

New Jet Methods for High-Multiplicity Environments

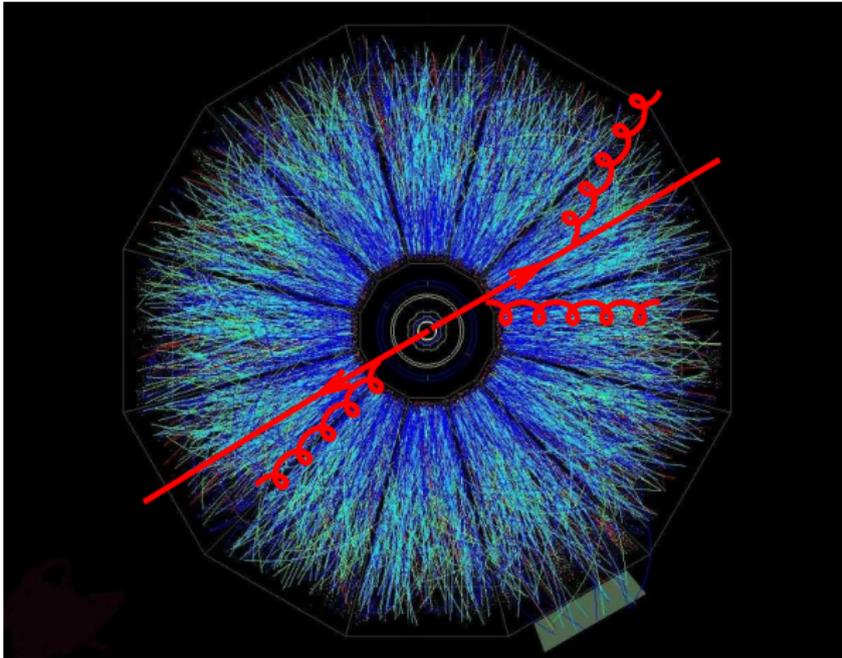
Gavin P. Salam

LPTHE, UPMC Paris 6 & CNRS

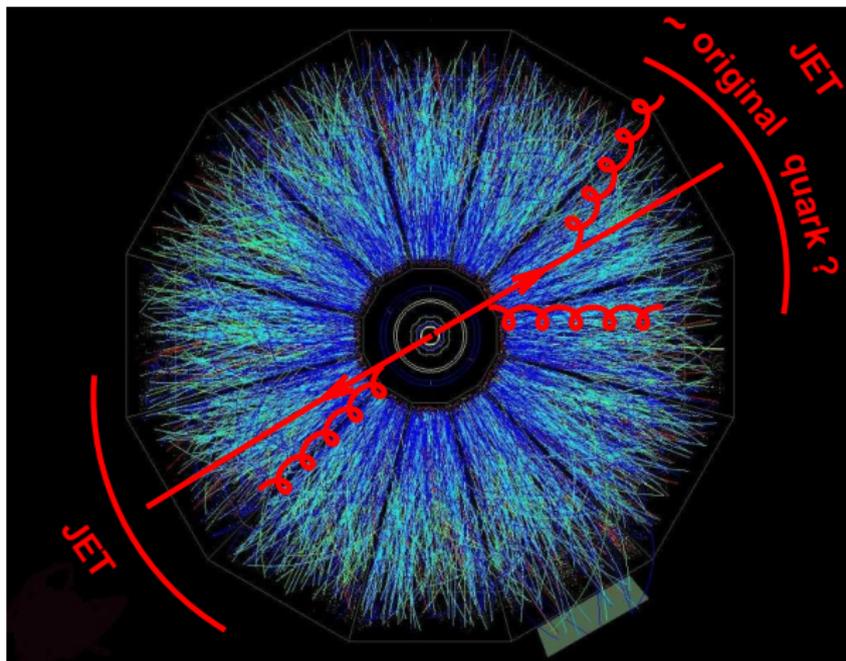
APS April Meeting

Washington D.C., 14 February 2010

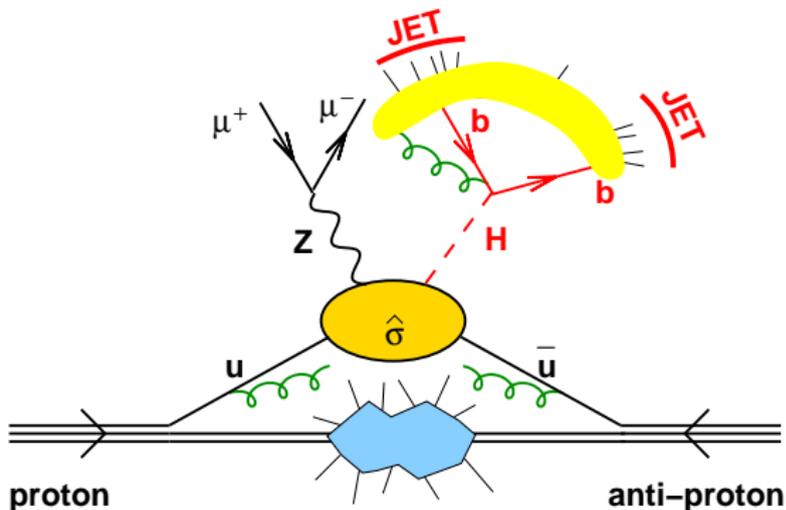
Based on work (some preliminary) with
Matteo Cacciari, Juan Rojo, Sebastian Sapeta, Gregory Soyez



Radiation from high-momentum quarks & gluons traversing hot medium can tell us about the medium



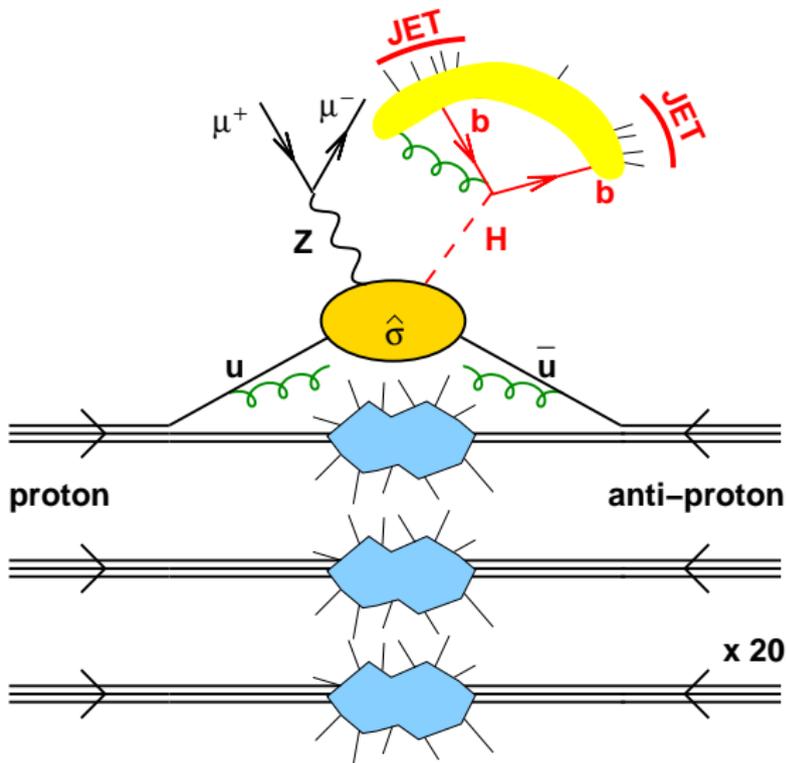
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Use jets to reconstruct quarks from decay of some new heavy object

e.g. a Higgs boson

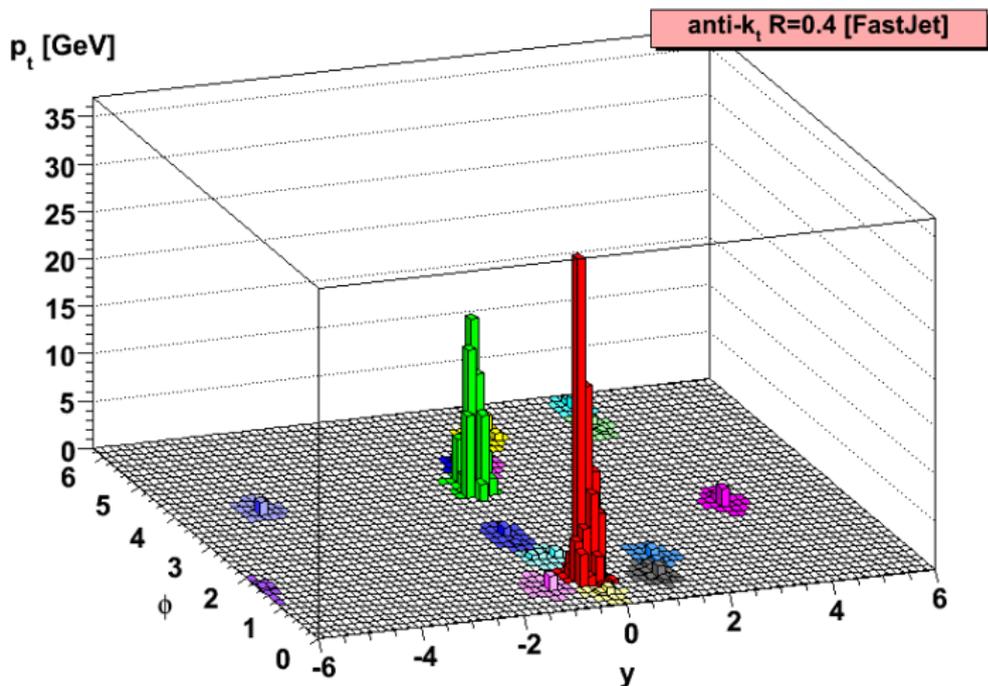
At high luminosity, many simultaneous pp collisions – not unlike AuAu/PbPb collision



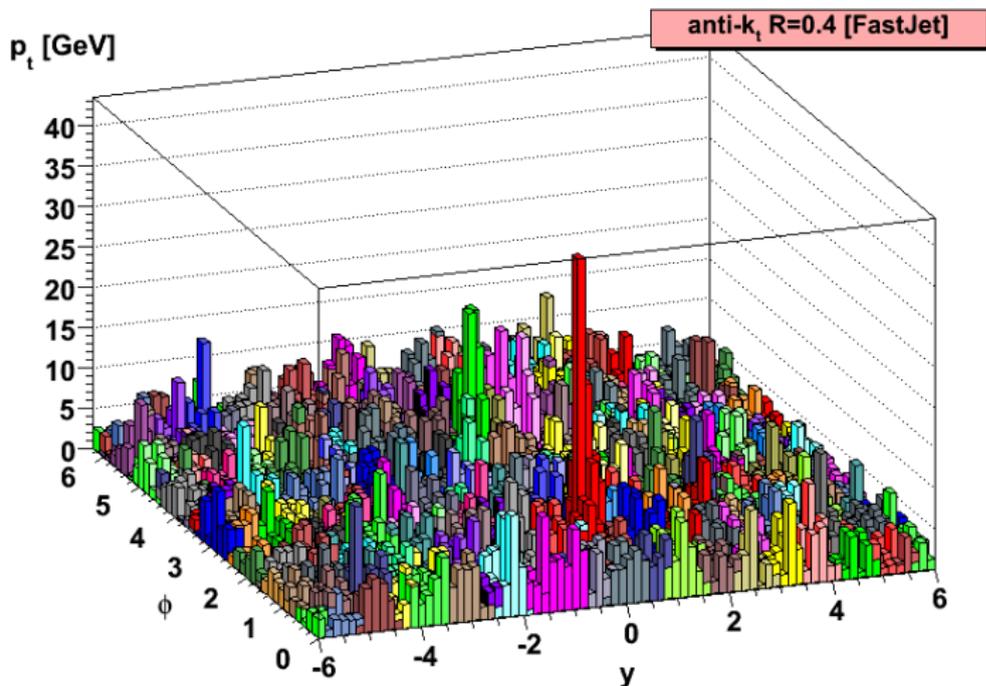
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A pp event (LHC 5.5 TeV, Pythia)



Contamination in jet

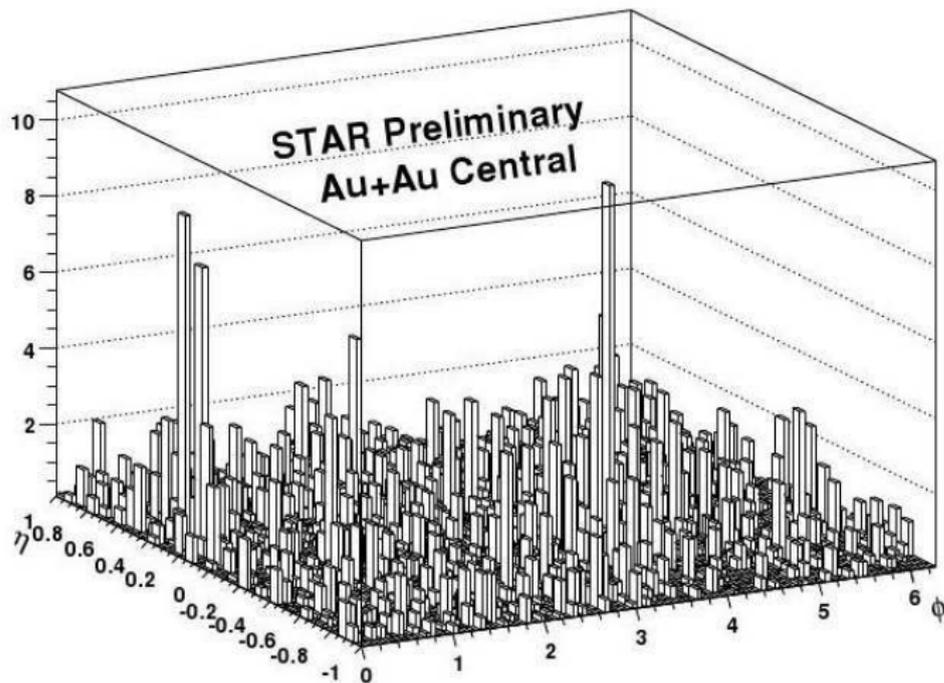
RHIC AuAu:
 $\mathcal{O}(40 \text{ GeV})$

LHC PbPb:
 $\mathcal{O}(100 \text{ GeV})$

LHC pp
(hi-lumi)
 $\mathcal{O}(5 - 40 \text{ GeV})$

A pp event (LHC 5.5 TeV, Pythia), embedded in a HI collision background (Hydjet 1.5)

Common challenge: large contamination



Contamination in jet

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LHC pp
(hi-lumi)
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A pp event (LHC 5.5 TeV, Pythia), **embedded in a HI collision background (Hydjet 1.5)** and an actual STAR event

What are ingredients of jet finding in noisy environments?

