

# Underlying Event Subtraction in Heavy Ion Collisions

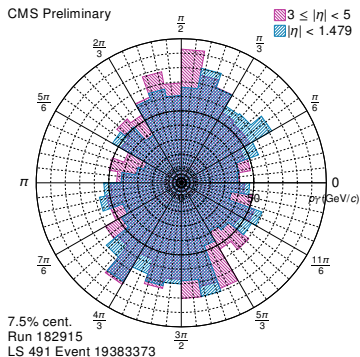
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MIT LNS

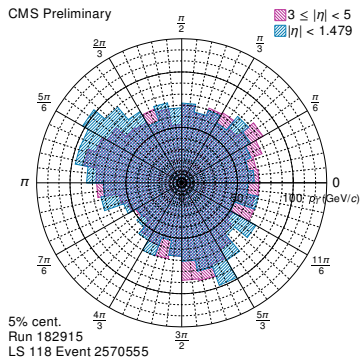
Mitigation of PU Effects at the LHC

- 1 The “old” CMS iterative pile-up algorithm
  - 2 The new CMS HF/Voronoi algorithm
- Emphasis is on pushing out physics results:
    - Little particle-level studies, write-up, public code
    - A lot of CMS GEANT results
    - Re-reconstruction campaign for 2011 PbPb data
    - ... plus first physics result to be shown next week (Quark Matter 2014 in Darmstadt, Germany)

# Heavy Ion Underlying Event



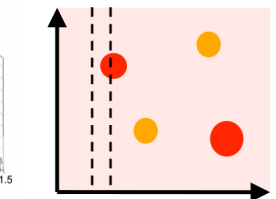
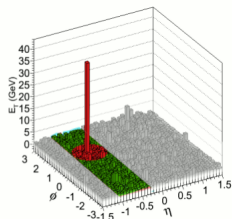
2nd harmonic



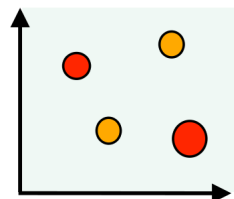
3rd harmonic

- $3 \leq |\eta| < 5$  corresponds to the CMS forward hadronic calorimeter (HF)
- Underlying event (UE) is not azimuthally symmetric, but has higher Fourier components correlated over  $\eta$
- The constant UE is parametrized in heavy ion vs. the impact parameter  $b$ , quoted in percentage of total cross section ( $0\% \Rightarrow b \approx 0$ , maximum UE;  $100\% \Rightarrow$  pp-like,  $b \rightarrow \infty$ )

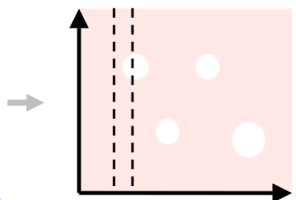
# CMS Iterative PU in Picture



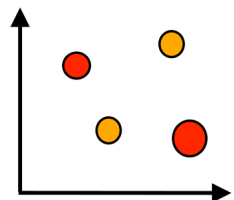
1.  $\langle E_T \rangle$  calculated in strips of  $\eta$ .  
Subtract  $\langle E_T \rangle + \sigma$



2. Run anti- $k_T$  algorithm on background-subtracted towers



3. Exclude reconstructed jets and re-estimate background



4. Re-run anti- $k_T$  algorithm to get final jets

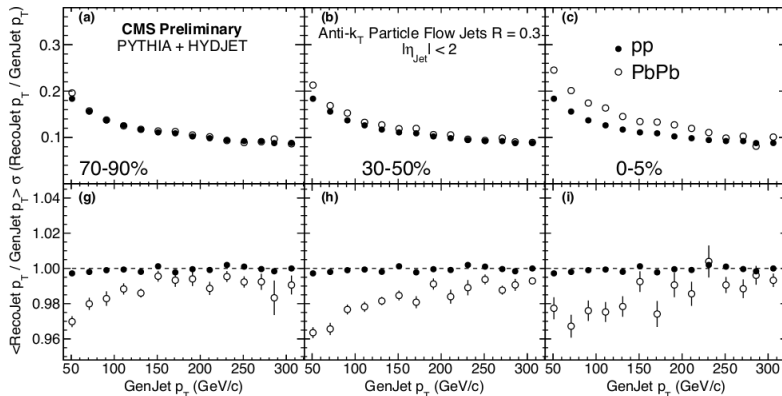
For details see:

- CMS, [arXiv:1102.1957](https://arxiv.org/abs/1102.1957)
- Kodolova et al., [EPJC 50 \(2007\) 117](https://arxiv.org/abs/0705.3802)

(M. Nguyen, Quark Matter 2012)

- Subtraction on “pseudotower” (= HCAL granularity) level
- Iteration 1:
  - Determination of  $\langle p_T \rangle$  and  $\sigma(p_T)$  of the UE
  - Subtract  $\langle p_T \rangle + n\text{SigmaPU} \times \sigma(p_T)$  (“noise suppression”), zero negative cells
- Iteration 2:
  - Exclude towers within  $\text{radiusPU}$  of a jet from iteration 1, with  $p_T > \text{jetPtMin}$ , redetermine  $\langle p_T \rangle$  and  $\sigma(p_T)$
  - Subtract  $\langle p_T \rangle + n\text{SigmaPU} \times \sigma(p_T)$
- $n\text{SigmaPU}$  serves dual purpose of actual noise suppression and a nonlinearity correction (caused by detector zero suppression).

# CMS Iterative PU Performance



- Jets are always oversubtracted by  $\approx 2 \text{ GeV}/c$
- Performance even poorer for calorimetric jets

CMS PAS-HIN-12-004

- Really two separate problems:

- 1 What amount of UE to subtract
- 2 How to subtract

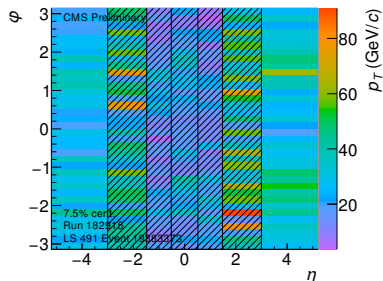
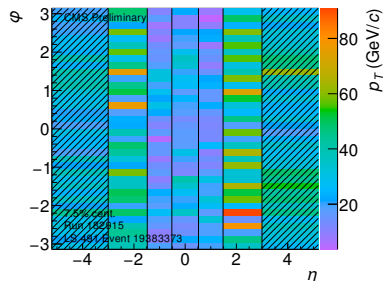
- Solution

- 1 Singular vector decomposition (SVD) modeling of the UE
- 2 Subtraction by Voronoi diagram
- 3 Equalization of positive/negative fluctuation

- Advantages:

- Arbitrary  $v_n$  subtraction (not limited to  $n = 2$  in data, but currently limited by MC)
- Full particle flow hard event, not restricted to reconstructed jets
  - Jet ID, substructure
  - Event shape, missing  $E_T$
  - Photon isolation
- Best PbPb JER currently observed in CMS GEANT

# Step 1: SVD Modeling of the Underlying Event



- Predict midrapidity UE from measured UE in the forward detectors (little jet contamination)
- Forward calorimeter have vastly different radiation depth and segmentation
- Need to model the nonlinear difference

CMS DP-2013/018



# Step 1: SVD Modeling of the Underlying Event

- Goal (RHS) is the Fourier expansion for an arbitrary midrapidity range

$$\frac{dp_T}{d\eta d\phi} = \text{Re} \left( \sum_{k=0}^{N_{\text{Fourier}}} a_k e^{ik\phi} \right) \quad (1)$$

- Approximate (predict)  $a_k$  as function of the forward HCAL (HF) Fourier decomposition

$$a_k = f_k(a_0^{\text{HF}}, a_1^{\text{HF}}, \dots, a_{N_{\text{Fourier}}}^{\text{HF}}). \quad (2)$$

- $f_k$  is expressed as

$$f_k = \sum_{l=0}^{N_{\text{Fourier}}} \sum_{l=0}^{N_{\text{order}}} c_l \frac{a_l^{\text{HF}}}{s_k} \quad (3)$$

