ and $d_{iB} = k_i^2$
- ▶ Recombine i, j
- ▶ Repeat

Bottom-up jets: Sequential recombination



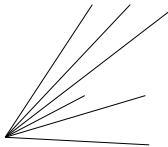
NB: hadron collider variables

- ▶ $\Delta R_{ij}^2 = (\phi_i - \phi_j)^2 + (y_i - y_j)^2$
- ▶ rapidity $y_i = \frac{1}{2} \ln \frac{E_i + p_{zi}}{E_i - p_{zi}}$
- ▶ ΔR_{ij} is boost invariant angle

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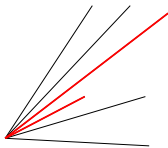
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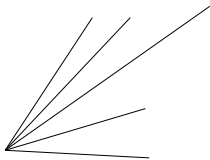
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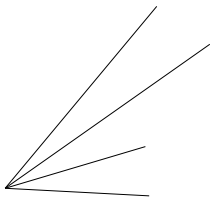
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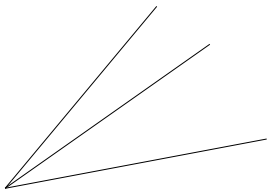
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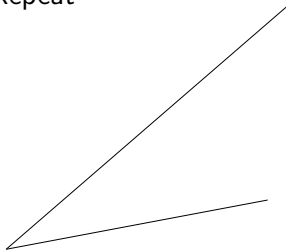
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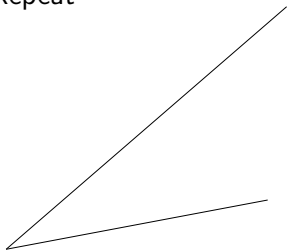
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k_t distance measures

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

are closely related to structure of divergences for QCD emissions

$$[dk_j] |M_{g \rightarrow g_i g_j}^2(k_j)| \sim \frac{\alpha_s C_A}{2\pi} \frac{dk_{tj}}{\min(k_{ti}, k_{tj})} \frac{d\Delta R_{ij}}{\Delta R_{ij}}, \quad (k_{tj} \ll k_{ti}, \Delta R_{ij} \ll 1)$$

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k_t algorithm attempts approximate inversion of branching process

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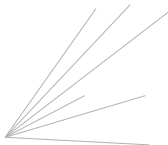
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- ▶ Find some/all stable cones
 - ≡ cone pointing in same direction as the momentum of its contents
- ▶ Resolve cases of overlapping stable cones

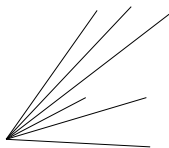
By using a 'split-merge' procedure

**Top-down jets:
cone algorithms**



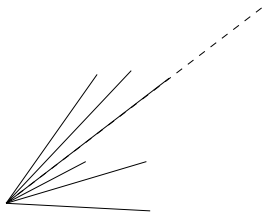
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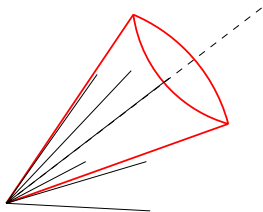
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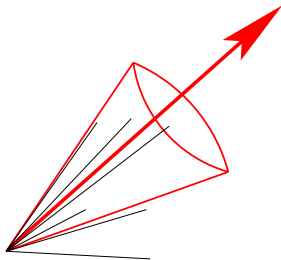
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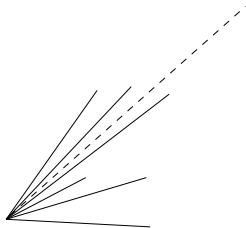
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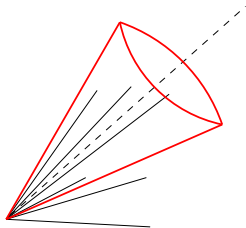
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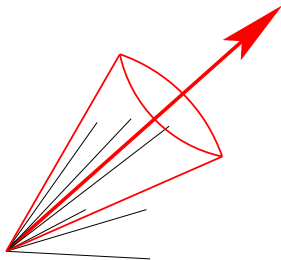
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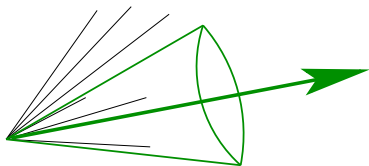
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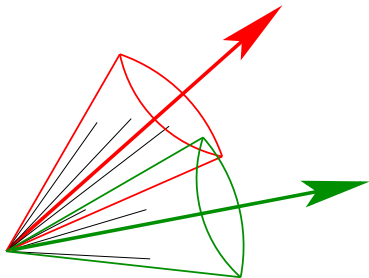
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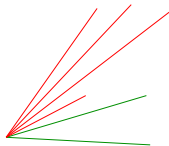
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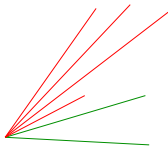


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How do you find the stable cones?

- ▶ Iterate from 'seed' particles
Done originally [JetClu, Atlas]
- ▶ Iterate from 'midpoints' between cones from seeds
Midpoint cone [Tevatron Run II]

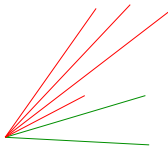


Tevatron & ATLAS cone algs have two main steps:

- ▶ Find some/all stable cones
≡ cone pointing in same direction as the momentum of its contents
- ▶ Resolve cases of overlapping stable cones
By running a 'split-merge' procedure

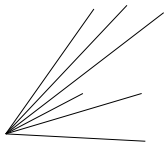
How do you find the stable cones?

- ▶ Iterate from 'seed' particles
Done originally [JetClu, Atlas]
- ▶ Iterate from 'midpoints' between cones from seeds
Midpoint cone [Tevatron Run II]



Procedure:

- ▶ Find one stable cone By iterating from hardest seed particle
- ▶ Call it a jet; remove its particles from the event; repeat

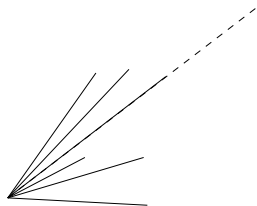


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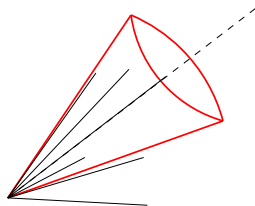


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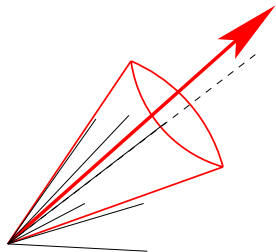


Procedure:

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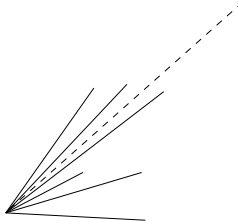
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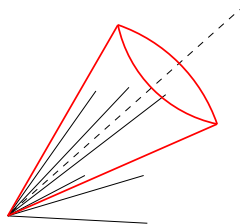
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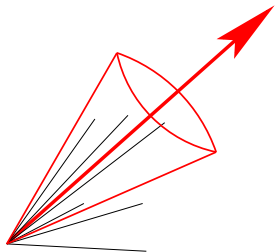
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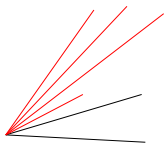
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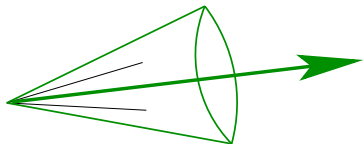
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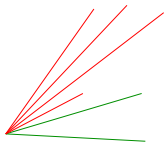
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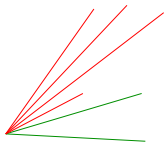
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Iterative Cone with Progressive Removal (IC-PR)

e.g. CMS it. cone, [Pythia Cone, GetJet], ...

- ▶ NB: not same type of algorithm as Atlas Cone, MidPoint, SIScone



Readying jet “technology” for the LHC era

[a.k.a. satisfying Snowmass]

Snowmass Accord (1990):

FERMILAB-Conf-90/249-E
[E-741/CDF]

Toward a Standardization of Jet Definitions *

Several important properties that should be met by a jet definition are [3]:

1. Simple to implement in an experimental analysis;
2. Simple to implement in the theoretical calculation;
3. Defined at any order of perturbation theory;
4. Yields finite cross section at any order of perturbation theory;
5. Yields a cross section that is relatively insensitive to hadronization.

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Property 1 \Leftrightarrow speed. (+other aspects)

- ▶ LHC events may have up to $N = 4000$ particles (at high-lumi)
- ▶ Sequential recombination algs. (k_t) slow, $\sim N^3 \rightarrow 60s$ for $N = 4000$

k_t not practical for $\mathcal{O}(10^9)$ events

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Property 4 \equiv Infrared and Collinear (IRC) Safety. It helps ensure:

- ▶ Soft (low-energy) emissions & collinear splittings don't change jets
- ▶ Each order of perturbation theory is smaller than previous (at high p_t)

Wasn't satisfied by the cone algorithms

'Trivial' computational issue:

▶ for N particles: N^2 d_{ij} searched through N times N^3

▶ 4000 parti

▶ Heavy lons. 50000 particles. 10 hours/event

(2000 CPU years)

Snowmass issue #1

The k_t algorithm and its speed

As far as possible physics choices should not be limited by computing.

Even if we're clever about repeating the full search each time, we still have $\mathcal{O}(N^2)$ d_{ij} 's to establish

'Trivial' computational issue:

- ▶ for N particles: $N^2 d_{ij}$ searched through N times = N^3
- ▶ 4000 particles (or calo cells): **1 minute**
NB: often study $10^7 - 10^9$ events (20-2000 CPU years)
- ▶ Heavy Ions: 30000 particles: **10 hours/event**

As far as possible physics choices should not be limited by computing.

Even if we're clever about repeating the full search each time, we still have $\mathcal{O}(N^2)$ d_{ij} 's to establish

There are $N(N - 1)/2$ distances d_{ij} — surely we have to calculate them all in order to find smallest?

k_t distance measure is partly *geometrical*:

$$\begin{aligned} \min_{i,j} d_{ij} &\equiv \min_{i,j} (\min\{k_{ti}^2, k_{tj}^2\} \Delta R_{ij}^2) \\ &= \min_{i,j} (k_{ti}^2 \Delta R_{ij}^2) \\ &= \min_i (k_{ti}^2 \min_j \Delta R_{ij}^2) \end{aligned}$$

In words: for each i look only at the k_t distance to its 2D geometrical nearest neighbour (GNN).

k_t distance need only be calculated between GNNs

Each point has 1 GNN \rightarrow need only calculate N d_{ij} 's

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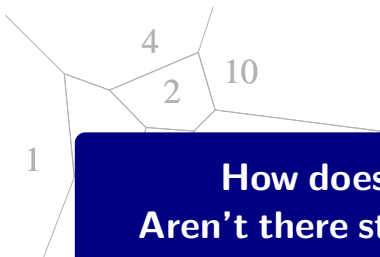
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Given a set of vertices on plane (1...10) a *Voronoi diagram* partitions plane into cells containing all points closest to each vertex.

How does use of GNN help?
Aren't there still $\frac{N^2}{2} \Delta R_{ij}^2$ to check... ?

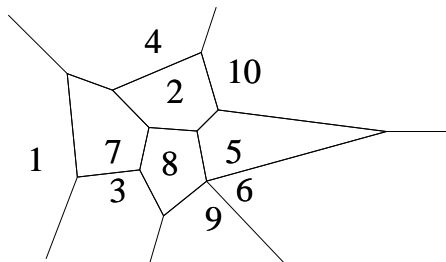
Geometrical nearest neighbour finding is a classic problem in the field of Computational Geometry

Deville's '99 [+ related work by other authors]
 Convenient C++ package available: CGAL, <http://www.cgal.org>

With help of CGAL, k_t clustering can be done in $N \ln N$ time.

Coded in the FastJet package (v1), Cacciari & GPS '05

2d nearest-neighbours



Given a set of vertices on plane (1...10) a *Voronoi diagram* partitions plane into cells containing all points closest to each vertex

Dirichlet '1850, Voronoi '1908

A vertex's nearest other vertex is always in an adjacent cell.

