

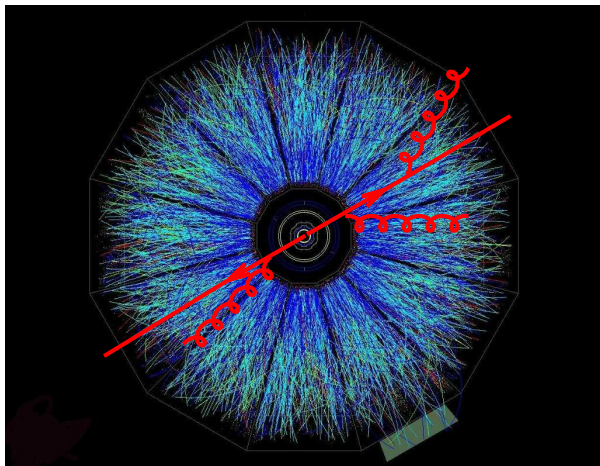
Jet reconstruction with FastJet

Gavin P. Salam

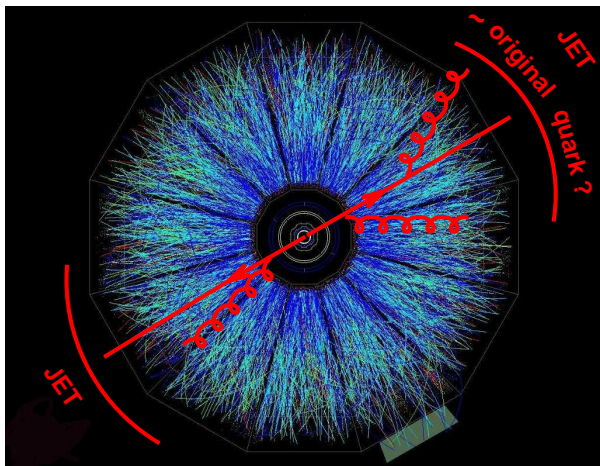
LPTHE, UPMC Paris 6 & CNRS

Jets in Proton-Proton and Heavy-Ion Collisions
Prague, 12 August 2010

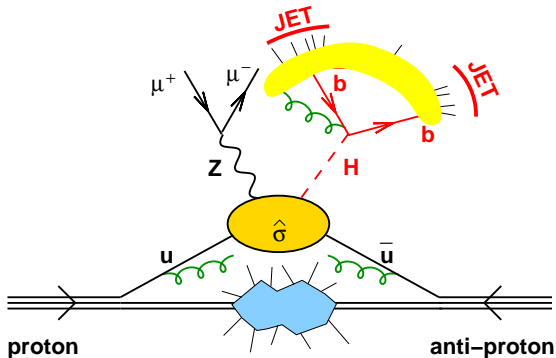
Based on work (some preliminary) with
Matteo Cacciari, Juan Rojo, Sebastian Sapeta, Gregory Soyez



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



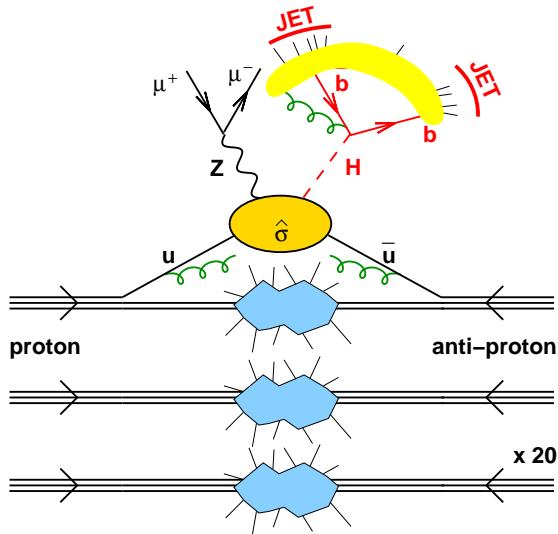
Radiation from high-momentum quarks & gluons traversing hot medium can tell us about the medium



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

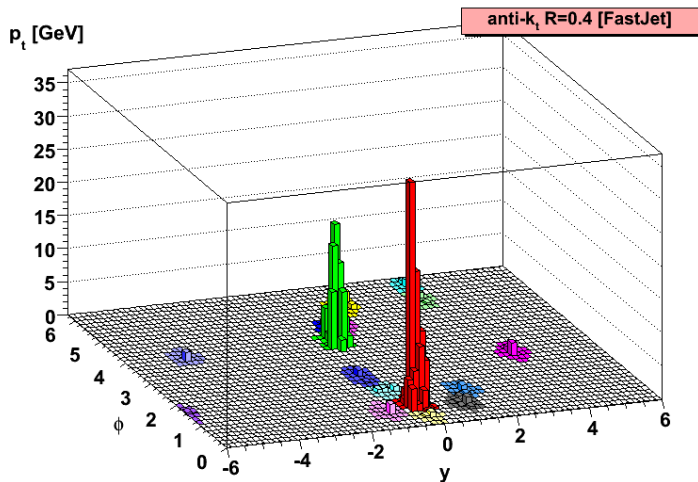


Use jets to reconstruct quarks from decay of some new heavy object

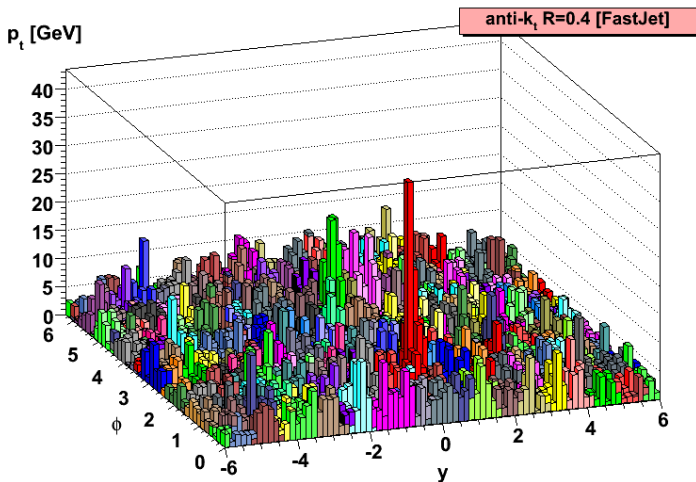
e.g. a Higgs boson

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

Common challenge: large contamination



A pp event (LHC 5.5 TeV, Pythia)



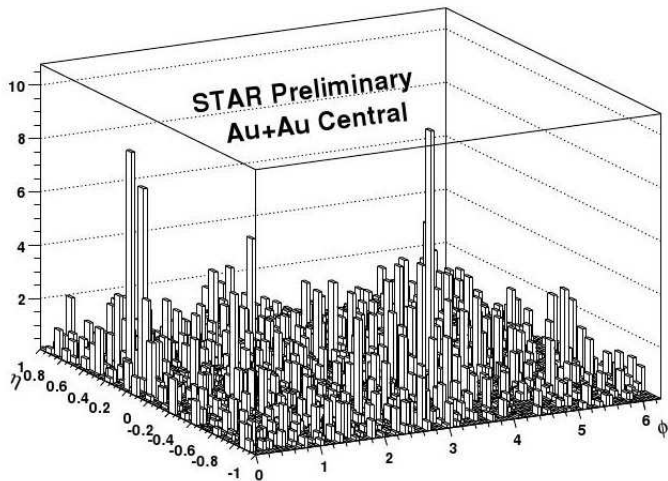
Contamination in jet

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)



