



Nonflow Subtraction Beyond Two-Particle Correlations

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Outline

- Flow vs nonflow.
- Nonflow subtraction method in two-particle correlation
- Motivation for nonflow subtraction in multi-particle correlations
- Subtraction framework
 - Superposition principle and practical considerations
 - Guiding principles and classification of methods
 - Constructing the scaling factors using pp as baseline
- Performance in O+O and d+Au collisions with $\langle v_2^2 \rangle$, $\langle v_2^2 \delta p_T \rangle$, $c_2\{4\}$
- Summary and outlook

Flow vs nonflow

- Notoriously subtle issue in heavy ion collisions.
- Flow: ideally correlations arise from hydrodynamic response to the collision geometry.
 - Not a one-body distribution: factorization breaking in p_T and η require subleading flow modes.
 - Can be hydro but not geometry driven: jets thermalized in the medium creating jet wake.
 - Non-thermalized modes/incomplete geometry response: path-length dep. of v_n for high- p_T jets.
- Nonflow: ideally correlations of particles unrelated with the collision geometry.
 - Rely on practical definition of sources: jets, resonance decays, HBT....
 - All respect Global Momentum Conservation (GMC), appearing as away-side long-range dipolar shape.

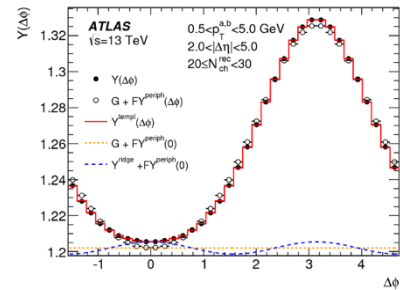
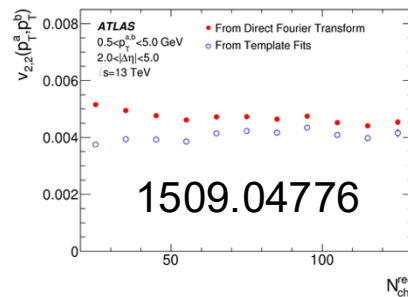
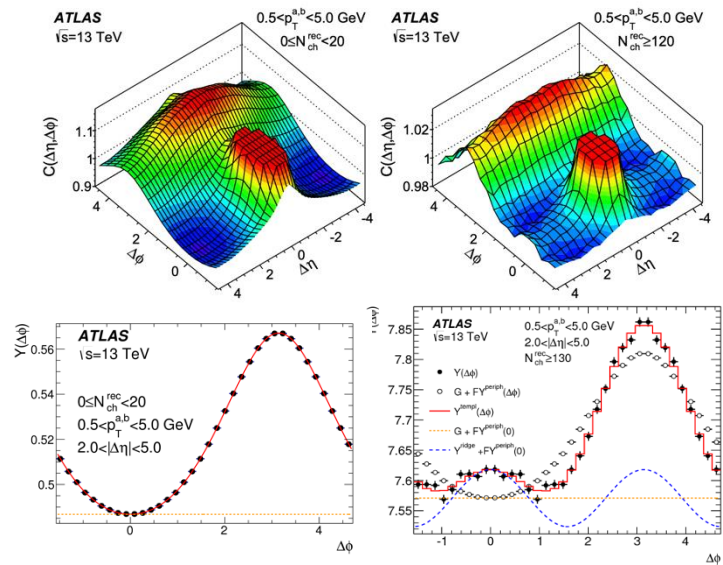
Flow vs nonflow

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- Nonflow: ideally correlations of particles unrelated with the collision geometry.
 - Rely on practical definition of sources: jets, resonance decays, HBT....
 - All respect Global Momentum Conservation (GMC), appearing as away-side long-range dipolar shape.
- Separation of flow and nonflow.
 - Particles usually participate both flow and nonflow: a flowing ρ^0 decays to two-pions.
 - Must be defined on m-particle correlation statistically: some multiplets attributed to flow and others to nonflow.
- Nonflow subtraction assume is unmodified
 - Usually implies that nonflow subtraction works at leading-order, usually breaks down if nonflow signal dominates

How to handle nonflow

- First, separate correlation into short- and long-range correlations: SRC and LRC
 - SRC contains complex thermalization physics, flow and nonflow separation not easy.
 - LRC cleaner but still affected by away-side jets.
- Second, subtract nonflow on LRC: get rid of jets, assuming unmodified.
 - Subtraction can't be trusted when nonflow signal is very large: might be ok to subtract 50% nonflow to get 50% flow signal. But should be extremely careful when you subtract 90% and interpret the 10% as flow (method systematics are important).
 - Cause interpretation issues: Multi-particle correlation from CMS, ATLAS high p_T , e^+e^- ridge.
- Applicability of subtraction in small systems: low p_T region, high-multiplicity events.

Caution on subtraction

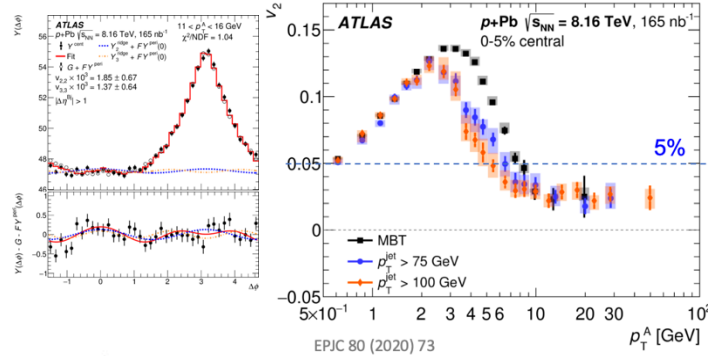


Dominated by nonflow from c_1 , nonflow in c_2 is small

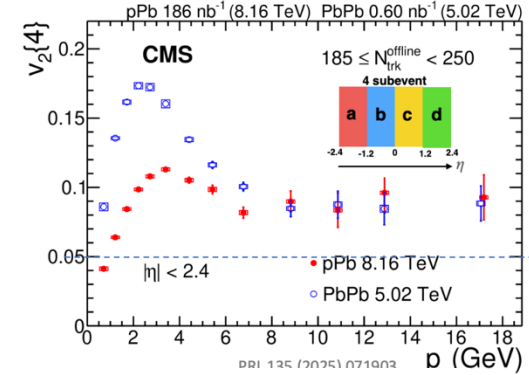
Caution on subtraction

Potential method biases everywhere, need to be carefully studied!

2-particle corr. w/ template nonflow sub.



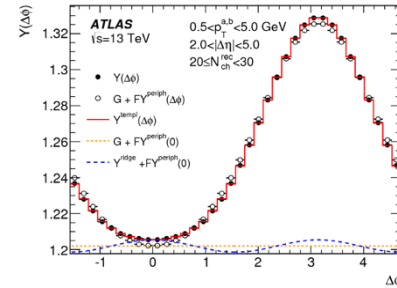
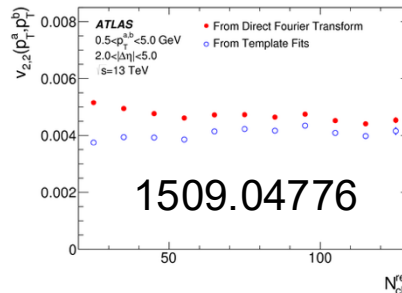
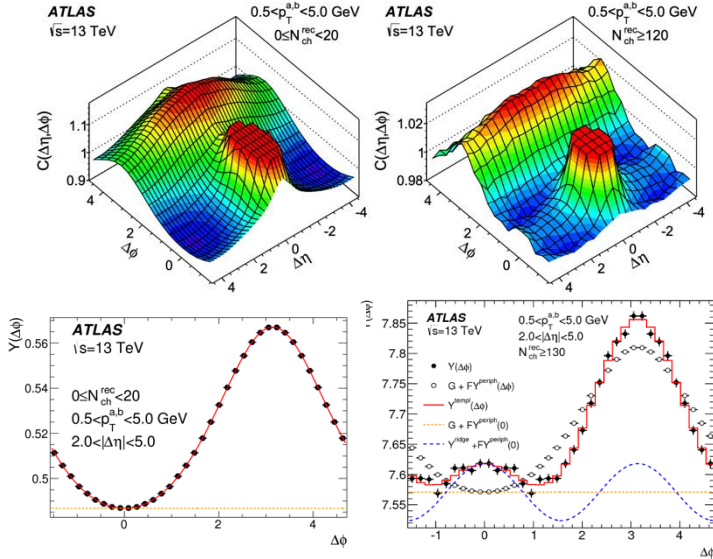
4-particle corr. w/ 4-subevent