E.g. GNN of point 7 must be among 1,4,2,8,3 (it is 3)

Construction of Voronoi diagram for N points: $N \ln N$ time Fortune '88

Update of 1 point in Voronoi diagram: expected $\ln N$ time

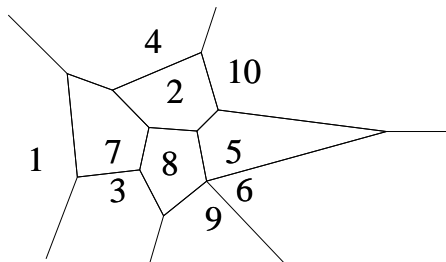
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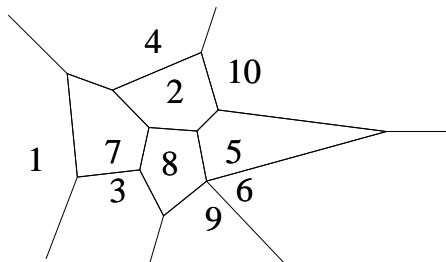
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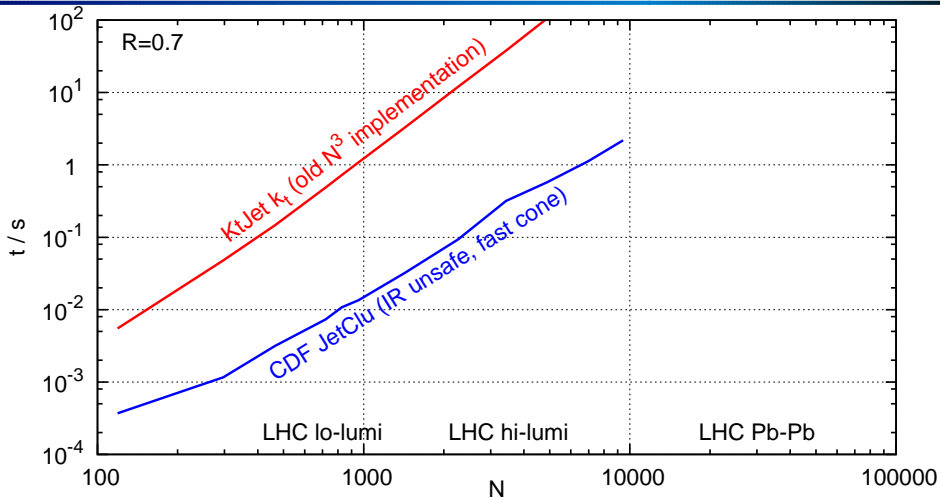
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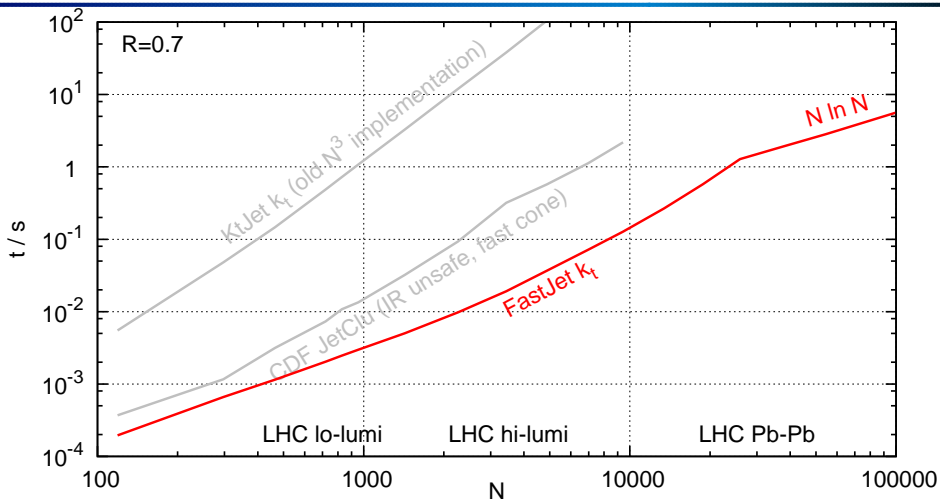
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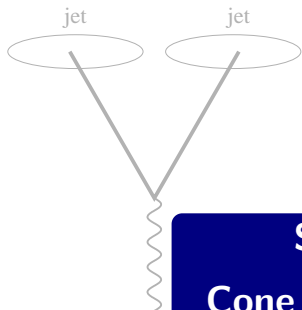
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k_t algorithm speed: old & new

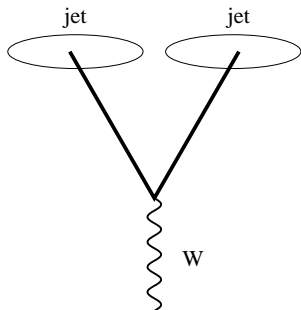
Factorisation of momentum & geometry
 → **2–3 orders of magnitude gain in speed!**

Speed competitive with fast cone algorithms



Snowmass issue #4
Cone algorithms and IR safety

	$\alpha_s^2 \alpha_{EW}$	$\alpha_s^3 \alpha_{EW}$	$\alpha_s^3 \alpha_{EW}$
1-jet			$+\infty$
2-jet	$\mathcal{O}(1)$	$-\infty$	0



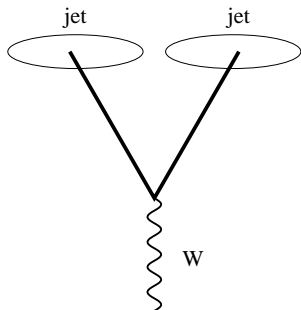
1-jet $\alpha_s^2 \alpha_{EW}$
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$\alpha_s^3 \alpha_{EW}$
 $-\infty$

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With these (& most) cone algorithms, perturbative infinities fail to cancel at some order \equiv IR unsafety

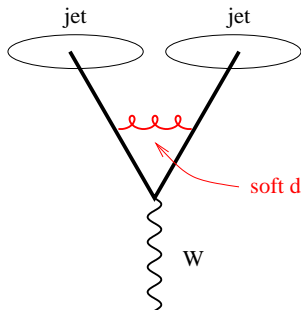
JetClu (& Atlas Cone) in Wjj @ NLO



$$\alpha_s^2 \alpha_{EW}$$

1-jet

2-jet $\mathcal{O}(1)$



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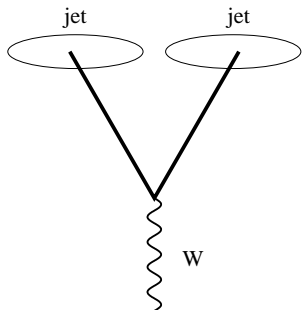
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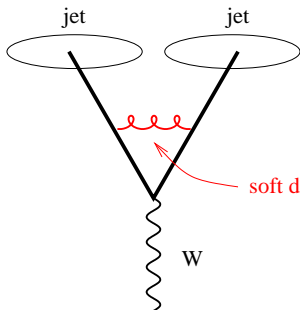
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1-jet

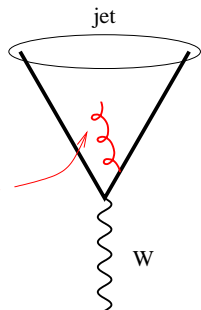
2-jet $\mathcal{O}(1)$



$$\alpha_s^3 \alpha_{EW}$$

$-\infty$

soft divergence



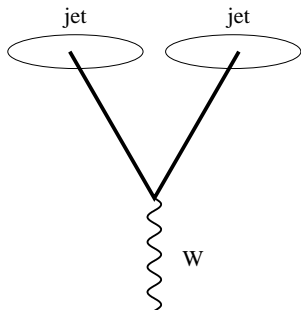
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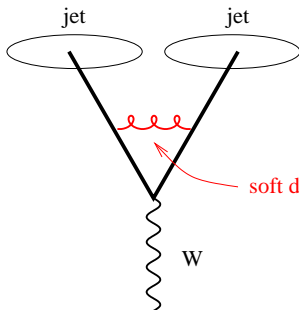
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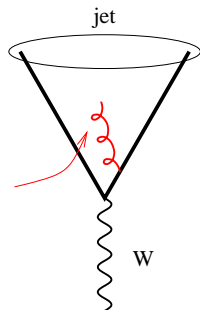
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0

With these (& most) cone algorithms, perturbative infinities fail to cancel at some order \equiv **IR unsafety**

Real life does not have infinities, but pert. infinity leaves a real-life trace

$$\alpha_s^2 + \alpha_s^3 + \alpha_s^4 \times \infty \rightarrow \alpha_s^2 + \alpha_s^3 + \alpha_s^4 \times \ln p_t/\Lambda \rightarrow \alpha_s^2 + \underbrace{\alpha_s^3 + \alpha_s^3}_{\text{BOTH WASTED}}$$

Among consequences of IR unsafety:

	<i>Last meaningful order</i>			Known at
	JetClu, ATLAS cone [IC-SM]	MidPoint [IC _{mp} -SM]	CMS it. cone [IC-PR]	
Inclusive jets	LO	NLO	NLO	NLO (→ NNLO)
W/Z + 1 jet	LO	NLO	NLO	NLO
3 jets	none	LO	LO	NLO [nlojet++]
W/Z + 2 jets	none	LO	LO	NLO [MCFM]
m_{jet} in $2j + X$	none	none	none	LO

NB: 50,000,000\$/£/CHF/€ investment in NLO

Multi-jet contexts much more sensitive: **ubiquitous at LHC**

And LHC will rely on QCD for background double-checks
 extraction of cross sections, extraction of parameters

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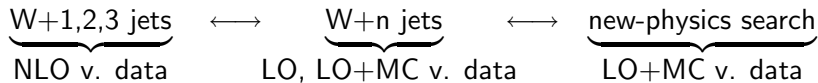
I do searches, not QCD. Why should I care about IRC safety?

- ▶ Are you looking for a mass-peak? ➡ you needn't care much
- ▶ Are you looking for an excess over bkgd? ➡ you need control samples, validated against QCD



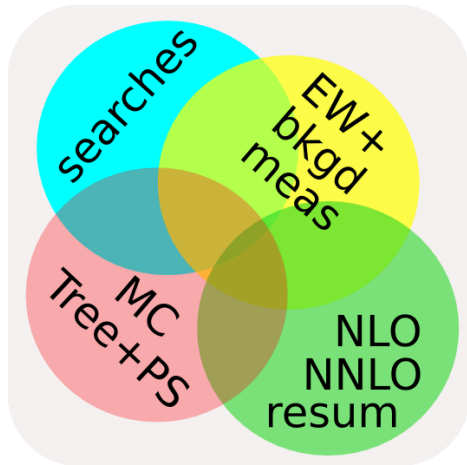
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$W+1,2,3$ jets
 NLO v. data
 IR safe alg.



$W+n$ jets
 LO, LO+MC v. data
 IR safe alg.

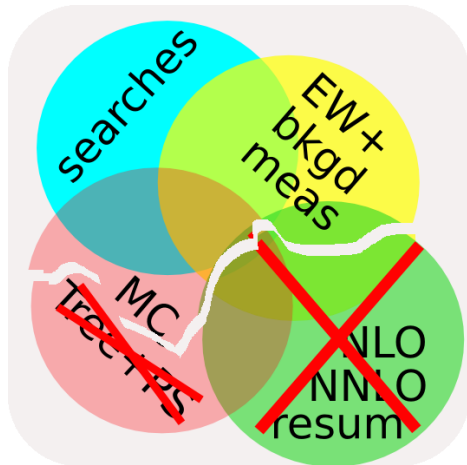


new-physics search
 LO+MC v. data
 IR safe alg.

Does lack of IRC safety matter?

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$W+1,2,3$ jets
 NLO v. data
 IR safe alg.

↔

$W+n$ jets
 LO, LO+MC v. data
 IR safe alg.

↔

new-physics search
 LO+MC v. data
 IR **unsafe** alg.

How do we solve
cone IR safety
problems?

Fix stable-cone finding



SISCone

GPS & Soyez '07

Same family as Tev. Run II alg

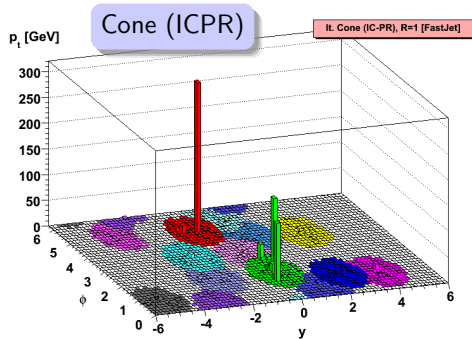
Invent "cone-like" alg.



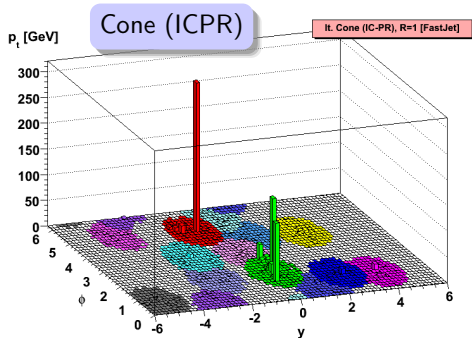
anti-kt

Cacciari, GPS & Soyez '08

Essential characteristic of cones?



