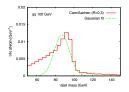
Quantifying the performance of jet definitions for kinematic reconstruction at the LHC M. Cacciari, J. Rojo, G. P. Salam, G. Soyez arXiv: 0810.1304v1 [hep-ph] 7 Oct 2008

> Olga Bessidskaia and Nicolo' Vladi Biesuz HASCO summer school Goettingen

> > 20th July 2012

Aim

- There exist many jet definitions and reconstruction algorithms.
- What is the best jet definition for kinematic reconstructions?
- Example: search for a narrow resonance from a new boson.
- The distribution of the reconstructed mass after parton shower is often non-Gaussian:



• Use physical parameters rather than unphysical Monte Carlo partons to find a quantitive measure of jet performance.

Analysis

The basic idea

- $\bullet\,$ Parton pairs produce ficticious bosons with a narrow resonance width of $1\,{\rm GeV}$
- The boson decays, producing quarks or gluons, which hadronize
- The hadrons are modelled as jets
- The mass of the boson is reconstructed
- Z' bosons are associated with jets from quarks: q ar q o Z' o q ar q
- H bosons are associated with jets from gluons: $gg \rightarrow H \rightarrow gg$
- We scan the masses of the ficitious bosons Z', H in mass range 100 GeV 2 TeV. We show only a few examples.

Jet Definition (1)

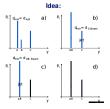
For k_t algorithm we define:

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

 d_{ij} - distance between the jets p_{ti}, p_{tj} - jet transverse momenta $\Delta R_{ij}^2 = (\phi_i - \phi_j)^2 + (y_i - y_j)^2$ - jet distance in azimuthal-rapidity plane between the 2 jets R - the minimum jet separation we impose ("focus parameter")

Then the algorithm proceeds;

- Evaluate d_{ij} for every couple of particles;
- **2** Find smallest of d_{ij} ;
- If $d_{ij} > d_{iB}$, stop clustering. Otherwise recombine i and j;
- Repeat from step 1;



• • = • • = •

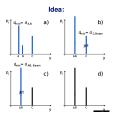
Jet definition (2)

For $anti - k_t$ algorithm we define:

$$d_{ij} = \min\left(\frac{1}{(p_{ti})^2}, \frac{1}{(p_{tj})^2}\right) \frac{\Delta R_{ij}}{R}; \quad d_{iB} = \frac{1}{(p_{ti})^2}$$
 (6)

Then the algorithm proceeds;

- Evaluate d_{ij} for every couple of particles;
- **2** Find smallest of d_{ij} ;
- If $d_{ij} > d_{iB}$, stop clustering. Otherwise recombine i and j;
- Repeat from step 1;



< ∃ ►

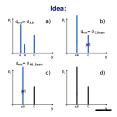
Jet definition (2)

For $anti - k_t$ algorithm we define:

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

Then the algorithm proceeds;

- Evaluate d_{ij} for every couple of particles;
- **2** Find smallest of d_{ij} ;
- 3 If $d_{ij} > d_{iB}$, stop clustering. Otherwise recombine i and j;
- Repeat from step 1;

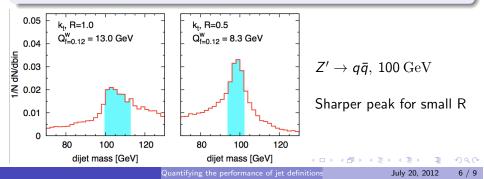


 K_t algorithm starts recombining the clusters from the smallest jets. Anti – K_t algorithm starts recombining the clusters from the biggest jets.

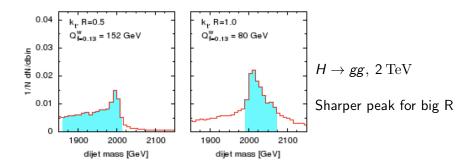
Figure of merit

$Q^{\rm w}_{\rm f=z}:$ a measure of the performance of jet definitions

- If two peaks have similar numbers of events, the **narrower** one **is better**.
- $Q_{f=z}^w$: the width of the smallest reconstructed mass window that contains a fraction f of the generated massive objects
- $Q_{\mathrm{f}=z}^{\mathrm{w}}$ has a dependence on R
- Low value of $Q_{f=z}^{w}$ = narrower peak = better jet definition

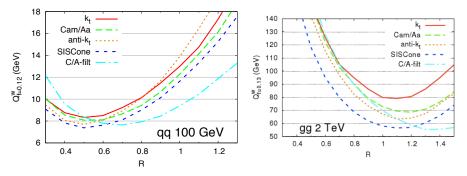


Results 1



< 3 >

Results 2



- For low-energetic quark jets (\sim 100 GeV), small values of $R~(\sim 0.5)$ are preferred
- For high-energetic gluon jets ($\sim 2~{\rm TeV}),$ larger values of $R~(\sim 1.2)$ are preferred
- Production of **secondary gluons** highest for high-energetic gluon pairs larger *R* required

Conclusions

- Definition of a data-driven physical quantitative measure $Q^{\rm w}_{f=z}$ characterizing the sharpness of a peak
- Use $Q_{f=z}^{w}$ to describe the performance of a jet definition
- Well adapted for non-Gaussian reconstructed masses
- Sharpness of the peak is related to the sensitivity of a measurement
- An optimal choice of R can increase the discovery potential of new particles