( $s_k$  are numerical factors to scale  $100 \text{ GeV}/c \lesssim a_k^{\text{HF}} \lesssim 5000 \text{ GeV}/c$  to  $\approx 1$ )

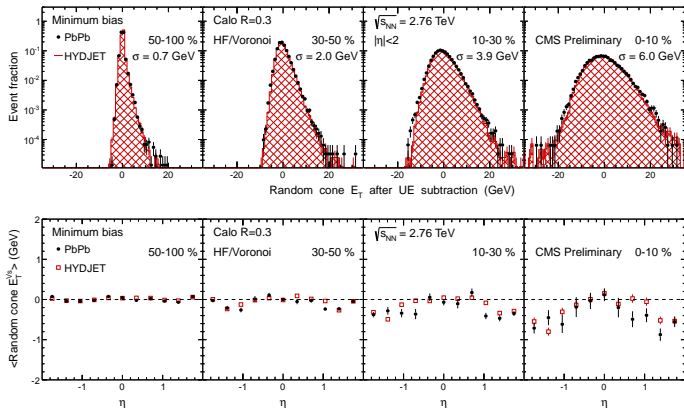
# Step 1: SVD Modeling of the Underlying Event

- Regression problem solved by SVD

$$\mathbf{f} = \mathbf{C}\hat{\mathbf{a}}^{\text{HF}}, \quad \hat{\mathbf{a}}^{\text{HF}} = \left( \frac{a_k^{\text{HF}}}{s_k} \right)_k. \quad (4)$$

- This structure is intentionally chosen to be extended later to midrapidity “extrapolation beneath the jet”  $\Rightarrow$  e.g. for pp PU.

# Data vs. MC Performance

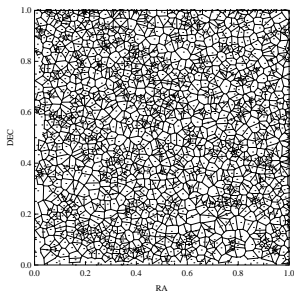


- For data, a separate training is used in step 1 (CMS GEANT is not accurate enough in forward region)
- The 0–10% there is a mean offset  $\approx 50$  GeV/c being subtracted
- Some fluctuation difference mainly from the MC event generator tuning

## Step 2: Subtraction by Voronoi Diagram

- Existing methods to avoid this problem:
  - 1 Cluster event including UE, remove soft scale at jet level (FastJet, ATLAS iterative mean)
  - 2 Convert final-state particles into a density, make the jet reconstruction commutative wrt. UE subtraction (RHIC-PHENIX Gaussian filter)
  - 3 Subtraction using fixed geometry (CMS iterative pile-up)  
Notice that this method also attempts produce a clustering that is restricted to the hard event.
- For particle-flow, need to extend (3) to:
  - 4 Associate an area to each final state particle (HF/Voronoi the first algorithm to attempt it)

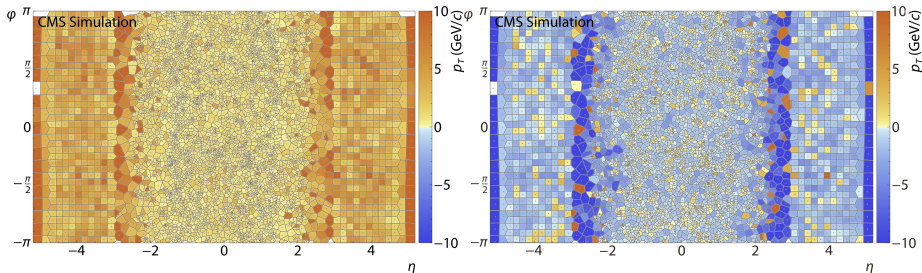
## Step 2: Subtraction by Voronoi Diagram



M. Soares-Santos et al., *Astrophys. J.* 727, 45 (2011)

- Density estimation by Voronoi diagram (point-like galaxies vs. mostly empty space) established for more than two decades in observational cosmology