Dominated by nonflow from c_1 , but nonflow in c_2 is also large

Factorization assumption broken

$$v_n\{4\} = -d_n\{4\}/(-c_n\{4\})^{3/4}$$



Dominated by nonflow from c_1 , nonflow in c_2 is small

Nonflow subtraction method in 2PC

- Assume nonflow shape is unmodified and can be estimated from LM or pp events

$$\langle v_n^2 \rangle (N) = \langle v_n^2 \rangle^{\text{obs}} (N) - f(N) \times \langle v_n^2 \rangle_{\text{LM}}^{\text{obs}}$$

- $f(N) \sim 1/N_{\text{ch}}$, a.la. independent source scenario for nonflow.
- Selection basis modify this scaling, especially if $p(N_{\text{ch}})$ has a sharp fall off:
 - Nonflow enhanced at large N_{ch} and suppressed at small N_{ch} . **Bias is stronger for near-side jet**
- Three nonflow subtraction methods assuming zero-flow in LM.

$$f_1 = N_{\text{LM}}/N$$

c_0 -scaling

$$f_2 = \langle v_1^2 \rangle^{\text{obs}} / \langle v_1^2 \rangle_{\text{LM}}^{\text{obs}} \approx f_1 \times \frac{Y_{\text{away}}}{Y_{\text{away,LM}}}$$

c_1 -scaling

$$f_3 = f_1 \times \frac{Y_{\text{near}}}{Y_{\text{near,LM}}}$$

Near_{jet}-scaling

Nonflow subtraction method in 2PC

■ LM may have flow → Template method

- Fourier expansion of correlation function:

$$Y(\Delta\phi) = N^a \left(1 + 2 \sum_{n=1}^{\infty} \langle v_n^2 \rangle^{\text{obs}} \cos n\Delta\phi \right)$$

- Peripheral subtraction removes the dipole component $Y(\Delta\phi) - F Y(\Delta\phi)_{\text{LM}} = G \left(1 + 2 \sum_{n=2}^{\infty} \langle v_n^2 \rangle \cos n\Delta\phi \right)$

$$N^a \langle v_1^2 \rangle^{\text{obs}} = F N_{\text{LM}}^a \langle v_1^2 \rangle_{\text{LM}}^{\text{obs}} \implies F = f_2/f_1 \quad G = N^a - F N_{\text{LM}}^a = N^a (1 - f_2)$$

- Template subtracted result $\langle v_n^2 \rangle^{\text{tmp}} = \frac{\langle v_n^2 \rangle^{\text{obs}} - f_2 \langle v_n^2 \rangle_{\text{LM}}^{\text{obs}}}{1 - f_2} = \frac{\langle v_n^2 \rangle - f_2 \langle v_n^2 \rangle_{\text{LM}}}{1 - f_2}$ If nonflow all subtracted

Nonflow subtraction method in 2PC

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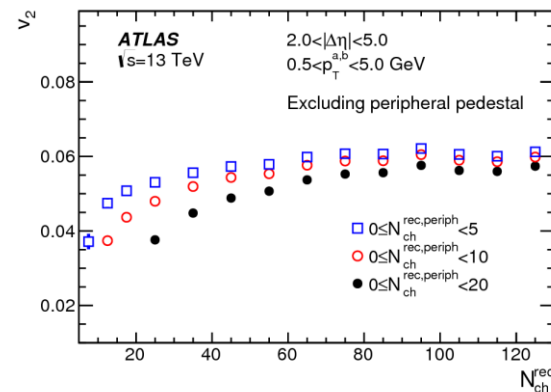
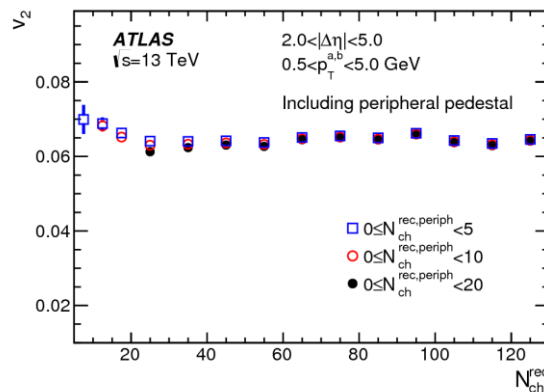
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■ Possible bias

- $v_n > v_{n\text{LM}}$, underestimate
- $v_n < v_{n\text{LM}}$, overestimate

- In pp: systematic trends of lower LM, shows that flow is $\sim N_{\text{ch}}$ independent in LM $\rightarrow$ no bias.

- Not the case in pPb and PbPb



Nonflow subtraction method in 2PC

■ LM may have flow → Template method

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■ Rewrite the equation

$$\langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{tmp}} - f_2 \left(\langle v_n^2 \rangle^{\text{tmp}} - \langle v_n^2 \rangle_{\text{LM}} \right)$$

■ Improved template: correct N_{ch} -dependence of flow starting with the 2nd bin

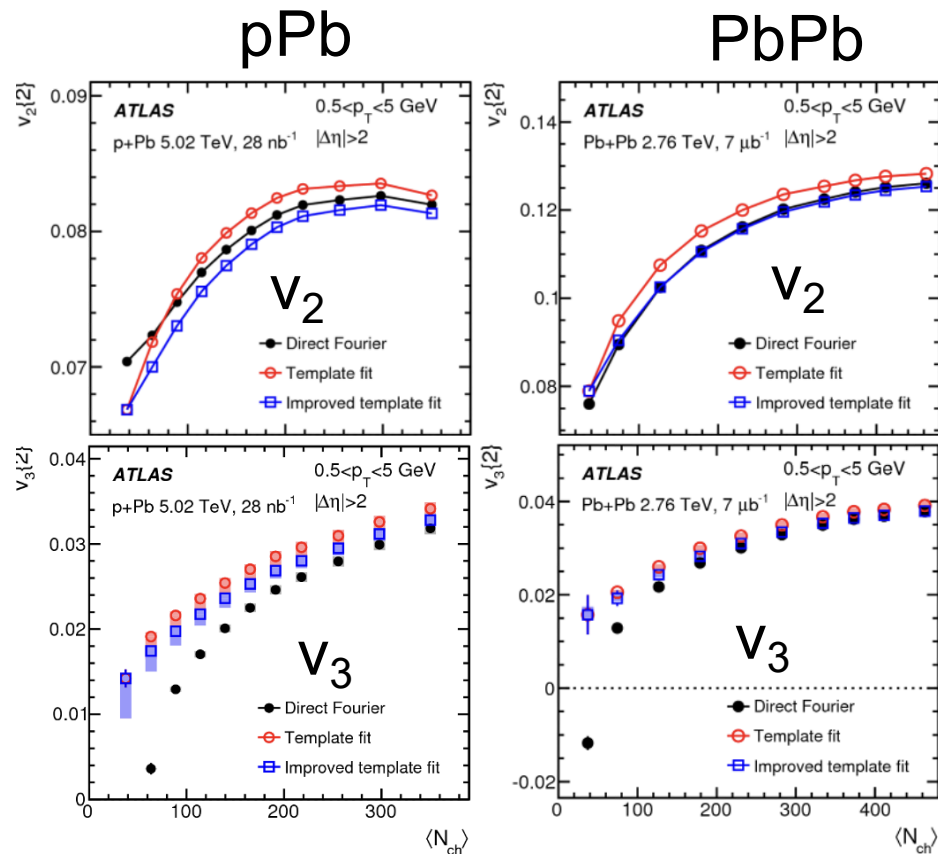
$$\langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{tmp}} - f_{2,\text{nextLM}} \left(\langle v_n^2 \rangle^{\text{tmp}} - \langle v_n^2 \rangle_{\text{nextLM}}^{\text{tmp}} \right)$$

Nonflow subtraction method in 2PC

$$\langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{tmp}} - f_{2,\text{nextLM}} \left(\underbrace{\langle v_n^2 \rangle^{\text{tmp}} - \langle v_n^2 \rangle_{\text{nextLM}}^{\text{tmp}}}_{\text{Account for } N_{\text{ch}} \text{ dependence of } v_n} \right)$$

