Probing the structure of the initial state of heavy-ion collisions with $p_T$-dependent flow fluctuations

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The connection between initial-state geometry and anisotropic flow can be quantified through a well-established mapping between $p_T$-integrated flow harmonics and cumulants of the initial transverse momentum density [1]. Flow responses to initial-state fluctuations at different scales [8–10]. The purpose of the present paper is to investigate the connection between subleading modes of anisotropic flow fluctuations [11–20] and the aspects of the initial state of heavy-ion collisions, especially at smaller scales.

The azimuthal flow can be characterized by flow harmonics $V_n$, which are defined as the Fourier coefficients of the azimuthal distribution of particles in a given event:

$$
\frac{dN}{dy \, p_T \, dp_T \, d\phi} = \frac{1}{2\pi} N(p_T, y) \sum_{n=-\infty}^{\infty} V_n(p_T, y) e^{-in\phi},
$$

where we consider particles of transverse momentum $p_T$, rapidity $y$ and energy $E$, corresponding to a particle density $N(p_T, y)$ in momentum space. Here, the harmonics $V_n$ are defined as complex numbers of modulus and phase $V_n \equiv |V_n| e^{i\epsilon_n}$ by

$$
V_n \simeq \kappa_n \epsilon_n,
$$

where the properties of the QGP are encoded in the single constant $\kappa_n$, and $\epsilon_n$ is an eccentricity characterizing the initial geometry, the precise definition of which may vary [2–7, 21–29]. For $n = 2$, for instance, we take

$$
\epsilon_2 \equiv -\frac{\{r^2 e^{2i\phi}\} - \{r e^{i\phi}\}^2}{\{r^2\} - \{r e^{i\phi}\} \{r e^{-i\phi}\}},
$$

where we define the spatial average

$$
\{\cdots\} \equiv \frac{\int d^2\vec{x} \rho(\vec{x}) (\cdots)}{\int d^2\vec{x} \rho(\vec{x})},
$$

in which $\rho(\vec{x})$ is the initial transverse energy density at the position $\vec{x}$ in the transverse plane. While the relation (2) is usually employed for integrated flow vectors, we here extend it to the differential flow $V_n(p_T)$, by considering independent values of $\kappa_n(p_T)$ in each momentum bin [30]. An extension of Eq. (2) to rapidity-dependent hydrodynamic response has been considered in [31, 32].

Despite the success of Eq. (2), anisotropic flow may also respond to other features of the initial state. In Fig. 1, the solid blue squares represent values of $V_2$ and $\epsilon_2$ for a set of events simulated in a state-of-the-art hydrodynamic model [33, 34] using TRENTo+UrQMD [35, 36].

I. INTRODUCTION

In high-energy heavy-ion collisions, the hydrodynamic expansion of the quark-gluon plasma (QGP) is driven by large pressure gradients that convert the anisotropic initial-state geometry into final-state momentum anisotropies, or anisotropic flow [1]. In fact, a quantitative, event-by-event mapping between features of the initial geometry and the resulting anisotropic flow can be established in hydrodynamic models of heavy-ion collisions [2, 7]. Within this framework, it is possible to estimate how — and to what extent — anisotropic flow observables respond to initial-state fluctuations at different scales [8–10]. The purpose of the present paper is to investigate the connection between subleading modes of anisotropic flow fluctuations [11–20] and the aspects of the initial state of heavy-ion collisions, especially at smaller scales.

The azimuthal flow can be characterized by flow harmonics $V_n$, which are defined as the Fourier coefficients of the azimuthal distribution of particles in a given event:

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in which $\rho(\vec{x})$ is the initial transverse energy density at the position $\vec{x}$ in the transverse plane. While the relation (2) is usually employed for integrated flow vectors, we here extend it to the differential flow $V_n(p_T)$, by considering independent values of $\kappa_n(p_T)$ in each momentum bin [30]. An extension of Eq. (2) to rapidity-dependent hydrodynamic response has been considered in [31, 32].

Despite the success of Eq. (2), anisotropic flow may also respond to other features of the initial state. In Fig. 1, the solid blue squares represent values of $V_2$ and $\epsilon_2$ for a set of events simulated in a state-of-the-art hydrodynamic model [33, 34] using TRENTo+UrQMD [35, 36].
The remarkable correlation between the two quantities visibly supports the approximation in Eq. (2) (dashed magenta line), indicating that elliptic flow fluctuations are mostly driven by a linear response to $\epsilon_2$. At the same time, the spread around linear correlation points to small corrections to the approximate linear response, which may originate from the finer details of the initial transverse energy distribution $N(p_T)/\langle N \rangle$, or from nonlinear response [4, 8, 43–48]. Extensions of Eq. (2) to contemplate such corrections were proposed and studied in Refs. [2, 7, 27, 30, 49].