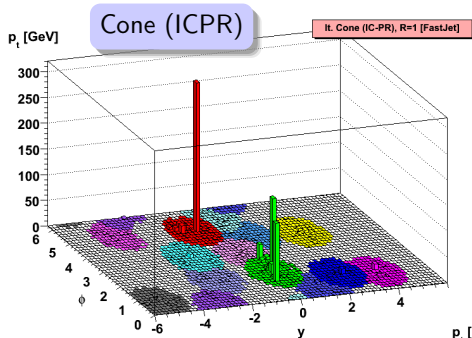
Essential characteristic of cones?



(Some) cone algorithms give **circular** jets in $y - \phi$ plane

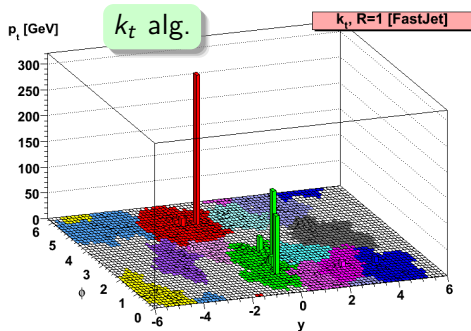
Much appreciated by experiments
 e.g. for acceptance corrections

Essential characteristic of cones?

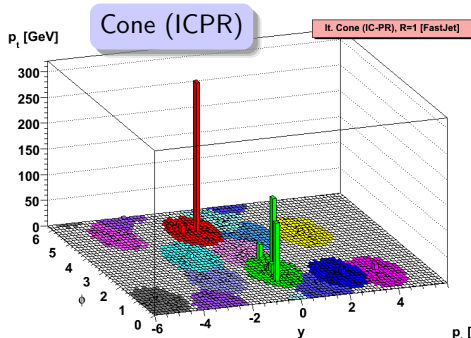


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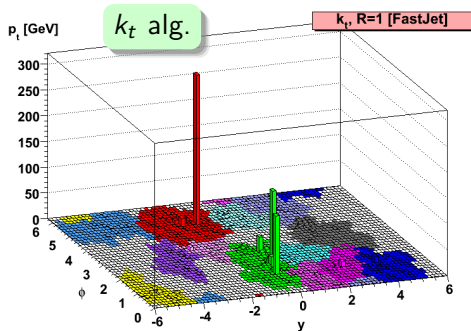
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k_t jets are **irregular**

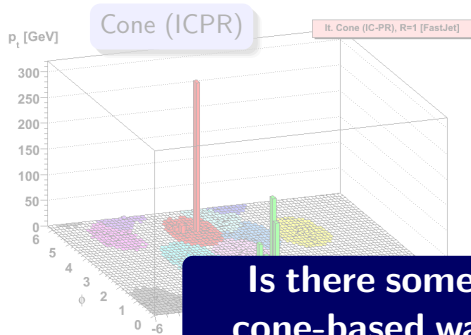
Because soft junk clusters together first:

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

Regularly held against k_t



Essential characteristic of cones?



(Some) cone algorithms give **circular** jets in $y - \phi$ plane

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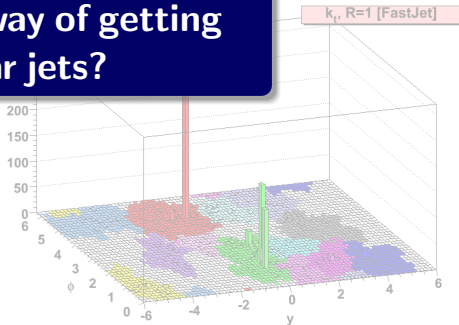
Is there some other, non cone-based way of getting circular jets?

k_t jets are

Because soft junk clusters together first:

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

Regularly held against k_t



Soft stuff clusters with nearest neighbour

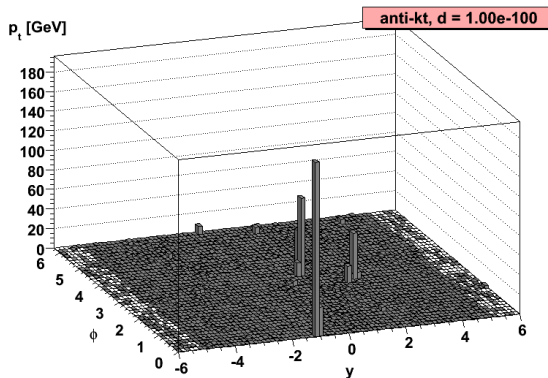
$$k_t: d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 \longrightarrow \text{anti-}k_t: d_{ij} = \frac{\Delta R_{ij}^2}{\max(k_{ti}^2, k_{tj}^2)}$$

Hard stuff clusters with nearest neighbour
Privilege collinear divergence over soft divergence

Soft stuff clusters with nearest neighbour

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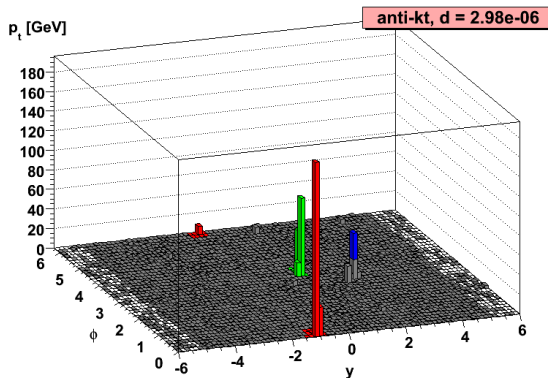
Hard stuff clusters with nearest neighbour
 Divergence over soft divergence



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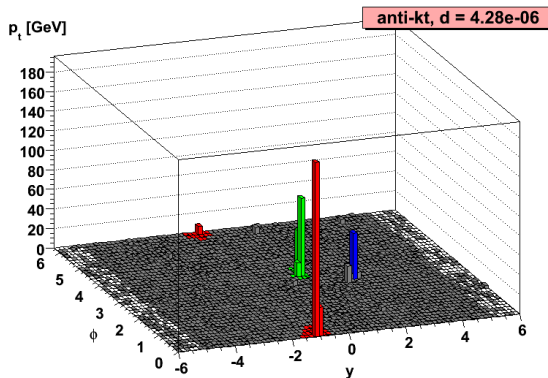
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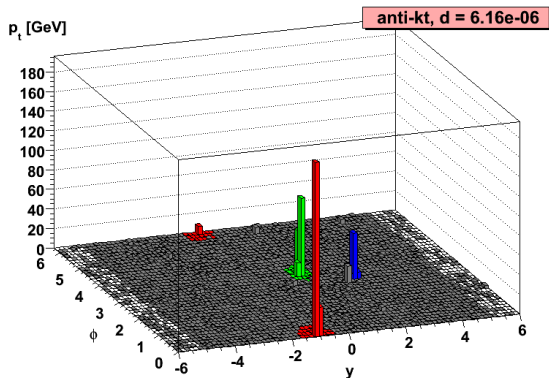
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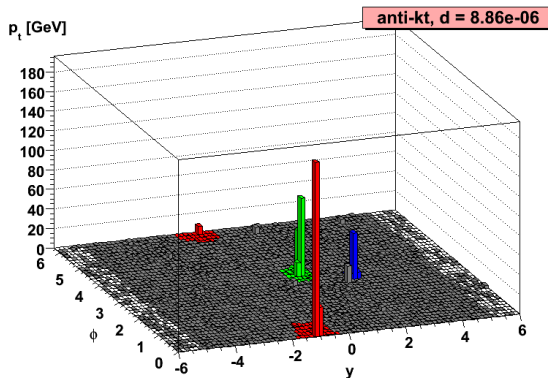
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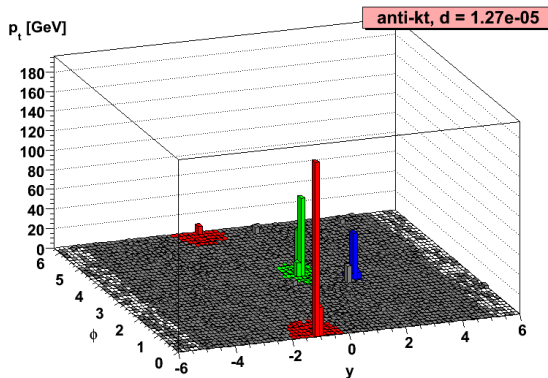
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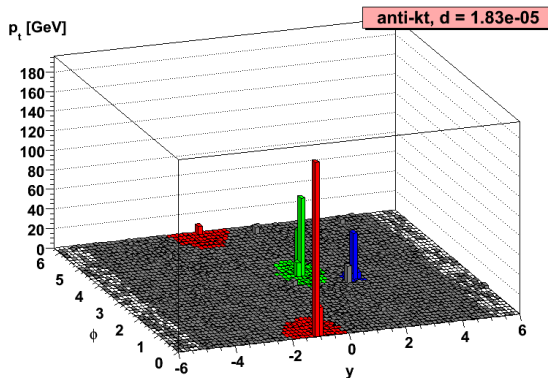
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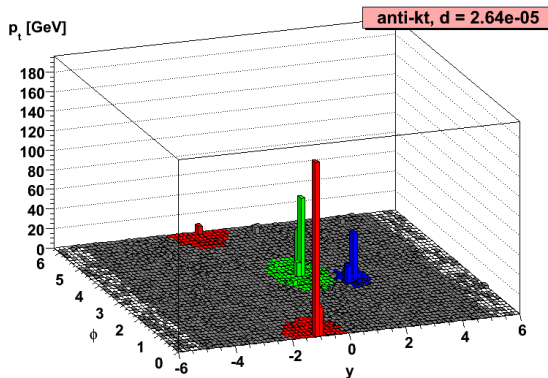
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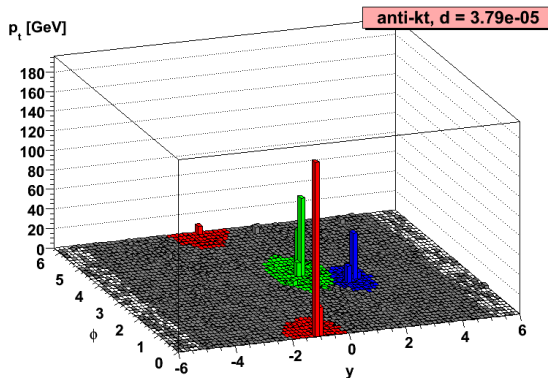
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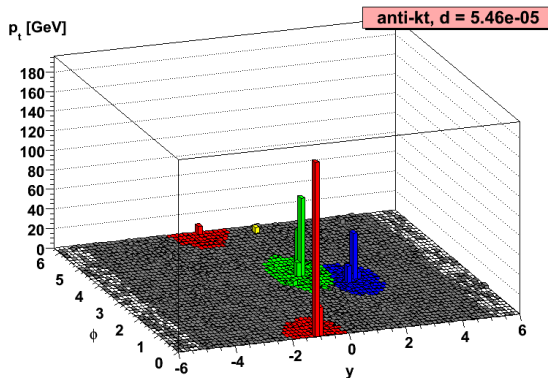
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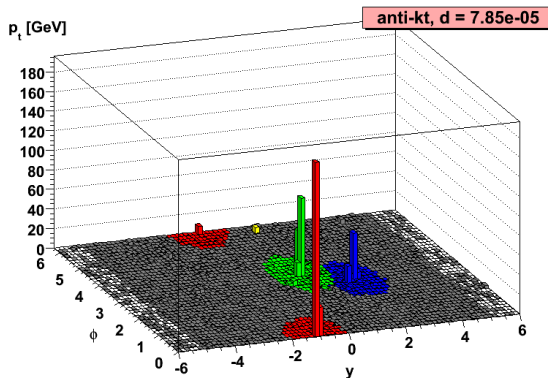
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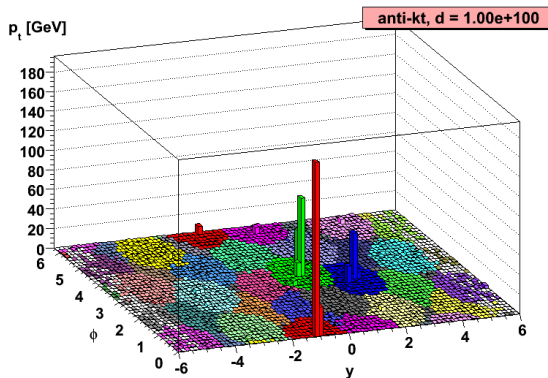
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Hard stuff clusters with nearest neighbour
Divergence over soft divergence



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

Generalise inclusive-type sequential recombination with

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

	Alg. name	Comment	time
$p = 1$	k_t CDOSTW '91-93; ES '93	Hierarchical in rel. k_t	$N \ln N$ exp.
$p = 0$	Cambridge/Aachen Dok, Leder, Moretti, Webber '97 Wengler, Wobisch '98	Hierarchical in angle Scan multiple R at once ↔ QCD angular ordering	$N \ln N$
$p = -1$	anti- k_t Cacciari, GPS, Soyez '08 ~ reverse- k_t Delsart	Hierarchy meaningless, jets like CMS cone (IC-PR)	$N^{3/2}$
SC-SM	SISCone GPS Soyez '07 + Tevatron run II '00	Replaces JetClu, ATLAS MidPoint (xC-SM) cones	$N^2 \ln N$ exp.