Correction is more important for v_3

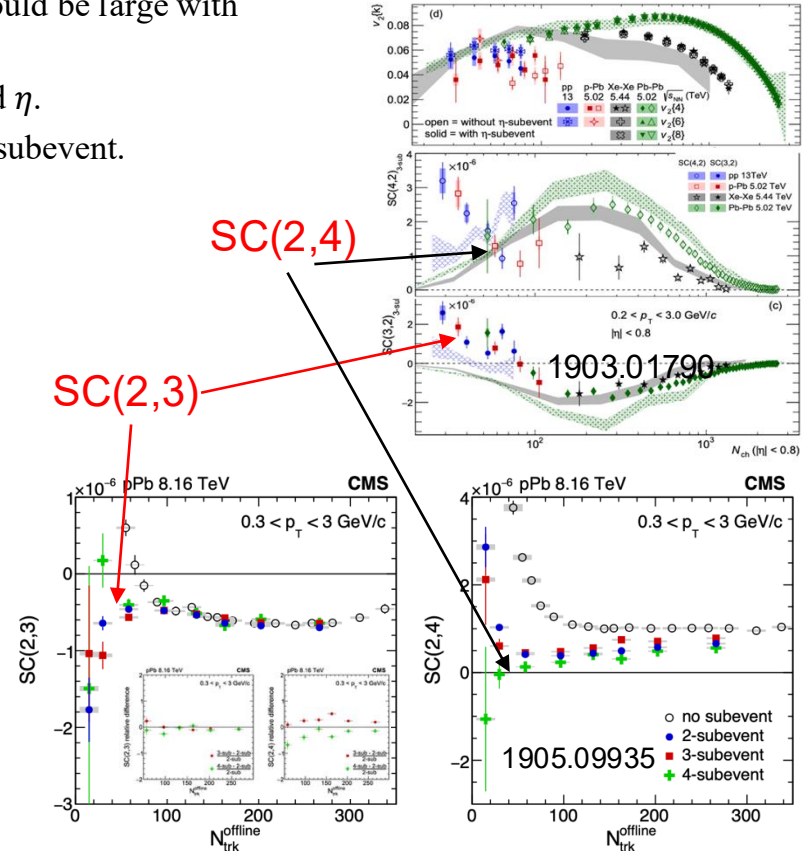
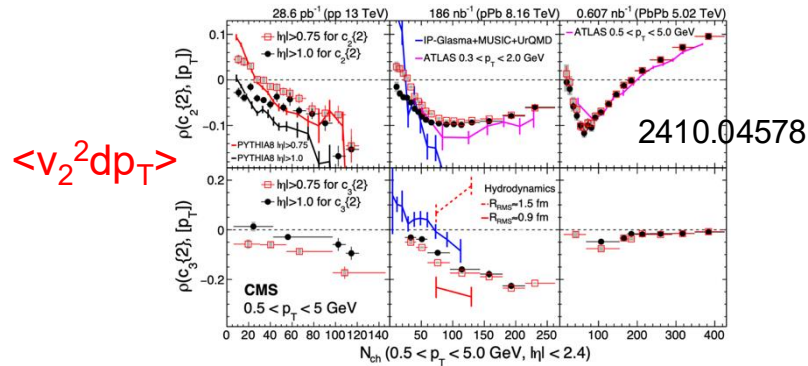
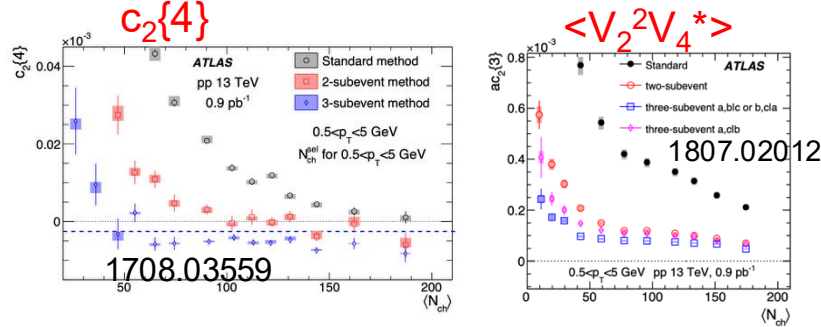
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Why nonflow subtraction in multi-particle correlations

■ Use multi-particle correlation to show flow is collective

- Quantify residual nonflow after subevent requirement, which could be large with complex behavior, results uninterpretable.
- Distinguish residual nonflow from flow decorrelations in p_T and η .
- Subevent also throw away statistics $\rightarrow$ a factor of $4^4/4!=11$ in 4 subevent.



Extend subtraction to m-particle correlations

- Extends along the five nonflow sub methods based on scaling: $1/N^{m-1}$.
- Nonflow estimator: correlators containing v_1 is dominated by nonflow (due to the sign change around $\langle p_T \rangle$, genuine p_T -integrated v_1 signal should be small).

$$\begin{aligned}
 \langle v_1^2 \rangle &\sim 1/N & \langle v_1^2 \delta p_T \rangle &\sim 1/N^2 & c_1\{4\} &= 2\langle v_1^2 \rangle^2 - \langle v_1^4 \rangle \sim 1/N^3 \\
 \langle V_1^2 V_2^* \rangle &\sim 1/N^2 & \langle V_1 V_2 V_3^* \rangle &\sim 1/N^2 & sc\{1, n\} &= \langle v_1^2 v_n^2 \rangle - \langle v_1^2 \rangle \langle v_n^2 \rangle \sim 1/N^3
 \end{aligned}$$

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- Method1: nonflow scales as $1/N^{m-1}$:

$$\begin{aligned} \langle v_n^2 \rangle &= \langle v_n^2 \rangle^{\text{obs}} - N_{2,\text{pp}}/N_2 \langle v_n^2 \rangle_{\text{pp}}^{\text{obs}}, \\ \langle v_n^2 \delta p_T \rangle &= \langle v_n^2 \delta p_T \rangle^{\text{obs}} - (N_{3,\text{pp}}/N_3)^2 \langle v_n^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}, \\ c_n\{4\} &= c_n\{4\}^{\text{obs}} - (N_{4,\text{pp}}/N_4)^3 c_n\{4\}_{\text{pp}}^{\text{obs}}. \end{aligned}$$

Extend subtraction to multi-particle correlations

■ Method2a: using native estimators

$$c_n\{4\} = c_n\{4\}^{\text{obs}} - \frac{c_1\{4\}^{\text{obs}}}{c_1\{4\}_{\text{LM}}^{\text{obs}}} c_n\{4\}_{\text{LM}}^{\text{obs}}$$

$$\langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{obs}} - \frac{\langle v_1^2 \rangle^{\text{obs}}}{\langle v_1^2 \rangle_{\text{LM}}^{\text{obs}}} \langle v_n^2 \rangle_{\text{LM}}^{\text{obs}} \quad \langle v_n^2 \delta p_T \rangle = \langle v_n^2 \delta p_T \rangle^{\text{obs}} - \frac{\langle v_1^2 \delta p_T \rangle^{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}}} \langle v_n^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}}$$

■ Method2b: using estimators of same order

$$\langle V_2^2 V_4^* \rangle = \langle V_2^2 V_4^* \rangle^{\text{obs}} - \frac{\langle V_1^2 V_2^* \rangle^{\text{obs}}}{\langle V_1^2 V_2^* \rangle_{\text{LM}}^{\text{obs}}} \langle V_2^2 V_4^* \rangle_{\text{LM}}^{\text{obs}} \quad \langle V_2 V_3 V_5^* \rangle = \langle V_2 V_3 V_5^* \rangle^{\text{obs}} - \frac{\langle V_1 V_2 V_3^* \rangle^{\text{obs}}}{\langle V_1 V_2 V_3^* \rangle_{\text{LM}}^{\text{obs}}} \langle V_2 V_3 V_5^* \rangle_{\text{LM}}^{\text{obs}}$$

■ Method2c: using mixed-order estimators

$$\langle v_n^2 \delta p_T \rangle = \langle v_n^2 \delta p_T \rangle^{\text{obs}} - \frac{\left(\langle v_1^2 \rangle^{\text{obs}} \right)^2}{\left(\langle v_1^2 \rangle_{\text{LM}}^{\text{obs}} \right)^2} \langle v_n^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}} \quad \langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{obs}} - \left(\frac{\langle v_1^2 \delta p_T \rangle^{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}} \right)^{1/2} \langle v_n^2 \rangle_{\text{pp}}^{\text{obs}}$$

