

Fast clustering of jets

Grégory Soyez

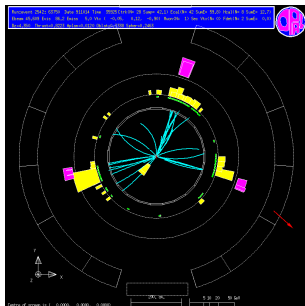
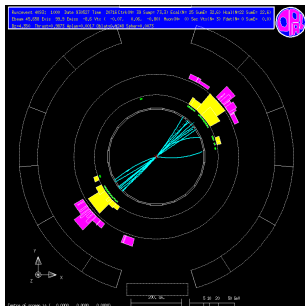
IPhT, CEA Saclay

GDR QCD — progress in algorithms and numerical tools
May 16 2017

Brief plan

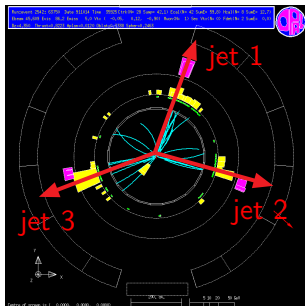
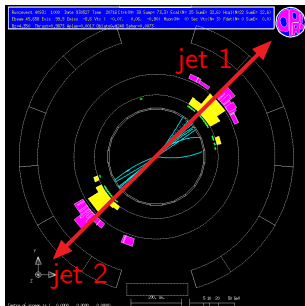
- Brief intro about the physics concepts: what are jets
- Clustering algorithms: Cambridge/Aachen, k_t , anti- k_t
- Main part: Nearest-neighbours and fast clustering
- If time left: Enumerating circles

- Final-state events are pencil-like
already observed in e^+e^- collisions:



- Consequence of the collinear divergence
QCD (quark & gluon) branching proba: $\frac{dP}{d\theta} \propto \frac{\alpha_s}{\theta}$

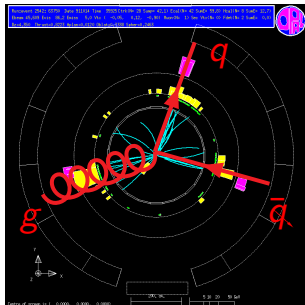
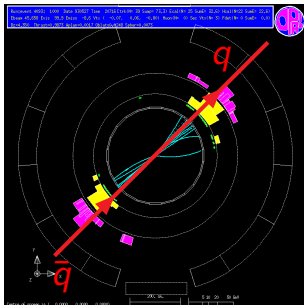
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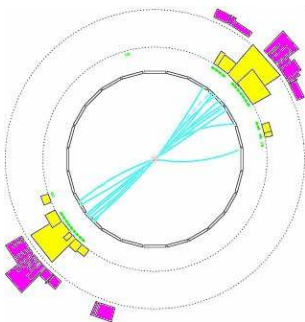
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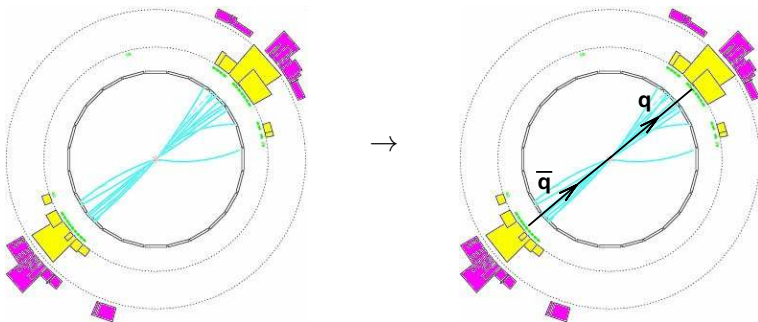
How many jets?