In this paper, we set out to investigate how corrections to Eq. (2), indicated by the spread of the blue squares in Fig. 1, might be experimentally studied through a principal component analysis (PCA) of anisotropic flow fluctuations [11–20]. In particular, we explore which features of the initial geometry are most relevant for understanding this analysis. The connection between subleading anisotropic flow and initial-state anisotropies was previously investigated in Refs. [12, 13], where different methods were employed to interpret the original observables of Ref. [11].

In Section IV, we present the PCA of flow fluctuations to be employed in our analysis [18]. In Section IV, we discuss a mapping of hydrodynamic response which extends Eq. (2) to encompass nonlinear response and finer details of the initial geometry [9]. Then, we apply this mapping to simulated hydrodynamic events and employ it to understand the PCA of anisotropic flow. Results are presented and discussed in Section IV and our main conclusions are summarized in Section V.

II. PRINCIPAL COMPONENT ANALYSIS OF FLOW FLUCTUATIONS

We wish to find measurable consequences of corrections to relation (2). However, only the left-hand side of this relation is accessible in experiments. Thus, fluctuations of the initial geometry must be inferred from fluctuations of anisotropic flow. One way this can be achieved is by exploring correlations between flow harmonics at different momentum bins. For concreteness, the blue squares in Fig. 2 displays a scatter plot of the elliptic flow coefficients $V_2(p_T^a)$ and $V_2(p_T^b)$, measured from particles of two different bins $a$ and $b$, with $p_T^a \neq p_T^b$, for a set of simulated events in a hybrid event-by-event hydrodynamic model [33, 34]. In this figure, correlations predicted by Eq. (2) are represented by the magenta dashed line, of slope $\kappa_2(p_T^b)/\kappa_2(p_T^a)$. Once again, the spread of points around the linear expectation implies that fluctuations of anisotropic flow are not entirely determined by fluctuations of $\epsilon_n$ alone, and in fact, elliptic flow coefficients at different transverse momentum fluctuate slightly differently from one another.

This deviation from perfect correlation in Fig. 2 can be quantified by a principal component analysis. In fact, this analysis can be carried out considering correla-

![Fig. 1. Scatter plot of the the elliptic flow harmonic, versus the corresponding eccentricity $\epsilon_2$, in modulus, for 40%–50% centrality. Solid blue squares correspond to values of the “flow per particle” $V_2(p_T)$, as defined in Eq. (1), while empty red circles include the effect of multiplicity fluctuations by showing instead the combination $N(p_T) V_2(p_T)/\langle N(p_T) \rangle$. The dashed magenta line shows the elliptic flow predicted from relation Eq. (2). As expected, the flow per particle is better correlated to the geometry of the system, in comparison to the “total flow” $V_2(p_T)$.](Image)

![Fig. 2. Scatter plot of the real part of the elliptic flow harmonic, $V_2(p_T)$, as defined in Eq. (1), in two separate transverse-momentum bins $p_T^a$ and $p_T^b$ (blue squares), for 40%–50% centrality. It is noticeable that the flow in the two bins are strongly correlated by fluctuations of the eccentricity $\epsilon_2$, as made clear by the prediction from Eq. (2) (dashed magenta line). Also shown are the first (“PC 1”, dot-dashed, orange line) and second (“PC 2”, solid, dark red line) principal components of elliptic flow fluctuations, projected onto the subspace described by $(V_2(p_T^a), V_2(p_T^b))$. The first component is associated to fluctuations of $\epsilon_2$, while the second one is related to corrections to Eq. (2).](Image)
tions among all the different transverse-momentum bins $[1, 2]$. Principal component analysis is a standard multivariate method that allows one to isolate linear combinations of variables which are linearly uncorrelated. By ordering the eigenvectors of the covariance matrix according to the eigenvalues, one can sort out which are the main directions of fluctuation — or principal components — within a given space of correlated variables [2, 3].

Fig. 2 shows the projections of the first ("PC 1") and second ("PC 2") principal components of elliptic flow onto the subspace spanned by $(V_2(p_T^a), V_2(p_T^b))$. The first, or leading, component of elliptic flow lies along the expectations from Eq. (2), indicating that this component is related to fluctuations of $\varepsilon_2$. On the other hand, this is not the case for the subleading component, which should be linked to other sources of fluctuation.