Contamination in jet

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

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

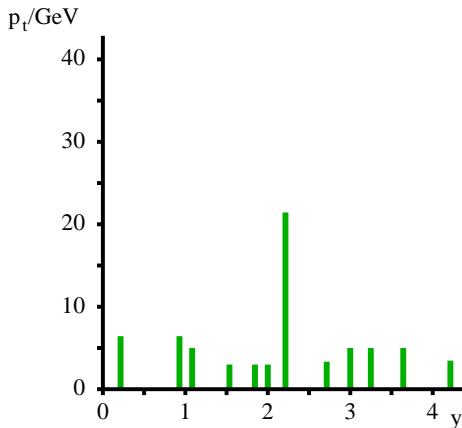
LHC pp
 (hi-lumi)
 $\mathcal{O}(5 - 40 \text{ GeV})$

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

1. Jet algorithms



A jet algorithms provides a mapping:

$$\text{particles} \xrightarrow[\text{jet.def.}]{} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

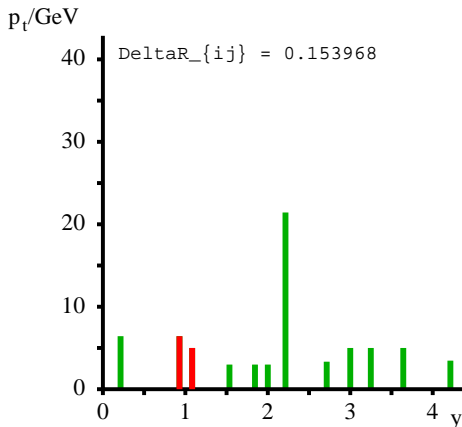
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

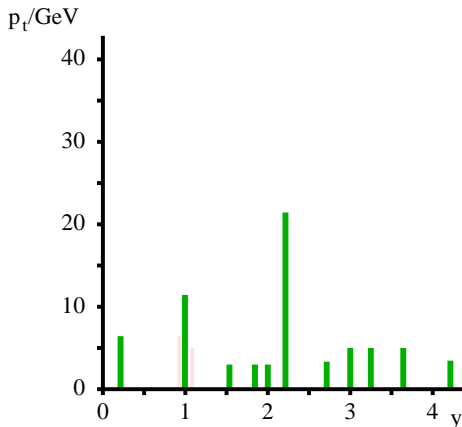
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

$$\text{particles} \xrightarrow[\text{jet.def.}]{} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

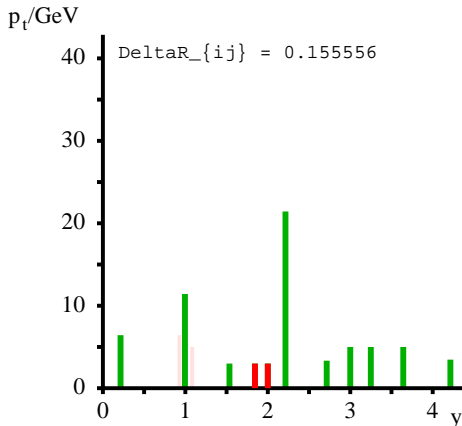
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

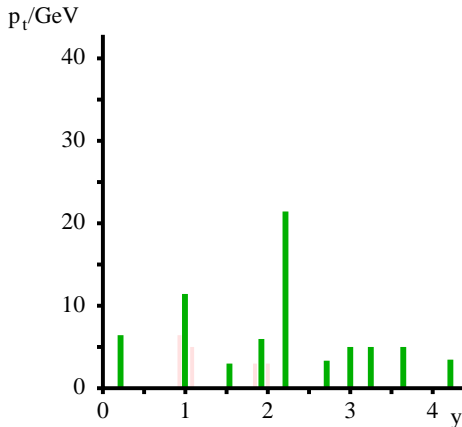
Dokshitzer et al '97

Wengler & Wobisch '98

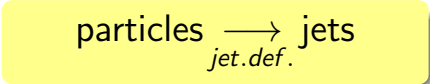
Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:



Simplest pp jet algorithm is “Cambridge/Aachen”

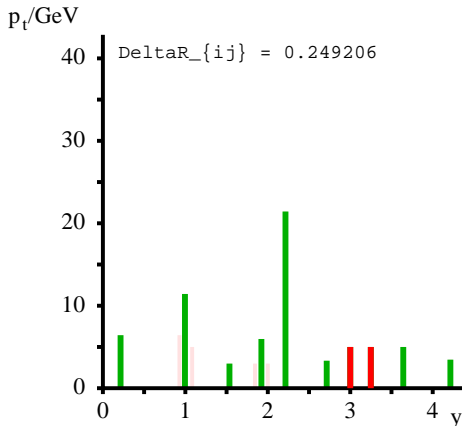
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

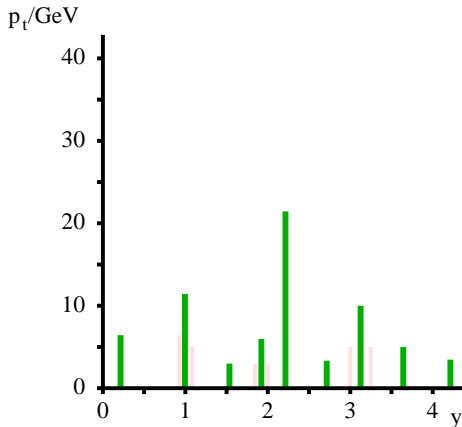
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

$$\text{particles} \xrightarrow{\text{jet.def.}} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

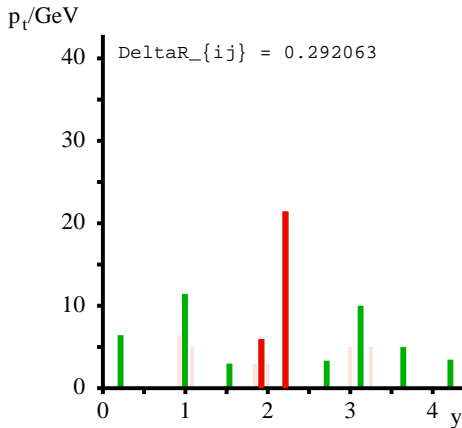
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

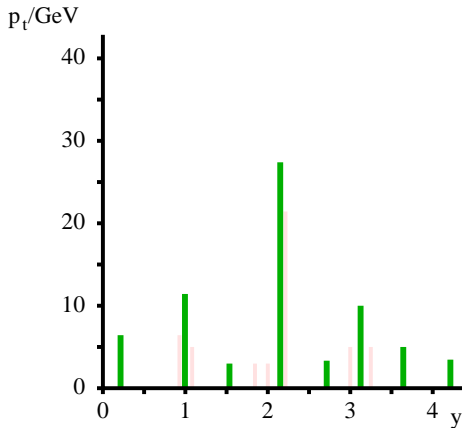
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

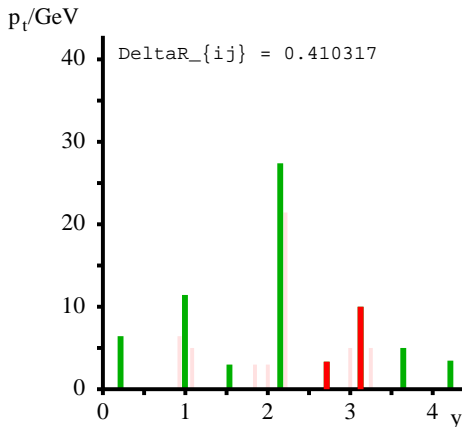
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

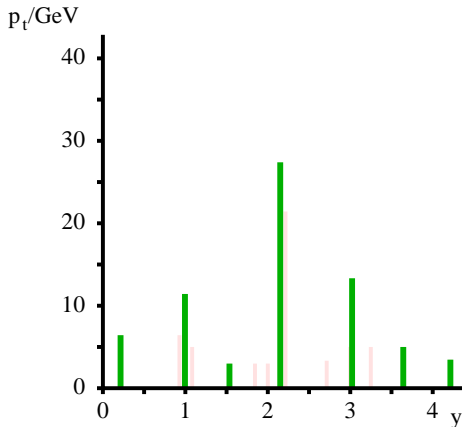
Dokshitzer et al '97

Wengler & Wobisch '98

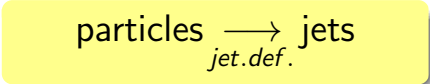
Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:



Simplest pp jet algorithm is “Cambridge/Aachen”

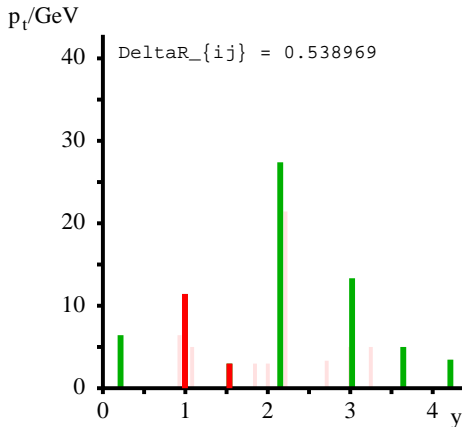
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

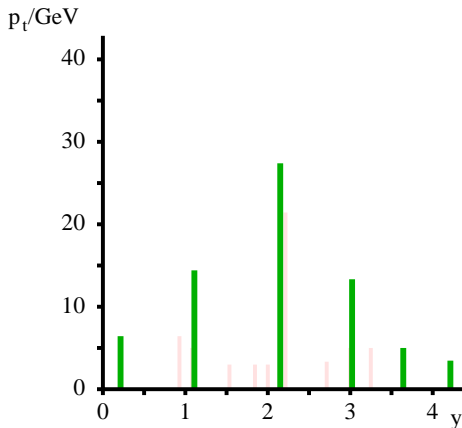
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

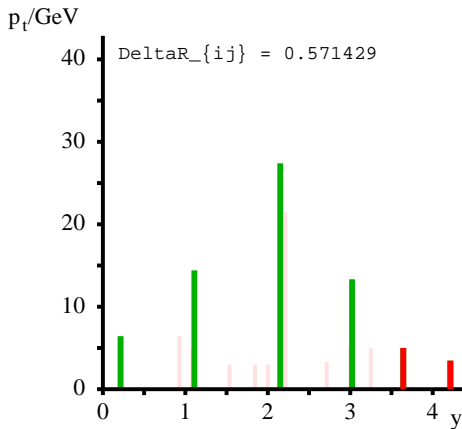
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

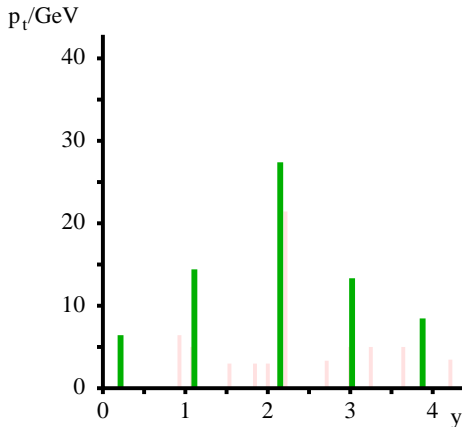
Dokshitzer et al '97

Wengler & Wobisch '98

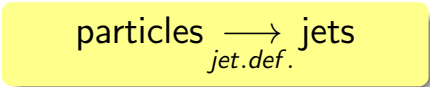
Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:



Simplest pp jet algorithm is “Cambridge/Aachen”

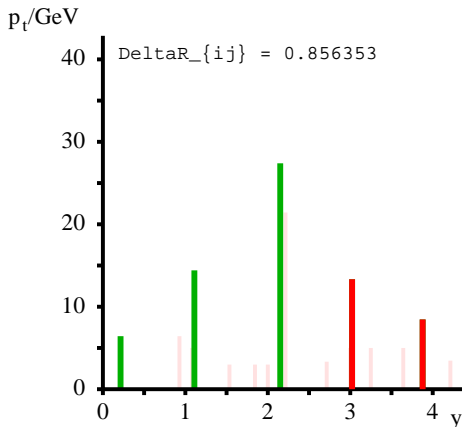
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

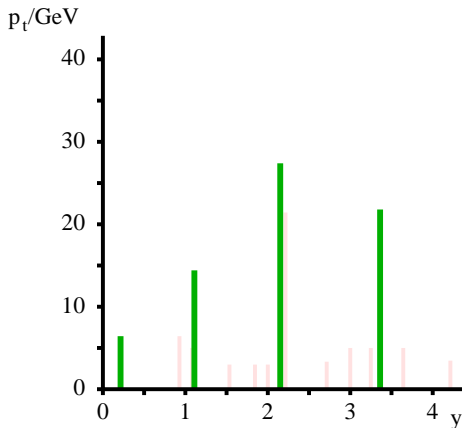
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

$$\text{particles} \xrightarrow[\text{jet.def.}]{} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

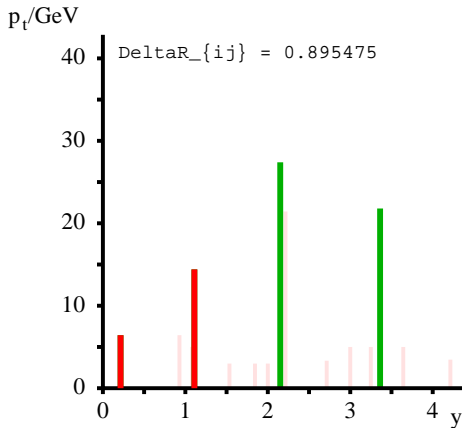
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

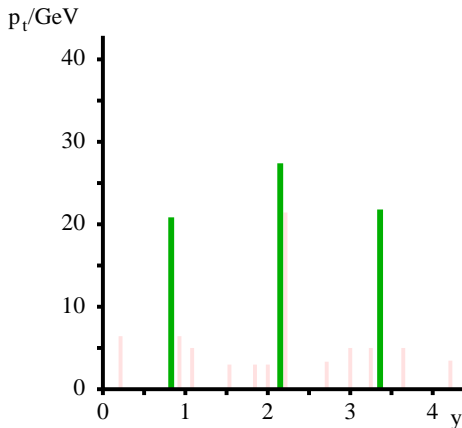
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

$$\text{particles} \xrightarrow[\text{jet.def.}]{} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

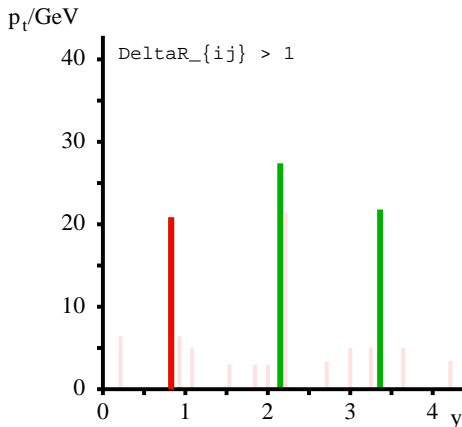
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

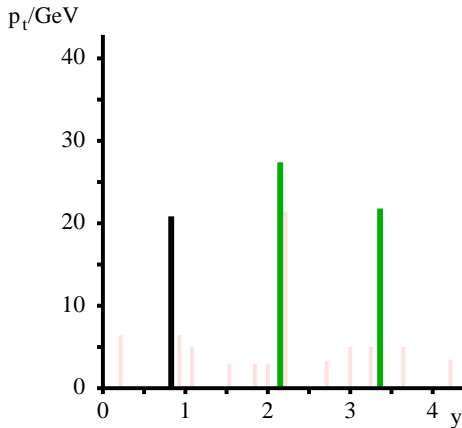
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithm provides a mapping:

$$\text{particles} \xrightarrow[\text{jet.def.}]{} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