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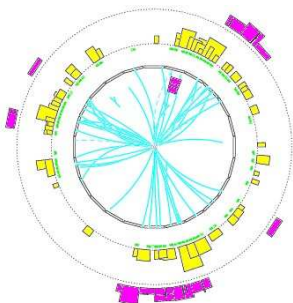
obviously 2 jets



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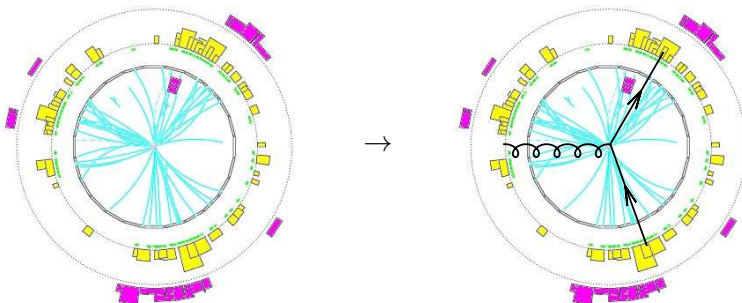
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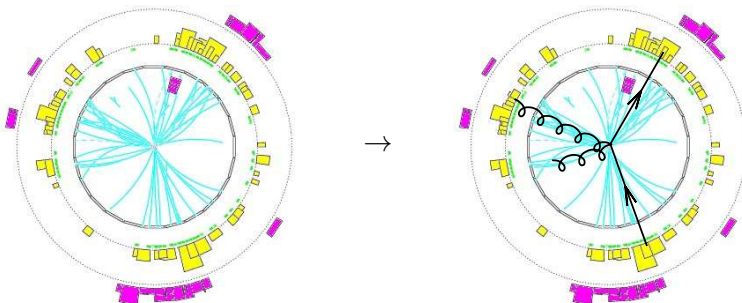
3 jets



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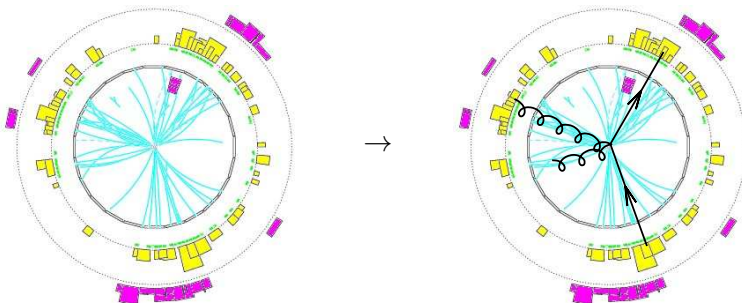
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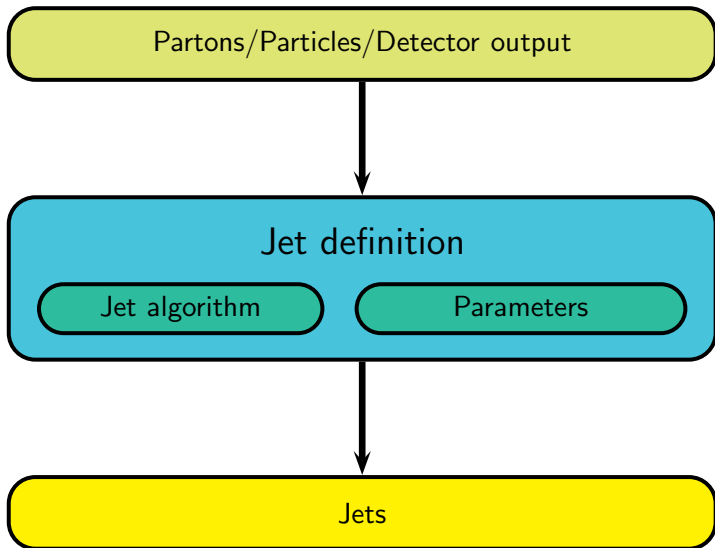
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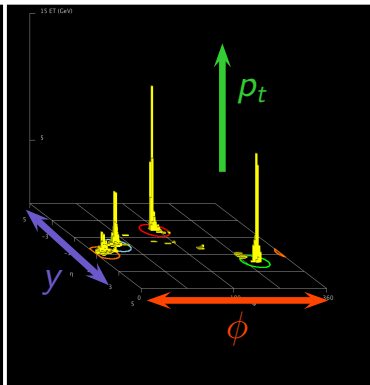
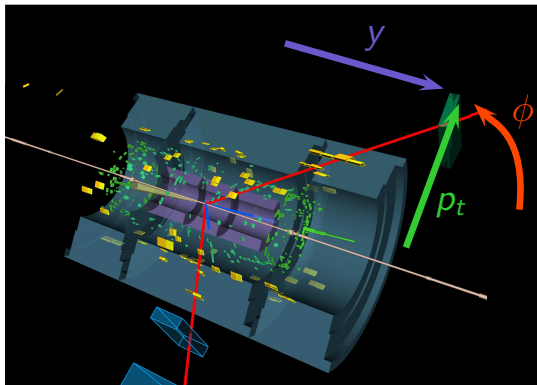
- “collinear” is arbitrary
- “parton” concept strictly valid only at LO