The PCA of anisotropic flow was first proposed in Ref. [11]. This original proposal was further explored in several papers [12–19] and experimentally measured by the CMS Collaboration [20]. While an event-by-event determination of the azimuthal distribution of particles, and thus of the flow harmonics $V_n(p_T)$, is severely hindered by the limited number of particles, a covariance matrix reflecting correlations among different bins can be safely extracted from two-particle correlations. The principal components $V_n^{(\alpha)}(p_T)$ can be found from the spectral decomposition of this matrix:

$$
\langle V_n(p_T^a) V_n^*(p_T^b) \rangle = \sum_{\alpha} \lambda^{(\alpha)} \psi_n^{(\alpha)}(p_T^a) \psi_n^{(\alpha)}(p_T^b) 
= \sum_{\alpha} V_n^{(\alpha)}(p_T^a) V_n^{(\alpha)}(p_T^b),
$$

(5)

where $M$ is the number of transverse-momentum bins, and $\lambda^{(\alpha)}$ and $\psi_n^{(\alpha)}(p_T)$ are the eigenvalues and eigenvectors, ordered in descending order $\lambda^{(\alpha)} \geq \lambda^{(\alpha+1)}$, and

$$
V_n^{(\alpha)}(p_T) \equiv \sqrt{\lambda^{(\alpha)}} \psi_n^{(\alpha)}(p_T).
$$

(6)

Since the covariance matrix is Hermitian, positive semi-definite and, assuming symmetry under parity transformations, also real, $V_n^{(\alpha)}(p_T)$ can be defined as real functions of the transverse momentum. While a precise measurement of the covariance matrix in Eq. (5) might be a challenge, this matrix was shown to be nearly equivalent to the alternative one introduced in [18], which in turn should be straightforward to measure.

In Fig. 2 one observes a clear hierarchy between the first or leading principal component ($\alpha = 1$) — corresponding to the dominant source of fluctuations — and the much smaller subleading principal component ($\alpha = 2$) — related to subdominant fluctuations [11]. By truncating Eq. (4) at $\alpha_{\text{max}} \leq M$, such that $\lambda^{(\alpha_{\text{max}})} \ll \lambda^{(1)}$, one can characterize the covariance matrix, a two-variable function, by only a few functions $V_n^{(\alpha \leq \alpha_{\text{max}})}(p_T)$ of a single variable, representing the projection of the principal components upon each momentum bin. In general, even across the entire measured momentum range, there is a strong hierarchy such that the matrix can be accurately represented by 2 or 3 principal components. Thus, the PCA allows for an optimal, compact visualization of two-particle correlations from fluctuations of $V_n(p_T)$ [11].

In Eq. (5), the obtained components depend on how the spectral condition is defined. Here, we write the eigenvalue equation as

$$
\sum_{b=1}^{M} V_n\Delta(p_T^a, p_T^b) V_n^{(\alpha)}(p_T^a) W(p_T^b) \Delta p_T^b = \lambda^{(\alpha)} V_n^{(\alpha)}(p_T^a),
$$

(7)

where $V_n\Delta(p_T^a, p_T^b) \equiv \langle V_n(p_T^a) V_n(p_T^b) \rangle$ and the index $b$ is summed over all transverse-momentum bins, each with a weight $W(p_T^b)$ [18]. The weight function can be chosen so as to emphasize different parts of the spectrum. Natural choices of weight include $W(p_T) = 1$, for uniform emphasis across $p_T$, and $W(p_T) = \langle N(p_T) \rangle$, focusing on more occupied momentum bins. In this work, we adopt the former choice and take $W = 1$.

In the original flow PCA of Ref. [11], the covariance matrix of the “total flow” $N(p_T,y) V_n(p_T,y)$ at each momentum bin was considered. However, a covariance matrix of the “flow per particle” $V_n(p_T,y)$ at each momentum bin was considered. However, a covariance matrix of the “flow per particle” [18] is better suited to our needs [15, 18]. In fact, in [18], an important difference between subleading fluctuations of the “total” and the “per-particle” anisotropic flow was found. This difference is clearly visible in Fig. 1, where empty red circles represent values of $N V_2/(N)$ in different events. Fluctuations of $V_2$ correlate better with the geometry of the events, while fluctuations of $N V_2$ are affected by fluctuations of particle number [18]. It is noteworthy that the deviation is larger for higher values of $|c_2|$, corresponding to more peripheral collisions, where multiplicity fluctuations are more important.

The decorrelation among flow fluctuations at different values of the momentum can also be explored using the factorization breaking coefficient $r_n(p_T^a, p_T^b)$ [3, 5, 6, 12, 60]. However, this approach reveals the importance of subleading fluctuations only in relative terms. In case the dominant flow fluctuations stemming from eccentricity fluctuations become too large, as is the case for peripheral collisions, subleading flow fluctuations will only weakly impact the value of $r_n(p_T^a, p_T^b)$.