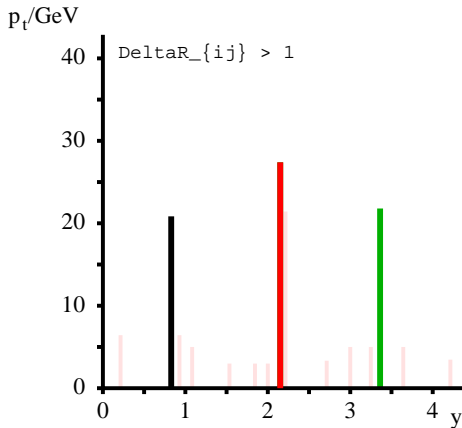
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

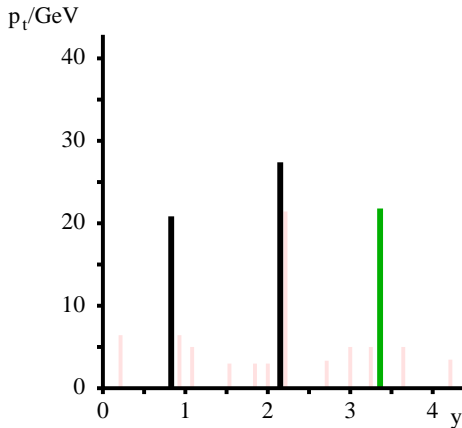
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

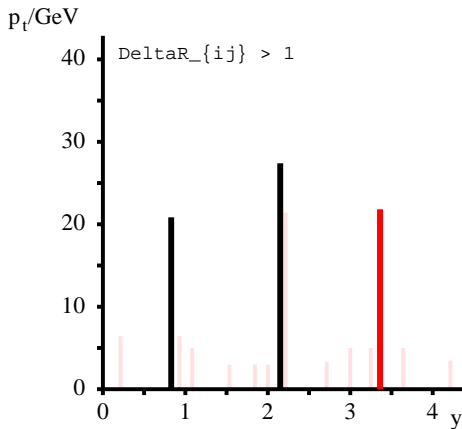
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

particles $\xrightarrow{\text{jet.def.}}$ jets

Simplest pp jet algorithm is “Cambridge/Aachen”

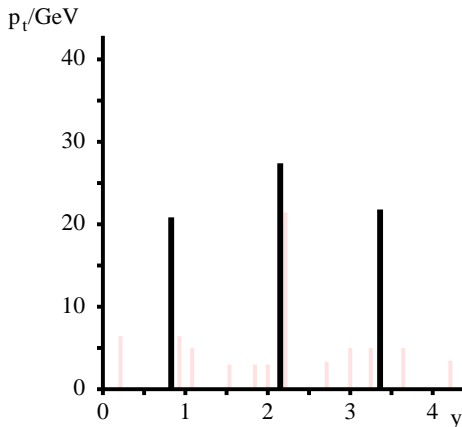
Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers



A jet algorithms provides a mapping:

$$\text{particles} \xrightarrow{\text{jet.def.}} \text{jets}$$

Simplest pp jet algorithm is “Cambridge/Aachen”

Dokshitzer et al '97

Wengler & Wobisch '98

Repeatedly recombine closest pair of objects, until all separated by $\Delta R_{ij}^2 = \Delta y_{ij}^2 + \Delta \phi_{ij}^2 > R^2$.

R parameter sets angular resolution

ϕ assumed 0 for all towers

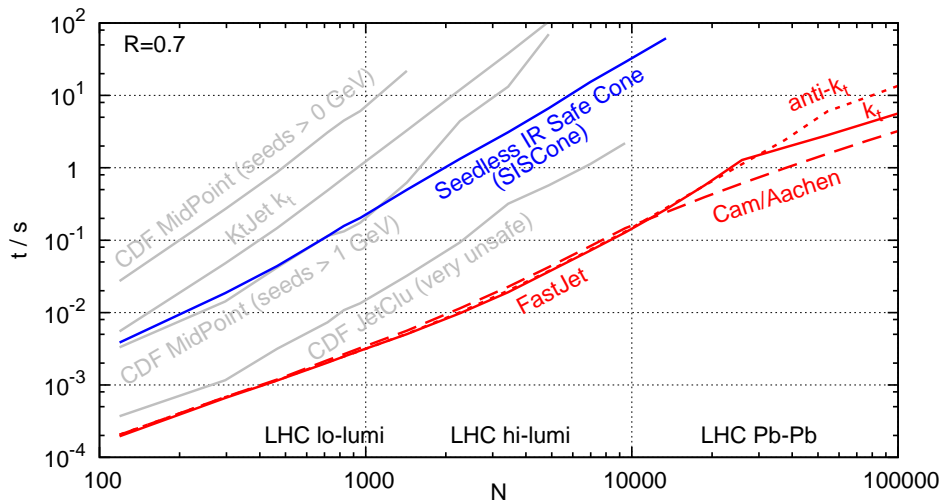
Generalise inclusive-type sequential recombination with

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

	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 [& much more] coded in (efficient) C++ at
<http://fastjet.fr/> (Cacciari, GPS & Soyez '05-'10)

FastJet: time to cluster N particles



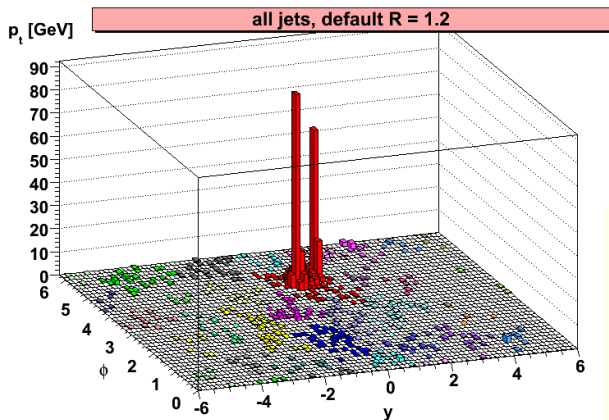
2. Jet areas

Measure jets' susceptibility to contamination by noise

Jets are made of finite number of pointlike particles.

Area not unambiguous concept

Jet areas must be defined



Add many soft particles to event

$10\text{--}100\text{ GeV}$ each

$A \propto \#$ inside jet

Cacciari, GPS & Soyez '08
measure of jet's susceptibility to contamination from soft radiation

Jets are made of finite number of pointlike particles.

Area not unambiguous concept

Jet areas must be defined

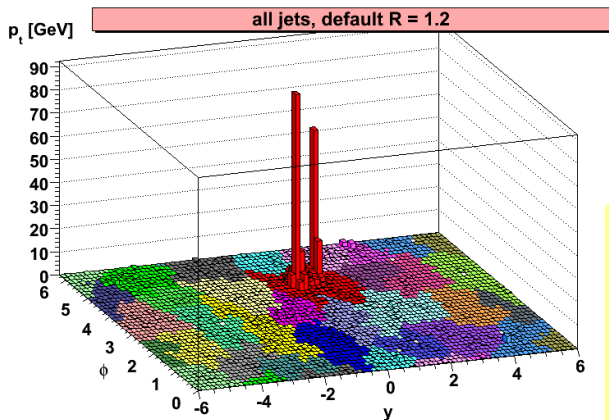
Add many soft particles to event

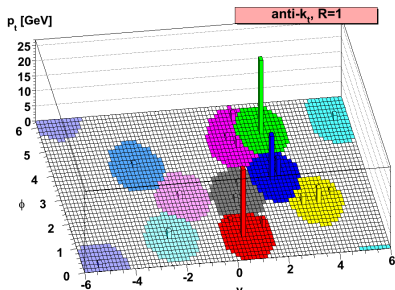
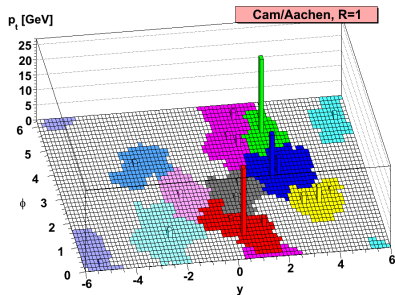
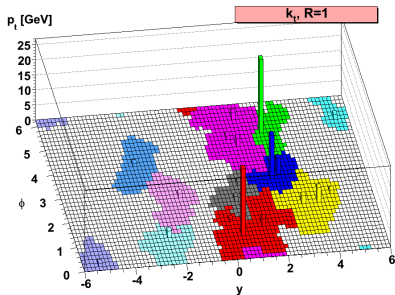
10^{-100} GeV each