1. Jets
2. Jet areas
3. Noise estimation
4. Noise subtraction
- [5. Noise suppression]

A jet algorithms provides a mapping:

particles \longrightarrow jets
jet.def.



Simplest pp jet algorithm is
“Cambridge/Aachen”

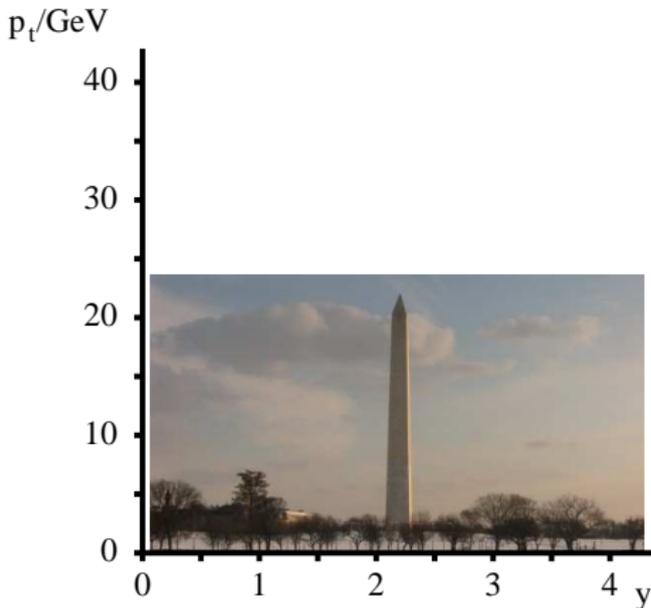
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of objects, until all separated by
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R parameter sets angular resolution