III. MAPPING HYDRODYNAMIC RESPONSE

Having built some intuition on the PCA of anisotropic flow, we now turn to a more quantitative study of its precise physical content. More specifically, we aim at determining which features of the fluctuating initial geometry are essential to the second principal component. To that end, we employ an approach based on Refs. [3, 5], explained below.
Let us assume that the QGP evolves deterministically, starting from early times, \( \tau \lesssim \tau_0 \). The energy-momentum tensor \( T^{\mu\nu}(\tau \geq \tau_0, \vec{x}) \) at later times is, thus, fully determined by its components at \( \tau = \tau_0 \). As a consequence, the final single-particle distribution is a functional of \( T^{\mu\nu}(\tau_0, \vec{x}) \)[1]

\[
\frac{dN}{dy_p d\vec{x} d\vec{p}} = \mathcal{F}[T^{\mu\nu}(\tau_0, \vec{x})].
\] (8)

Our purpose is to model the azimuthal dependence of \( \mathcal{F}[T^{\mu\nu}(\tau_0, \vec{x})] \) in a systematic manner. This can be achieved by employing a cumulant expansion of the initial conditions to define eccentricities \( \epsilon_{n,m} \). Thus, one can establish phenomenological relations \( V_n \approx \mathcal{F}_n[\{\epsilon_{n',m'}\}] \), where \( \mathcal{F}_n \) can be approximated by a power series in \( \epsilon_{n',m'} \). This series is restricted to terms with the correct symmetries and ordered according to a hierarchy of scales, in a Ginzburg-Landau fashion. As will become clear, the leading lowest-order term in such a series, \( \propto \epsilon_{n,n} \), is related to the usual eccentricity scaling of Eq. (2), while corrections give rise to the subleading principal components of anisotropic flow.

A. Characterizing the initial geometry

For simplicity, we assume a “static” transverse energy distribution at \( \tau = \tau_0 \) and neglect components of \( T^{\mu\nu}(\vec{x}, \tau_0) \) other than the energy density \( \rho(\vec{x}) \equiv T^{\tau\tau}(\tau_0, \vec{x}) \)[2]. It proves useful to take its Fourier transform,

\[
\rho(\vec{k}) = \int d^2 x \, \rho(\vec{x}) e^{i\vec{k} \cdot \vec{x}},
\] (9)

so that different values of \( |\vec{k}| \) probe \( \rho(\vec{x}) \) at different scales. In fact, \( \rho(\vec{k}) \) can be interpreted as a moment generating function, from which eccentricities might be extracted [2]. Because of their transformation properties under rotation, it is convenient to define \( z \equiv x + i y \) and \( k_z \equiv k_x + i k_y \). Moments of \( z \) and \( z^* \) are given by

\[
\{z^j z^{*\ell}\} = \frac{(-2i)^{j+\ell}}{\rho_0} \left. \frac{\partial^{j+\ell} \rho(\vec{k})}{\partial k_z^j \partial k_z^{\ell}} \right|_{k=0},
\] (10)

where \( \rho_0 = \rho(\vec{k} = 0) \) and we use the definition in Eq. (4).

The moments \( \{z^j z^{*\ell}\} \) are not invariant under translations and depend on the choice of coordinate system [5]. This is related to the fact that \( |k_x|^{-1} \) and \( |k_y|^{-1} \) are actually scales of distance to an arbitrary origin, not of separation between points. This issue can be solved by using, instead, the function

\[
W(\vec{k}) \equiv \log \left( \frac{\rho(\vec{k})}{\bar{\rho}} \right),
\] (11)

where \( \bar{\rho} \) sets an arbitrary scale with the same units as \( \rho(\vec{k}) \). Notice that under a translation,

\[
\vec{x} \to \vec{x} + \vec{d}: \quad W(\vec{k}) \to W(\vec{k}) + i \vec{k} \cdot \vec{d},
\] (12)

so that all but the first derivatives of \( W(\vec{k}) \) are invariant under translations. Cumulants can be computed from

\[
\{z^j z^{*\ell}\}_{cml} = \left. (-2i)^{j+\ell} \frac{\partial^{j+\ell} W(\vec{k})}{\partial k_z^j \partial k_z^{\ell}} \right|_{k=0}.
\] (13)

In order to study transformation properties under rotations, it is convenient to employ polar coordinates, where \( z \equiv r e^{i\phi} \). We thus define

\[
\rho_{j-\ell, j+\ell} \equiv \{z^j z^{*\ell}\} = \{r^{j+\ell} e^{i(j-\ell)\phi}\},
\] (14)

\[
W_{j-\ell, j+\ell} \equiv \{z^j z^{*\ell}\}_{cml} = \{r^{j+\ell} e^{i(j-\ell)\phi}\}_{cml},
\] (15)

which transform as the harmonics \( V_n \) under rotations,

\[
\phi \to \phi + \delta: \quad \rho_{n,m} \to \rho_{n,m} e^{in\delta}, \quad W_{n,m} \to W_{n,m} e^{in\delta}.
\] (16)

Because of Eqs. (14) and (15), \( \rho_{n,m} \) and \( W_{n,m} \) are defined only for even, non-negative values of \( m - |n| \). They are taken to vanish otherwise.