All these algorithms coded in (efficient) C++ at
<http://fastjet.fr/> (Cacciari, GPS & Soyez '05-08)

Algorithm	Type	IRC status	Evolution
exclusive k_t	SR $_{p=1}$	OK	$N^3 \rightarrow N \ln N$
inclusive k_t	SR $_{p=1}$	OK	$N^3 \rightarrow N \ln N$
Cambridge/Aachen	SR $_{p=0}$	OK	$N^3 \rightarrow N \ln N$
Run II Seedless cone	SC-SM	OK	\rightarrow SIScone
CDF JetClu	IC $_r$ -SM	IR $_{2+1}$	[\rightarrow SIScone]
CDF MidPoint cone	IC $_{mp}$ -SM	IR $_{3+1}$	\rightarrow SIScone
CDF MidPoint searchcone	IC $_{se,mp}$ -SM	IR $_{2+1}$	[\rightarrow SIScone]
D0 Run II cone	IC $_{mp}$ -SM	IR $_{3+1}$	\rightarrow SIScone [with p_t cut?]
ATLAS Cone	IC-SM	IR $_{2+1}$	\rightarrow SIScone
PxCone	IC $_{mp}$ -SD	IR $_{3+1}$	[little used]
CMS Iterative Cone	IC-PR	Coll $_{3+1}$	\rightarrow anti- k_t
PyCell/CellJet (from Pythia)	FC-PR	Coll $_{3+1}$	\rightarrow anti- k_t
GetJet (from ISAJET)	FC-PR	Coll $_{3+1}$	\rightarrow anti- k_t

SR = seq.rec.; IC = it.cone; FC = fixed cone;

SM = split-merge; SD = split-drop; PR = progressive removal

Snowmass is solved

But it was a problem from the 1990s

What are the problems we *should* be
trying to solve for LHC?

Which jet definition(s) for LHC?

Choice of algorithm (k_t , SISCone, ...)

Choice of parameters (R , ...)

Can we address this question scientifically?

Jetography

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Jetography

Jet definitions differ mainly in:

alg + R

1. How close two particles must be to end up in same jet
[discussed in the '90s, e.g. Ellis & Soper]
2. How much perturbative radiation is lost from a jet
[indirectly discussed in the '90s (analytic NLO for inclusive jets)]
3. How much non-perturbative contamination
(hadronisation, UE, pileup) a jet receives
[partially discussed in '90s — Korchemsky & Sterman '95, Seymour '97]

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The question's dangerous: a "parton" is an ambiguous concept

Three limits can help you:

- ▶ Threshold limit e.g. de Florian & Vogelsang '07
- ▶ Parton from color-neutral object decay (Z')
- ▶ Small- R (radius) limit for jet

One simple result

$$\frac{\langle p_{t,jet} - p_{t,parton} \rangle}{p_t} = \frac{\alpha_s}{\pi} \ln R \times \begin{cases} 1.01 C_F & \text{quarks} \\ 0.94 C_A + 0.07 n_f & \text{gluons} \end{cases} + \mathcal{O}(\alpha_s)$$

only $\mathcal{O}(\alpha_s)$ depends on algorithm & process
cf. Dasgupta, Magnea & GPS '07

Hadronisation: the “parton-shower” \rightarrow hadrons transitionMethod:

- ▶ “infrared finite α_s ” à la Dokshitzer & Webber '95
- ▶ **prediction** based on e^+e^- event shape data
- ▶ could have been deduced from old work Korchensky & Sterman '95
Seymour '97

Main result

$$\langle p_{t,jet} - p_{t,parton-shower} \rangle \simeq -\frac{0.4 \text{ GeV}}{R} \times \begin{cases} C_F & \text{quarks} \\ C_A & \text{gluons} \end{cases}$$

cf. Dasgupta, Magnea & GPS '07
coefficient holds for anti- k_t ; see Mrinal's talk for k_t alg.

“Naive” prediction (UE \simeq colour dipole between pp):

$$\Delta p_t \simeq 0.4 \text{ GeV} \times \frac{R^2}{2} \times \begin{cases} C_F & q\bar{q} \text{ dipole} \\ C_A & \text{gluon dipole} \end{cases}$$

DWT Pythia tune or ATLAS Jimmy tune tell you:

$$\Delta p_t \simeq \mathbf{10 - 15 \text{ GeV}} \times \frac{R^2}{2}$$

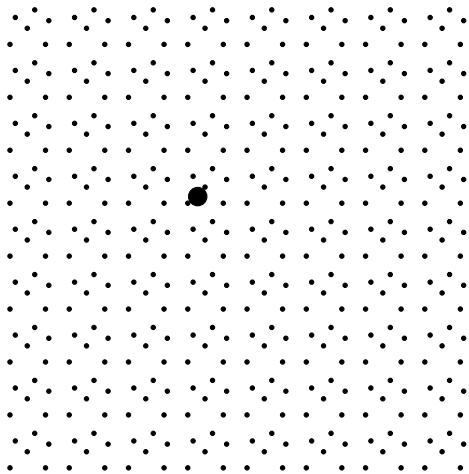
This big coefficient motivates special effort to understand interplay between jet algorithm and UE: “jet areas”

How does coefficient depend on algorithm?

How does it depend on jet p_t ? How does it fluctuate?

cf. Cacciari, GPS & Soyez '08

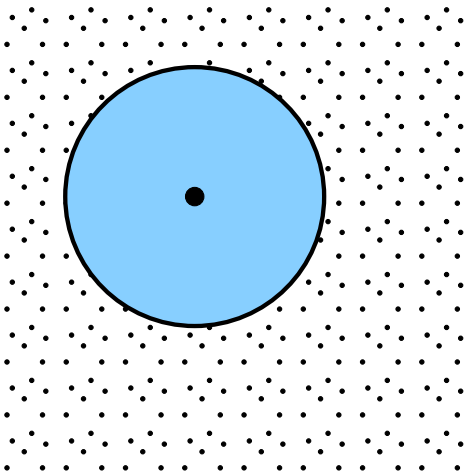
1. One hard particle, many soft

SIScone, any R , $f \gtrsim 0.391$ **Jet area =**

Measure of jet's susceptibility to
uniform soft radiation

Depends on details of an
algorithm's clustering dynamics.

2. One hard stable cone, area = πR^2



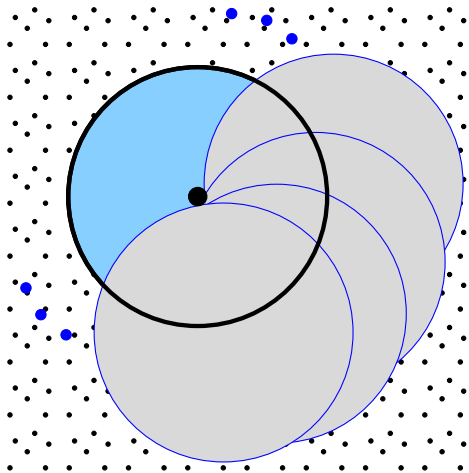
SIScone, any R , $f \gtrsim 0.391$

Jet area =

Measure of jet's susceptibility to
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Depends on details of an
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3. Overlapping “soft” stable cones



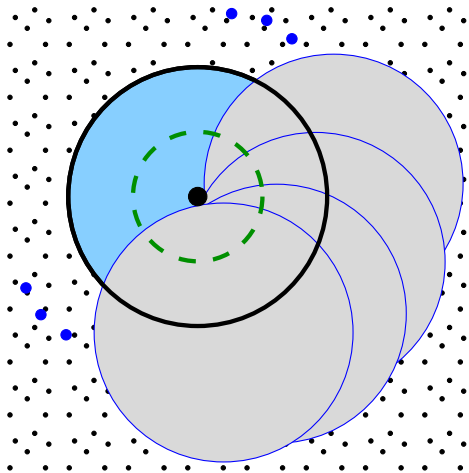
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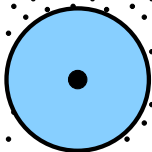
4. "Split" the overlapping parts

SIScone, any $R, f \gtrsim 0.391$ **Jet area =**

Measure of jet's susceptibility to
uniform soft radiation

Depends on details of an
algorithm's clustering dynamics.

5. Final hard jet (reduced area)

SIScone, any R , $f \gtrsim 0.391$ **Jet area =**

Measure of jet's susceptibility to uniform soft radiation

Depends on details of an algorithm's clustering dynamics.

SIScone's area (1 hard particle)

$$= \frac{1}{4} \pi R^2$$

Jet algorithm properties: summary

	k_t	Cam/Aachen	anti- k_t	SISCone
reach	R	R	R	$(1 + \frac{p_{t2}}{p_{t1}})R$
$\Delta p_{t,PT} \simeq \frac{\alpha_s C_i}{\pi} \times$	$\ln R$	$\ln R$	$\ln R$	$\ln 1.35R$
$\Delta p_{t,hadr} \simeq -\frac{0.4 \text{ GeV} C_i}{R} \times$	0.7	?	1	?
area = $\pi R^2 \times$	0.81 ± 0.28	0.81 ± 0.26	1	0.25
$+ \pi R^2 \frac{C_i}{\pi b_0} \ln \frac{\alpha_s(Q_0)}{\alpha_s(Rp_t)} \times$	0.52 ± 0.41	0.08 ± 0.19	0	0.12 ± 0.07

In words:

- ▶ k_t : area fluctuates a lot, depends on p_t (bad for UE)
- ▶ Cam/Aachen: area fluctuates somewhat, depends less on p_t
- ▶ anti- k_t : area is constant (circular jets)
- ▶ SISCone: reaches far for hard radiation (good for resolution, bad for multijets), area is smaller (good for UE)

Jet momentum significantly affected by R

So what R should we choose?

*Examine this in context of reconstruction
of dijet resonance*

What R is best for an isolated jet?

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

PT radiation:

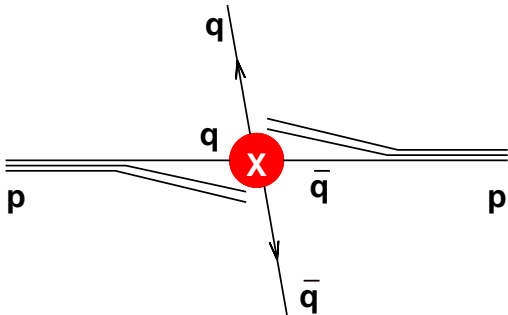
$$q : \langle \Delta p_t \rangle \simeq \frac{\alpha_s C_F}{\pi} p_t \ln R$$

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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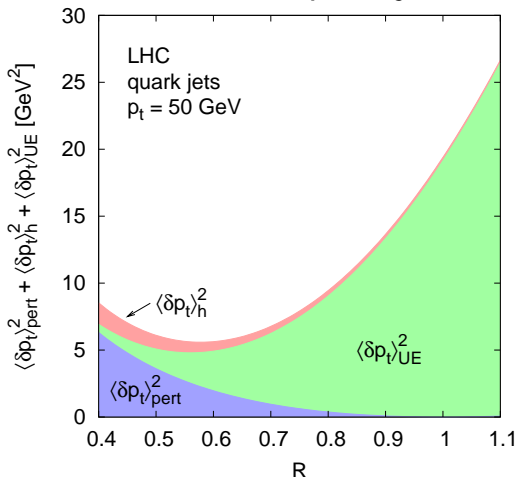
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50 GeV quark jet



in small- R limit (!)

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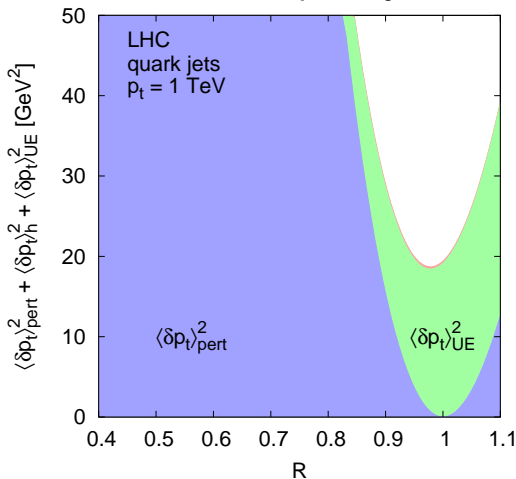
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1 TeV quark jet



in small- R limit (!)