$A \propto \#$ inside jet

Cacciari, GPS & Soyez '08

measure of jet's susceptibility to contamination from soft radiation



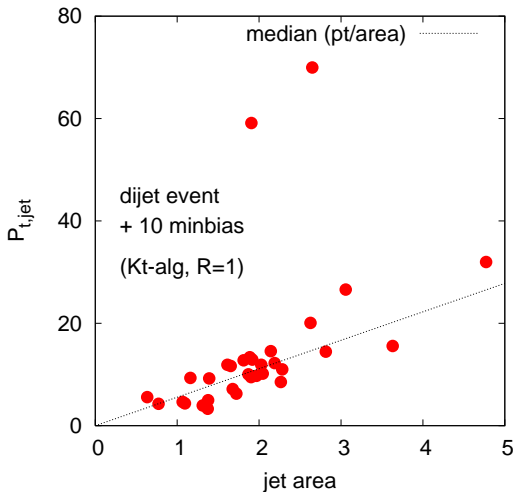


A family of algorithms, all cluster pair with smallest d_{ij} :

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

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

3. Noise estimation



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

4. Noise subtraction

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

A_{jet} = jet area

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

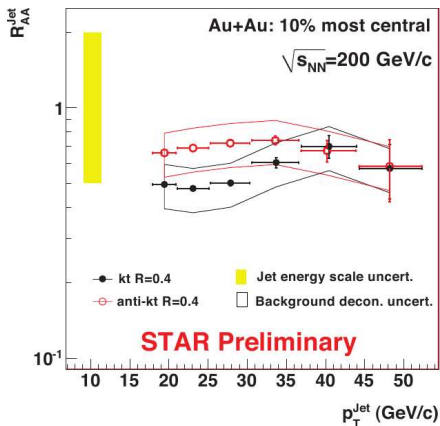
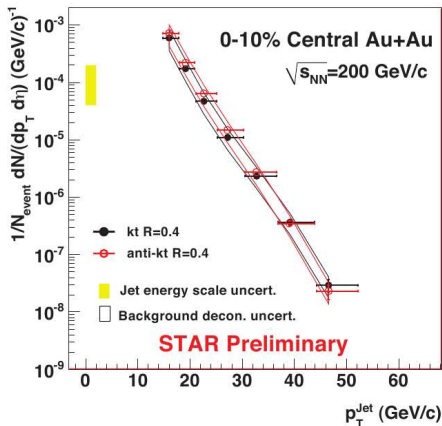
This procedure is intended to be common to pp, pp with pileup (multiple simultaneous minbias) and HIC

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?

Let's examine some of the issues

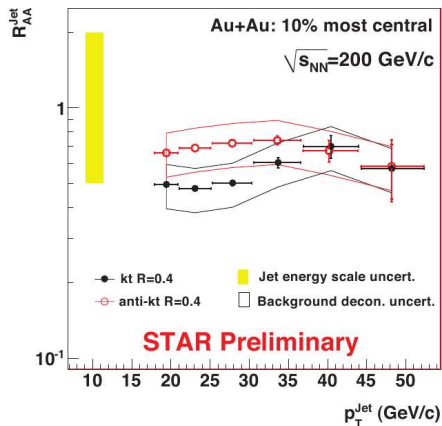
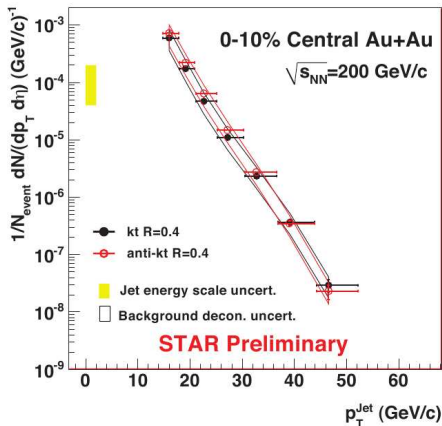
Area/median method → STAR jet results



Method designed to minimise biases, but some still persist.

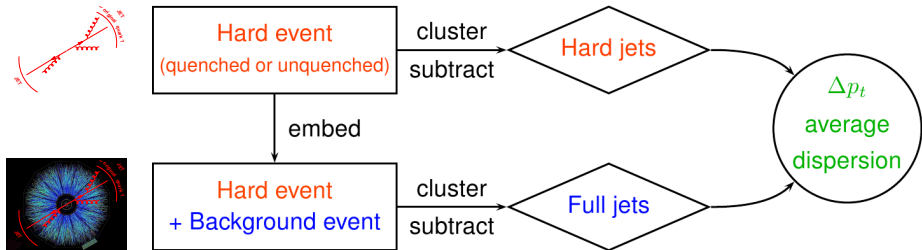
Question: can we calculate size of biases? Can we further reduce them?

Area/median method → STAR jet results



Method designed to minimise biases, but some still persist.

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



Define:

$$\Delta p_t = p_t^{\text{full(subtracted)}} - p_t^{\text{hard}}$$

Study:

average of Δp_t , $\langle \Delta p_t \rangle$
dispersion of Δp_t , $\sigma_{\Delta p_t}$

We don't have real background (or hard!) events, so use

- ▶ **Hard event:** Pythia 6.4, and optionally PyQyen 1.5 / QPythia
- ▶ **Background:** Hydjet 1.6 (Hydjet++ 2.1 for cross-checks)
- ▶ **Analysis:** FastJet 2.4 (& 2.5-devel), $R = 0.4$ for all main jet finders (k_t with $R = 0.5$ for bkgd estimation).

Cacciari, Rojo, GPS & Soyez, to appear soon...

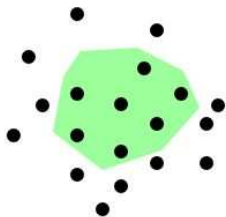
Example #1: non-zero $\langle \Delta p_t \rangle$

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

BACK REACTION

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

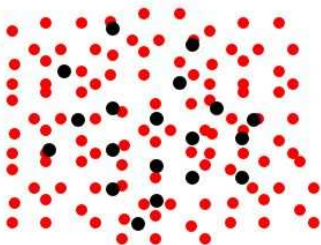
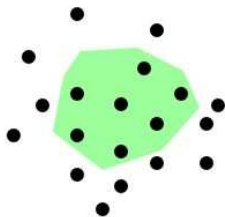
Without
background



BACK REACTION

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

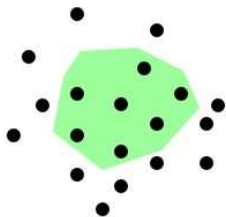
Without
background



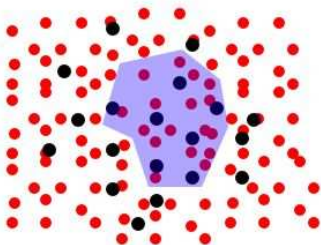
BACK REACTION

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

Without
background



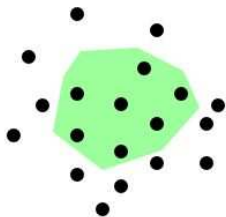
With
background



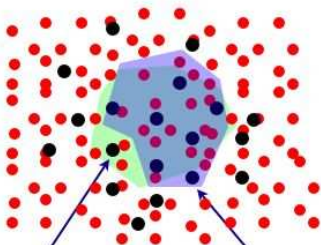
BACK REACTION

“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

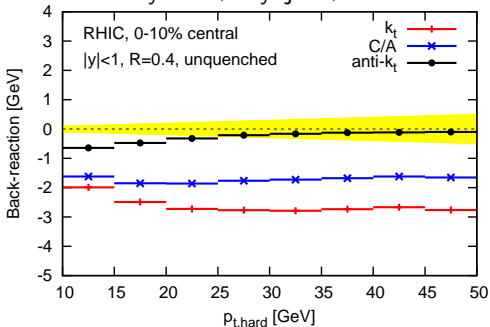
Soft & collinear approximation:

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

Cacciari, GPS & Soyez '08
 + large corrections

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

Pythia + Hydjet + FastJet



Cacciari, Rojo, GPS & Soyez, prelim.