Jet definition



A bit of useful kinematics

[Both: ATLAS public events ($H \rightarrow 2\mu 2e$ & 4 jets)]



- Rapidity y : longitudinal component (along the beam axis)
- Azimuthal angle ϕ : around the beam axis
- Transverse momentum p_t : “energy” transverse to the beam

2 big approaches to jet clustering

- 30 years of history and debates
- All introduce a parameter R
 - ▶ “Jet radius”
 - ▶ distance of “collinearity” in $y - \phi$
- 2 big categories:
 - ▶ find circles (cones) containing flows of energy
 - ▶ undo a branching process by successive pairwise recombinations
- See Gavin’s review from 2009 for details

Most common approach today: recombination algorithms

Generalised- k_t algorithm

- From all the objects to cluster, define the distances

$$d_{ij} = \min(p_{t,i}^{2p}, p_{t,j}^{2p})(\Delta y_{ij}^2 + \Delta \phi_{ij}^2), \quad d_{iB} = p_{t,i}^{2p} R^2$$

- repeatedly find the minimal distance
 - if d_{ij} : recombine i and j into $k = i + j$
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- repeatedly find the minimal distance
 - if d_{ij} : recombine i and j into $k = i + j$
 - if d_{iB} : call i a jet
- Parameter p is (typically) one of
 - ▶ $p = 1$: k_t algorithm (closest to QCD)
[Catani,Dokshitzer,Seymour,Weber,Ellis,Soper,1993]
 - ▶ $p = 0$: Cambridge/Aachen (geometrical distance)
[Dokshitzer,Leder,Moretti,Webber,1997]
 - ▶ $p = -1$: anti- k_t (the LHC choice) [M.Cacciari,G.Salam,GS,2008]

Main question for today

1. Cambridge/Aachen

Given a set of N points (with weights p_t) in a plane
($y - \phi$) repeatedly find the closest pair

2. (anti-) k_t

Same, but use (anti)- k_t distance for measuring closeness

Things to keep in mind

- N is in the 1000-50000 range \Rightarrow look at large N
- stopping distance R

Geometrical (Camb./Aachen) case: Naive approach

compute all d_{ij}	N^2
find minimum	N^2
recombine $i + j$	1
iterate	$\times N$
total	$\mathcal{O}(N^3)$

- works for all algs
- prohibitively slow

Observations:

- No need to keep track of all the distances:

$$\min_{i,j}\{d_{ij}\} = \min_i\{d_{i,NN(i)}\} \quad \text{with} \quad NN(i) = \min_j\{d_{ij}\}$$

only keep track of the **nearest neighbour (NN)** of each particle

- Do not recalculate all NNs at each step; if $i + j \rightarrow k$, we need $NN(k)$ and $NN(\ell)$ when $NN(\ell) = i$ or j