ϕ assumed 0 for all towers



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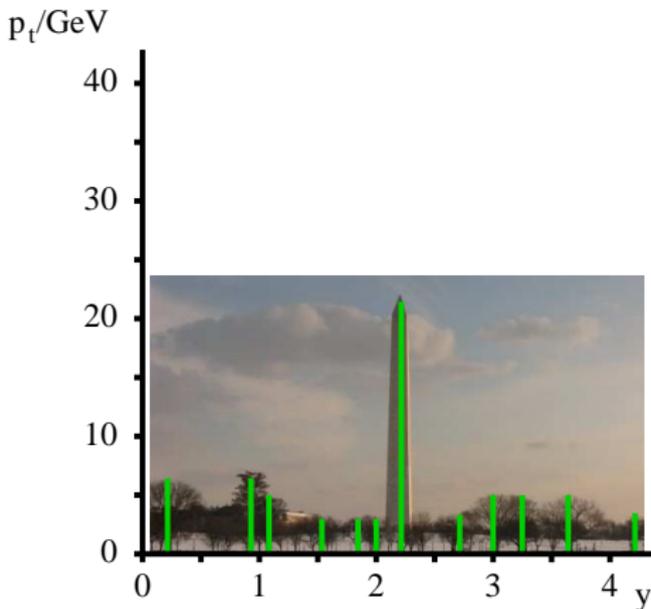
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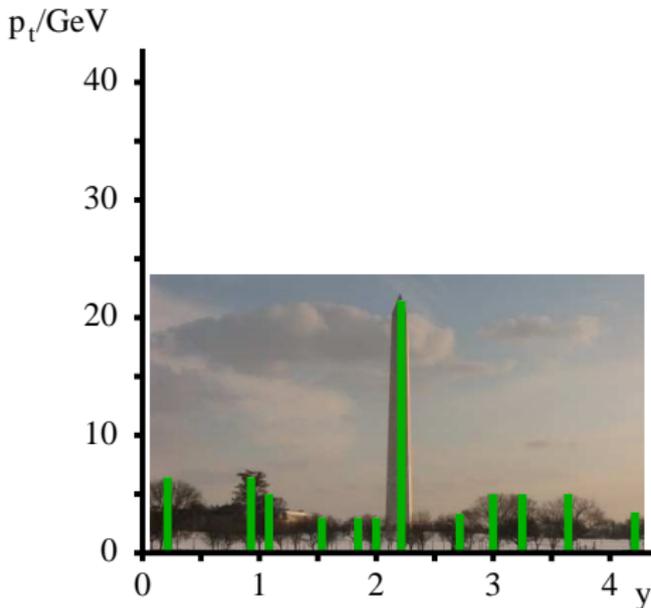
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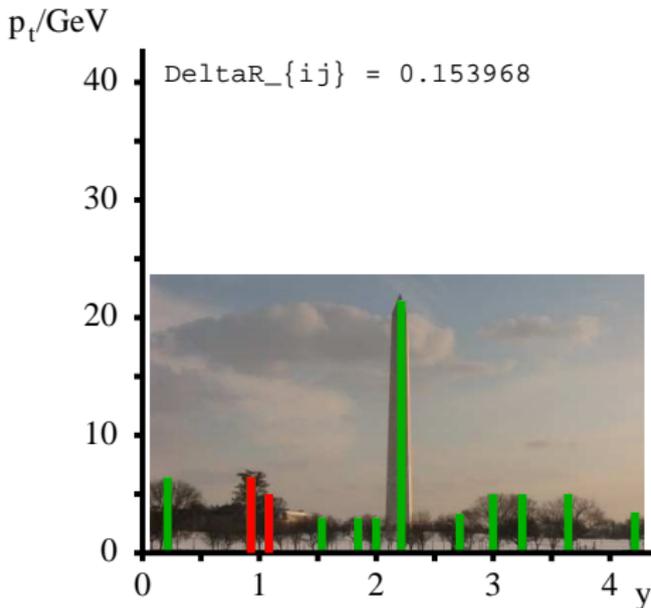
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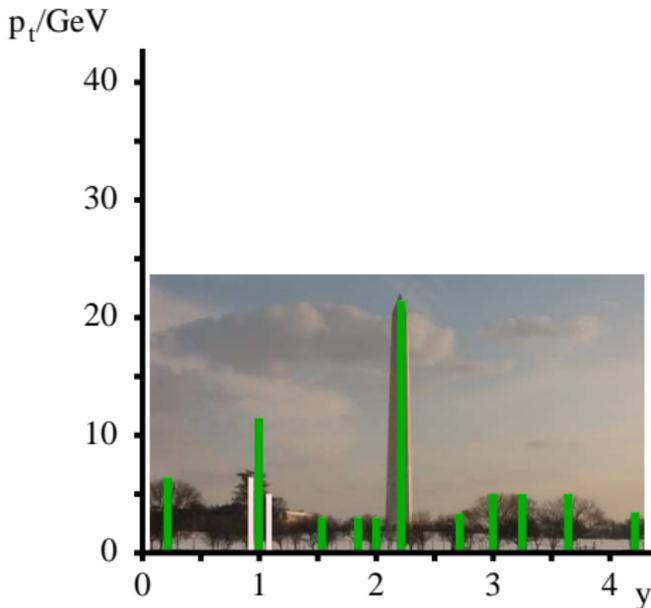
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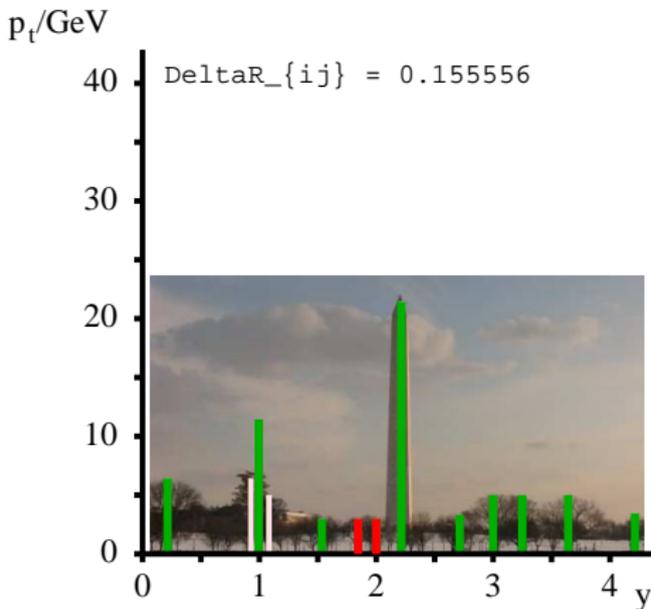
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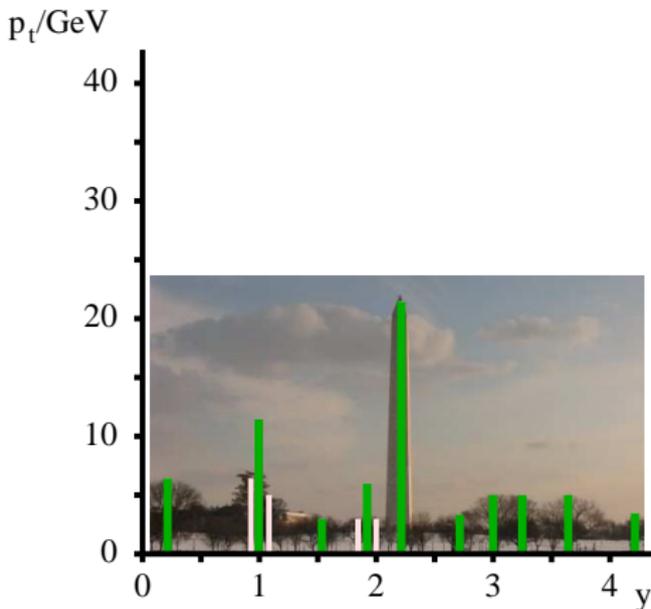
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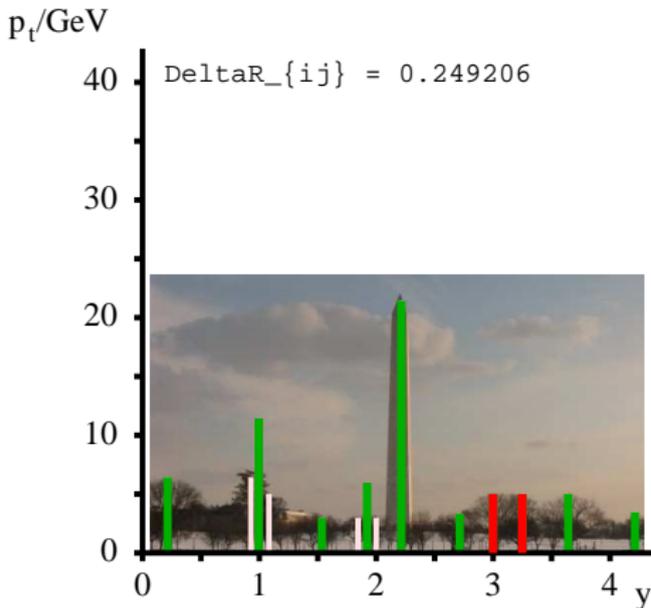
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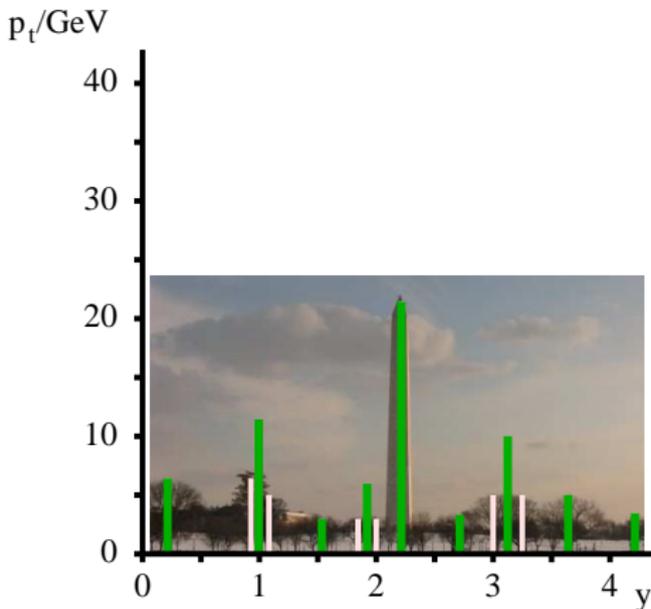
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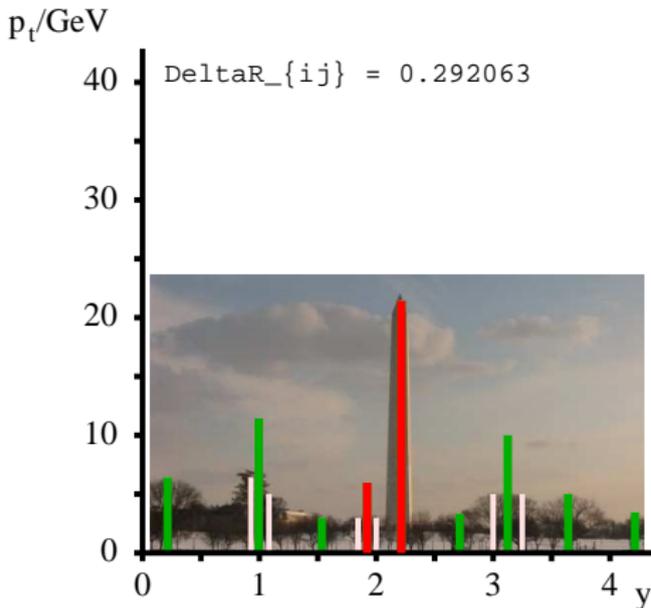
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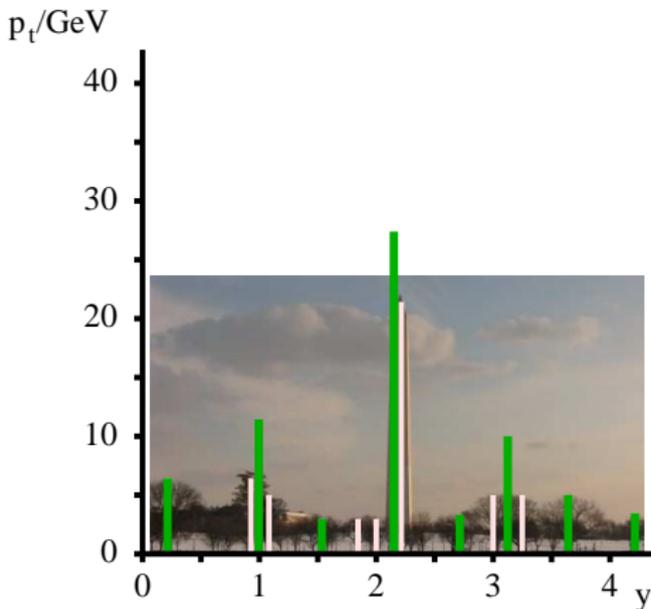
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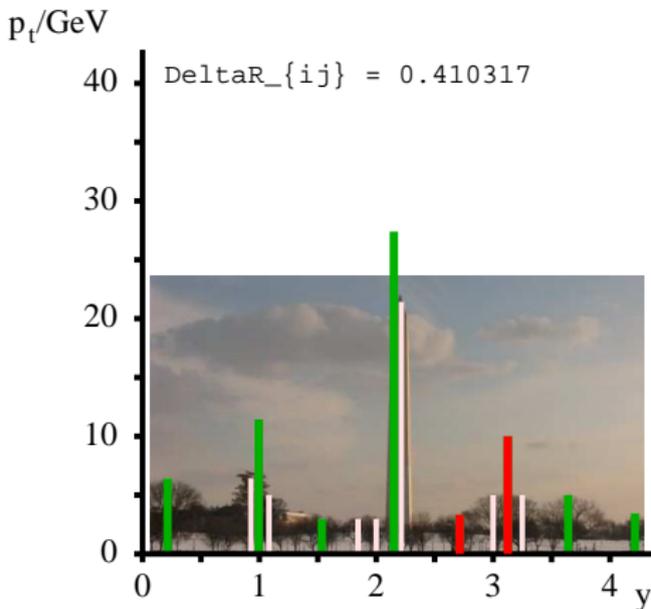
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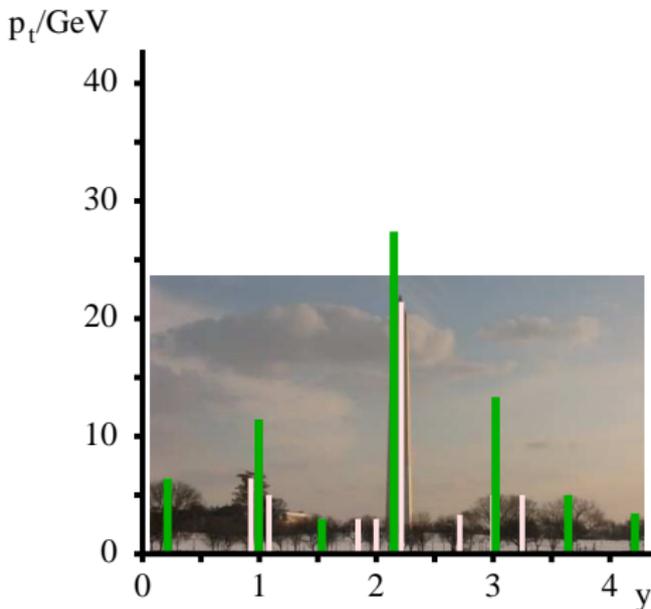
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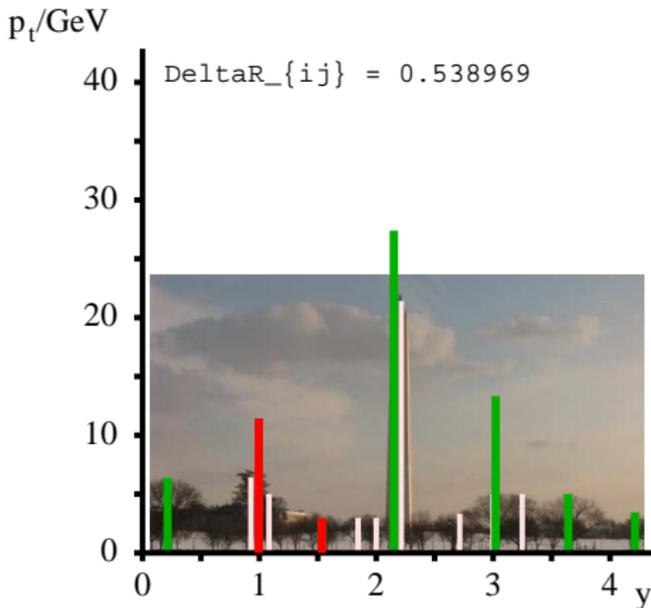
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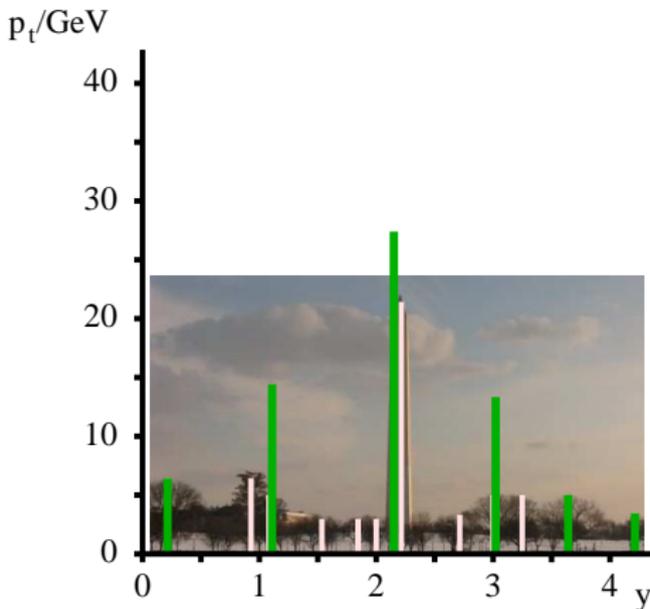