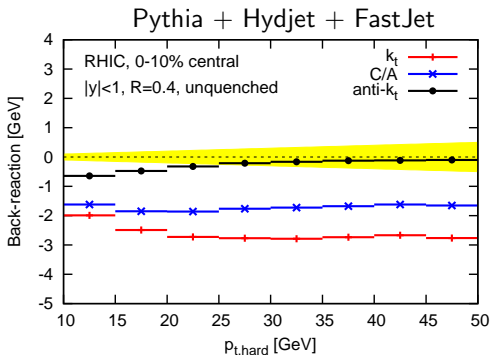
anti- k_t bias = 0, as expected

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



Cacciari, Rojo, GPS & Soyez, prelim.

anti- k_t bias = 0, as expected

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

Example #2: 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

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

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

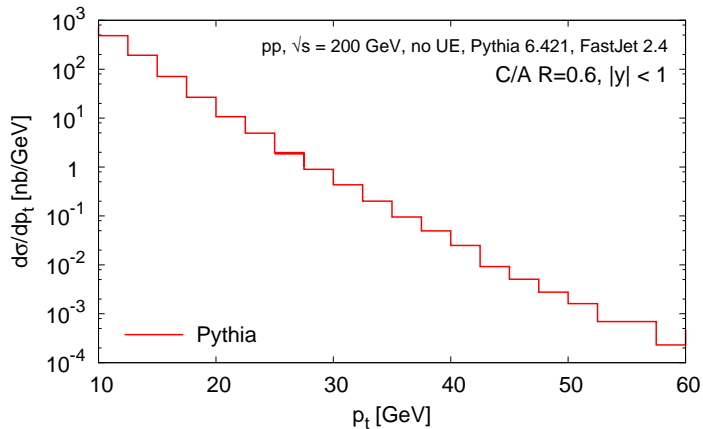
Put in numbers:

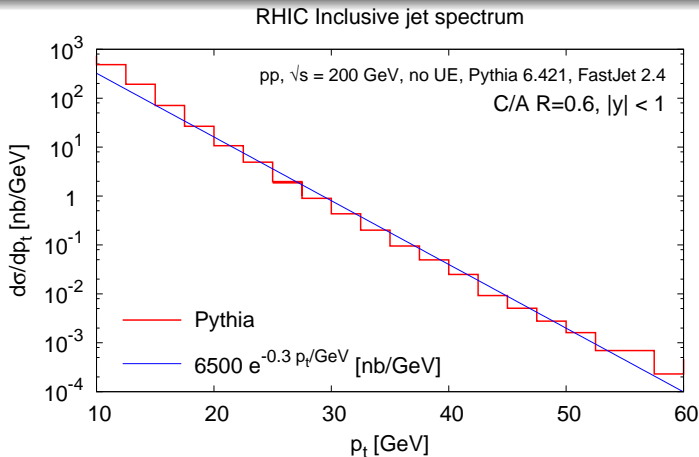
$\sigma_{UE} \simeq 8 - 10 \text{ GeV}$

→ $\sigma_{\Delta p_t} \sim \mathbf{6 - 7 \text{ GeV}}$

What impact does this have?

RHIC Inclusive jet spectrum





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

$$\exp(-ap_t) \otimes \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{\Delta p_t^2}{2\sigma^2}\right)$$

Real fluctuations not quite Gaussian
Real spectrum not quite exponential (especially at low & high p_t)
But simple approximations give instructive analytical answers

Convolution rescales spectrum by factor:

$$\exp\left(\frac{1}{2}a^2\sigma^2\right) \sim 10 \text{ for } \sigma \sim 7 \text{ GeV}$$

Convolution migrates p_t 's by

$$a\sigma^2 \sim 15 \text{ GeV for } \sigma \sim 7 \text{ GeV}$$

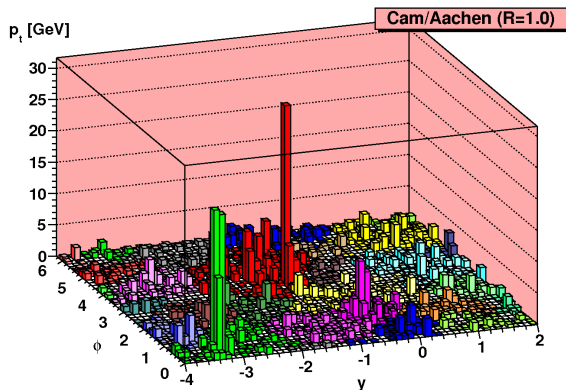
Reducing fluctuations, while
limiting bias:

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



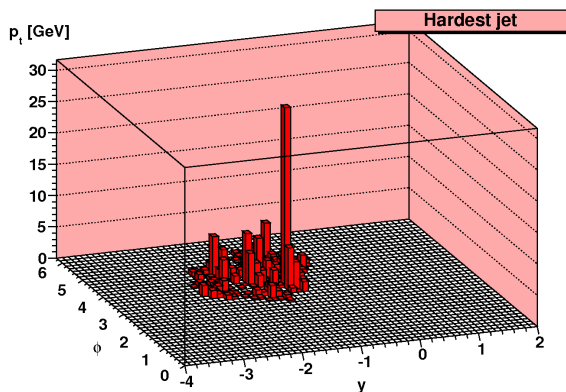
1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest "subjets" leading quark + 1 gluon
4. The result is a "filtered" jet

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

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



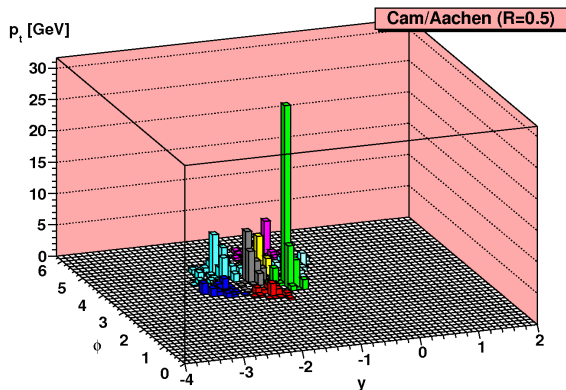
1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest "subjets" leading quark + 1 gluon
4. The result is a "filtered" jet

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

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



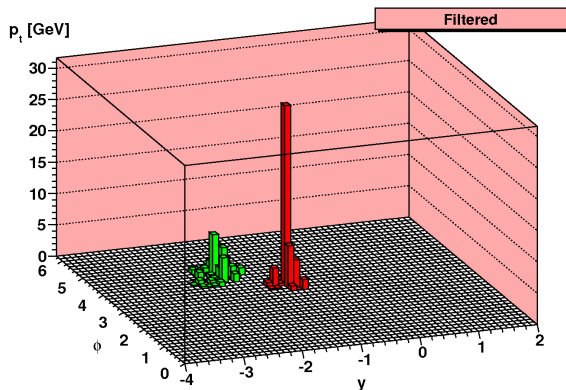
1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest "subjets" leading quark + 1 gluon
4. The result is a "filtered" jet