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What R is best for an isolated jet?

1 TeV quark jet

PT radiation:

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

At low p_t , small R limits relative impact of UE

Hadronization

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

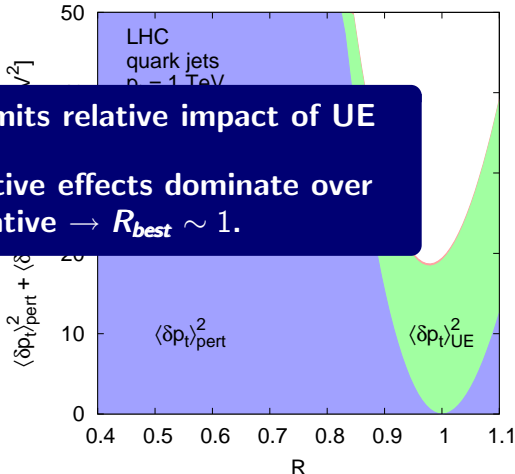
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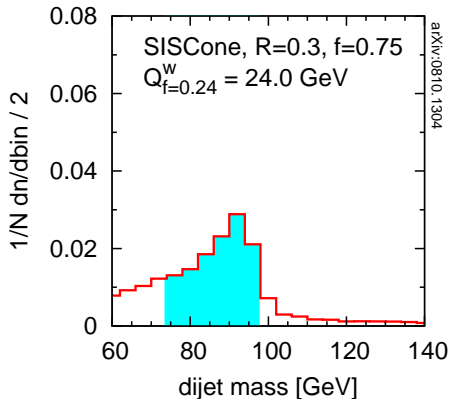
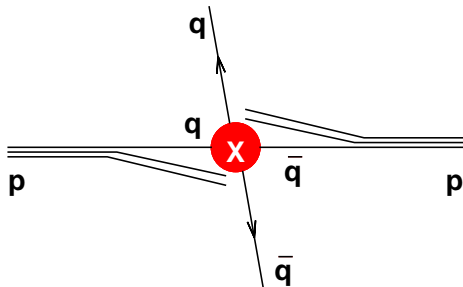
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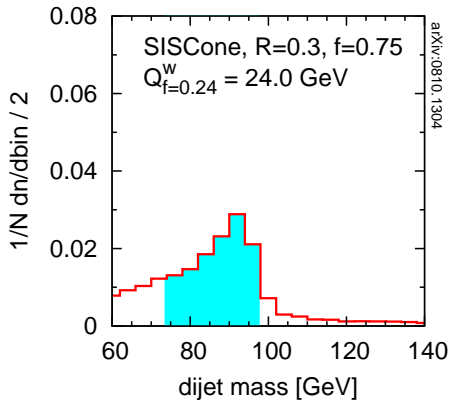
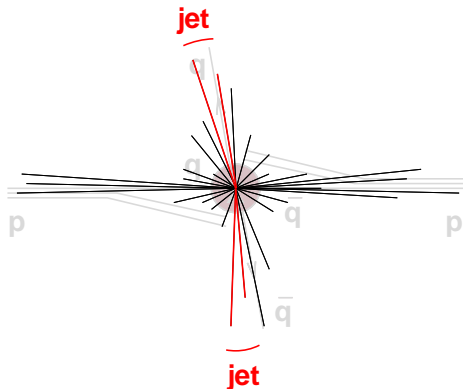
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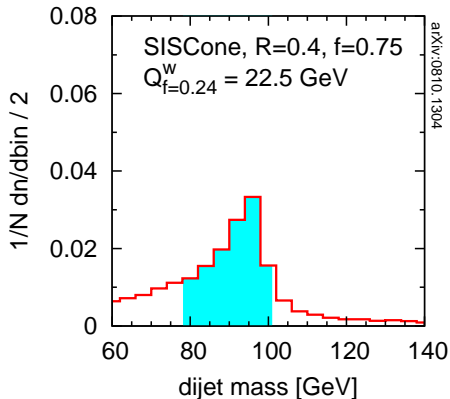
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Dijet mass: scan over R [Pythia 6.4] $R = 0.3$ $qq, M = 100 \text{ GeV}$ Resonance $X \rightarrow$ dijets

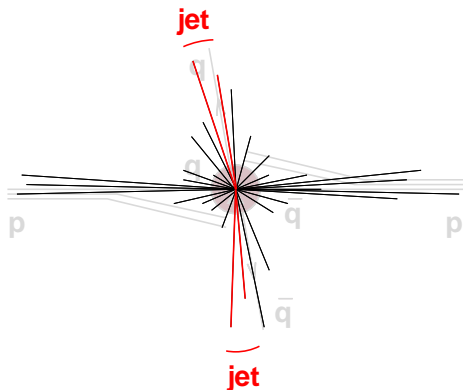
Dijet mass: scan over R [Pythia 6.4] $R = 0.3$ qq, $M = 100$ GeVResonance X \rightarrow dijets

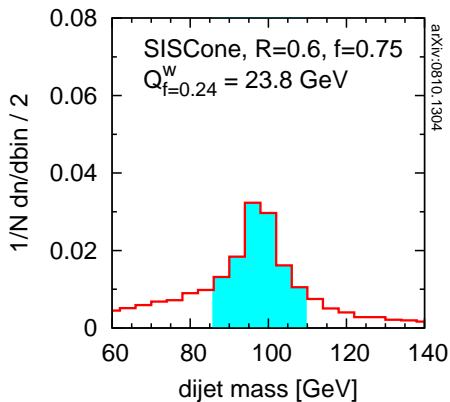
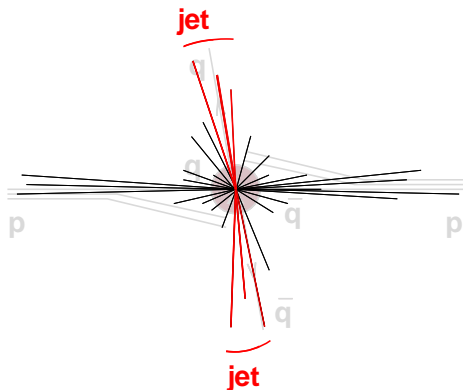
$R = 0.4$

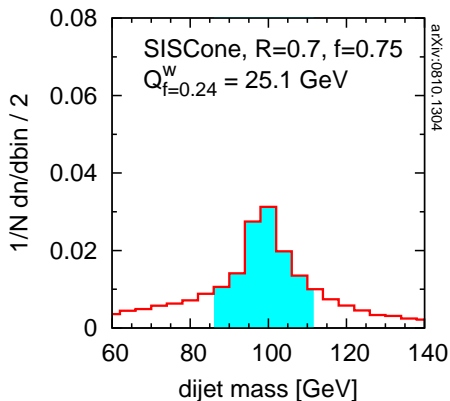
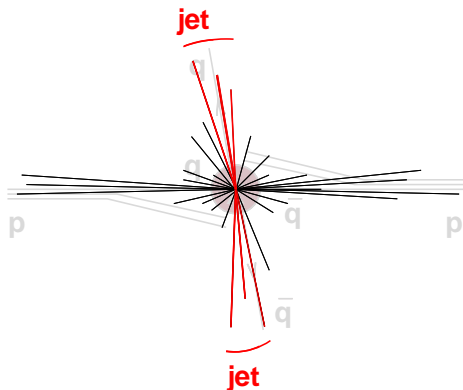
$qq, M = 100 \text{ GeV}$



Resonance X \rightarrow dijets

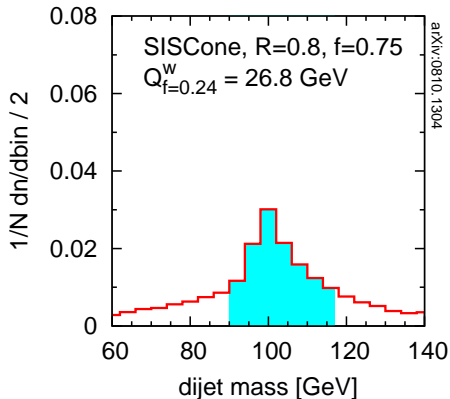


$R = 0.6$ qq, $M = 100$ GeVResonance X \rightarrow dijets

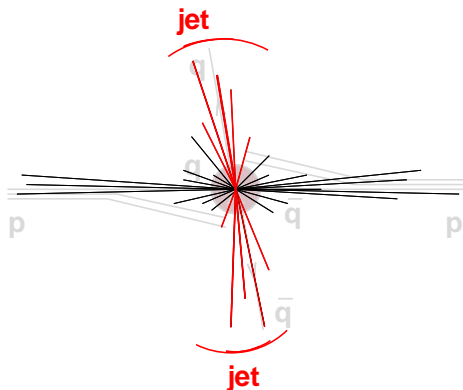
$R = 0.7$ qq, $M = 100$ GeVResonance X \rightarrow dijets

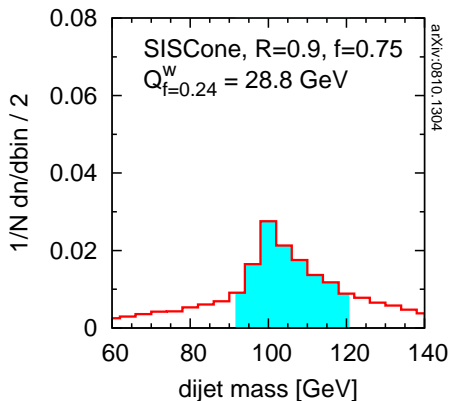
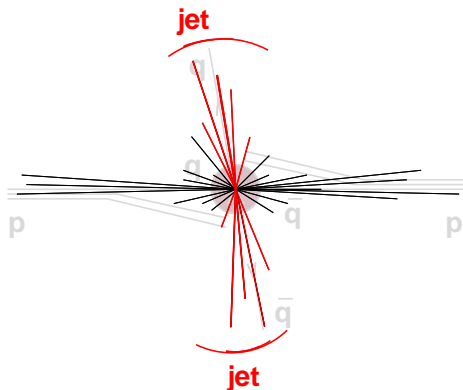
$R = 0.8$

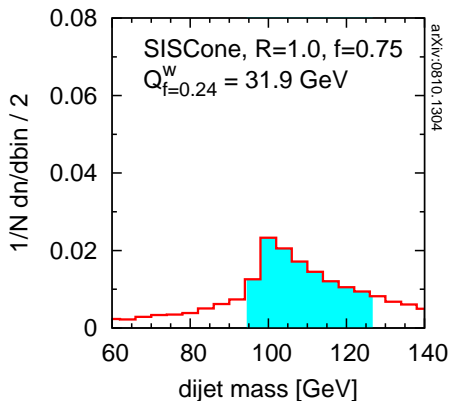
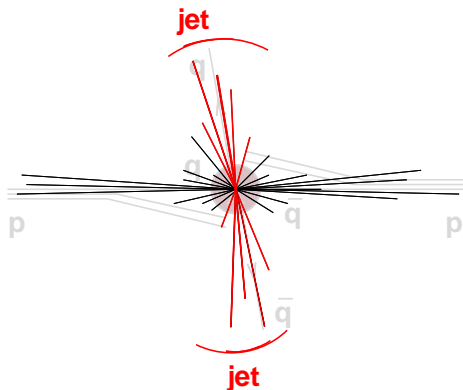
$qq, M = 100 \text{ GeV}$

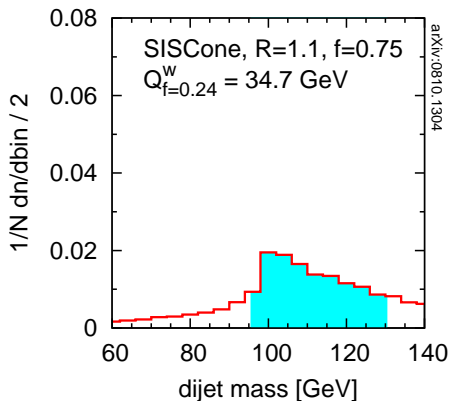
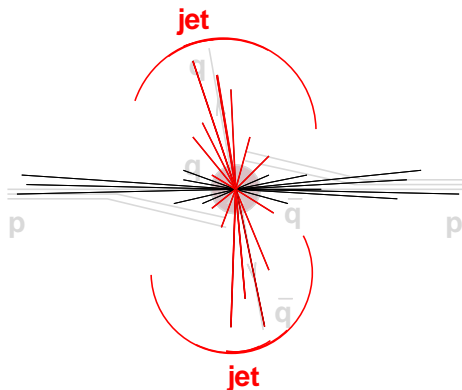