## Step 2: Subtraction by Voronoi Diagram



Before UE subtraction

After UE subtraction

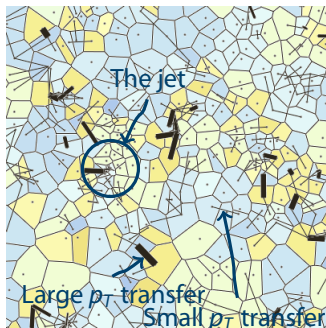
- Subtraction algorithm matches the particle position with the area where the nearest neighbor of a given point is that particle-flow candidate
- Note the 2nd Fourier component with  $\Psi^{\text{EP}} \approx 0$  that is subtracted away

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## Step 3: Equalization

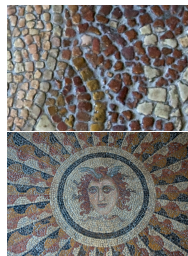
- An idea that is inspired from ATLAS early “noise tool” (now abandoned)
- Try to get a right “noise tool” without the issues how ATLAS implemented it:
  - Make it non-iterative
  - Define the goal as same energy at given (jet algorithm) distance scale, and not just remove negativity

# Step 3: Equalization in Pictures



- Ingredient #3: “Equalization” to produce positive event
- Purely a band-aid for iterative, positive- $p_T$  jet reconstruction algorithms
- Balance out negative cells to represent the same event using positive cells only
- Implement a delocalized underlying event on particle level

- Similar in idea to a mosaic:
  - Manipulation of color tiles on a small scale
  - “Reconstructs” to a correct visual impression on a coarse scale





# Step 3: Equalization in Equations

- Set of variables describing the equalization problem:

- 1 Geometry

- $R_{kj}$  is the distance between cells  $k, j$

- 2 Energy/energy transfer

- $p_j$  is the  $p_T$  within a positive cell  $j$
    - $\hat{p}_j$  is the equalized  $p_T$
    - $n_k$  is the  $p_T$  within a negative cell  $k$
    - $t_{kj}$  is the energy transfer cell  $k \leftarrow j$

- 3 "Service variables"

- $c$  is minimax cost variable
    - $u$  is the globally subtracted residual energy

## Step 3: Equalization in Equations

- Rewrite this as linear optimization

$$\underset{c, t_{kj}}{\text{minimize}} \quad c + \varepsilon \sum_{k, j, \Delta R_{kj} < R_{\text{jet}}} t_{kj} \Delta R_{kj}$$

$$\text{subject to} \quad \sum_{R_{\text{jet}}} \hat{p}_i - \sum_{R_{\text{jet}}} p_i + \sum_{R_{\text{jet}}} n_i \leq c \quad \text{around all positive regions}$$

$$\sum_{R_{\text{jet}}} \hat{p}_i - \sum_{R_{\text{jet}}} p_i + \sum_{R_{\text{jet}}} n_i \geq -c \quad \text{around all positive regions}$$

$$p_j - \sum_k t_{kj} - u = \hat{p}_i \quad \text{for positive cells } p_j$$

$$n_k + \sum_j t_{kj} - u \leq 0 \quad \text{for negative cells } n_k$$

$$c \geq 0$$

$$\hat{p}_i \geq 0$$

$$t_{kj} \geq 0$$

$$u \geq 0$$

## Step 3: Equalization in Equations

- Target function

$$\underbrace{c}_{p_T \text{ nonclosure}} + \underbrace{\varepsilon \sum_{k,j, \Delta R_{kj} < R_{\text{jet}}} t_{kj} \Delta R_{kj}}_{\text{Energy transfer cost over distance}} \quad (6)$$

- The  $\varepsilon$  term is a “tie-breaker” to handle the degeneracy, i.e. recombination over large distances are (slightly) more costly. Value e.g.  $\varepsilon = 10$  MeV (hide it in a  $p_T$  range we don't really care about).
- The minimax condition

$$-c \leq \underbrace{\sum_{R_{\text{jet}}} \hat{p}_i - \sum_{R_{\text{jet}}} p_i + \sum_{R_{\text{jet}}} n_i}_{p_T \text{ nonclosure}} \leq c \quad (7)$$

automatically finds the worst nonclosure and attempts to remove it

- With the CMS PbPb nonlinearity the minimax condition is modified to:

$$-c \leq \dots \leq \frac{c}{4} \quad (8)$$

## Step 3: Equalization in Equations

- Energy transfer and residual subtraction

$$\underbrace{p_j - \sum_k t_{kj}}_{p_T \text{ after transferring out}} - \underbrace{u}_{\text{residual subtraction}} = \underbrace{\hat{p}_i}_{\text{positive-only event } p_T} \quad \text{for positive cells } p_j$$

$$\underbrace{n_k + \sum_j t_{kj}}_{p_T \text{ after transferring in}} - \underbrace{u}_{\text{residual subtraction}} \leq \underbrace{0}_{\text{any } p_T \text{ that will be truncated to 0}} \quad \text{for negative cells } n_k$$

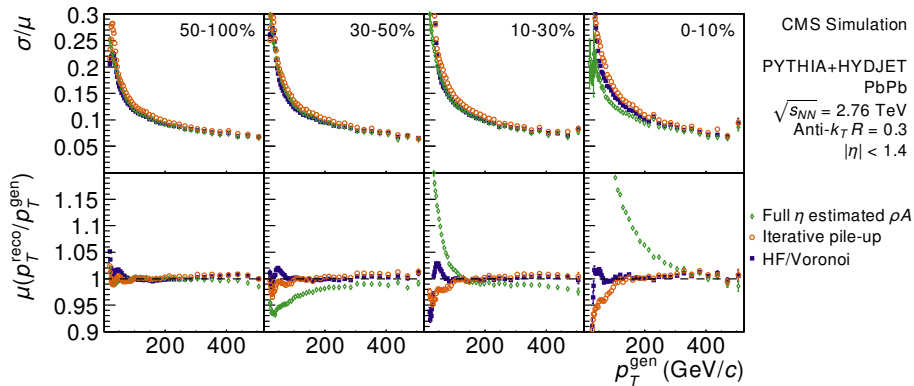
(9)

- $\hat{p}_i$  is the solution
- $c, \hat{p}_i, t_{kj}, u$  all variables that are positively constrained
- In practice we segment the event into staggered tiles to allow different  $\eta$  region having different nonclosure (staggered tiling is the “cost trick” applied twice)

# Free Parameters

- Step 1: Many ( $\approx 50k$ ) parameters in the UE model
- Step 2–3: Exactly two parameters
  - $R_{\text{jet}}$ , same as in the jet reconstruction algorithm
  - The  $\frac{1}{4}$  that is for the CMS nonlinearity

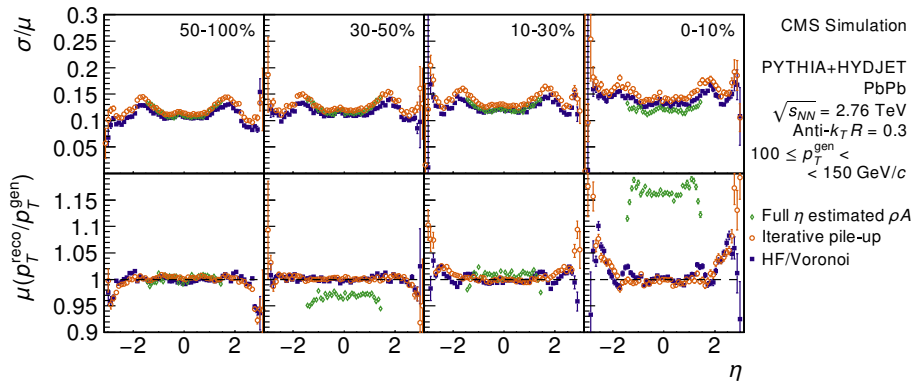
# MC Generator + GEANT Performance: Vs. $p_T^{\text{truth}}$



- $\approx 10\%$  improved mid-rapidity jet energy resolution for the new HF/Voronoi algorithm vs. old CMS' iterative pileup (PU)
- Median- $\rho_A$  is the "global rescaled" variant (there is a still-CMS-internal version with FastJet-"doughnut")