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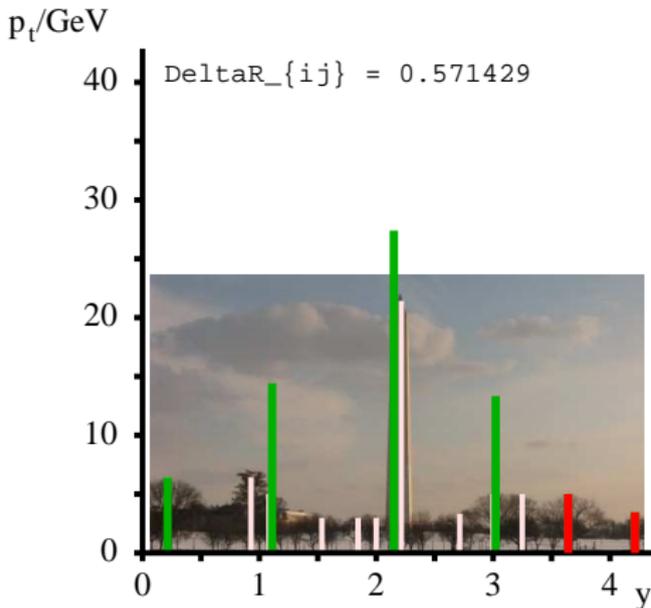
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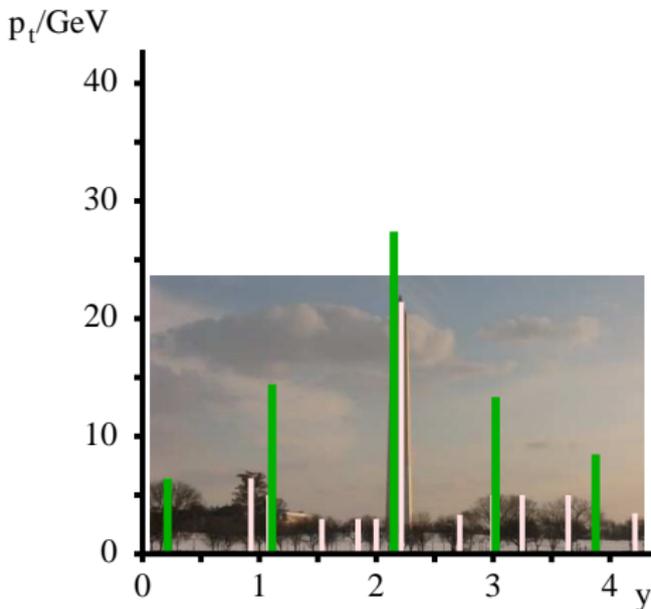
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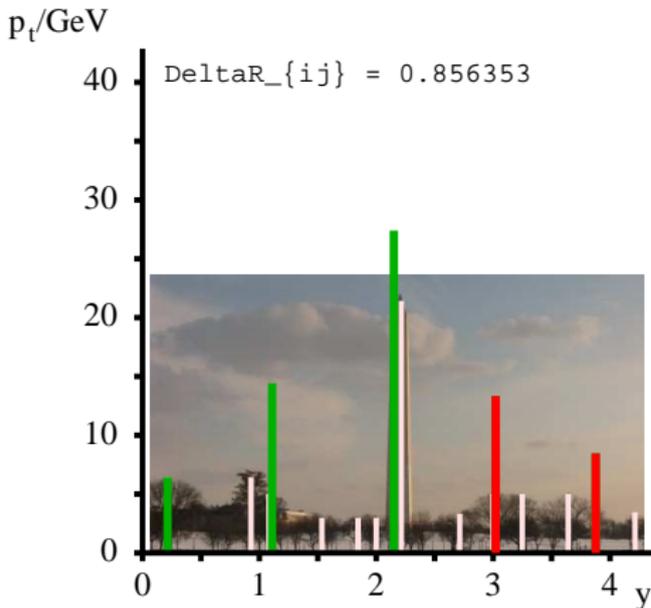
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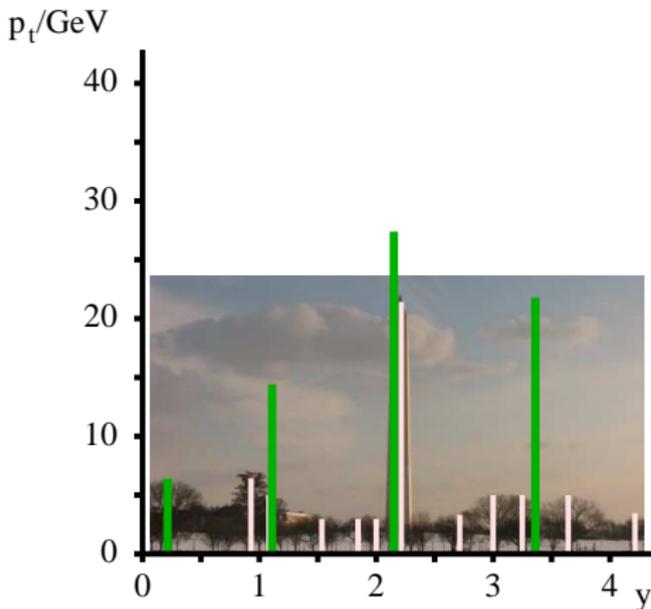
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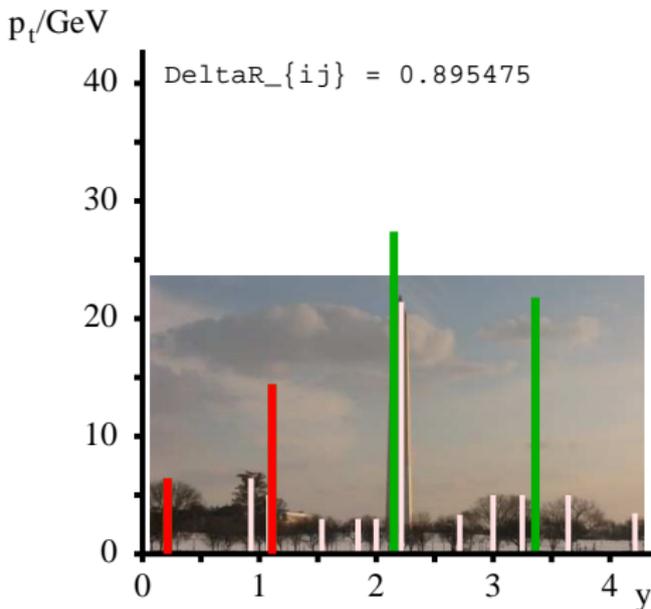
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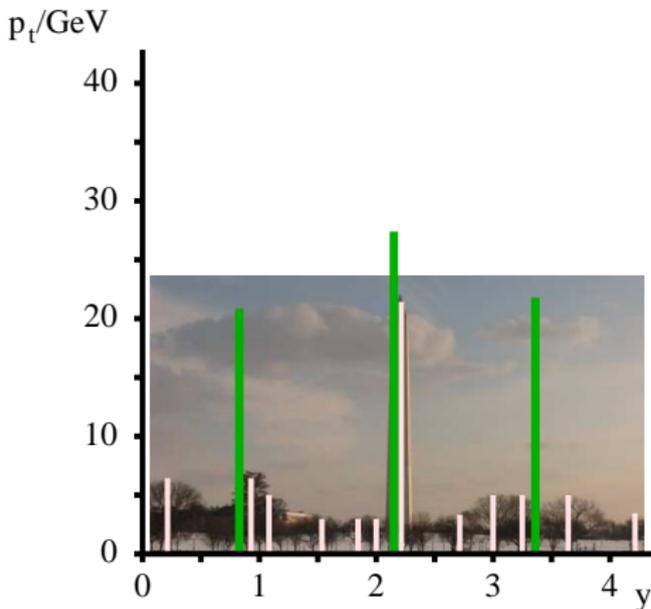
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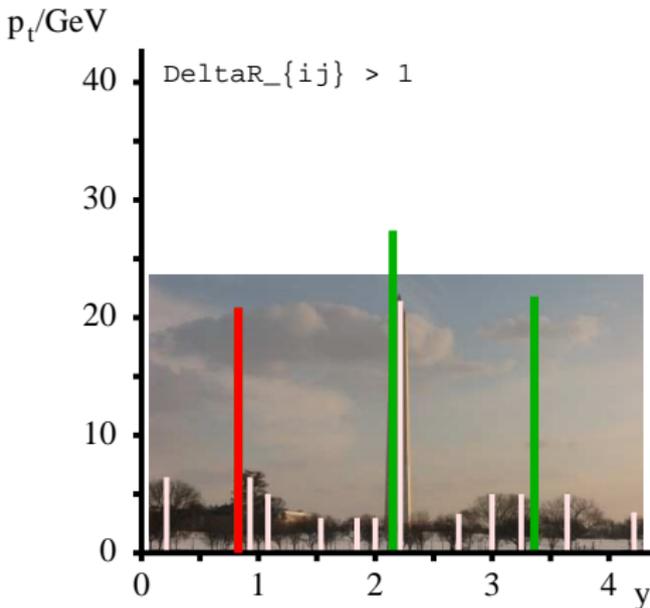
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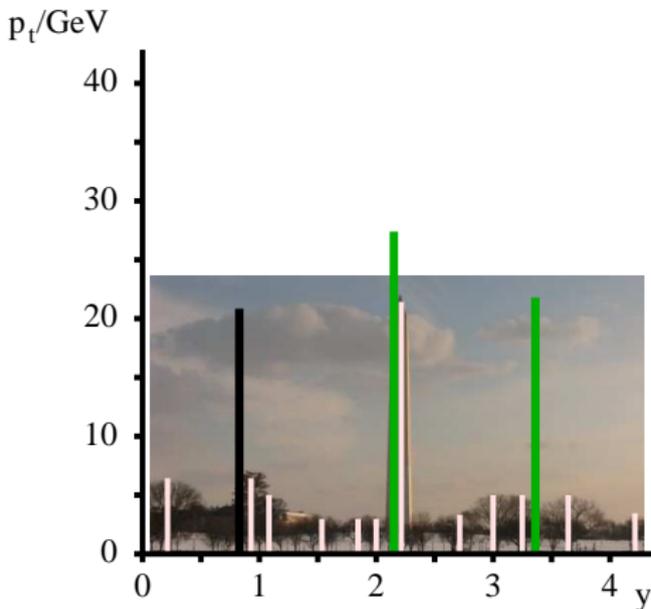
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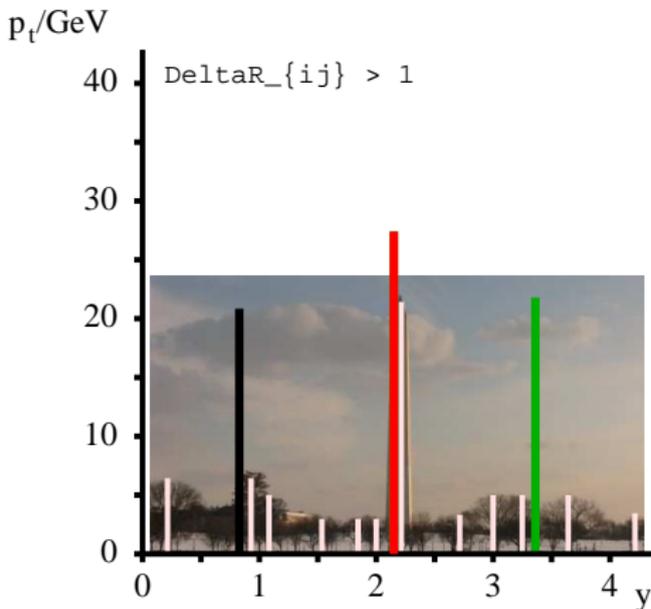
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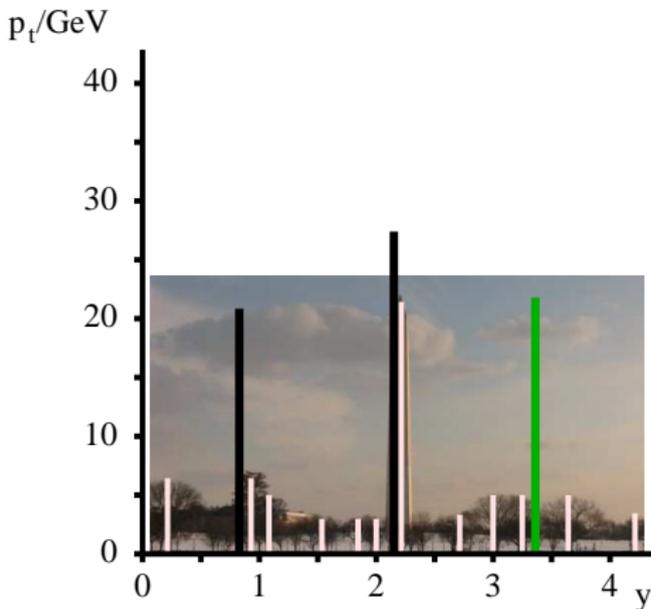
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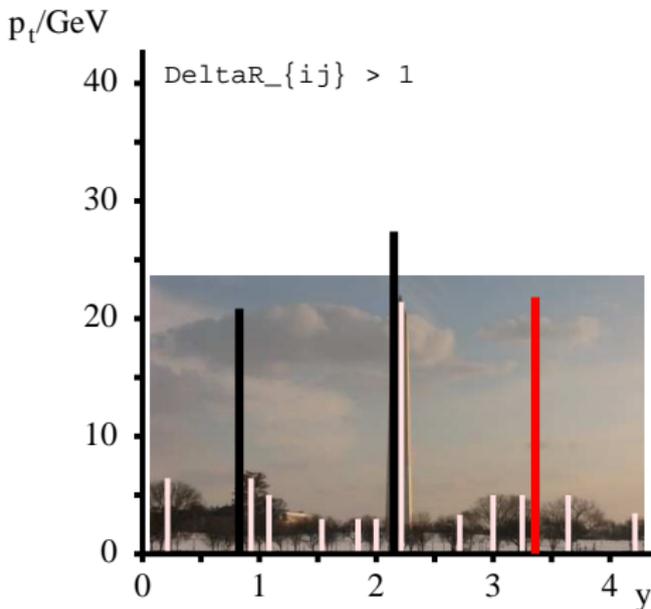
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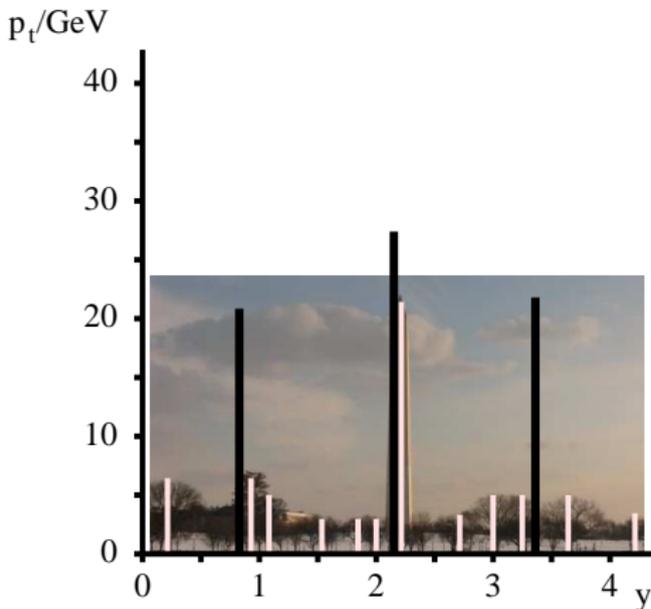
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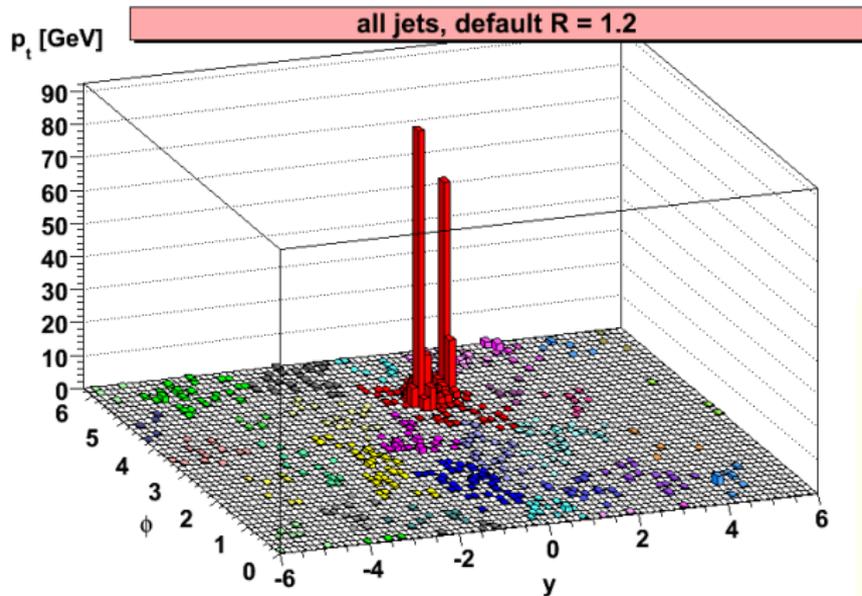
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Jets are made of finite number of pointlike particles.