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

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



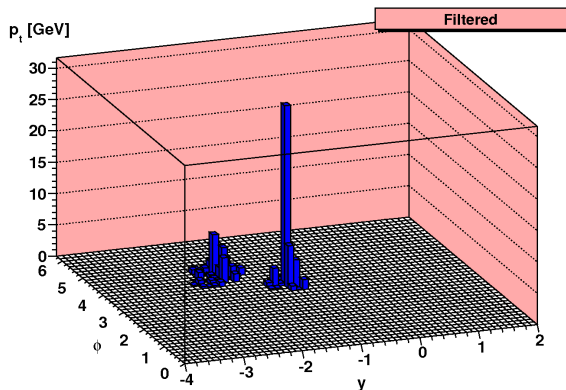
1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest “subjets”
leading quark + 1 gluon
4. The result is a “filtered” jet

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

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

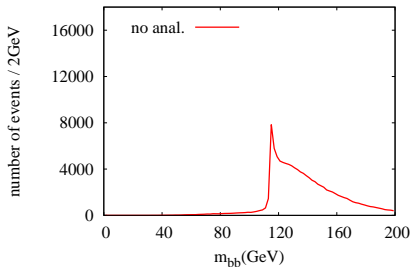


1. Consider a jet
2. View it on smaller angular resolution scale R_{filt}
3. Take (e.g.) 2 hardest “subjets”
leading quark + 1 gluon
4. The result is a “filtered” jet

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

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

Without Filtering



With Filtering

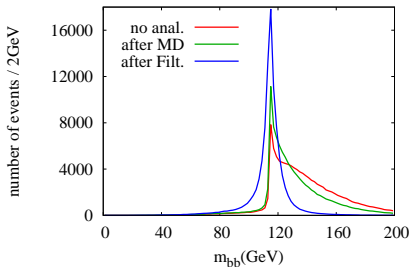
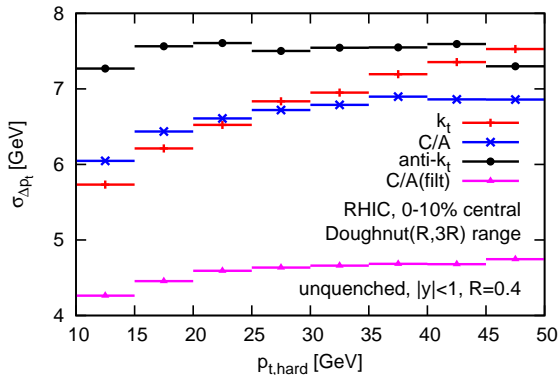


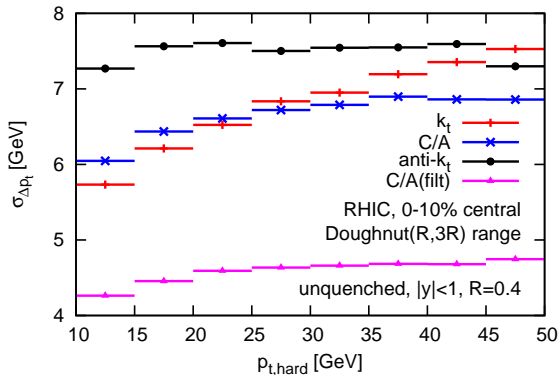
Figure from Rubin

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



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

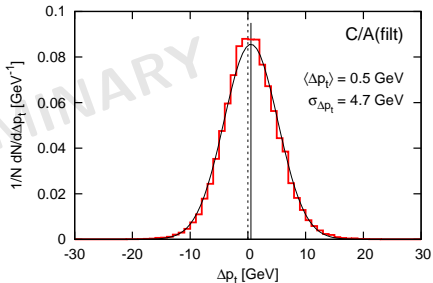
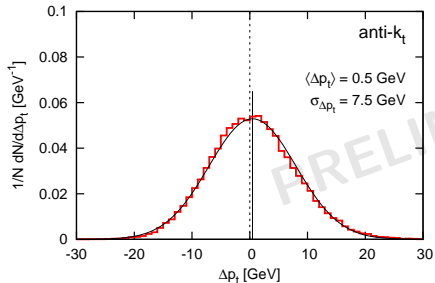
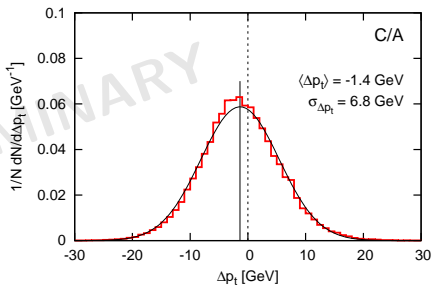
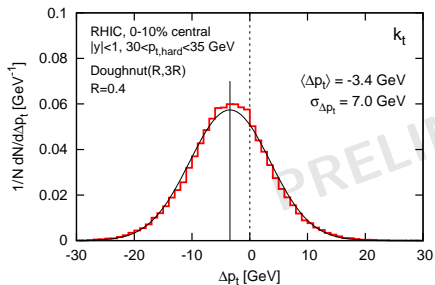


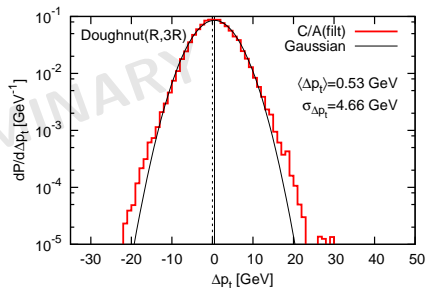
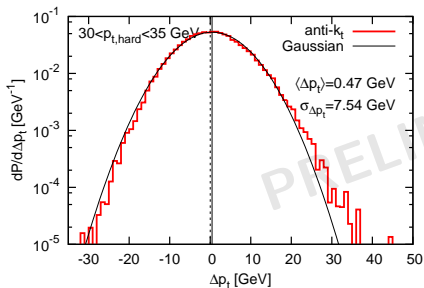
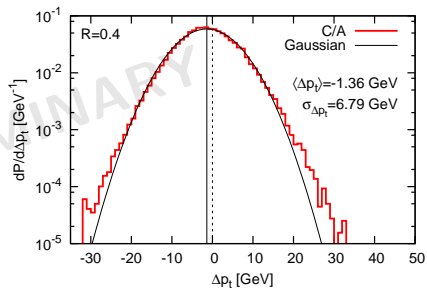
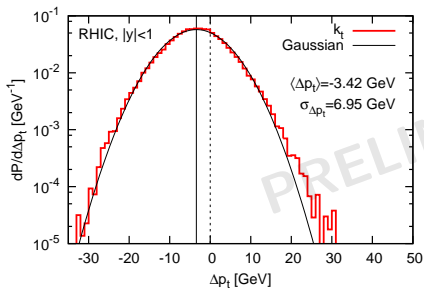
Filtering reduces jet area
by $\sim \frac{1}{2}$

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
 NB: Gaussian filtering (Cole & Lai '08) not the same thing



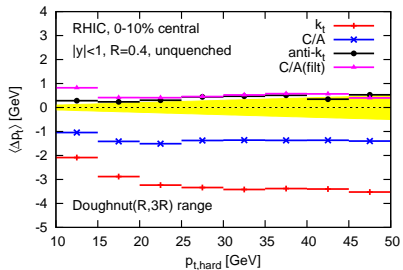


Does filtering introduce new biases in jets in quenched case?

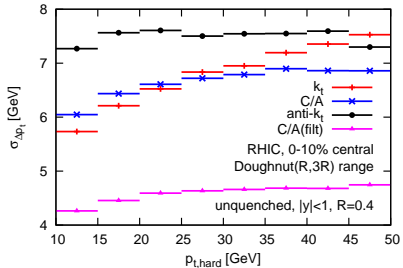
Vacuum QCD: we know how much gluon radiation we lose
QCD in medium: extra medium-induced radiation lost?