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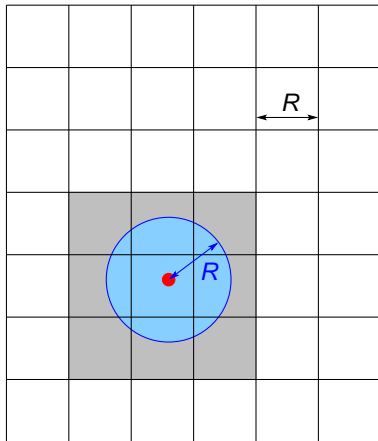
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New implementation:

Init: compute all $NN(i)$	N^2
find smallest $d_{i,NN(i)}$	N
recombine $i + j$	1
compute $NN(k)$ and $NN(\ell)$'s	N
iterate	$\times N$
total	$\mathcal{O}(N^2)$

- works for all algs
- efficient for N not too large

Geometrical (Camb./Aachen) case: Tiling



- NN only in current or neighbouring tile
- $\Rightarrow NN$ search is $\mathcal{O}(n = N/N_{\text{tiles}})$

Geometrical (Camb./Aachen) case: Tiling

Init: create tiling	N
Init: compute all $NN(i)$	Nn
Init: sort the $d_{i,NN(i)}$	$N \log(N)$
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- Valid for Cambr./Aachen ($d_{ij} = \Delta R_{ij}^2$)
- Variants for finding $\min\{d_{i,NN(i)}\}$.
- Tricks to avoid neighbour tiles when possible

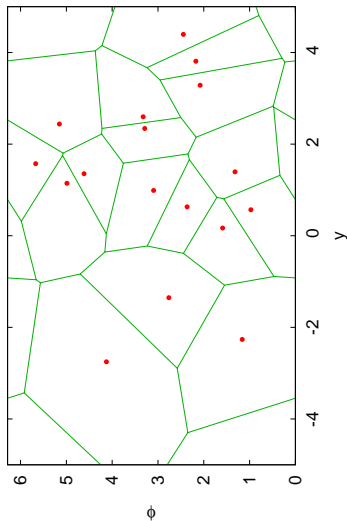
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- **Optimal for**
 $30 \lesssim N \lesssim 5 \cdot 10^5$

Still can do better

[M.Cacciari,G.Salam,2005]



- Voronoi graph: bisectors between pairs of points
- $NN(i)$ is one of the $\mathcal{O}(\log N)$ adjacent cells
- Construct: $\mathcal{O}(N \log N)$
- Add/remove: $\mathcal{O}(\log N)$

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FastJet lemma

If the pair (i, j) minimises d_{ij} and $p_{ti}^{2p} < p_{tj}^{2p}$,
then j is the geometrical NN of i .

Proof: Assume there is k s.t. $\Delta R_{ik} < \Delta R_{ij}$. We would have

$$\begin{aligned} d_{ik} &= \min(p_{ti}^{2p}, p_{tk}^{2p}) \Delta R_{ik}^2 \\ &< p_{ti}^{2p} \Delta R_{ij}^2 = d_{ij}, \end{aligned}$$

a contradiction.

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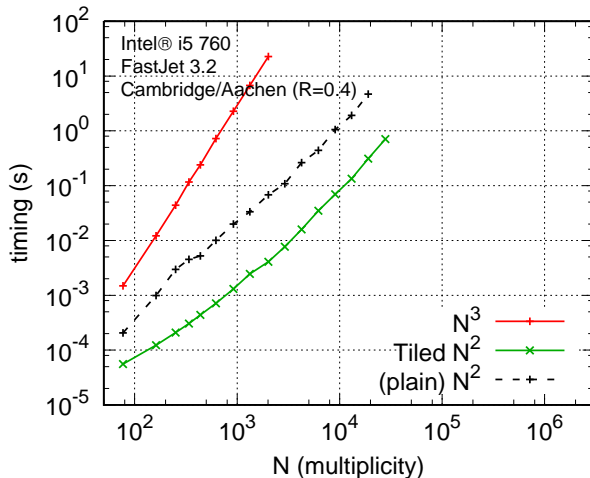
⇒ all the above strategy (working with geometrical NN) work