The set of all \( \rho_{n,m} \), or all \( W_{n,m} \), is sufficient to fully recover the shape of the initial condition \( \rho(\vec{x}) \). In fact, expanding \( \rho(\vec{k}) \) and \( W(\vec{k}) \) in powers of \( k_z \) and \( k_z^* \), and using Eqs. (10) and (13), one finds that these moments and cumulants can be interpreted as series coefficients [2]:

\[
\rho(\vec{k}) = \rho_0 \sum_{m=n=-m}^{\infty} \sum_{n=-m}^{m} \frac{(i/2)^m \rho_{n,m}}{(m-n/2)! (m+n/2)!} k_z^m e^{-in\phi_k},
\] (17)

\[
W(\vec{k}) = \sum_{m=n=-m}^{\infty} \sum_{n=-m}^{m} \frac{(i/2)^m W_{n,m}}{(m-n/2)! (m+n/2)!} k_z^m e^{-in\phi_k},
\] (18)

where it becomes clear that larger values of \( m \) become important at higher values of \( k \) and, thus, at smaller spatial scales. Also, if cumulants with \( m \geq 3 \) are neglected, one obtains a simple Gaussian distribution. Notice as well that only coefficients with positive \( n \) are required, since \( \rho_{-n,m} = \rho^*_{n,m} \) and \( W_{-n,m} = W^*_{n,m} \).

More details, including explicit, general expressions for the cumulants \( W_{n,m} \) in terms of the moments \( \rho_{n,m} \), can be found in Appendix A. As an example, we write down the expressions for \( m = 2 \):

\[
W_{0,2} = \{r^2\} - \{r e^{-i\phi}\} \{e^{i\phi}\},
\] (19)

\[
W_{2,2} = \{r^2 e^{2i\phi}\} - \{r e^{i\phi}\}^2,
\] (20)

---

1 We here assume boost-invariant initial conditions. A study of hydrodynamic response beyond 2+1-dimensional hydrodynamics can be found in [31, 32].

2 A similar treatment including other components of \( T^{\mu\nu} \) can be found in Ref. [31].
and $m = 3$:

$$W_{1,3} = \left\{ r^3 e^{i \phi} \right\} - \left\{ r^2 e^{2i \phi} \right\} - 2 \left\{ r e^{i \phi} \right\} + 2 \left\{ r e^{-i \phi} \right\} + 2 \left\{ r e^{i \phi} \right\}^2 \right\} ;$$

$$W_{3,3} = \left\{ r^3 e^{3i \phi} \right\} - 3 \left\{ r^2 e^{2i \phi} \right\} + 2 \left\{ r e^{i \phi} \right\}^3 .$$

**B. Hydrodynamic response to initial geometry**

The cumulant expansion above allows one to characterize the initial geometry of the system with a set of complex numbers $W_{n,m}$. The index $n$ specifies a harmonic of the azimuthal distribution of energy, while $m$ indirectly determines the length scales contributing to each cumulant, as well as the scaling with the typical transverse size $L$:

$$W_{n,m} = \left\{ r^m e^{i n \phi} \right\} \propto L^m . \tag{23}$$

Because of the oscillating exponential $e^{i n \phi}$, realistic initial conditions are expected to have $W_{n,0,m} \ll L^m$. Thus, we can characterize the initial-state anisotropies with typically small, dimensionless, system-size independent eccentricities $\epsilon_{n,m}$:

$$\epsilon_{n,m} = -\frac{W_{n,m}}{R^m} , \tag{24}$$

where we use $R = \sqrt{W_{0,2}}$ as a measure of system size. Besides having well defined rotational symmetries, all eccentricities except $\epsilon_{\pm 1,1}$ are invariant under translation, as can be seen from Eq. (11).

From Eq. (9), $V_n(p_T)$ is a function of the eccentricities in Eq. (24), which we can expand as a power series. This power series is restricted to terms which transform as $V_n$ under rotations, which considerably simplifies its form. Up to linear response, we have:

$$V_n(p_T) \approx \sum_{m=1}^{m_{\text{max}}} \sum_{n=1}^{n_{\text{max}}} \kappa^{(n)}_{m} (p_T) \epsilon_{n,m} + \mathcal{O}(\epsilon_{n,m_{\text{max}}}) + \mathcal{O}(\epsilon^2) , \tag{25}$$

where, assuming that larger scales contribute the most, we neglect eccentricities with $m > m_{\text{max}}$. By enforcing translational invariance, we have excluded $m = 1$ from the series expansion. Including higher powers of $\epsilon_{n,m}$, up to $p_{\text{max}}$, we find

$$V_n(p_T) \approx \sum_{p=1}^{p_{\text{max}}} \sum_{n'=1}^{n_{\text{max}}} \sum_{m'=1}^{m_{\text{max}}} \kappa^{(n)}_{m'} (p_T) \prod_{i=1}^{p} \epsilon_{n'_i,m'_i} + \mathcal{O}(\epsilon_{n,m_{\text{max}}}) + \mathcal{O}(\epsilon^{p_{\text{max}}+1}) , \tag{26}$$

where the sum includes negative values of $n'$ and is restricted to terms with the correct rotational and translational symmetries.