UNQUENCHED

PT SHIFT



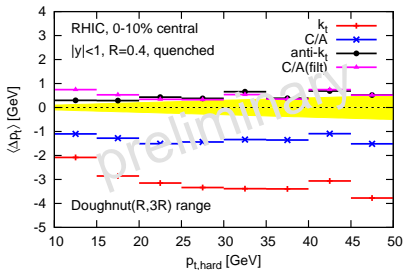
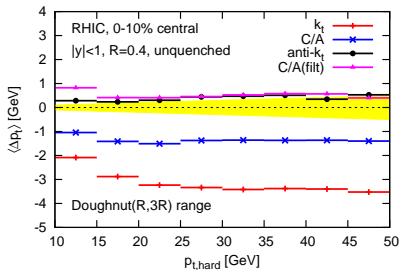
DISPERSION



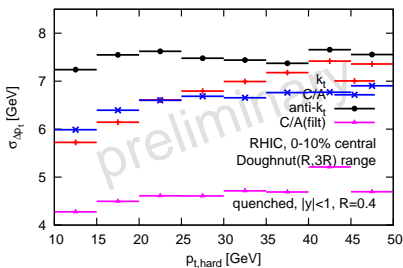
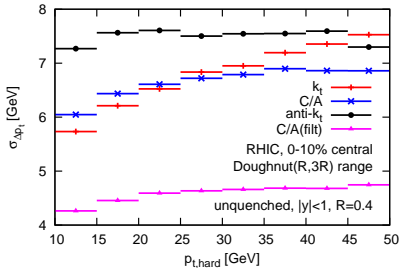
UNQUENCHED

QUENCHED

PT SHIFT

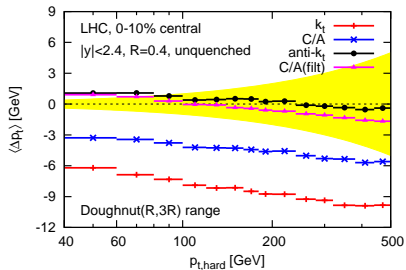


DISPERSION

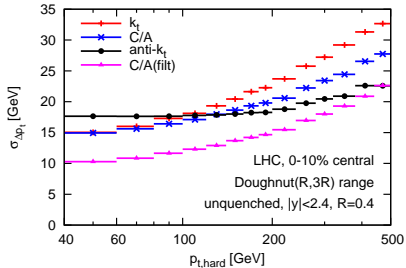


UNQUENCHED

PT SHIFT

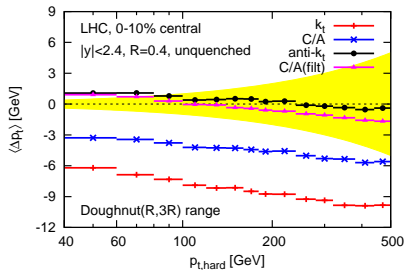


DISPERSION

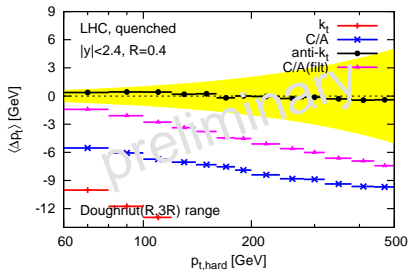


PT SHIFT

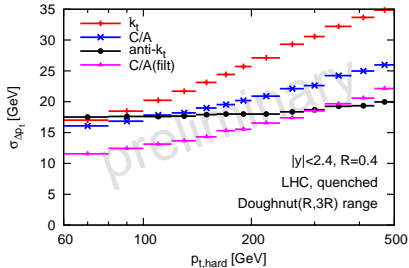
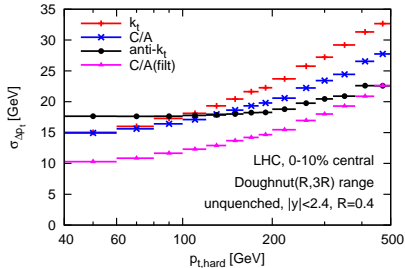
UNQUENCHED



QUENCHED



DISPERSION



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

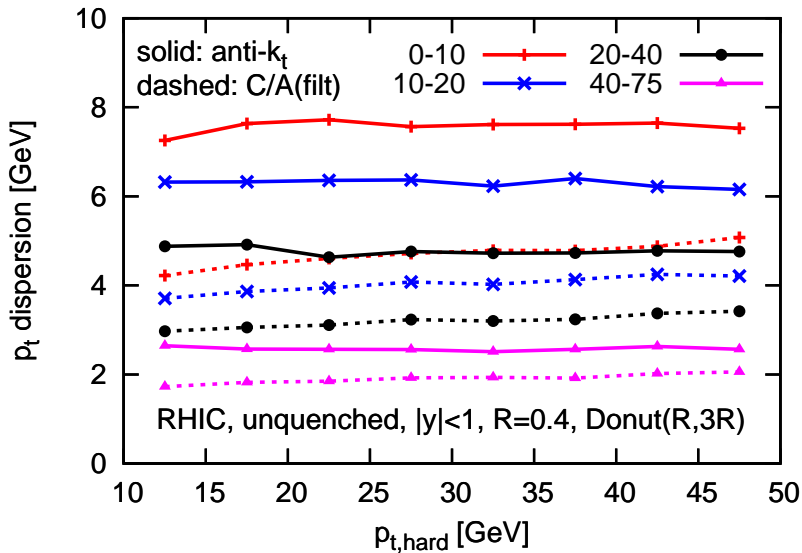
It's a tough job to accurately remove 40 GeV of noise from a 40 GeV hard jet in the context of a steeply falling cross-section.

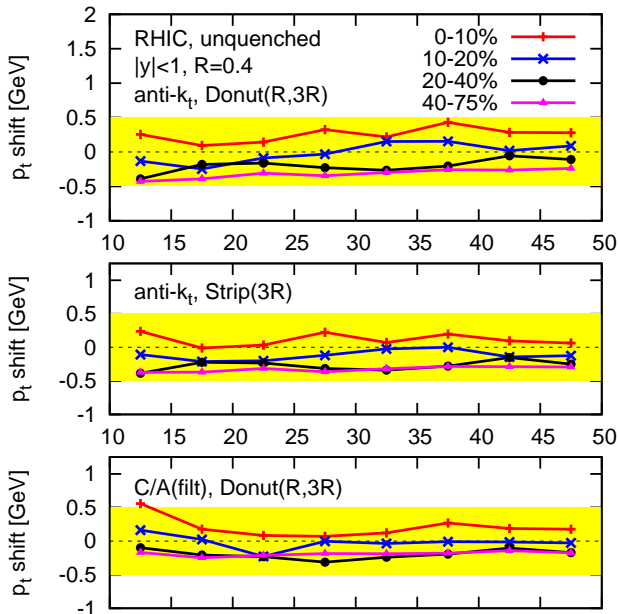
Theory calculations can guide the choices one makes

- ▶ 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



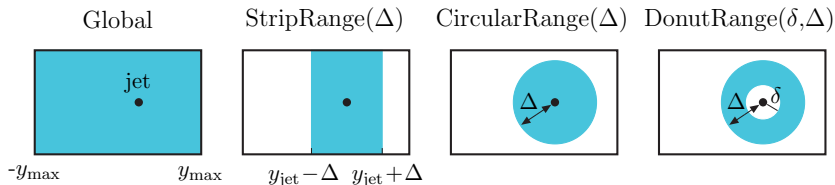


Example #2: another bias

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:

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

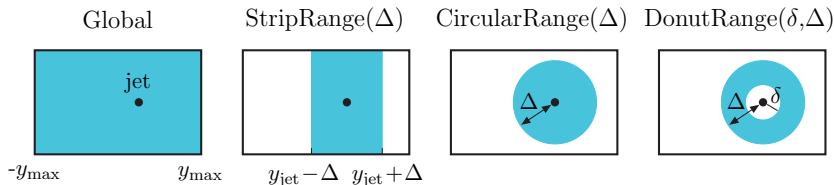
ν = number of particles / unit area

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!

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!