Area not unambiguous concept

Jet areas must be defined



Add many soft particles to event

10^{-100} GeV each

$A \propto \#$ inside jet

Cacciari, GPS & Soyez '08

measure of jet's susceptibility to contamination from soft radiation

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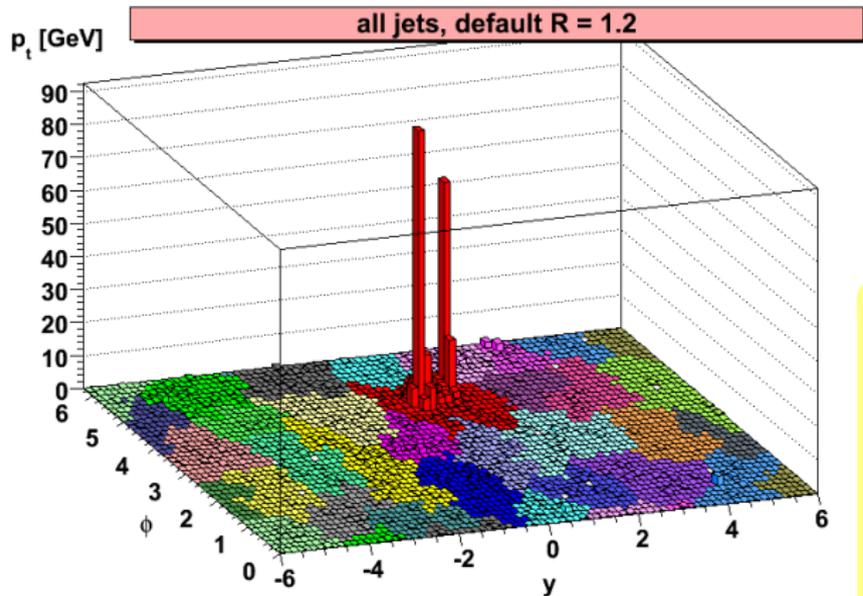
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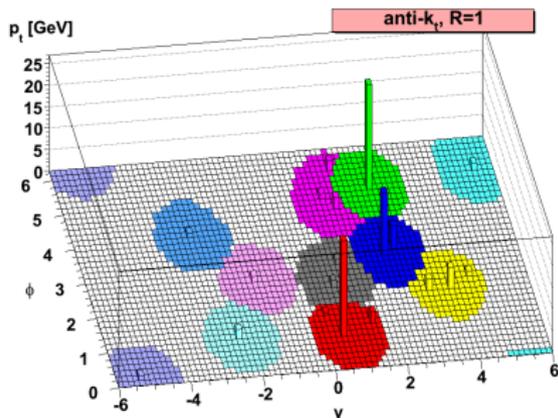
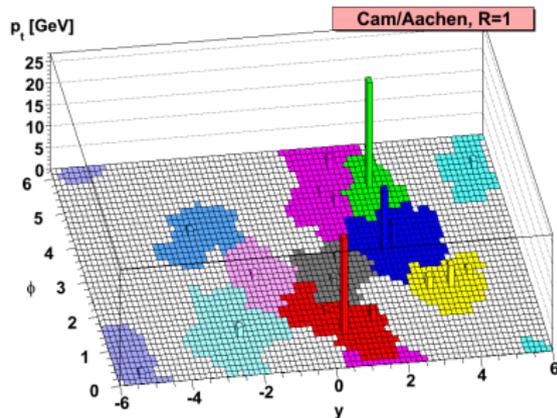
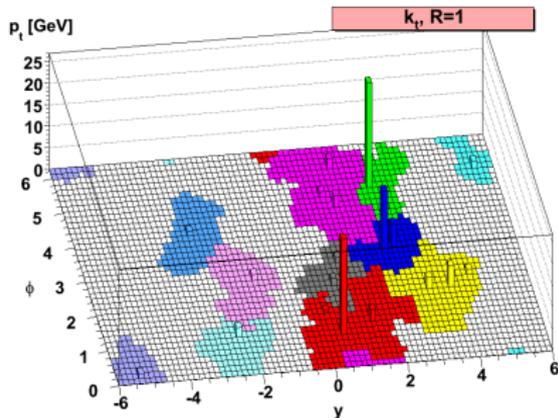
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Areas for 3 jet algorithms

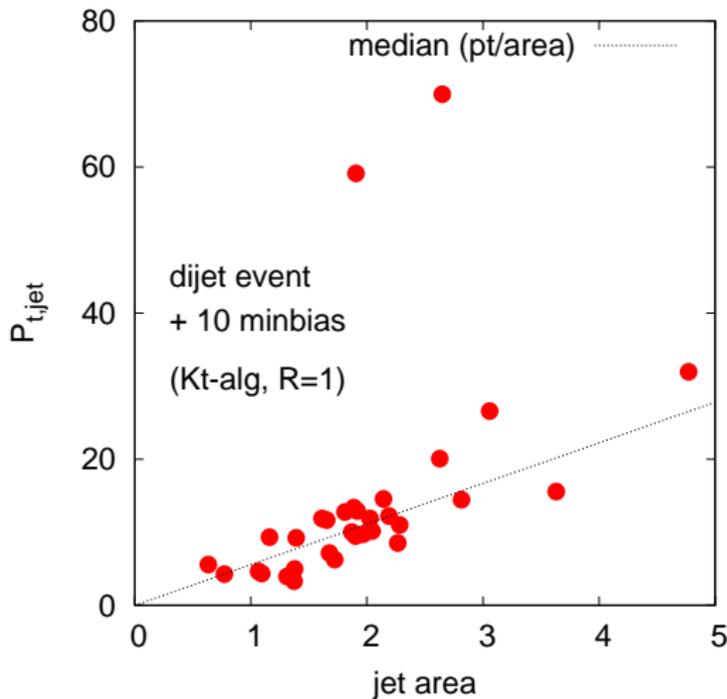


A family of algorithms, all cluster pair with smallest d_{ij} :

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

$$p = \begin{cases} 1 & k_t \\ 0 & \text{C/A} \\ -1 & \text{anti-}k_t \end{cases}$$

Estimating $\rho \equiv$ background noise level



Most jets in event are “background”

Their p_t is correlated with their area.

Estimate ρ :

$$\rho \simeq \text{median}_{\{jets\}} \left[\frac{p_{t,jet}}{A_{jet}} \right]$$

Median limits bias
from hard jets
Cacciari & GPS '07

$$p_{t,jet}^{\text{subtracted}} = p_{t,jet} - \rho \times A_{jet}$$

A_{jet} = jet area

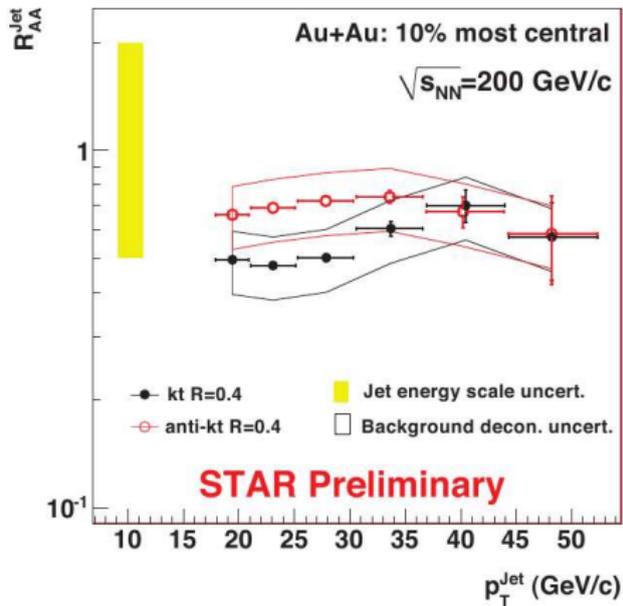
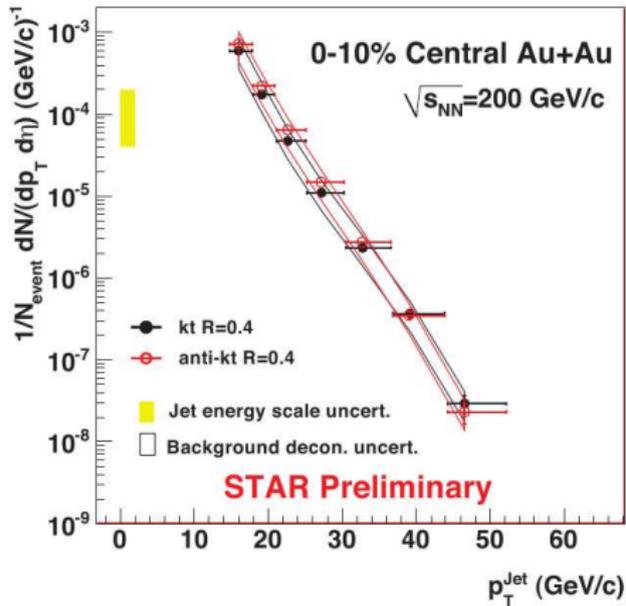
$\rho = p_t$ per unit area from underlying event
(or “background”)

This procedure is intended to be common to pp, pp with pileup (multiple simultaneous minbias) and HIC

NB in AuAu at RHIC: $p_{t,jet}^{\text{subtracted}} = 20 - 50$ GeV, $\rho \simeq 80$ GeV and $A_{jet} \simeq 0.5$

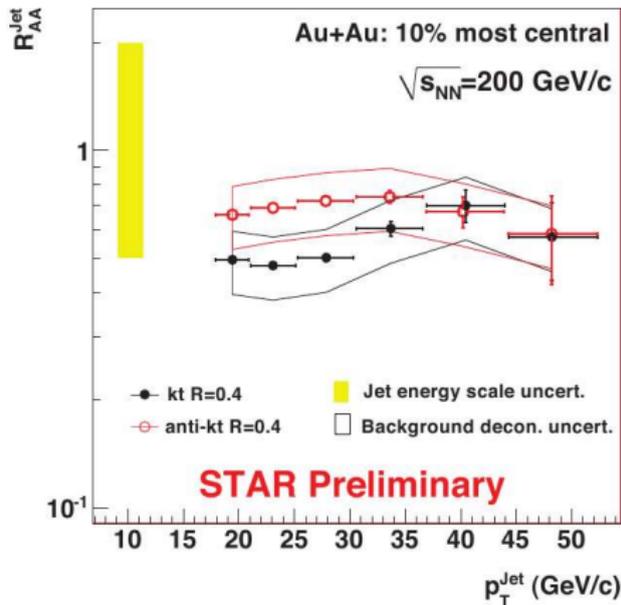
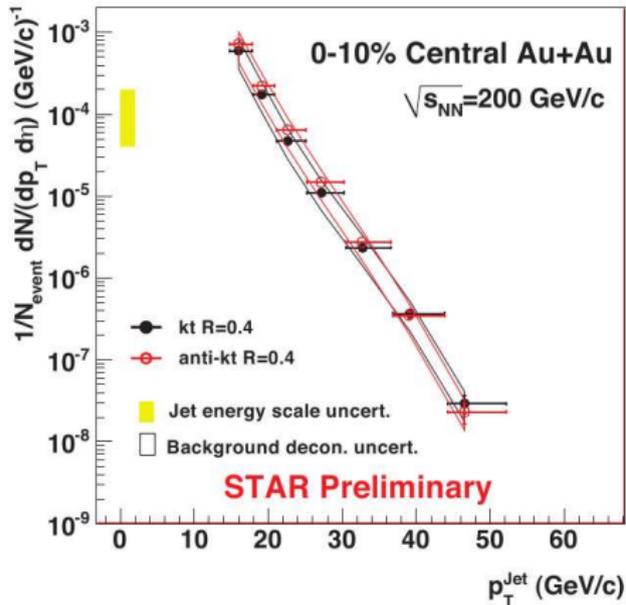
Use at RHIC