The coefficients $\kappa(p_T)$ in Eqs. (25) and (26) are responsible for encoding all the information on the relevant QGP properties, e.g. equation of state and transport coefficients. They can be obtained by minimizing the squared norm of the residuals $\delta_n$:

$$\delta_n = V_n^{(\text{hydro})} - V_n^{(\text{test})} \tag{27}$$

in each transverse-momentum bin, where $V_n^{(\text{hydro})}$ and $V_n^{(\text{test})}$ are the flow harmonics from full hydrodynamic simulations and estimates obtained from the power series in Eq. (26), respectively. By taking the derivative of $\langle |\delta_n|^2 \rangle$ with respect to $\kappa$, we arrive at the system of equations:

$$\sum_{n'=n}^{n_{\text{max}}} \sum_{m'=m}^{m_{\text{max}}} \Re \langle \varepsilon_{n,m}^* \varepsilon_{n',m'} \rangle \kappa^{(n)}_{n',m'} = \Re \langle V_n^* \varepsilon_{n,m} \rangle , \tag{28}$$

where $\varepsilon_{n,m} \equiv \prod \varepsilon_{n_i,m_i}$. Solving Eq. (28) yields optimal values of $\kappa$, which can be employed to predict the flow harmonic $V_n(p_T)$. Any dependence of the final flow harmonics on $W_{0,m}$ must be incorporated in the coefficients $\kappa$, which for this reason are mildly centrality dependent.

We emphasize that this prescription treats each momentum bin independently. As such, a description of correlated fluctuations between different momentum bins, as measured with PCA, is a non-trivial test of the framework.

Both Eqs. (25) and (26) are generalizations of Eq. (2). Similar expressions have been presented in Refs. [2–5]. Unlike previous approaches, however, we here undertake the description of the differential flow harmonics $V_n(p_T)$, by also promoting the coefficients $\kappa$ to functions of $p_T$.

**IV. RESULTS**

We apply the mapping of Section III to $Pb + Pb$ collisions at center-of-mass energy $\sqrt{s_{NN}} = 2.76$ TeV simulated in an event-by-event hybrid model — the same simulated events shown in Figs. [1] and [2]. Our initial conditions are generated with the parametric model TRENTo and fed into relativistic viscous hydrodynamics as implemented in MUSIC. Model parameters for TRENTo (except for the normalization factor) and for the parameterization of the hydro viscosities are taken from the Bayesian analysis of [42], where they were optimized to describe LHC data. More details and results from this model can be found in Refs. [13, 33, 34]. Hadrons are sampled from the freeze-out hypersurface using iSpectraSampler (ISS) [39] and their interactions in the hadron gas phase are described with UrQMD [40, 41]. A direct event-by-event determination of $V_n(p_T)$, which would be otherwise impractical, is enabled by applying an oversampling procedure, in which the freeze-out hypersurface of each hydrodynamic event is converted into particles multiple times, until a threshold number of particles is achieved. This artificial increase of the number of
particles also has the advantage of dissolving correlations from hadronic interactions and resonance decays.

Our aim is to understand which terms and eccentricities in Eq. (25) provide the most important corrections to Eq. (2). In particular, we are interested in understanding the importance of linear response to higher-order eccentricities such as \( \epsilon_{2,4} \). We start by writing 7-term predictors for \( V_2 \) and \( V_3 \), with three subdominant linear terms and three nonlinear terms:

\[
V_2 \simeq \kappa_2^{(2)} \epsilon_{2,2} + \kappa_4^{(2)} \epsilon_{2,4} + \kappa_6^{(2)} \epsilon_{2,6} + \kappa_8^{(2)} \epsilon_{2,8} + \mathcal{O}(m = 10)
\]
\[
+ \kappa_1^{(2,2,3)} |\epsilon_{2,2}|^2 \epsilon_{2,2} + \kappa_1^{(2,2)} |\epsilon_{2,4}|^2 \epsilon_{2,4}^* + \kappa_1^{(2,3)} |\epsilon_{2,6}|^2 \epsilon_{1,3}^2
\]
\[
+ \ldots + \mathcal{O}(\epsilon^3),
\]

(29)

\[
V_3 \simeq \kappa_3^{(3)} \epsilon_{3,3} + \kappa_5^{(3)} \epsilon_{3,5} + \kappa_7^{(3)} \epsilon_{3,7} + \kappa_9^{(3)} \epsilon_{3,9} + \mathcal{O}(m = 11)
\]
\[
+ \kappa_1^{(3,2,2)} \epsilon_{2,2} \epsilon_{1,3} + \kappa_1^{(3,2)} \epsilon_{2,2}^* \epsilon_{2,4} + \kappa_1^{(3,4)} \epsilon_{4,4} \epsilon_{1,3}
\]
\[
+ \ldots + \mathcal{O}(\epsilon^3),
\]

(30)

where the only cubic terms are those involving \( \epsilon_{2,2} \), expected to be the largest eccentricity. In order to understand the role of each term in Eqs. (29) and (30), we calculate, for different combinations of terms, predictions for the event-by-event \( V_2(p_T) \) and \( V_3(p_T) \), which we then compare to full hydrodynamic simulations. The coefficients \( \kappa(p_T) \) are fixed at their optimal values for each set of terms. A more comprehensive study, including results for \( V_4(p_T) \) and \( V_5(p_T) \), can be found in Appendix B.