Extend subtraction to multi-particle correlations

■ Method2a: using native estimators

$$c_n\{4\} = c_n\{4\}^{\text{obs}} - \frac{c_1\{4\}^{\text{obs}}}{c_1\{4\}_{\text{LM}}^{\text{obs}}} c_n\{4\}_{\text{LM}}^{\text{obs}}$$

$$\langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{obs}} - \frac{\langle v_1^2 \rangle^{\text{obs}}}{\langle v_1^2 \rangle_{\text{LM}}^{\text{obs}}} \langle v_n^2 \rangle_{\text{LM}}^{\text{obs}} \quad \langle v_n^2 \delta p_T \rangle = \langle v_n^2 \delta p_T \rangle^{\text{obs}} - \frac{\langle v_1^2 \delta p_T \rangle^{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}}} \langle v_n^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}}$$

■ Method2b: using estimators of same order

$$\langle V_2^2 V_4^* \rangle = \langle V_2^2 V_4^* \rangle^{\text{obs}} - \frac{\langle V_1^2 V_2^* \rangle^{\text{obs}}}{\langle V_1^2 V_2^* \rangle_{\text{LM}}^{\text{obs}}} \langle V_2^2 V_4^* \rangle_{\text{LM}}^{\text{obs}} \quad \langle V_2 V_3 V_5^* \rangle = \langle V_2 V_3 V_5^* \rangle^{\text{obs}} - \frac{\langle V_1 V_2 V_3^* \rangle^{\text{obs}}}{\langle V_1 V_2 V_3^* \rangle_{\text{LM}}^{\text{obs}}} \langle V_2 V_3 V_5^* \rangle_{\text{LM}}^{\text{obs}}$$

■ Method2c: using mixed-order estimators

$$\langle v_n^2 \delta p_T \rangle = \langle v_n^2 \delta p_T \rangle^{\text{obs}} - \frac{\left(\langle v_1^2 \rangle^{\text{obs}} \right)^2}{\left(\langle v_1^2 \rangle_{\text{LM}}^{\text{obs}} \right)^2} \langle v_n^2 \delta p_T \rangle_{\text{LM}}^{\text{obs}} \quad \langle v_n^2 \rangle = \langle v_n^2 \rangle^{\text{obs}} - \left(\frac{\langle v_1^2 \delta p_T \rangle^{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}} \right)^{1/2} \langle v_n^2 \rangle_{\text{pp}}^{\text{obs}}$$

- Nonflow subtraction on cumulants not moments (avoid flow-nonflow mixing).

Caveat:

- Proliferation of methods, what is the best method for each observable?
→ Need a detailed HIJING study to check the residual subtractions
- Account for selection bias on N

Multiplicity weight

Plot the scaled observables $\mathcal{O}\{m\}^{1/(m-1)} \sim 1/N$ $\mathcal{O} = 1/N, \langle v_2^2 \rangle, \langle v_2^2 \delta p_T \rangle, \langle v_1^2 \rangle, \langle v_1^2 \delta p_T \rangle, c_2\{4\}$

2PC scales with $1/N$, so we should plot at the event with the similar level of non-flow

$$\frac{1}{N_2} = \left\langle \frac{1}{N} \right\rangle = \frac{\sum_i 1/N_i N_i(N_i - 1)}{\sum_i N_i(N_i - 1)} \quad N_2 \approx \frac{\sum_i N_i^2}{\sum_i N_i}$$

m-particle correlation scales with $1/N^{m-1}$, so we should plot at the x-axis value N_m .

$$\frac{1}{N_m^{m-1}} = \left\langle \frac{1}{N^{m-1}} \right\rangle = \frac{\sum_i 1/N_i^{m-1} \times N_i(N_i - 1) \dots (N_i - m + 1)}{\sum_i N_i(N_i - 1) \dots (N_i - m + 1)} \quad N_m^{m-1} = \frac{\sum_i N_i(N_i - 1) \dots (N_i - m + 1)}{\sum_i N_i} \approx \frac{\sum_i N_i^m}{\sum_i N_i}$$

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Why they show up like this: consider N particles from N_s sources with k particles in each source

$$N_i = N_{s,i} k$$

$$\mathcal{O}\{m\} = \frac{\sum_i N_s k(k-1) \dots (k-m+1) \hat{o}}{\sum_i N_i(N_i-1) \dots (N_i-m+1)} \approx \frac{\sum_i N_{s,i} k_m \hat{o}}{\sum_i N_i^m} = \frac{\sum_i N_{s,i} k_m \hat{o}}{\sum_i N_{s,i}^m k^m}$$

hence
$$N_m^{m-1} \mathcal{O}\{m\} = \frac{\sum_i N_i^m}{\sum_i N_i} \mathcal{O}\{m\} = \frac{\sum_i N_i^m}{\sum_i N_i} \frac{\sum_i N_{s,i} k_m \hat{o}}{\sum_i N_i^m} = \frac{\sum_i N_{s,i} k_m \hat{o}}{\sum_i N_{s,i} k} \sim \text{const}$$

General formula for nonflow subtraction (using pp)

- Expected scaling behavior $N_m \mathcal{O}\{m\}^{1/(m-1)} \approx N_{m,\text{pp}} \mathcal{O}\{m\}_{\text{pp}}^{1/(m-1)}$

- Use ε (contains v_1) to estimate the scale factor for nonflow

$$\mathcal{O}\{m\} = \mathcal{O}\{m\}^{\text{obs}} - f_{\mathcal{O}}^{\varepsilon} \times \mathcal{O}\{m\}_{\text{pp}}^{\text{obs}} \quad f_{\mathcal{O}}^{\varepsilon} = \left(\left(\frac{\mathcal{E}\{k\}^{\text{obs}}}{\mathcal{E}\{k\}_{\text{pp}}^{\text{obs}}} \right)^{1/(k-1)} \underbrace{\frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \cdot \frac{N_k}{N_m}}_{W_{m,k}} \right)^{m-1}$$

- Multiplicity weight bias depends on the order of correlation $W_{m,k} \equiv \frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \cdot \frac{N_k}{N_m} \approx \frac{N_{m,\text{pp}}}{N_{k,\text{pp}}}$

- We can extend to the template method
$$\mathcal{O}\{m\}^{\text{tmp}} = \frac{\mathcal{O}\{m\}^{\text{obs}} - f_{\mathcal{O}}^{\varepsilon} \mathcal{O}\{m\}_{\text{pp}}^{\text{obs}}}{1 - f_{\mathcal{O}}^{\varepsilon}}$$

- as well as the improved template method

$$\mathcal{O}\{m\} = \mathcal{O}^{\text{sub}}\{m\}^{\text{tmp}} - f_{\mathcal{O}, \text{nextLM}}^{\varepsilon} \left(\mathcal{O}^{\text{sub}}\{m\}^{\text{tmp}} - \mathcal{O}^{\text{sub}}\{m\}_{\text{nextLM}}^{\text{tmp}} \right)$$

Dataset and analysis

- 5.36 TeV: pp and O+O; 200 GeV: pp, O+O and dAu.
 - 2000M events are generated for each datasets
- Selection: 0.2-2 GeV, $|\eta| < 2.5$ (LHC), $|\eta| < 1.5$ (RHIC)
 - Number of particles denoted as N.
- Two subevents: $\eta_a < -0.5$, $\eta_b > 0.5$.
- We focus on $\langle v_2^2 \rangle$, $\langle v_2^2 \delta p_T \rangle$, and $c_2\{4\}$

Three choices of reference ε

Stress test the subtraction method

$$\mathcal{O}\{m\} = \mathcal{O}\{m\}^{\text{obs}} - f_{\mathcal{O}}^{\varepsilon} \times \mathcal{O}\{m\}_{\text{pp}}^{\text{obs}}$$

$$f_{\mathcal{O}}^{\varepsilon} = \left(\left(\frac{\mathcal{E}\{k\}^{\text{obs}}}{\mathcal{E}\{k\}_{\text{pp}}^{\text{obs}}} \right)^{1/(k-1)} \underbrace{\frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \cdot \frac{N_k}{N_m}}_{W_{m,k}} \right)^{m-1}$$