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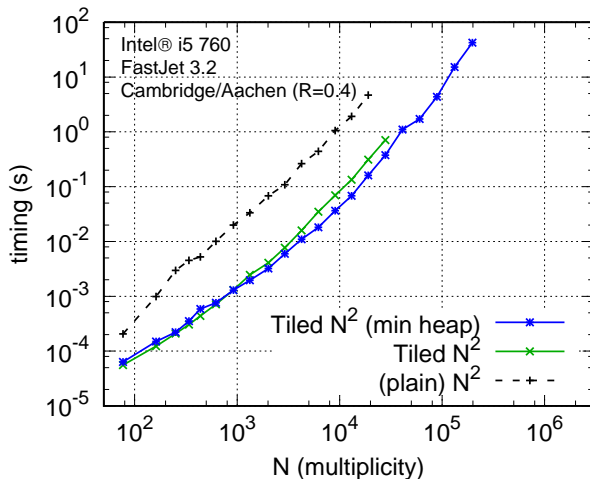
- Matteo Cacciari and Gavin Salam in 2005; I joined in 2008.
- <http://www.fastjet.fr>
- Software for fast jet clustering
- Now extended to reference software for jet clustering + manipulations
- Used by the whole LHC community

[M.Cacciari, G.Salam, 2005]
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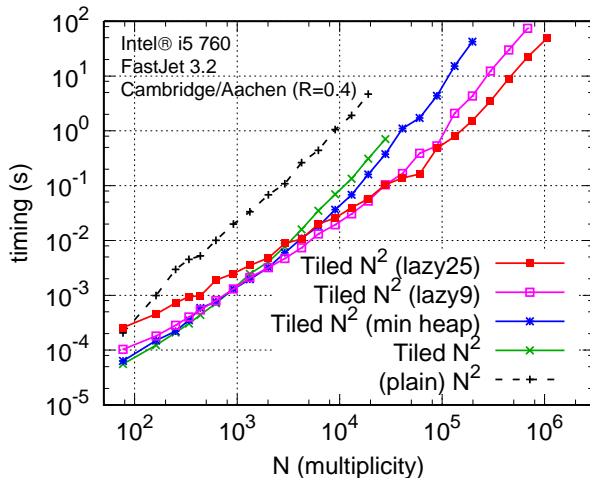
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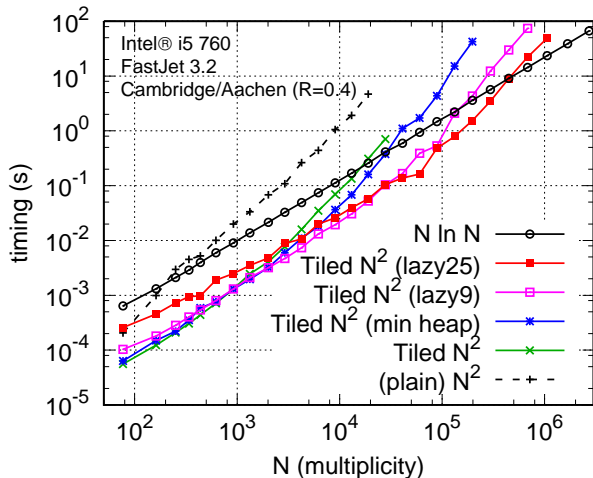
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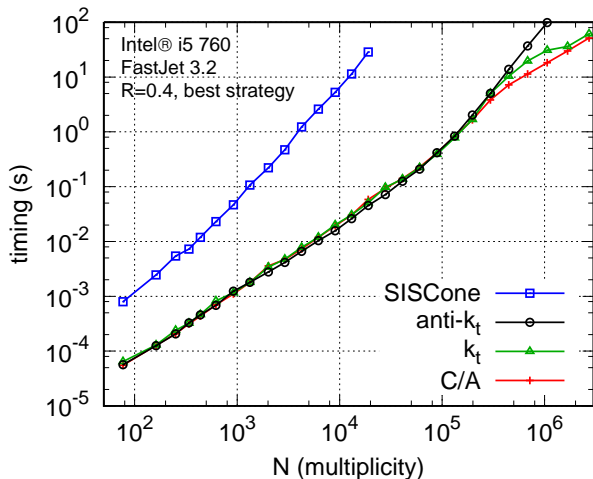