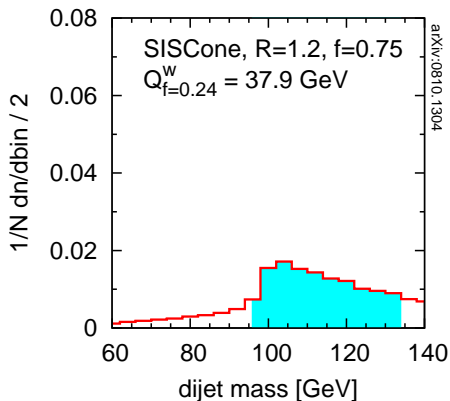
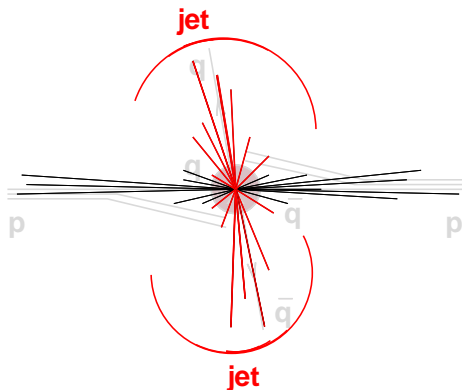
Resonance X \rightarrow dijets



$R = 0.9$ qq, $M = 100$ GeVResonance X \rightarrow dijets

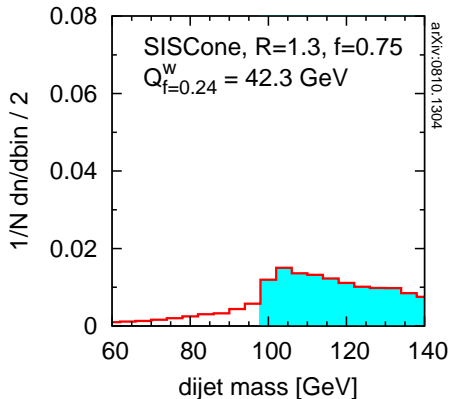
$R = 1.0$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

$R = 1.1$ $qq, M = 100 \text{ GeV}$ Resonance X \rightarrow dijets

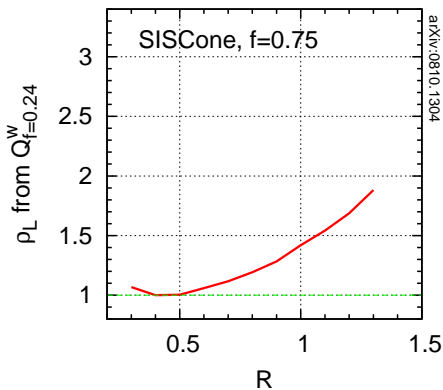
Dijet mass: scan over R [Pythia 6.4] $R = 1.2$ qq, $M = 100$ GeVResonance X \rightarrow dijets

$R = 1.3$

qq, $M = 100$ GeV



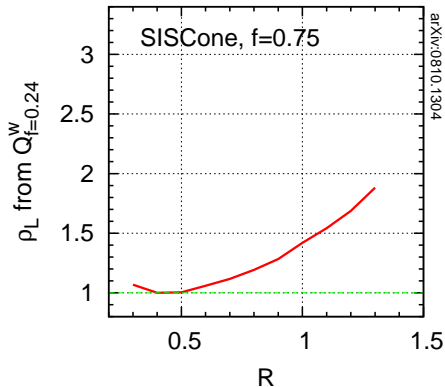
qq, $M = 100$ GeV



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

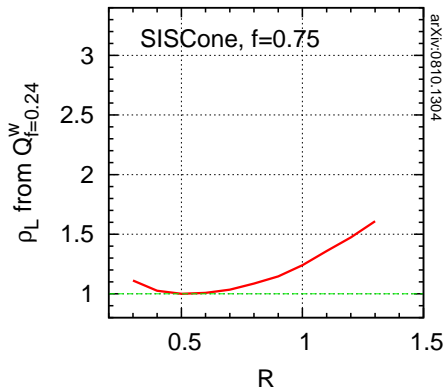
e.g. CMS arXiv:0807.4961

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

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

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



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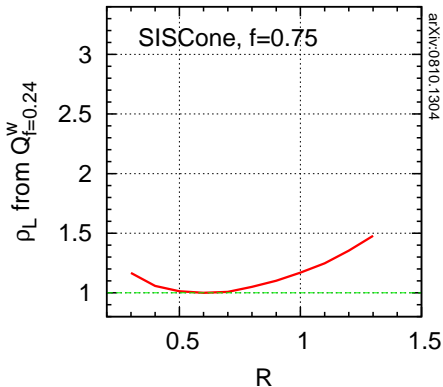
e.g. CMS arXiv:0807.4961

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

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

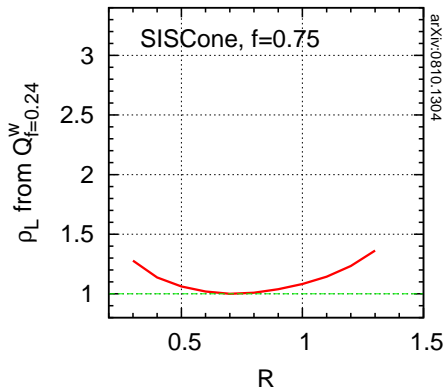
e.g. CMS arXiv:0807.4961

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

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

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



Best R is at minimum of curve

- ▶ Best R depends strongly on mass of system
- ▶ Increases with mass, just like crude analytical prediction
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BUT: so far, LHC's plans involve running with fixed smallish R values

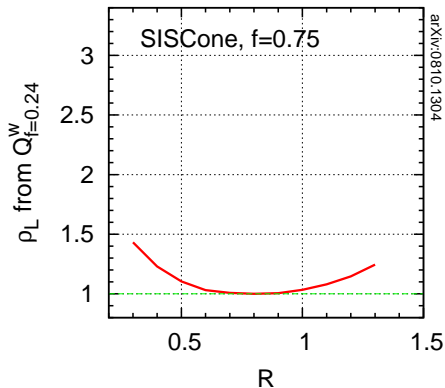
e.g. CMS arXiv:0807.4961

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

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

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



Best R is at minimum of curve

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- ▶ 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

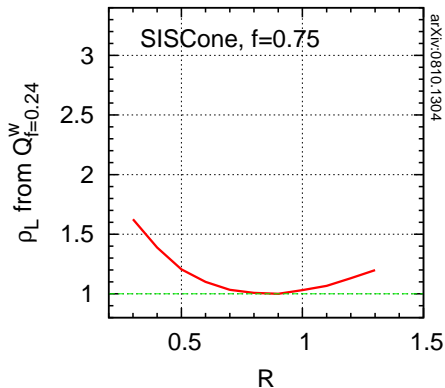
e.g. CMS arXiv:0807.4961

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

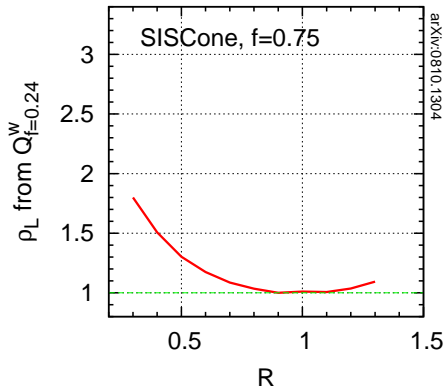
BUT: so far, LHC's plans involve running with fixed smallish R values

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

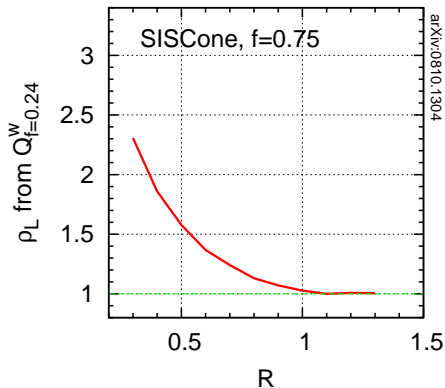
BUT: so far, LHC's plans involve running with fixed smallish R values

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

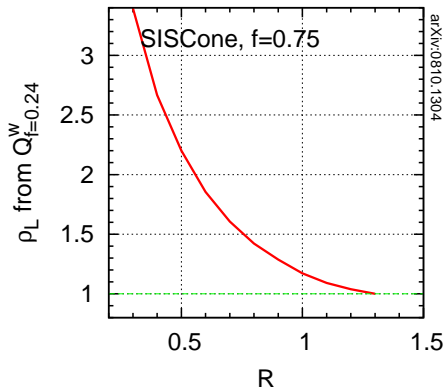
BUT: so far, LHC's plans involve running with fixed smallish R values

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

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- ▶ 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

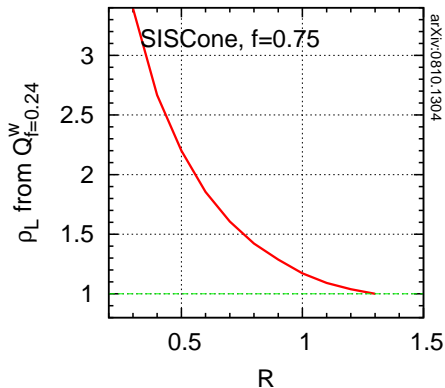
e.g. CMS arXiv:0807.4961

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Cacciari, Rojo, GPS & Soyez '08

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

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



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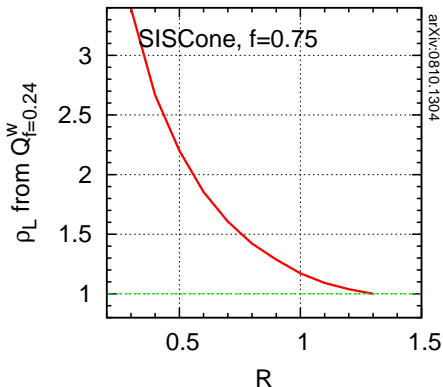
e.g. CMS arXiv:0807.4961

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

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

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



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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

File Edit View History Bookmarks Tools Help

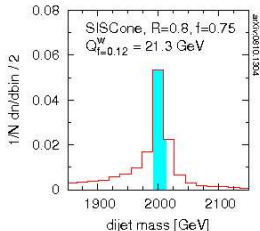
http://www.lpthe.jussieu.fr/~salam/jet-quality/

Testing jet definitions: qq & gg c...

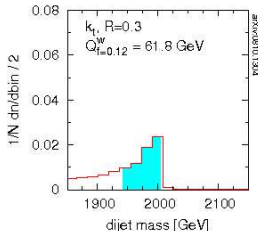
Testing jet definitions: qq & gg cases

by M. Cacciari, J. Rojo, G.P. Salam and G. Soyez, arXiv:0810.1304

qq, M = 2000 GeV



qq, M = 2000 GeV


 k_t C/A anti- k_t SIScone C/A-filt

 R = 0.8
 $Q_{f=z}^W$ $Q_{f=x\sqrt{M}}^W$ x 2

 rebin = 2
 qq gg

 mass = 2000

 pileup: none 0.05 0.25 mb^{-1}/ev

 subtraction:
 k_t C/A anti- k_t SIScone C/A-filt

 R = 0.3
 $Q_{f=z}^W$ $Q_{f=x\sqrt{M}}^W$ x 2

 rebin = 2
 qq gg

 mass = 2000

 pileup: none 0.05 0.25 mb^{-1}/ev

 subtraction:

This page is intended to help visualize how the choice of jet definition impacts a dijet invariant mass reconstruction at LHC.

The controls fall into 4 groups:

- the jet definition
- the binning and quality measures
- the jet-type (quark, gluon) and mass scale
- pileup and subtraction

The events were simulated with Pythia 6.4 (DWT tune) and reconstructed with FastJet 2.3.

For more information, view and listen to the **flash demo**, or click on individual terms.

This page has been tested with Firefox v2 and v3, IE7, Safari v3, Opera v9.5, Chrome 0.2.

How about task of resolving separate jets
from separate partons?

Illustrate in context of boosted $H \rightarrow b\bar{b}$
reconstruction

- ▶ Signal is $W \rightarrow \ell\nu$, $H \rightarrow b\bar{b}$.
- ▶ Backgrounds include $Wb\bar{b}$, $t\bar{t} \rightarrow \ell\nu b\bar{b}jj$, ...