A. Quality of the predictors

The quality of the estimators in Eq. (25) and (26) can be assessed by measuring how they correlate with \( V_n(p_T) \), at the optimum values of \( \kappa \). To measure the degree of correlation we employ the Pearson correlation coefficient

\[
Q_n(p_T) = \frac{\text{Re}(V_n^{\text{(hydro)}}(p_T) V_n^{\text{(est)}}(p_T))}{\sqrt{\langle |V_n^{\text{(hydro)}}(p_T)|^2 \rangle \langle |V_n^{\text{(est)}}(p_T)|^2 \rangle}},
\]

(31)

which is always between \(-1\), corresponding to perfect anticorrelation, and 1, corresponding to perfect correlation.

Results for the correlation coefficients \( Q_2 \) and \( Q_3 \) are displayed in Fig. 3 for predictors constructed from different sets of terms. We first note that, as long as the leading eccentricity \( \epsilon_{n,n} \) is included, all combinations of terms provide good predictors of \( V_n \), with \( Q_n \) consistently close to 1. However, it is also visible that including new eccentricities and terms can improve the correlation even further, especially for noncentral events and higher \( p_T \). In particular, we highlight that the full predictors in Eqs. (29) and (30) (dashed blue curve), containing seven terms, are consistently closer to the full simulation results, hinting at the convergence of the double expansion in Eq. (26). Our predictors have their worst perfor-

FIG. 3. Pearson correlation coefficient between the flow harmonics \( V_2(p_T) \) (upper panel) and \( V_3(p_T) \) (lower panel) and predictions of their event-by-event fluctuations from eccentricities of the initial transverse geometry. Different curves correspond to different predictors, with the blue dashed curve corresponding to the full expressions in Eqs. (29) and (30). Events are simulated for \( Pb+Pb \) collisions at \( \sqrt{s_{NN}} = 2.76 \text{ TeV} \), within a hybrid event-by-event hydrodynamic model (TRENTo+Music+UrQMD).
performance for central collisions and low transverse momentum, where eccentricities are expected to be smaller.

Different terms in the mapping of hydrodynamic response become more important depending on centrality and transverse momentum. It is to be expected that, in more peripheral collisions, increasing eccentricities render nonlinear terms more important. Furthermore, if larger values of \( p_T \) are associated to stronger pressure gradients, it should come as no surprise that nonlinear hydrodynamic response becomes more important as \( p_T \) increases. In fact, for \( V_2 \), the subleading linear term \( \epsilon_{2,4} \) is especially important in more central collisions, while the nonlinear term \( \epsilon_{2,3} \) becomes more important at higher centralities. The cubic term \( |\epsilon_{2,2}|^2\epsilon_{2,2} \) — often assumed to be the most important subdominant term in peripheral events — is typically less important than \( \epsilon_{2,3} \). It becomes relevant for non-central collisions and at low \( p_T \), where most of the particles are, but also where its effect is barely visible. For \( V_3 \), the nonlinear term proportional to \( \epsilon_{2,2}^2\epsilon_{1,3} \) also becomes more important in more peripheral events, but the \( p_T \) dependence is even stronger. The linear term proportional to \( \epsilon_{3,5} \) becomes less relevant at higher \( p_T \), being eclipsed by \( \epsilon_{2,2}\epsilon_{1,3} \) even in central collisions. As a caveat, we stress that the higher cumulants \( W_{n,m} \) are, in general, nonlinear on the moments \( \rho_{n,m} \), and the terms that we consider to be linear here might be viewed as nonlinear elsewhere — if eccentricities are defined from \( \rho_{n,m} \). This is especially important for understanding the higher harmonic results presented in Appendix B.

B. PCA from hydrodynamic response

In Sec. III we have argued that corrections to the usual eccentricity scaling in Eq. (2) are revealed by subleading principal components of the flow harmonics. By employing the results of Sec. III we can now make our argument quantitative and investigate how different terms and eccentricities in Eq. (2) affect the PCA of anisotropic flow. Besides verifying our claim, we are also able to reveal the information contained in subleading principal components at different centralities and transverse momenta for both elliptic and triangular flow.

It can be shown that the leading principal component is nearly unaffected by subleading terms, at least for \( n = 2, 3 \). Because the third principal component of anisotropic flow can be quite small, we thus focus on the second principal component \( V_n^{(2)}(p_T) \). Results are shown in Fig. 4 where we compare the subleading principal components extracted from full hydrodynamic simulations in our model (solid black curve) and from the predictors in Eqs. (29) and (30) (dashed blue curve). The vertical axis shows the projection \( V_n^{(2)}(p_T) \) of the second principal component on a given momentum bin, while the horizontal axis corresponds to increasing values of \( p_T \). A good agreement is found, especially for more central collisions.