Three observables

	$\mathcal{E} = 1/N$	$\langle v_1^2 \rangle$	$\langle v_1^2 \delta p_T \rangle_{\text{ling}}$
$\langle v_2^2 \rangle$	$\frac{N_{2,\text{pp}}}{N_2}$	$\frac{\langle v_1^2 \rangle_{\text{obs}}}{\langle v_1^2 \rangle_{\text{pp}}^{\text{obs}}}$	$\left(\frac{\langle v_1^2 \delta p_T \rangle_{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}} \right)^{1/2} W_{2,3}$
$\langle v_2^2 \delta p_T \rangle$	$\frac{N_{3,\text{pp}}^2}{N_3^2}$	$\left(\frac{\langle v_1^2 \rangle_{\text{obs}}}{\langle v_1^2 \rangle_{\text{pp}}^{\text{obs}}} W_{3,2} \right)^2$	$\frac{\langle v_1^2 \delta p_T \rangle_{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}}$
$c_2\{4\}$	$\frac{N_{4,\text{pp}}^3}{N_4^3}$	$\left(\frac{\langle v_1^2 \rangle_{\text{obs}}}{\langle v_1^2 \rangle_{\text{pp}}^{\text{obs}}} W_{4,2} \right)^3$	$\left(\frac{\langle v_1^2 \delta p_T \rangle_{\text{obs}}}{\langle v_1^2 \delta p_T \rangle_{\text{pp}}^{\text{obs}}} \right)^{3/2} W_{4,3}^3$

Scaling behavior in standard method

Plot the scaled observables $\mathcal{O}\{m\}^{1/(m-1)} \sim 1/N$

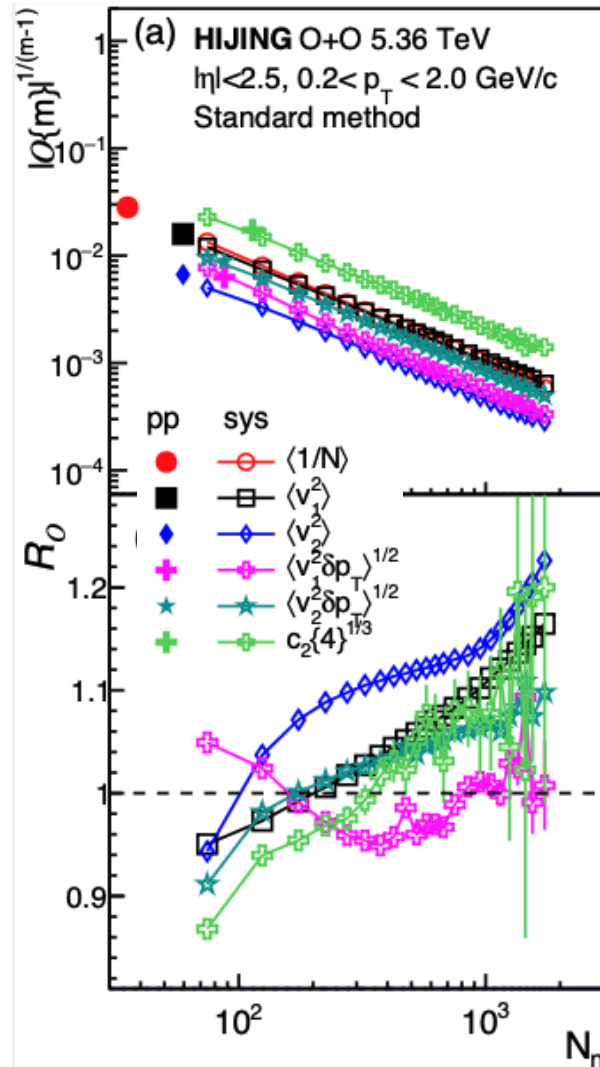
$$\mathcal{O} = \langle v_2^2 \rangle, \langle v_2^2 \delta p_T \rangle, c_2\{4\}, \quad 1/N, \langle v_1^2 \rangle, \langle v_1^2 \delta p_T \rangle$$

Since pp collision has a broad range in N , value of N_m where we plot the result depends on the observable $\mathcal{O}$ (solid symbols).

$$N_m \mathcal{O}\{m\}^{1/(m-1)} \approx N_{m,\text{pp}} \mathcal{O}\{m\}_{\text{pp}}^{1/(m-1)}$$

$$\Rightarrow R_{\mathcal{O}} \equiv \frac{N_m}{N_{m,\text{pp}}} \left(\frac{\mathcal{O}\{m\}}{\mathcal{O}\{m\}_{\text{pp}}} \right)^{1/(m-1)} \approx 1$$

Indeed, the $R_{\mathcal{O}}$ cluster around 1, here plotted for the standard method, >70% non-flow subtracted!



Scaling behavior in two subevent method

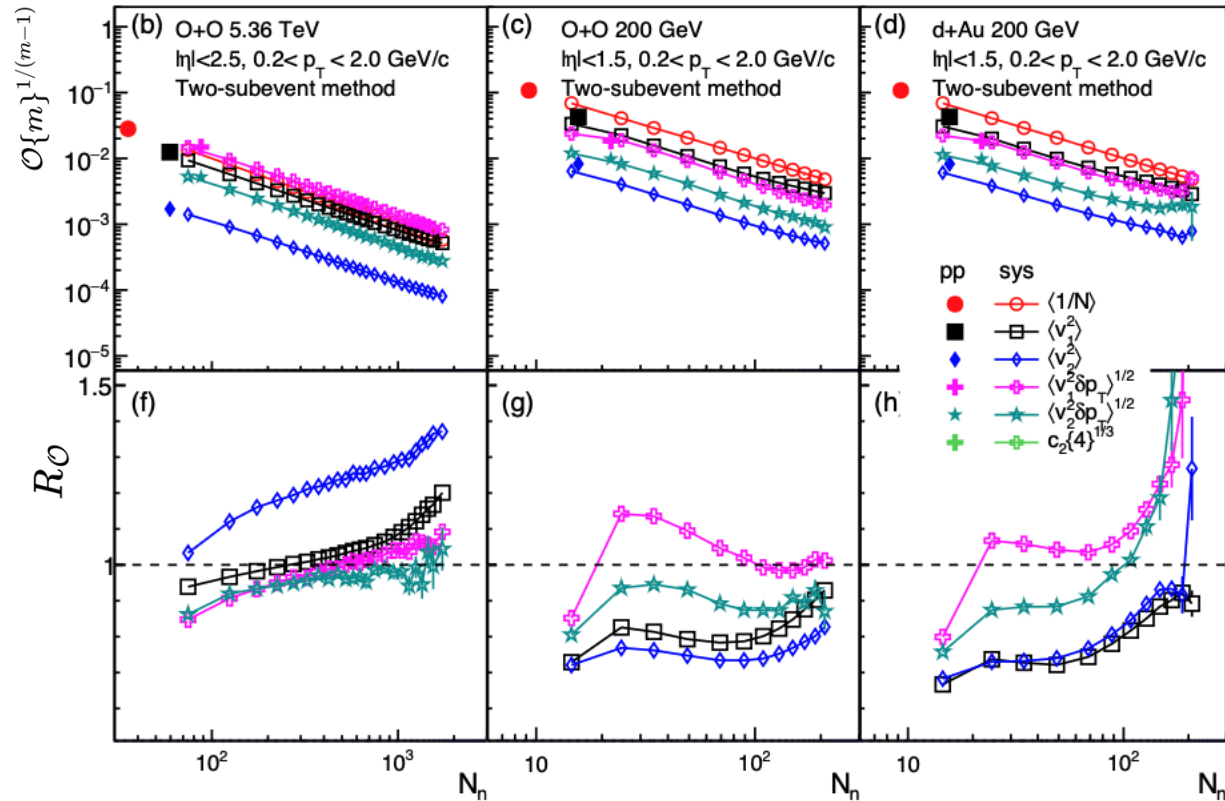
2sub-event results behaves like standard method.

The nonflow has somewhat different centrality dependence, but also generally within 20% from 1

Deviation from 1 reflects residual nonflow based on $1/N$ scaling.