This method is basis of STAR jet results

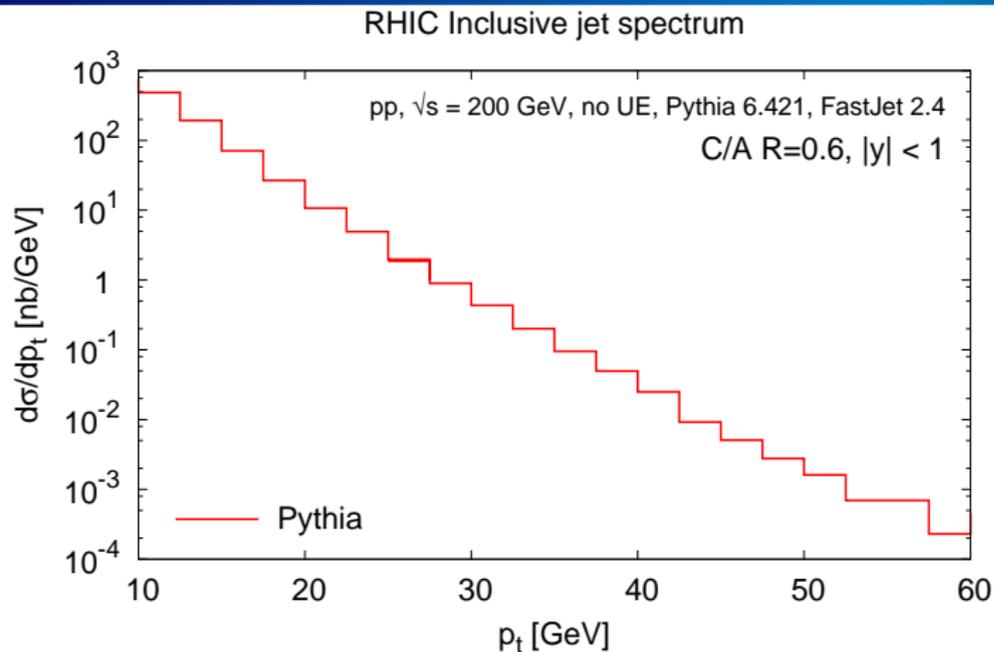


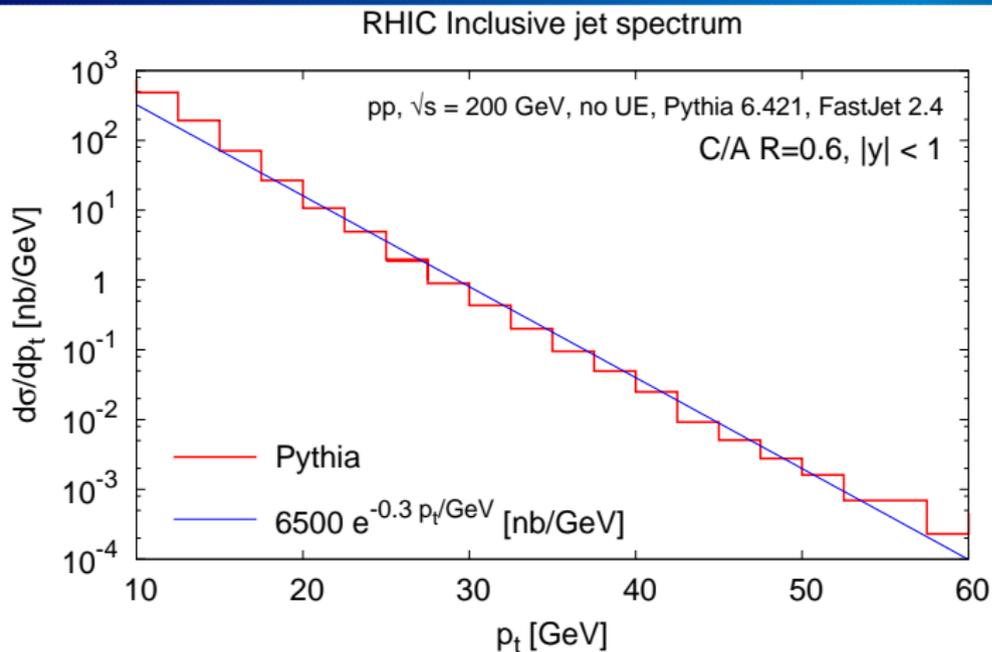
Method designed to minimise biases, but some still persist.
STAR corrects remaining biases based (partly) on Monte Carlo modelling.
Question: can we calculate size of biases? Can we further reduce them?
Identify complementary methods?

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 Question: can we calculate size of biases? Can we further reduce them?
 Identify complementary methods?





To help think about impact of falling cross section at RHIC, approximate it as:

$$\frac{d\sigma}{dp_t} \sim \exp(-0.3 p_t / \text{GeV})$$

Interplay of PDFs & $1/p_t^4$ matrix element

The problem is basically about **subtracting the correct** amount of “underlying event” from each jet, in order to reconstruct correct jet energy.

Take the model for the jet spectrum, $\exp(-ap_t)$ $a = 0.3 \text{ GeV}^{-1}$

Suppose you make a “mistake”:

- ▶ Systematic offset in p_t by $\delta p_{t,jet}$
 - mistake in spectrum by **factor** $\exp(a \delta p_{t,jet})$
If $\delta p_{t,jet} = 3 \text{ GeV}$, factor = 2.5
- ▶ Gaussian error of std.dev. σ_{jet} in subtraction
 - mistake in spectrum by **factor** $\exp(a^2 \sigma_{jet}^2 / 2)$
If $\sigma_{jet} = 5 \text{ GeV}$, factor = 3.1

You want to know R_{AA} to within a few tens of percent.

Residual systematic offsets must be understood to within 1 GeV.
Fluctuations must be as small as possible, and accurately known.

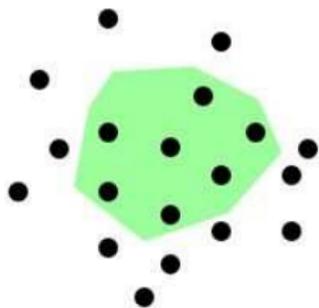
Example #1: a bias

(background does not *just* linearly add noise to jet)

BACK REACTION

“How (much) a jet changes when immersed in a background”

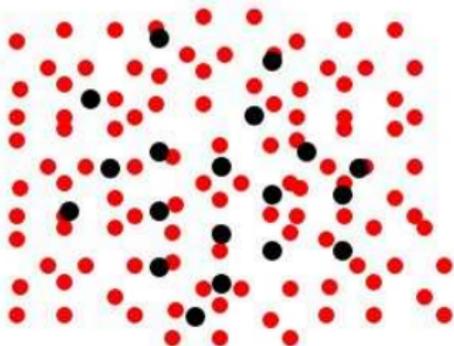
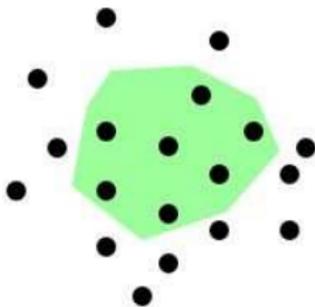
Without
background



BACK REACTION

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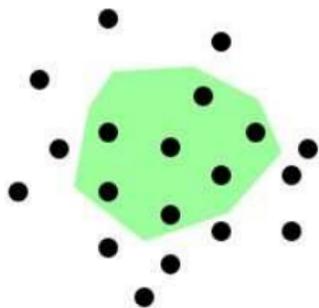
Without
background



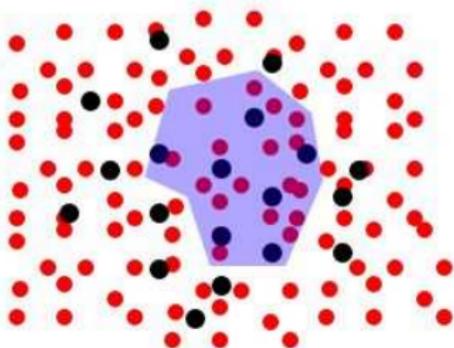
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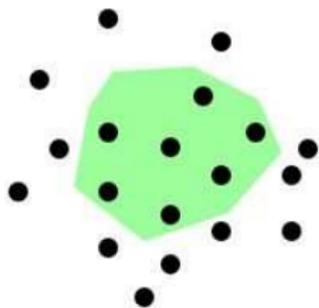
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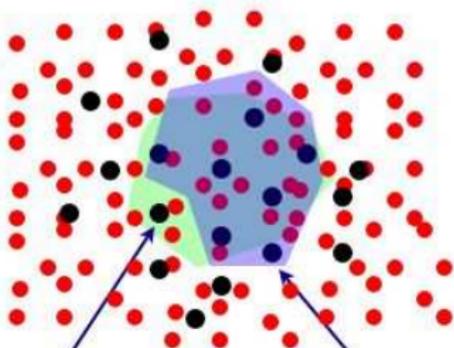
BACK REACTION

“How (much) a jet changes when immersed in a background”

Without
background



With
background



Backreaction **loss**

Backreaction **gain**

Soft & collinear approximation:

$$\delta p_t^{BR} = \mathcal{B}_{alg} \cdot \rho R^2 \frac{2C_i}{\pi} \alpha_s \ln \frac{p_t}{\rho R^2}$$

Cacciari, GPS & Soyez '08
+ large corrections

jet alg	\mathcal{B}_{alg}
k_t	-0.3
C/A	-0.3
anti- k_t	0

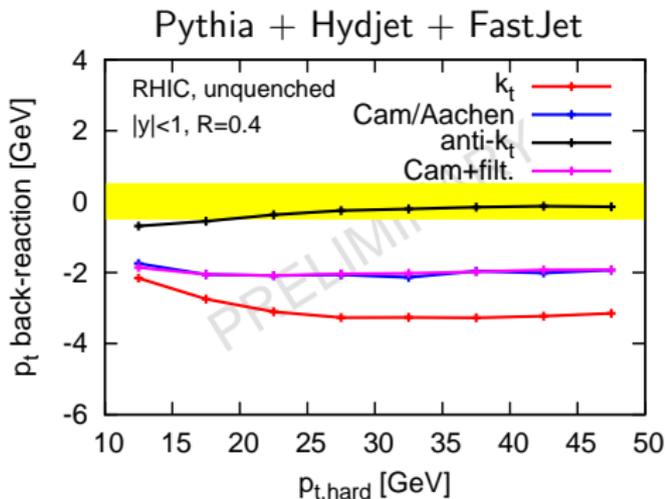
Backreaction can be calculated (sort of...)

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Cacciari, Rojo, GPS & Soyez, prelim.
anti- k_t bias = 0, as expected

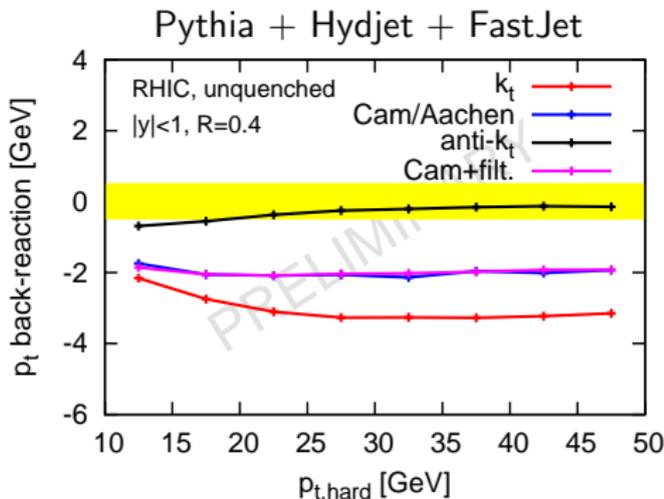
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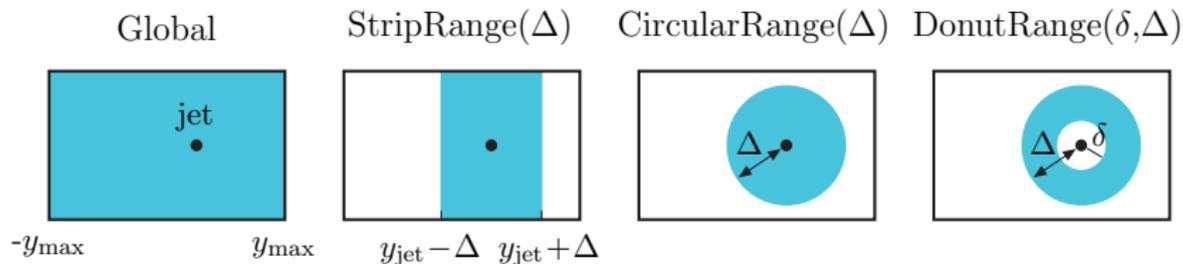
Different jet algorithms have different systematics
Use of more than one provides important cross-checks

Example #2: another bias

is ρ measured correctly?

What could go wrong?

- Rapidity and azimuth dependence of ρ distribution means ρ near jet $\neq \rho$ measured over large region. So try various regions:



- Median estimate \neq mean contamination. Can be studied in toy models:

$$\rho^{\text{median}} \simeq \rho^{\text{true}} \left(1 - \frac{1}{3\nu R^2} \right)$$

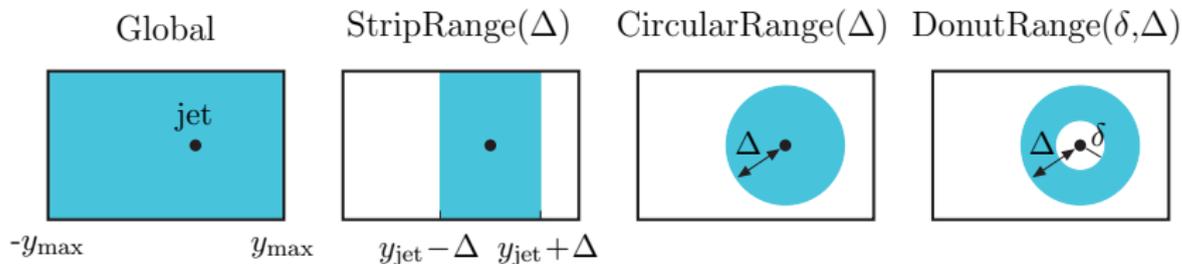
ν = number of particles / unit area

With $\nu = 100$, $R = 0.4$, $\mathcal{O}(2\%) \rightarrow \mathcal{O}(1 \text{ GeV})$ on jet p_t

Cacciari, GPS & Sapeta '09, for measuring $\rho \sim 2 \text{ GeV}$ in pp collisions!