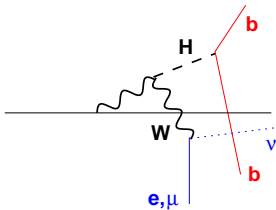
Studied e.g. in ATLAS TDR

Difficulties, e.g.

- ▶ $gg \rightarrow t\bar{t}$ has $\ell\nu b\bar{b}$ with **same intrinsic mass scale**, but much higher partonic luminosity
- ▶ Need exquisite control of bkgd shape

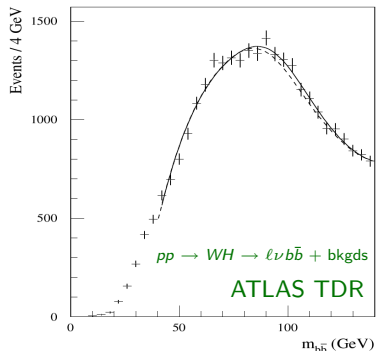
Try a long shot?

- ▶ Go to high p_t ($p_{tH}, p_{tV} > 200$ GeV)
- ▶ Lose 95% of signal, but more efficient?
- ▶ Maybe kill $t\bar{t}$ & gain clarity?



- ▶ Signal is $W \rightarrow \ell\nu, H \rightarrow b\bar{b}$.
- ▶ Backgrounds include $Wb\bar{b}, t\bar{t} \rightarrow \ell\nu b\bar{b}jj, \dots$

Studied e.g. in ATLAS TDR

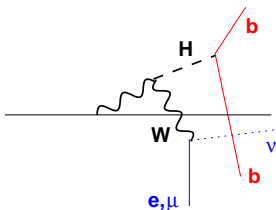


Difficulties, e.g.

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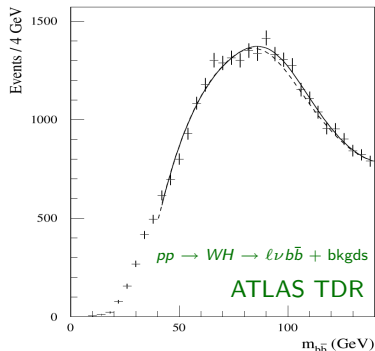
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Studied e.g. in ATLAS TDR

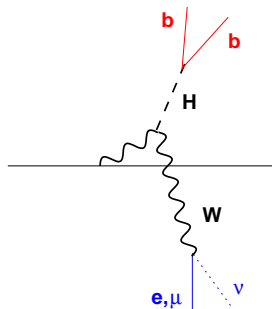


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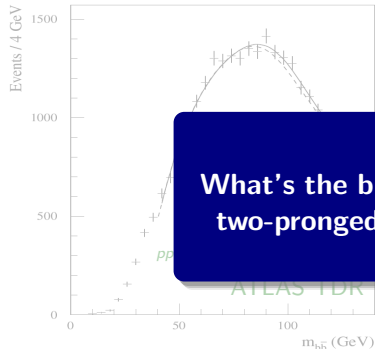
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- ▶ Signal is $W \rightarrow \ell\nu, H \rightarrow b\bar{b}$.
- ▶ Backgrounds include $Wb\bar{b}, t\bar{t} \rightarrow \ell\nu b\bar{b}jj, \dots$

Studied e.g. in ATLAS TDR

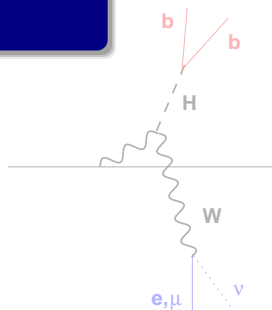


Difficulties, e.g.

- ▶ $gg \rightarrow t\bar{t}$ has $\ell\nu b\bar{b}$ with same intrinsic partonic

Question:
 What's the best strategy to identify the two-pronged structure of the boosted Higgs decay?

kgd shape



- Try a long shot?**
- ▶ Go to high p_t ($p_{tH}, p_{tV} > 200$ GeV)
 - ▶ Lose 95% of signal, but more efficient?
 - ▶ Maybe kill $t\bar{t}$ & gain clarity?

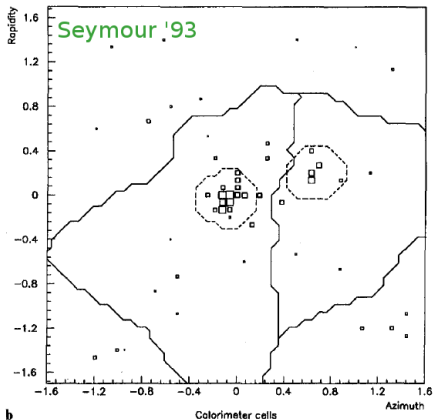
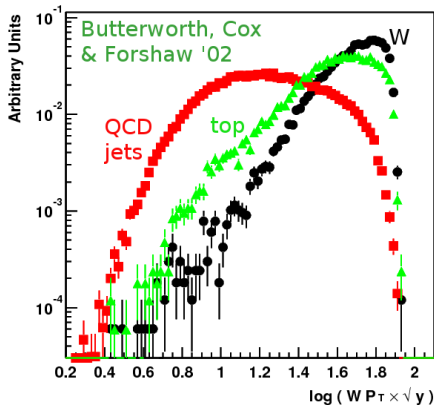


Fig. 2. A hadronic W decay, as seen at calorimeter level, **a** without, and **b** with, particles from the underlying event. Box sizes are logarithmic in the cell energy, lines show the borders of the sub-jets for infinitely soft emission according to the cluster (solid) and cone (dashed) algorithms

Use k_t jet-algorithm's hierarchy to split the jets



Use k_t alg.'s distance measure (rel. trans. mom.) to cut out QCD bkgd:

$$d_{ij}^{k_t} = \min(p_{ti}^2, p_{tj}^2) \Delta R_{ij}^2$$

Y-splitter

only partially correlated with mass

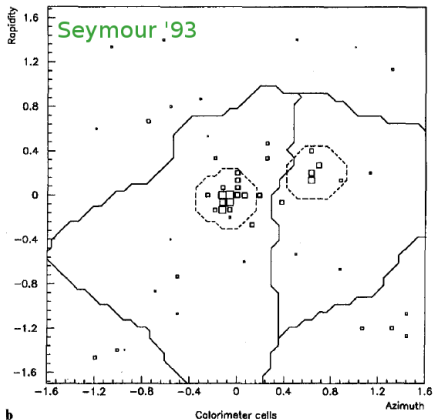
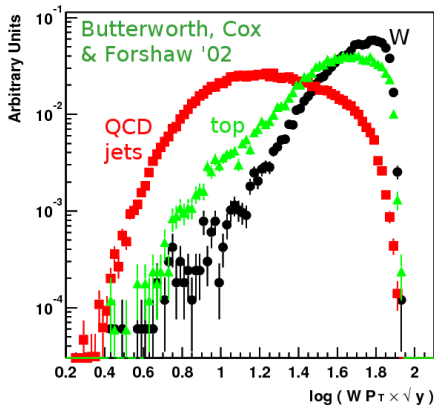


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Y-splitter

only partially correlated with mass

The Cambridge/Aachen jet alg.

Dokshitzer et al '97
Wengler & Wobisch '98

Work out $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2$ between all pairs of objects i, j ;

Recombine the closest pair;

Repeat until all objects separated by $\Delta R_{ij} > R$.

[in FastJet]

Gives “hierarchical” view of the event; work through it backwards to analyse jet

The Cambridge/Aachen jet alg.

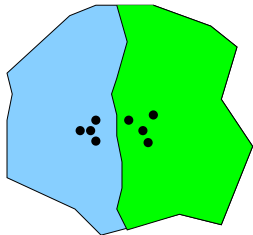
Dokshitzer et al '97
Wengler & Wobisch '98

*Work out $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2$ between all pairs of objects i, j ;
Recombine the closest pair;
Repeat until all objects separated by $\Delta R_{ij} > R$.*

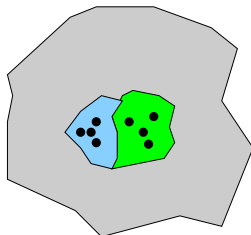
[in FastJet]

Gives “hierarchical” view of the event; work through it backwards to analyse jet

k_t algorithm



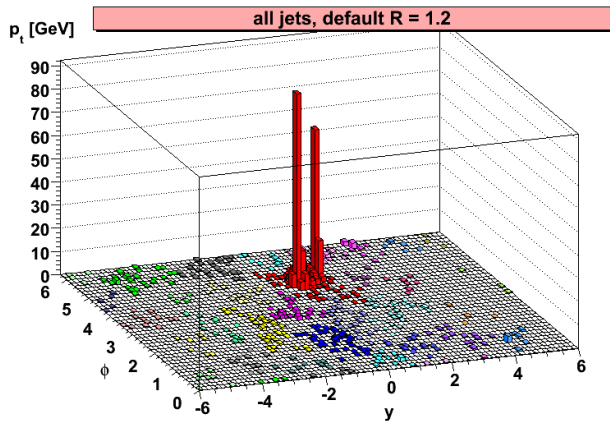
Cam/Aachen algorithm



Allows you to “dial” the correct R to keep perturbative radiation, but throw out UE

SIGNAL

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



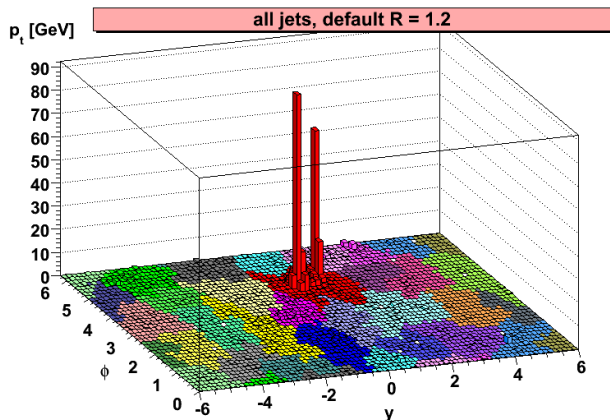
Zbb BACKGROUND

Cluster event, C/A, R=1.2

arbitrary norm.

SIGNAL

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



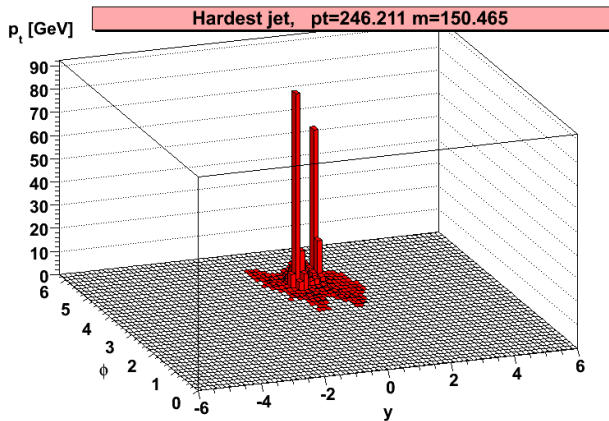
Zbb BACKGROUND

Fill it in, \rightarrow show jets more clearly

arbitrary norm.

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

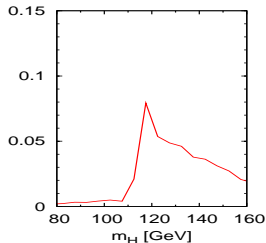
Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



Consider hardest jet, $m = 150$ GeV

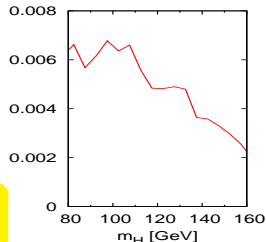
SIGNAL

$200 < p_{tZ} < 250$ GeV



Zbb BACKGROUND

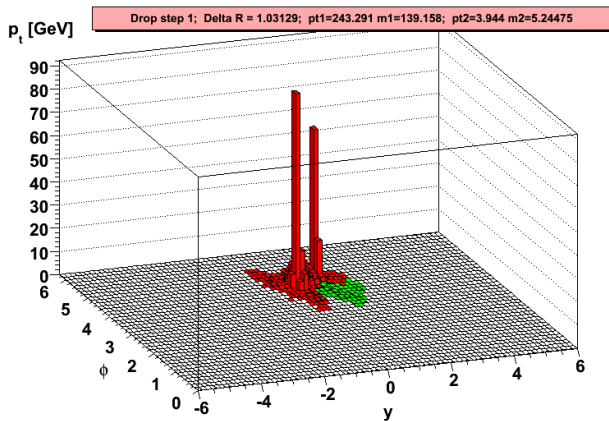
$200 < p_{tZ} < 250$ GeV



arbitrary norm.