It is clear some other reference observables may scale better, so need to explore other subtraction methods.

$$\mathcal{E} = 1/N \quad \langle v_1^2 \rangle \quad \langle v_1^2 \delta p_T \rangle$$

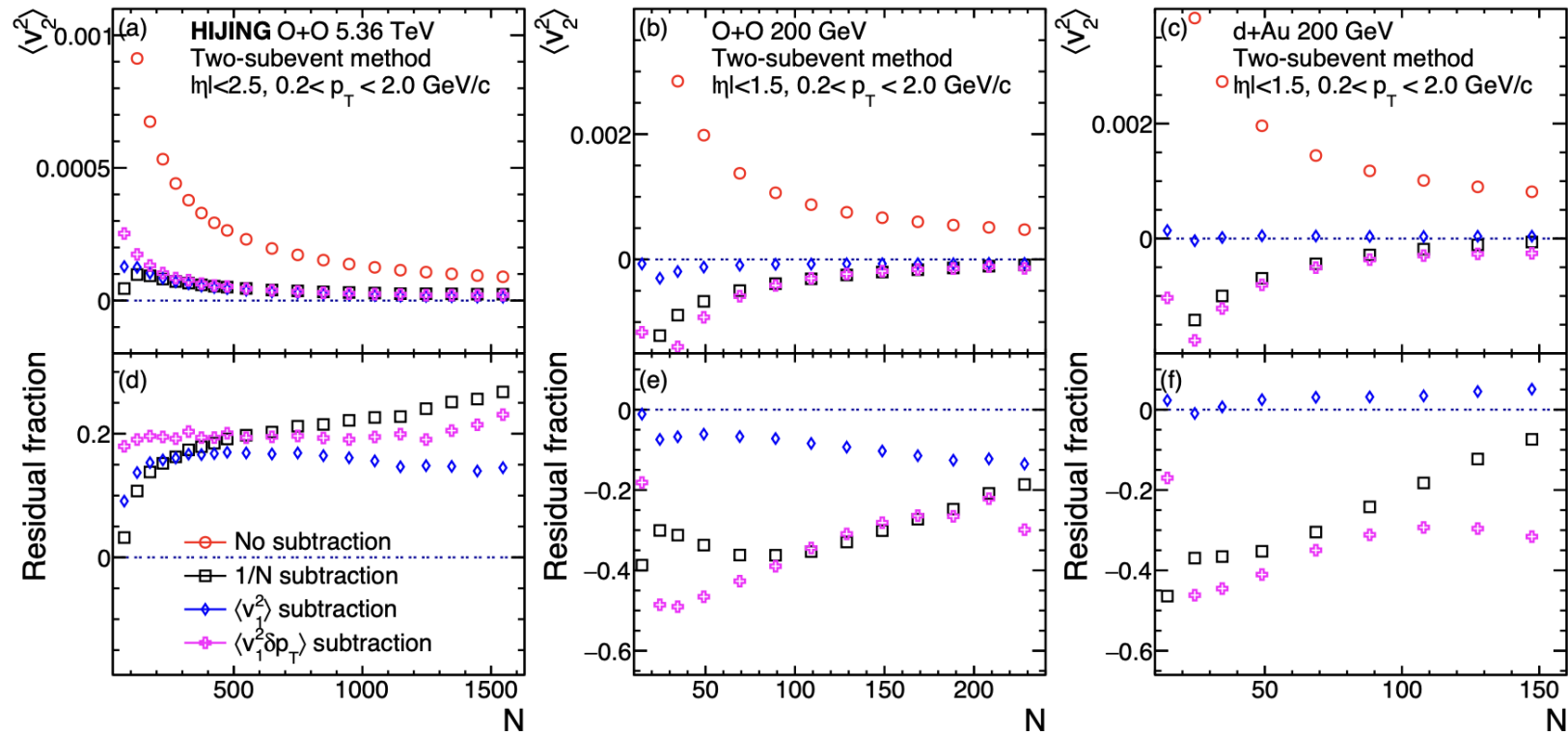


Result of subtraction for $\langle v_2^2 \rangle$

$\langle v_1^2 \rangle$ method consistently gives better result

$\langle v_1^2 \delta p_T \rangle$ scaling is worst and $1/N$ scaling is in between

$$\frac{\mathcal{O}\{m\}}{\mathcal{O}\{m\}^{\text{obs}}}$$

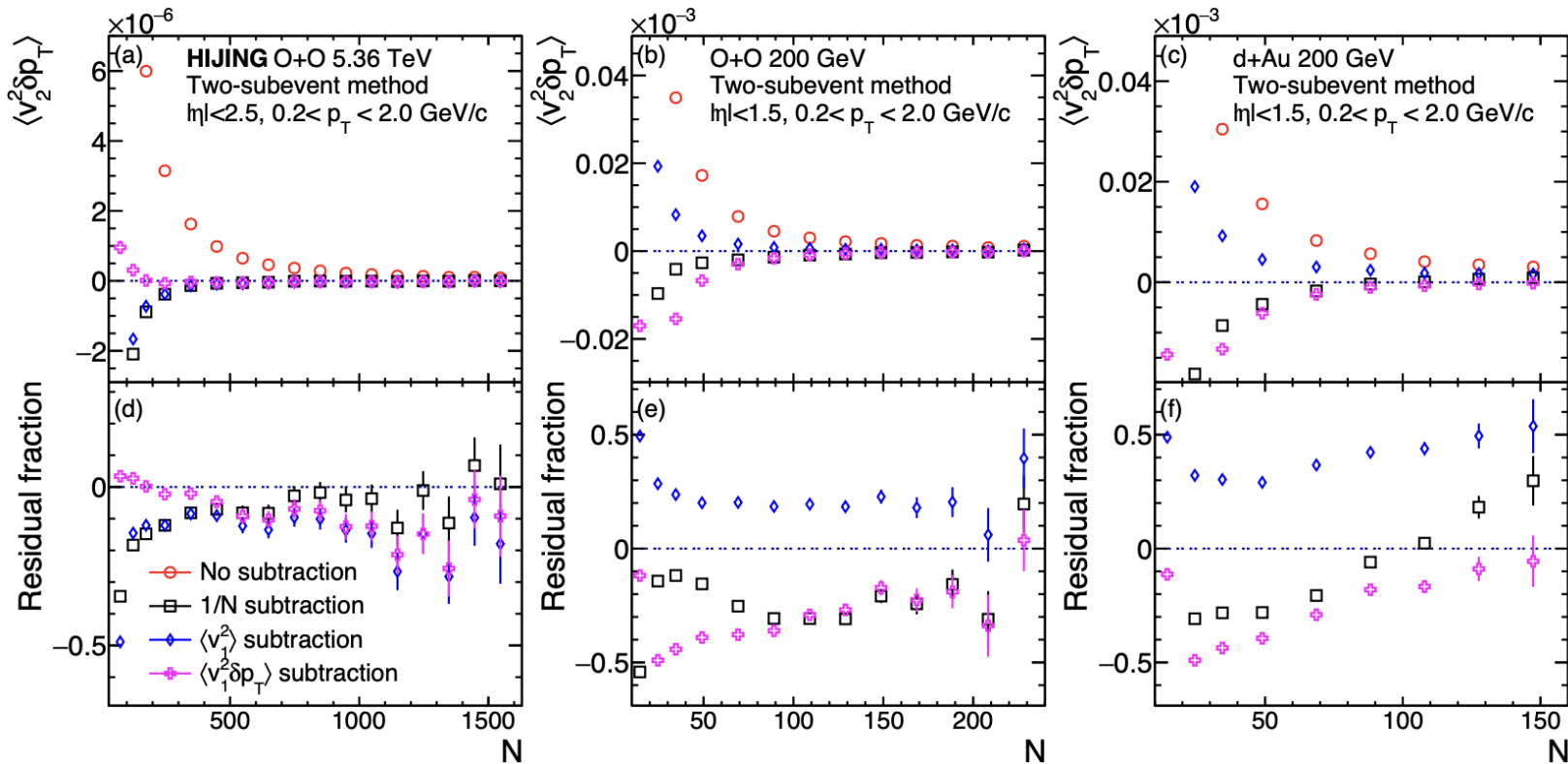


Result of subtraction for $\langle v_2^2 \delta p_T \rangle$

1/N scaling give better result

$\langle v_1^2 \rangle$ and $\langle v_1^2 \delta p_T \rangle$ scaling are comparable