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Example #3: fluctuations

Fluctuations of amount of background / underlying-event in a square of unit area can be characterised in terms of σ_{UE} , which is $\mathcal{O}(10 \text{ GeV})$ at RHIC.

Dispersion in jet subtraction, σ_{jet} is given by

$$\sigma_{jet} = \sigma_{UE} \times \sqrt{A_{jet}}$$

jet alg	$\langle A_{jet} \rangle$
k_t	$0.81\pi R^2$
C/A	$0.81\pi R^2$
anti- k_t	πR^2

+ p_t -dependent scaling
 violations for k_t and C/A

Put in numbers and find $\sigma_{jet} \sim 7 \text{ GeV}$.

This is dangerous

Steeply falling spectrum rescaled by $\times 10$?

Obvious solution: reduce R

But then lose gluon radiation

Can be very severe with quenching
 cf. STAR tried $R = 0.2$ instead of 0.4

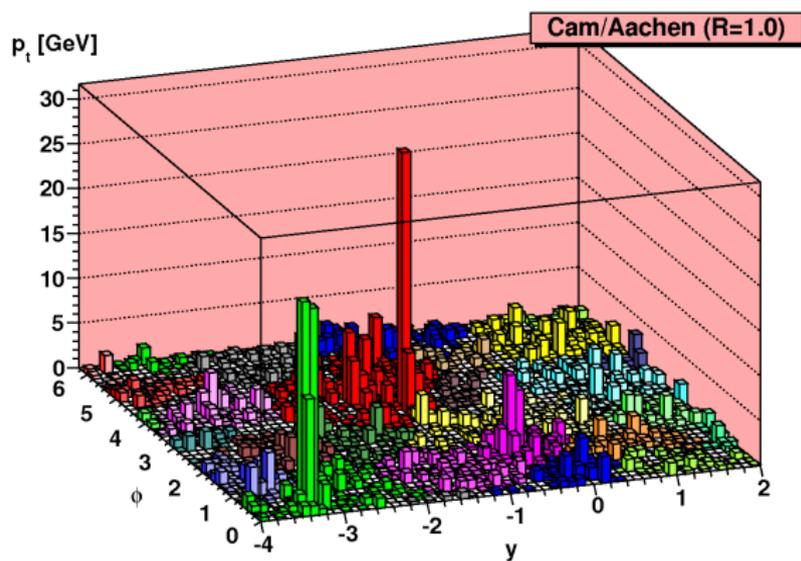
Reducing fluctuations, while
limiting bias:

filtering

Idea to improve resolution for an LHC Higgs search in $H \rightarrow b\bar{b}$ decay mode!

Keep hardest $\mathcal{O}(\alpha_s)$ gluon emission in jet, while throwing out soft "junk"

Butterworth, Davison, Rubin & GPS '08



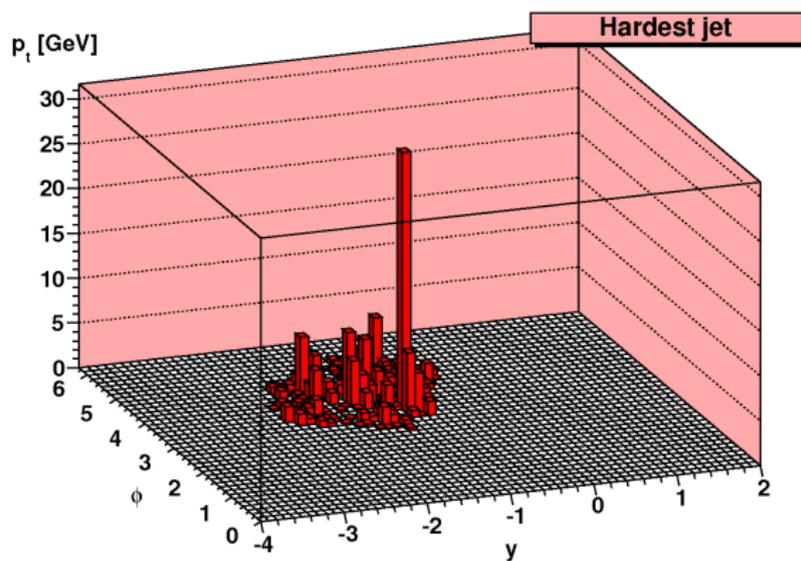
1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest "subjets" leading quark + 1 gluon
4. The result is a "filtered" jet

Related ideas by Ellis, Vermillion & Walsh '09 and Krohn, Thaler & Wang '09

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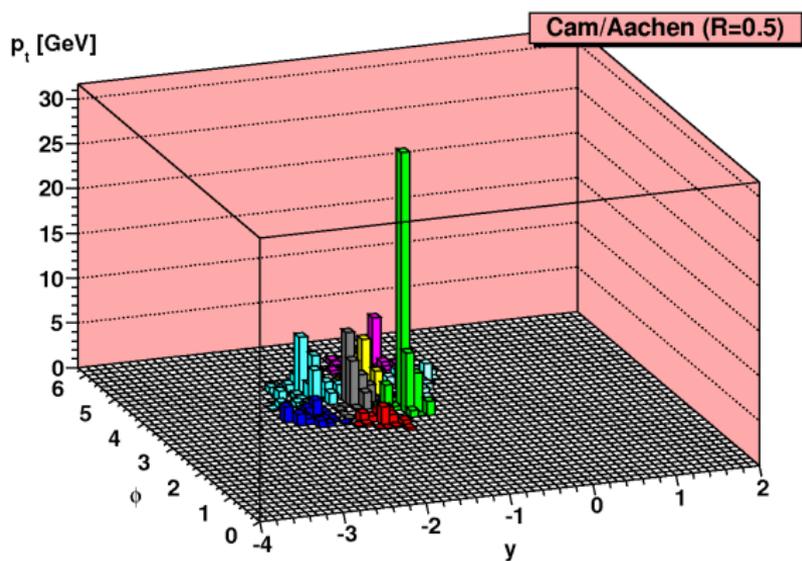
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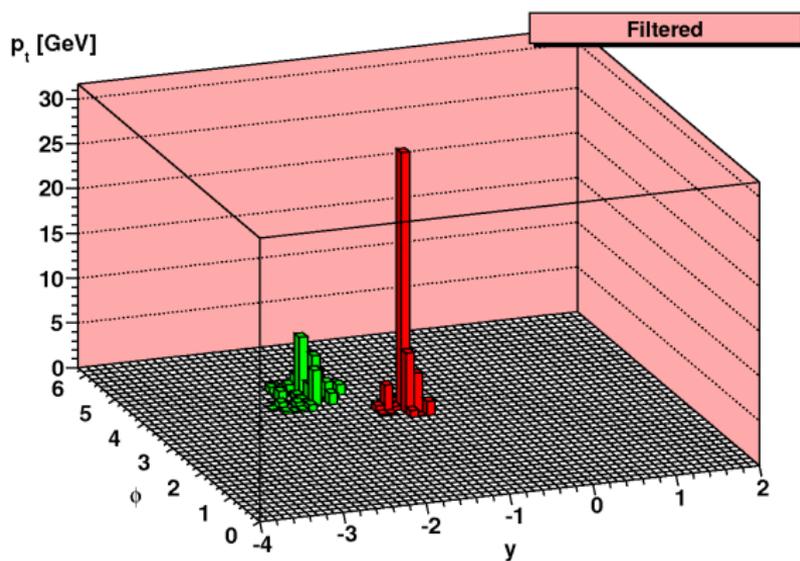
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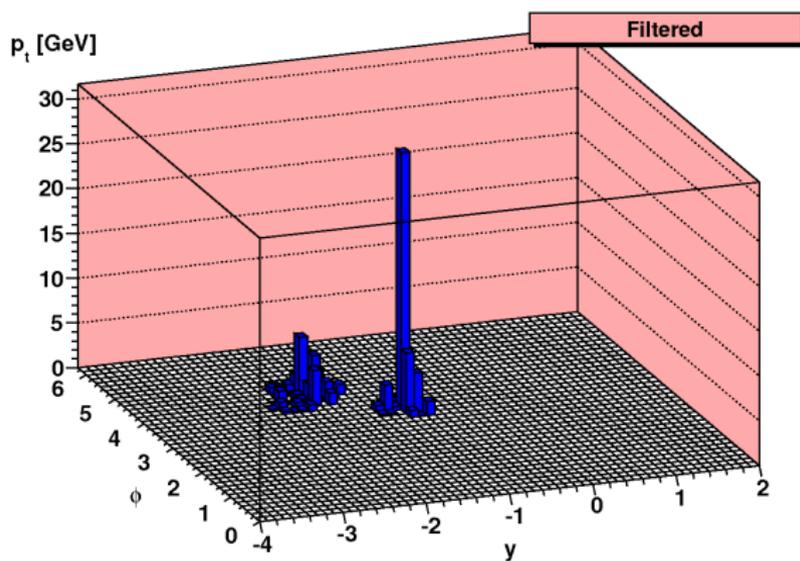
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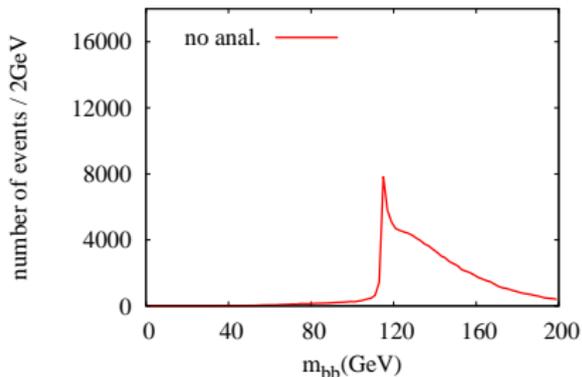


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Reconstructed mass for jets from decay of high- p_t Higgs-boson [without pileup]

Without Filtering



With Filtering

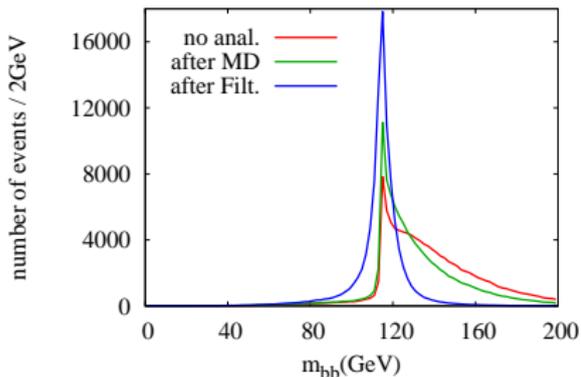
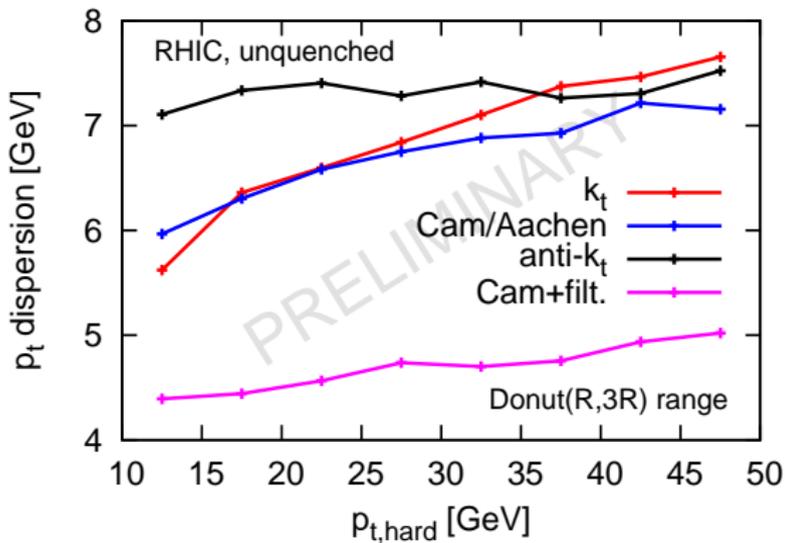


Figure from Rubin

Among the techniques adopted in search for $H \rightarrow b\bar{b}$ at LHC

Impact of filtering on dispersion in HIC

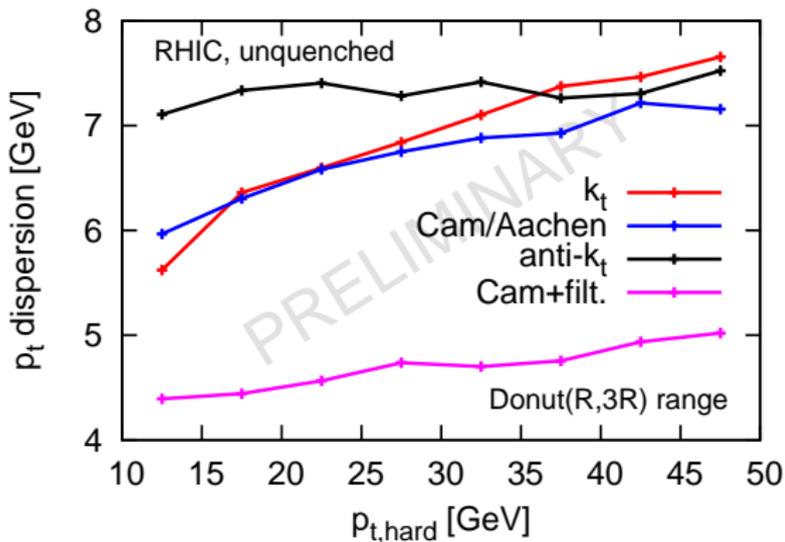


Filtering reduces jet area by $\sim \frac{1}{2}$

Fluctuations $\propto \sqrt{A}$
 should go down by $\sim \sqrt{\frac{1}{2}}$

And they do

Impact of filtering on dispersion in HIC



Filtering reduces jet area by $\sim \frac{1}{2}$

Fluctuations $\propto \sqrt{A}$
 should go down by $\sim \sqrt{\frac{1}{2}}$

And they do

Filtering's reduction of dispersion from 7 GeV to 5 GeV means experimental "unfolding" might be factor 3 instead of factor 10