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New in FJ-3.1

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- CGAL, $N \ln N$



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- works for $k_t(N \ln N)$ and anti- $k_t(N^{3/2})$
- SISCone: see next slides
- at LHC: 1000× faster than “KtJet”

Stale cones and related problems

- Cone algorithm: find directions of energy flow

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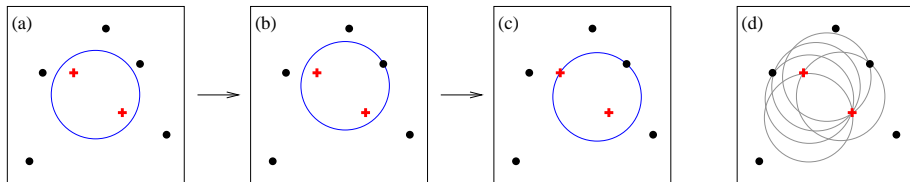
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 - ▶ reasonable speed ($\mathcal{O}(N^3)$)
 - ▶ Infrared (or collinear) unsafe \Rightarrow not good for theory

The SISCone solution

[G.Salam,GS,2007]

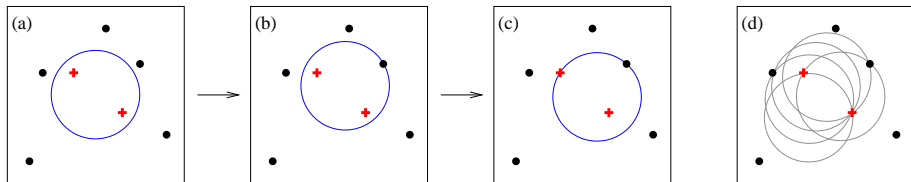


(a) start with a circle

see <https://siscone.hepforge.org/>

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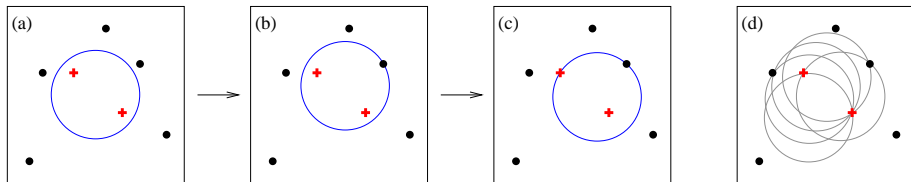
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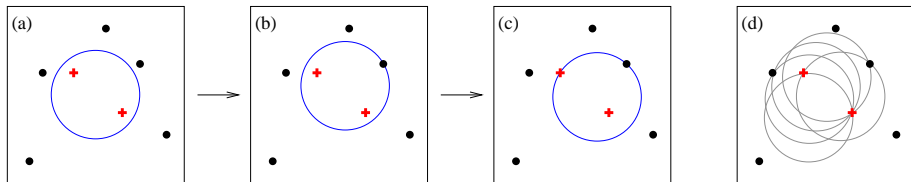
(c) it can be rotated until it touches a second

\Rightarrow enumerate circles by enumerating pairs of points
enumerate pairs: N^2 , check stability: $N \Rightarrow \mathcal{O}(N^3)$

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 \Rightarrow enumerate circles by enumerating pairs of points
enumerate pairs: N^2 , check stability: $N \Rightarrow \mathcal{O}(N^3)$

(d) order the circles in angle $N \Rightarrow \mathcal{O}(N^2 \ln N)$

see <https://siscone.hepforge.org/>

Conclusions

Geometrical constructions can help designing powerful algorithms

- ▶ tilings, Voronoi graphs for iterative clustering
- ▶ circles enumeration for cone algorithms

Thank You!