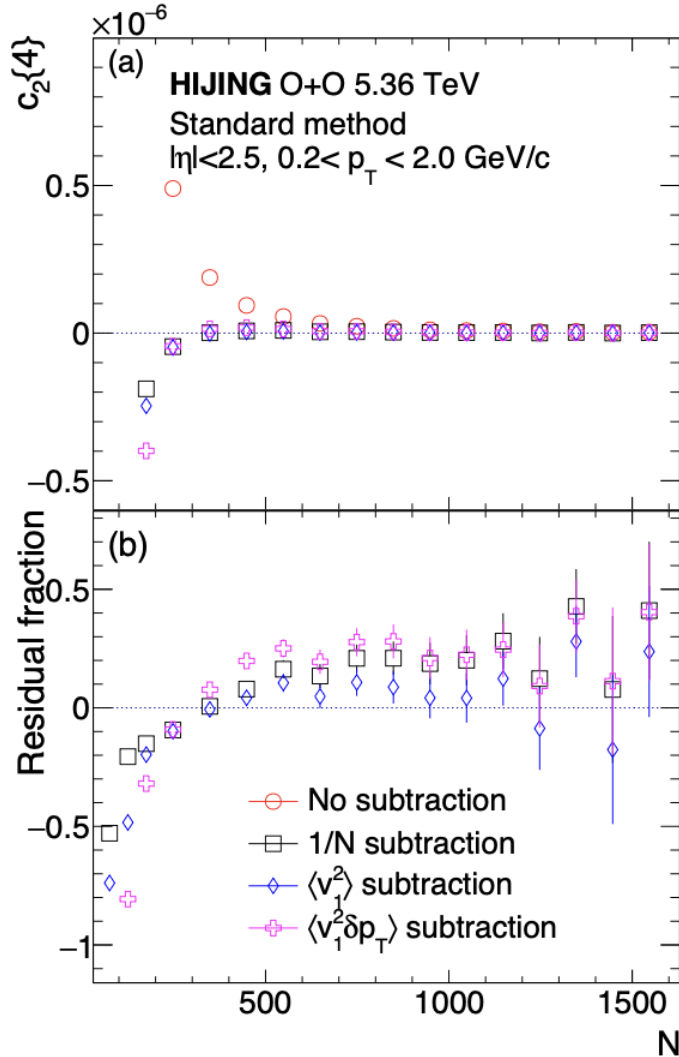
$$\frac{\mathcal{O}\{m\}}{\mathcal{O}\{m\}_{\text{obs}}}$$



Result of subtraction for $c_2\{4\}$

$$\frac{\mathcal{O}\{m\}}{\mathcal{O}\{m\}^{\text{obs}}}$$

$\langle v_1^2 \rangle$ scaling works the best



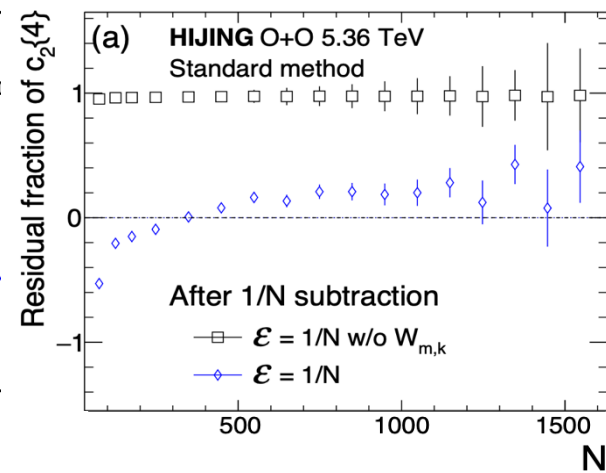
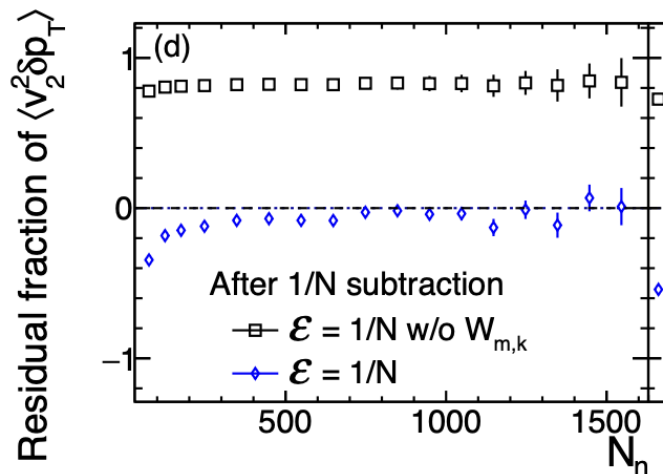
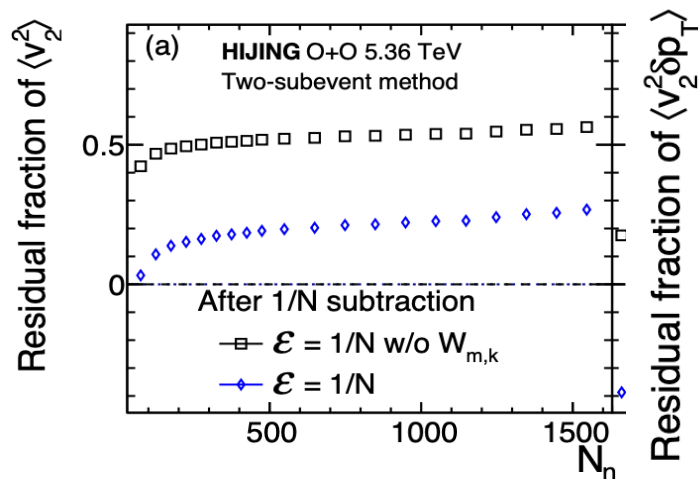
Importance of using the correct weighting factor

Considering reweighted N : $N_m = \left(\frac{\sum_i N_i^m}{\sum_i N_i} \right)^{1/(m-1)}$ vs $\langle N \rangle$

$$\frac{\langle v_2^2 \rangle_{\text{sub}}}{\langle v_2^2 \rangle} = 1 - \frac{N_{2,\text{pp}}}{N_2} \frac{\langle v_2^2 \rangle_{\text{pp}}}{\langle v_2^2 \rangle}$$

$$\frac{\langle v_2^2 \delta p_T \rangle_{\text{sub}}}{\langle v_2^2 \delta p_T \rangle} = 1 - \left(\frac{N_{3,\text{pp}}}{N_3} \right)^2 \frac{\langle v_2^2 \delta p_T \rangle_{\text{pp}}}{\langle v_2^2 \delta p_T \rangle}$$

$$\frac{v_2^4\{4\}_{\text{sub}}}{v_2^4\{4\}} = 1 - \left(\frac{N_{4,\text{pp}}}{N_4} \right)^3 \frac{v_2^4\{4\}_{\text{pp}}}{v_2^4\{4\}}$$



Again, considering reweighted N is essential!

Implication of residual non-closure

subtraction in HIJING does not give perfect closure.

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$$\frac{\mathcal{O}^{\text{sub}}\{m\}}{\mathcal{O}\{m\}} = 1 - \left(\left(\frac{\mathcal{E}\{k\}}{\mathcal{E}\{k\}_{\text{pp}}} \right)^{1/(k-1)} \frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \frac{N_k}{N_m} \right)^{m-1} \frac{\mathcal{O}\{m\}_{\text{pp}}}{\mathcal{O}\{m\}} \neq 0$$

But we can introduce a correct factor K to enforce zero

$$0 = 1 - K_{\mathcal{O}}^{\mathcal{E}} \left(\left(\frac{\mathcal{E}\{k\}}{\mathcal{E}\{k\}_{\text{pp}}} \right)^{1/(k-1)} \frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \frac{N_k}{N_m} \right)^{m-1} \frac{\mathcal{O}\{m\}_{\text{pp}}}{\mathcal{O}\{m\}} \quad \text{From HIJING}$$

This factor $K_{\mathcal{E},\mathcal{O}}(N)$ can be applied to data to check consistency of results:

HIJING

$$\mathcal{O}\{m\} = \mathcal{O}\{m\}^{\text{obs}} - K_{\mathcal{O}}^{\mathcal{E}} \times f_{\mathcal{O}}^{\mathcal{E}} \mathcal{O}\{m\}_{\text{pp}}^{\text{obs}} \quad f_{\mathcal{O}}^{\mathcal{E}} = \left(\left(\frac{\mathcal{E}\{k\}^{\text{obs}}}{\mathcal{E}\{k\}_{\text{pp}}^{\text{obs}}} \right)^{1/(k-1)} \underbrace{\frac{N_{m,\text{pp}}}{N_{k,\text{pp}}} \cdot \frac{N_k}{N_m}}_{W_{m,k}} \right)^{m-1}$$

Summary

- We developed a general nonflow subtraction strategy for m-particle correlations, inspired by approximate $1/N^{m-1}$ scaling behavior of nonflow, lead to a plethora of methods.
- Deviation from such scaling arise from selection bias, partially accounted in $\langle v_1^2 \rangle$ and $\langle v_1^2 \delta p_T \rangle$ method; Bias also arises from wide N range of reference, in part. for minbias pp.
- Estimate residual nonflow fraction in HIJING: $\lesssim 20-30\%$ when converted to 2PC level
→ In real data: if flow signal is 30% of nonflow, the nonflow systematics can reach 100%.
- These methods are useful for understanding the nature of nonflow, and for evaluating the residual nonflow systematics in the subevent methods.
- But given the large number of nonflow subtraction methods, each observable-method combination need to be studied carefully in both models (HIJING, PYTHIA) and data
→ identify the best method for each observable.