Numbers are rough – intended to give an idea of impact
 Alternative ideas: see Cole & Lai '08

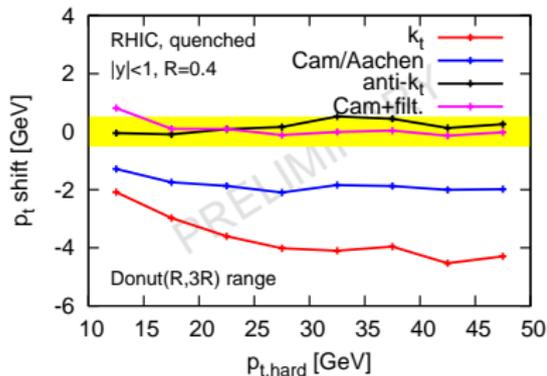
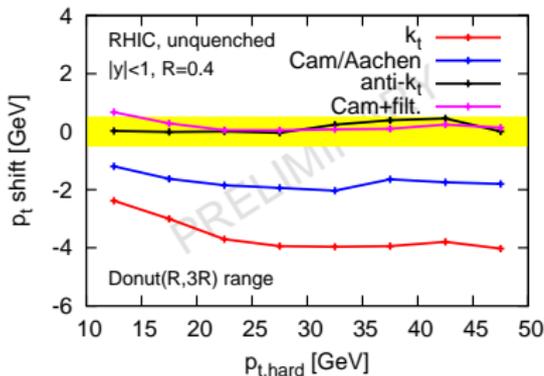
Does filtering introduce new biases in jets in quenched case?

Vacuum QCD: we know how much gluon radiation we lose
QCD in medium: extra medium-induced radiation lost?

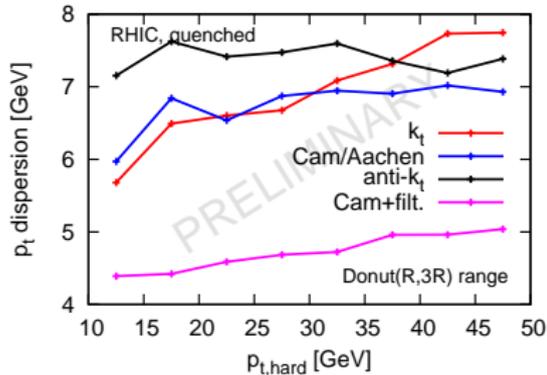
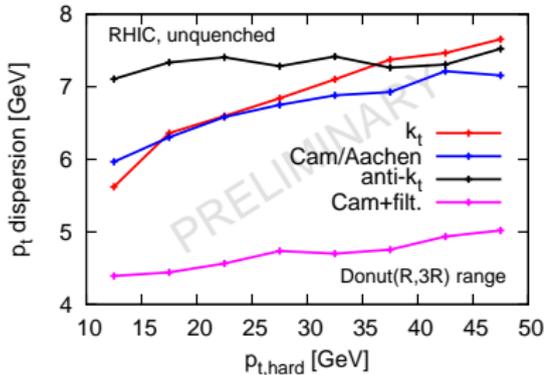
UNQUENCHED

QUENCHED

PT SHIFT

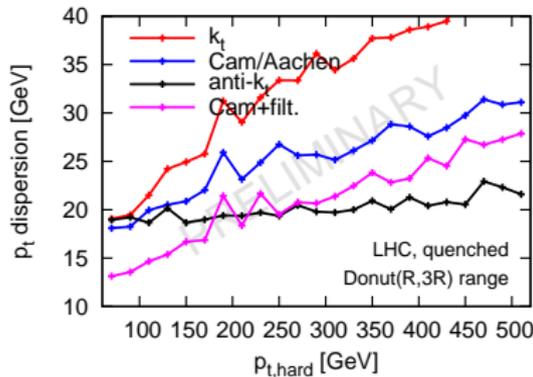
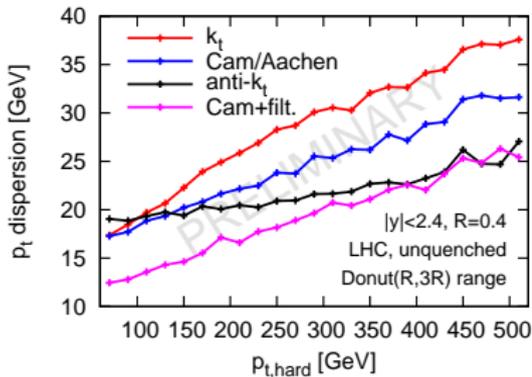
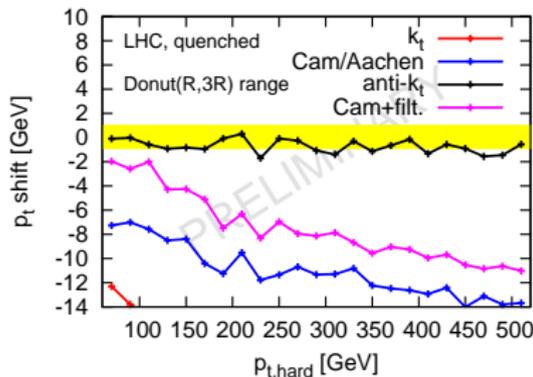
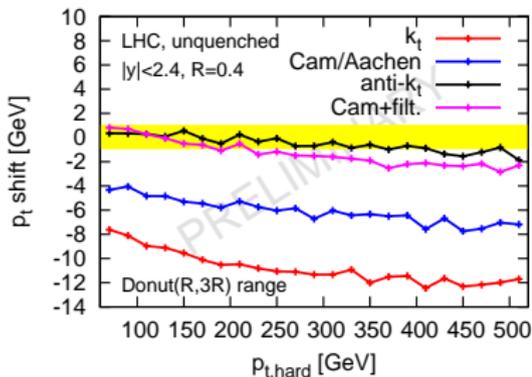


DISPERSION



UNQUENCHED

QUENCHED



It's still early days for jet-finding in HIC (& high-luminosity LHC)

It's a tough job to accurately remove 40 GeV of noise from a 40 GeV hard jet in the context of a steeply falling cross-section.

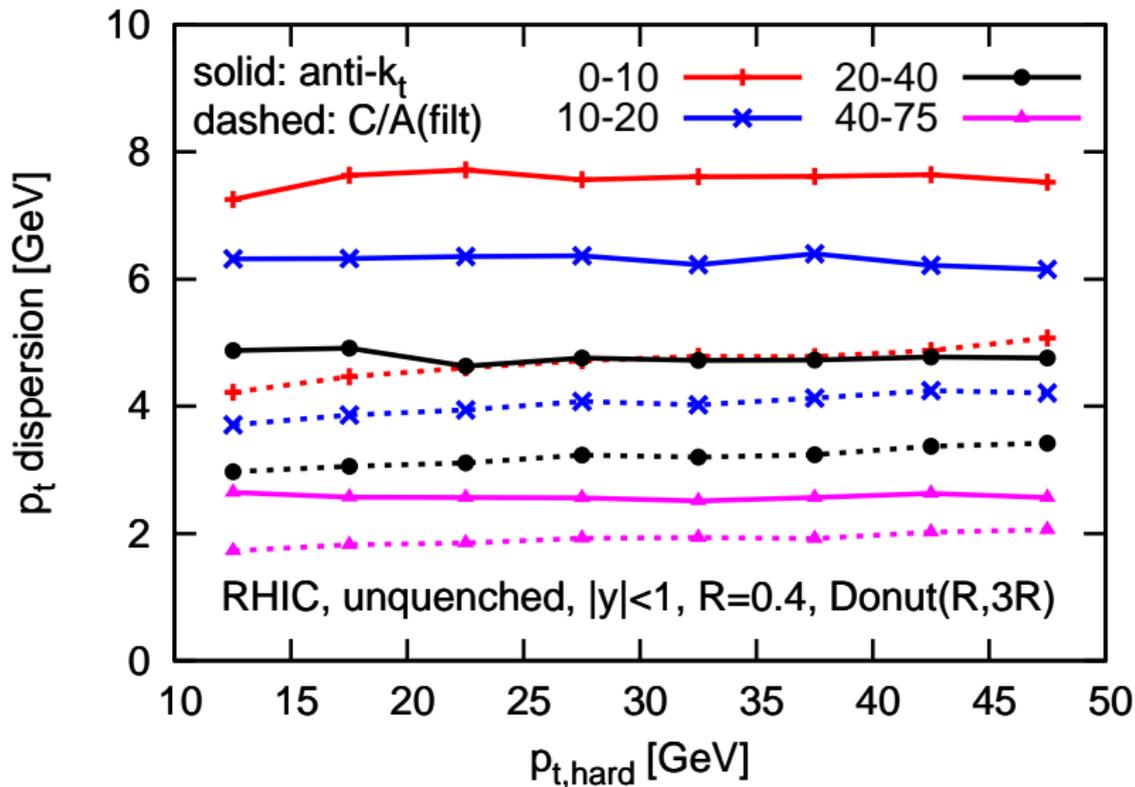
Theory calculations can guide the choices one makes

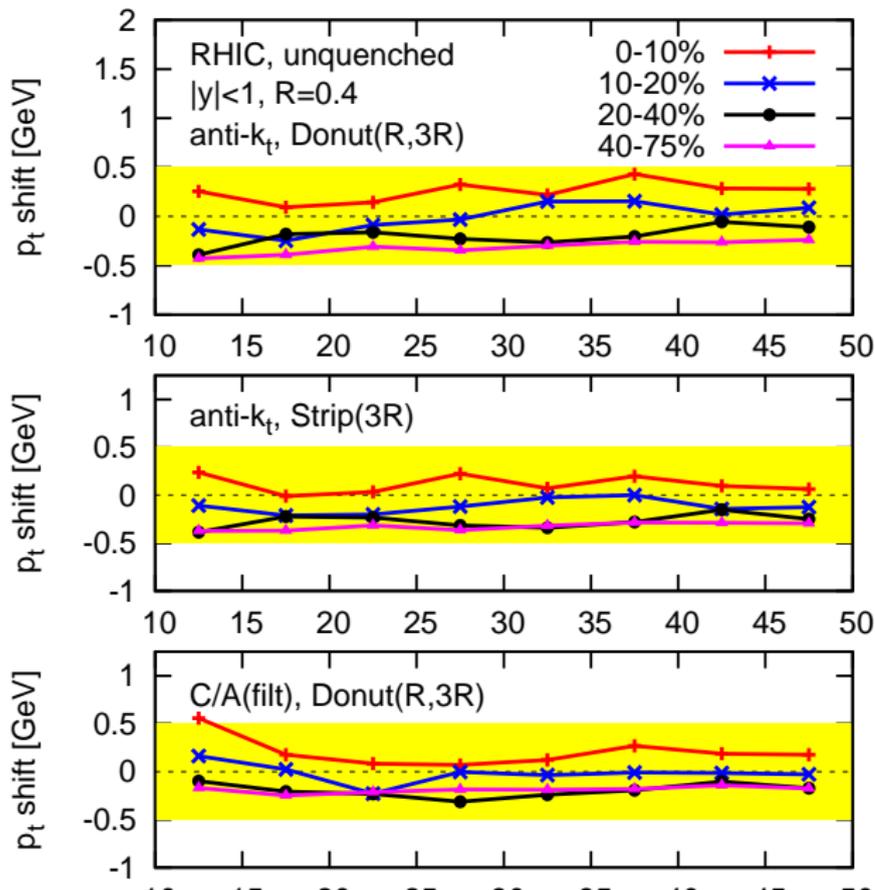
- ▶ Give us an idea of size of corrections semi-independently of Monte Carlo
Some of them are rather large
- ▶ Tell us which approaches are complementary in their systematics
Adding to robustness of experimental measurements, e.g. k_t v. anti- k_t
NB: it's still hard to estimate how quenching affects systematics
- ▶ Guide design of new tools that have smaller systematics
Like filtering, yet to be tried out at RHIC

Important potential for cross-fertilization between ideas in
HIC and LHC pp programs.

EXTRAS

Dispersion for non central AuAu





Anti- k_t jet spectrum, pp $\sqrt{s} = 200$ GeV