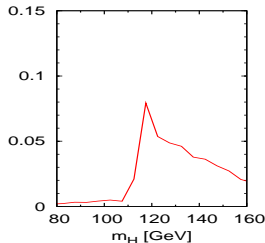
$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}, @14\text{ TeV}, m_H = 115\text{ GeV}$$

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



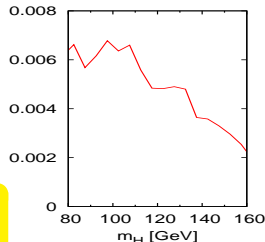
SIGNAL

$200 < p_{tZ} < 250\text{ GeV}$



Zbb BACKGROUND

$200 < p_{tZ} < 250\text{ GeV}$

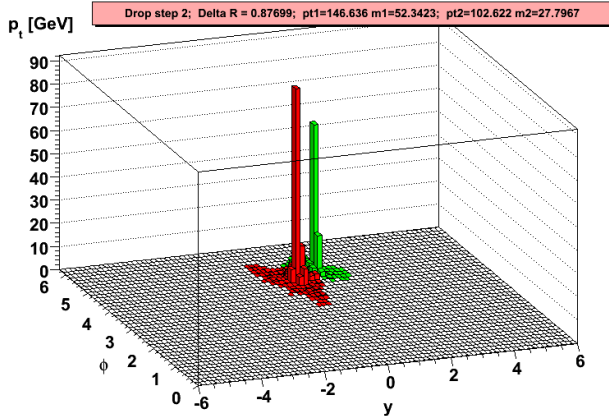


split: $m = 150\text{ GeV}, \frac{\max(m_1, m_2)}{m} = 0.92 \rightarrow \text{repeat}$

arbitrary norm.

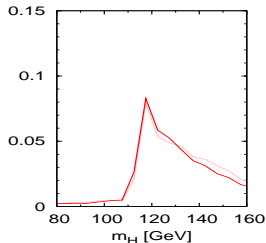
$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}, @14\text{ TeV}, m_H = 115\text{ GeV}$$

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



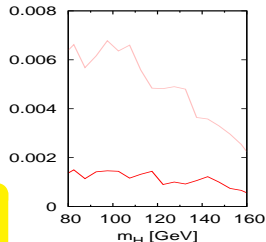
SIGNAL

$200 < p_{tZ} < 250\text{ GeV}$



Zbb BACKGROUND

$200 < p_{tZ} < 250\text{ GeV}$

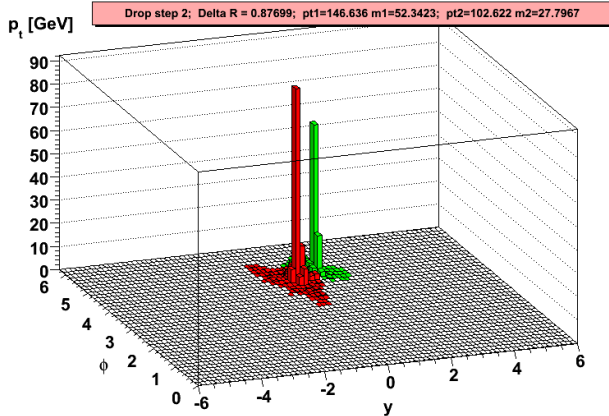


split: $m = 139\text{ GeV}$, $\frac{\max(m_1, m_2)}{m} = 0.37 \rightarrow$ mass drop

arbitrary norm.

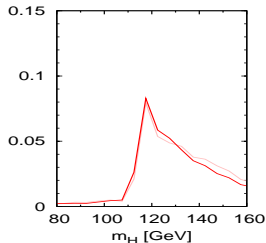
$$pp \rightarrow ZH \rightarrow \nu\bar{\nu}b\bar{b}, @14\text{ TeV}, m_H = 115\text{ GeV}$$

Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



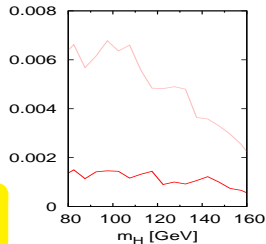
SIGNAL

$200 < p_{tZ} < 250\text{ GeV}$



Zbb BACKGROUND

$200 < p_{tZ} < 250\text{ GeV}$

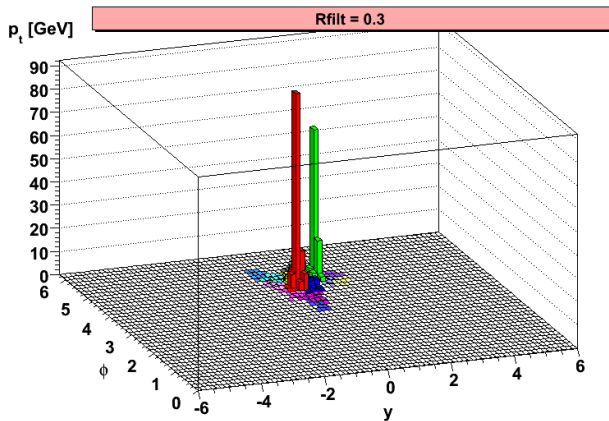


check: $y_{12} \simeq \frac{p_{t2}}{p_{t1}} \simeq 0.7 \rightarrow \text{OK} + 2\text{ } b\text{-tags (anti-QCD)}$

arbitrary norm.

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

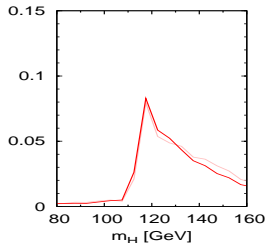
Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



$$R_{filt} = 0.3$$

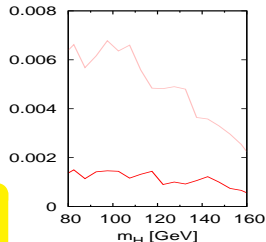
SIGNAL

$200 < p_{tZ} < 250$ GeV



Zbb BACKGROUND

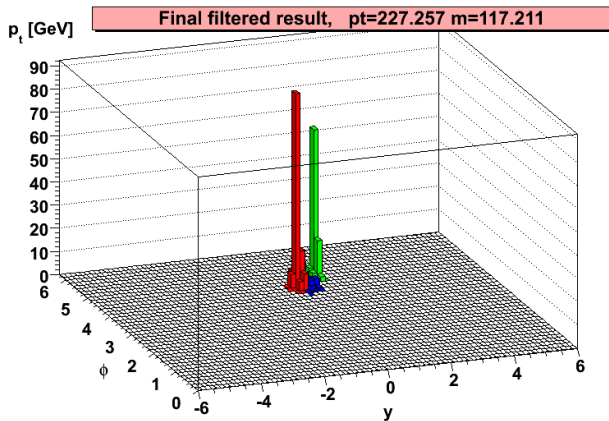
$200 < p_{tZ} < 250$ GeV



arbitrary norm.

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

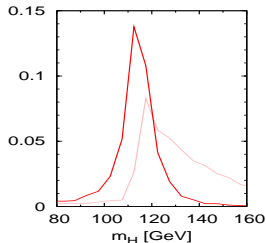
Herwig 6.510 + Jimmy 4.31 + FastJet 2.3



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

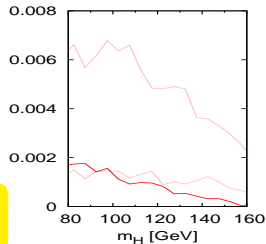
SIGNAL

$200 < p_{tZ} < 250$ GeV



Zbb BACKGROUND

$200 < p_{tZ} < 250$ GeV

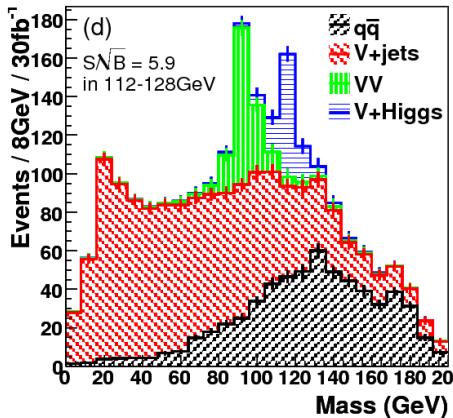


arbitrary norm.

Cross section for signal and the Z +jets background in the leptonic Z channel for $200 < p_{TZ}/\text{GeV} < 600$ and $110 < m_J/\text{GeV} < 125$, with perfect b -tagging; shown for our jet definition (C/A MD-F), and other standard ones close to their optimal R values.

Jet definition	σ_S/fb	σ_B/fb	$S/\sqrt{B \cdot \text{fb}}$
C/A, $R = 1.2$, MD-F	0.57	0.51	0.80
k_t , $R = 1.0$, y_{cut}	0.19	0.74	0.22
SISCone, $R = 0.8$	0.49	1.33	0.42
anti- k_t , $R = 0.8$	0.22	1.06	0.21

Combined HZ + HW, $p_t > 200$ GeV



- ▶ Take $Z \rightarrow \ell^+ \ell^-$, $Z \rightarrow \nu \bar{\nu}$,
 $W \rightarrow \ell \nu$ $\ell = e, \mu$
- ▶ $p_{tV}, p_{tH} > 200$ GeV
- ▶ $|\eta_V|, |\eta_H| < 2.5$
- ▶ Assume real/fake b -tag rates of 0.7/0.01.
- ▶ Some extra cuts in HW channels to reject $t\bar{t}$.
- ▶ Assume $m_H = 115$ GeV.

At $\sim 5\sigma$ for 30 fb^{-1} this looks like a competitive channel for light Higgs discovery. **Deserves serious exp. study!**

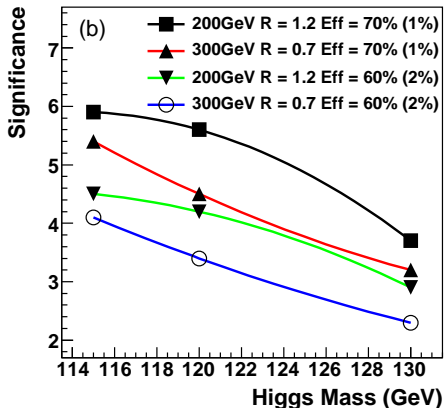
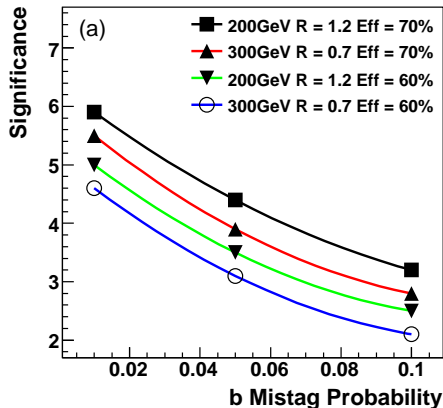
Conclusions

- ▶ There are no longer any valid excuses for using jet algorithms that are incompatible with the Snowmass criteria.
 - LHC experiments are adopting the new tools
 - Individual analyses need to follow suit
- ▶ It's time to move forwards with the question of how best to use jets in searches
- ▶ Examples here show two things:
 - ▶ Good jet-finding brings significant gains
 - ▶ There's room for serious QCD theory input into optimising jet use
 - Not the *only* way of doing things
 - But brings more insight than trial & error MC

This opens the road towards *Jetography*, QCD-based autofocus for jets

EXTRAS

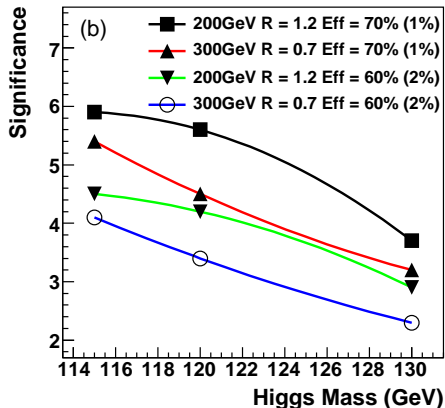
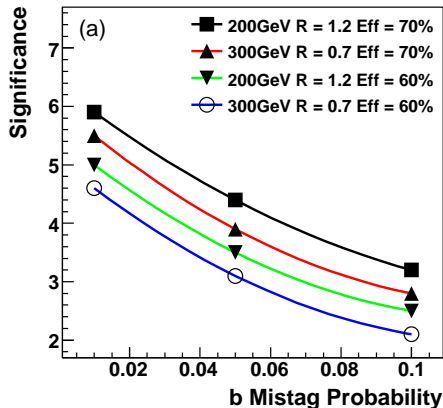
Impact of b -tagging, Higgs mass



Most scenarios above 3σ

For it to be a significant discovery channel requires decent b -tagging, lowish mass Higgs [and good experimental resolution]

In nearly all cases, looks feasible for extracting WH , ZH couplings



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