Back up

Importance of using the correct weighting factor $\mathcal{E} = 1/N^{\text{ong Jia}}$

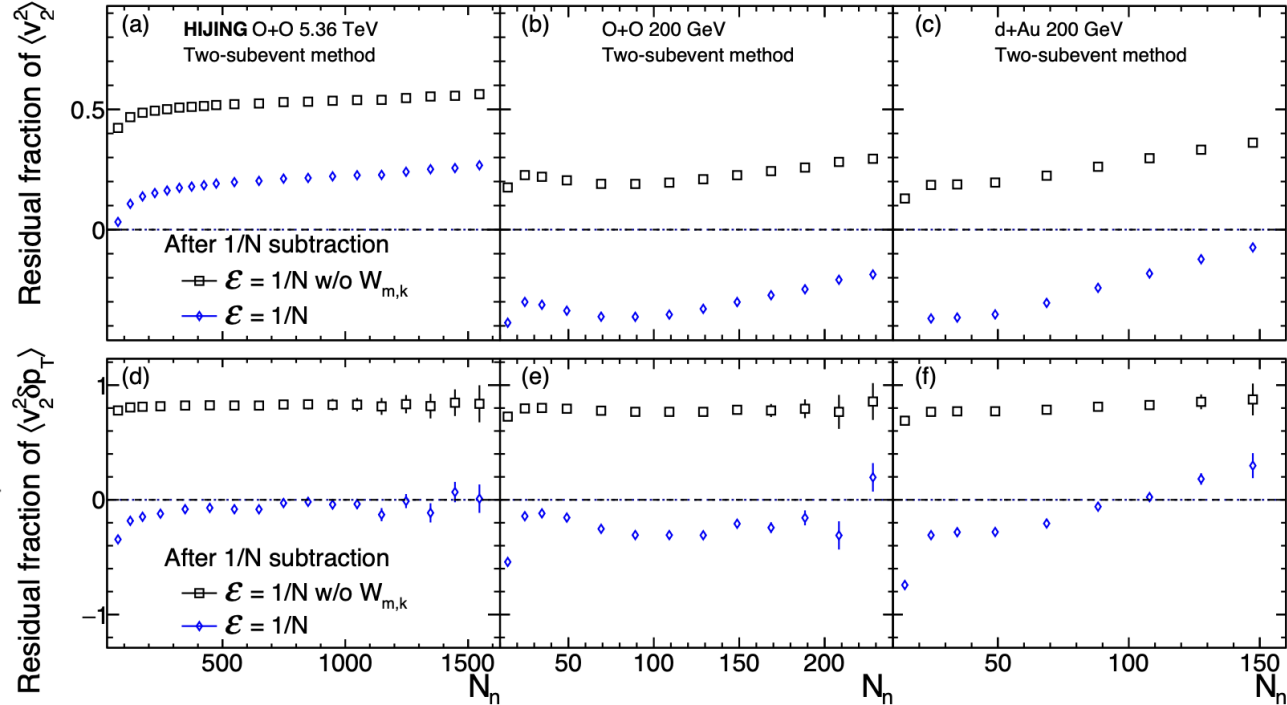
Considering reweighted N leads to improvement

$$N_m = \left(\frac{\sum_i N_i^m}{\sum_i N_i} \right)^{1/(m-1)} \quad \text{vs} \quad \langle N \rangle$$

$$\frac{\langle v_2^2 \rangle_{\text{sub}}}{\langle v_2^2 \rangle} = 1 - \frac{N_{2,\text{pp}}}{N_2} \frac{\langle v_2^2 \rangle_{\text{pp}}}{\langle v_2^2 \rangle}$$

We know 1/N-scaling
under-subtracts nonflow in
data if not using $N_{2\text{pp}}$

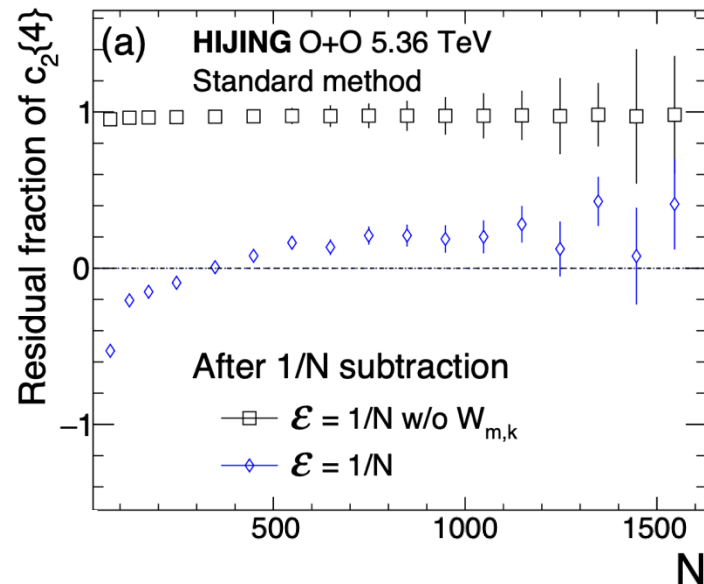
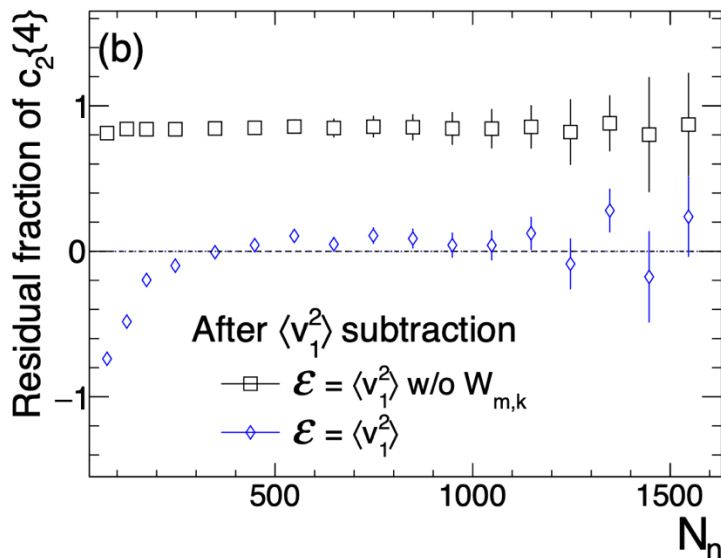
$$\frac{\langle v_2^2 \delta p_T \rangle_{\text{sub}}}{\langle v_2^2 \delta p_T \rangle} = 1 - \left(\frac{N_{3,\text{pp}}}{N_3} \right)^2 \frac{\langle v_2^2 \delta p_T \rangle_{\text{pp}}}{\langle v_2^2 \delta p_T \rangle}$$



Importance of using the correct weighting factor

$$\frac{v_2^4\{4\}_{\text{sub}}}{v_2^4\{4\}} = 1 - \left(\frac{\langle v_1^2 \rangle}{\langle v_1^2 \rangle_{\text{pp}}} \frac{N_{4,\text{pp}}}{N_{2,\text{pp}}} \frac{N_2}{N_4} \right)^3 \frac{v_2^4\{4\}_{\text{pp}}}{v_2^4\{4\}}$$

$$\frac{v_2^4\{4\}_{\text{sub}}}{v_2^4\{4\}} = 1 - \left(\frac{N_{4,\text{pp}}}{N_4} \right)^3 \frac{v_2^4\{4\}_{\text{pp}}}{v_2^4\{4\}}$$



Again, considering reweighted N is essential!

Importance of using the correct weighting factor

Considering the reweighted N leads to improvement