CMS DP-2013/018

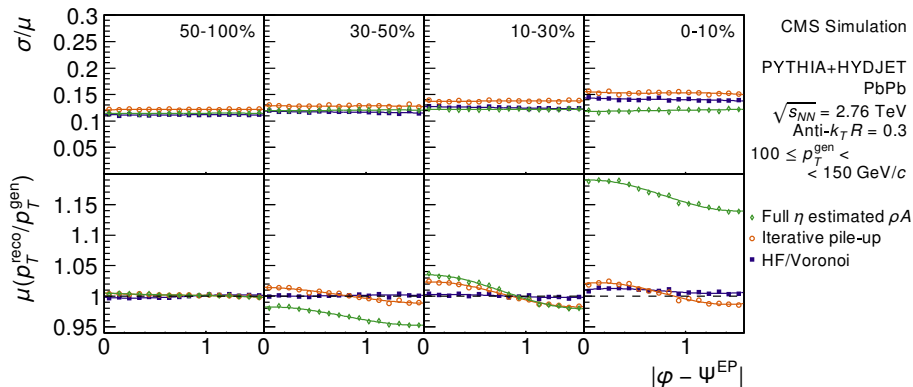
# MC Generator + GEANT Performance: Vs. $\eta$



- Nonclosure in 0–10%/endcap due to factorization between  $\eta$  and centrality dependent JEC

CMS DP-2013/018

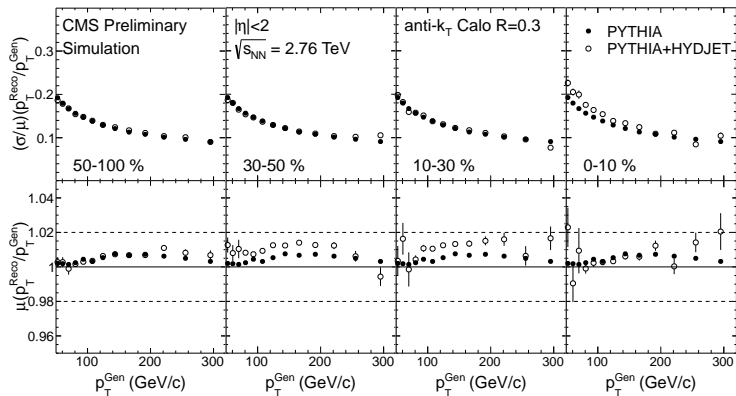
# MC Generator + GEANT Performance: Vs. $|\varphi - \Psi^{\text{EP}}|$



- $\Psi^{\text{EP}}$  is the arctan of the 2nd Fourier coefficients of  $dE_T/d\varphi$  in the HF (direction of “event plane” with maximum energy)
- Improves the previous 5–6% jet energy scale (JES) variation vs. orientation in iterative PU down to  $\approx 1\%$
- Fit to  $A + B|\varphi - \Psi^{\text{EP}}| + C \cos(2|\varphi - \Psi^{\text{EP}}|)$  for visual aid



# MC Generator + GEANT Performance: Calorimetric Jets



- The first analysis for 2014 that uses the HF/Voronoi algorithm happens to be jet-track correlation that needs calorimetric jets (track bias in particle-flow)
- Performance better than past iterative PU with particle-flow

CMS PAS-HIN-14-010

# Summary

- First algorithm for a “non-jet” UE subtraction in particle-flow environment
- New ideas for UE determination:
  - Data-driven prediction of midrapidity UE from forward detectors
  - Applicable for calorimeter with nonlinearity
- New ideas for UE subtraction:
  - Arbitrary geometry (e.g. particle-flow)
  - Full conversion to a hard event without negative particles, that clusters into the correct jet energy
- Studied and commissioned in CMS, first physics results already public
- Standalone/particle-level code and write-up in preparation (apologies!)

# Part I

## Backup

# The “Full $\eta$ estimated $\rho_A$ ” Being Compared to

- Stock “global rescaled”  $\rho_A$ -subtraction in FastJet 3.0.3
- $\rho$  estimation using  $k_T R = 0.6$ , within  $|\eta| < 5.191$
- $\rho$  rescaling for  $|\eta| < 1.479$  (largest acceptance still like an order-4 polynomial)
- Same calibration procedure (note: L1–L2 are factorized)