We stress that a nonvanishing subleading principal component can only be predicted if more than one term is included in the predictor for \( V_n \). In Fig. 4 this is illustrated by the flat pale-blue solid lines, corresponding...
to predictions from a single eccentricity $\epsilon_{n,m}$, in which case $V_n$ at each $p_T$ must fluctuate identically. Thus, subleading PCA modes uniquely isolate higher corrections in the cumulant series. Predictions for $V_n^{(2)}(p_T)$ including one subleading term for each harmonic are also presented. Once again, we find that nonlinear terms are more important for more peripheral collisions, where eccentricities become larger. For $V_2^{(2)}(p_T)$, the linear subleading term $\propto \epsilon_{2,4}$ provides the dominant contribution in central collisions, indicating a sensitivity to smaller scale structure of the initial state. In more peripheral centralities, both cubic and quadratic terms are equally important. For $V_3^{(2)}(p_T)$, we find nonlinear terms to be more important in general. Already at $0 - 10\%$ centrality, a competition between linear and nonlinear terms is found, with the linear $\propto \epsilon_{3,5}$ term dominating at low $p_T$, while the nonlinear $\propto \epsilon_1 \epsilon_2 \epsilon_{2,2}$ term dominates at higher $p_T$. For more peripheral collisions, the prediction from the linear subdominant term $\propto \epsilon_{3,5}$ looks qualitatively different from the full simulation results — it crosses the horizontal axis two times, more than required by orthogonality with $V_2^{(1)}$.

We have also checked that the predictors in Eqs. (29) and (30) provide a reasonable description of the third principal components of both elliptic and triangular flow. This confirms that the good description of subleading principal components is not fortuitous, and indicates that these predictors provide a surprisingly detailed description of flow fluctuations. Results can be found in Appendix B.

C. Granularity of the initial state

From Fig. 4 we find the subleading component of elliptic flow $V_2^{(2)}(p_T)$ to be especially sensitive to $\epsilon_{2,4}$, being dominated by its contributions in central collisions. Unlike the more familiar $\epsilon_{2,2}$, this eccentricity characterizes the initial state at smaller length scales, making $V_2^{(2)}(p_T)$ a promising probe of the granularity of the initial transverse energy distribution [8–10].

An important question is then, exactly how sensitive is the subleading PCA mode to the small-scale structure of the initial state, and what specific length scales can be probed?

To address these questions, we first employ a simple independent source model (ISM) [4]. In this model, $N$ identical sources are randomly (and independently) distributed in the transverse plane according to some probability distribution $\rho(\vec{x})$. Each event then has a density

$$\rho(\vec{x}) = \sum_{i=1}^{N} \rho_S(\vec{x} - \vec{x}_i),$$

(32)

where $\rho_S(\vec{x} - \vec{x}_i)$ is the density distribution of a single source [4]. Since $W_{n,m}$ are cumulants, the effect of the random source positions separates from the effect of the shape of each source $\rho_S$. That is, we can write the single event density as a convolution

$$\rho(\vec{x}) = \int d^2x' \rho_D(\vec{x}') \rho_S(\vec{x} - \vec{x}')$$

(33)

with

$$\rho_D(\vec{x}) \equiv \sum_{i=1}^{N} \delta(\vec{x} - \vec{x}_i).$$

(34)

Because Eq. (33) is a convolution and the generating function $W(\vec{k})$ is the logarithm of a Fourier transform, contributions from $\rho_D$ and $\rho_S$ combine additively and so do all cumulants:

$$W(\vec{k}) = \ln \left( \frac{\rho_D(\vec{k})}{\rho_D(0)} \right) + \ln \left( \frac{\rho_S(\vec{k})}{\rho_S(0)} \right)$$

$$= W_D(\vec{k}) + W_S(\vec{k}),$$

(35)

$$\implies W_{n,m} = W_{n,m}^D + W_{n,m}^S,$$

(36)

where, from now on, the superscripts $D$ and $S$ denote the distribution of sources and the average source shape, respectively. Assuming isotropic sources, $W_{n,m} = W_{n,m}^D$ for all $n \neq 0$, but the system size is still affected by the r.m.s. source radius $R_S \equiv \sqrt{\langle W_S^2 \rangle}$, which sets the scale for the granularity of the system:

$$W_{0,2} = R_D^2 + R_S^2 \equiv R^2.$$

(37)

Thus, from Eq. (3), and assuming $R_S \ll R_D$:

$$\epsilon_{n,m} = \frac{R_D^2 \epsilon_{n,m}^D}{R_m} = \frac{(R^2 - R_S^2)^{m/2} \epsilon_{n,m}^D}{R_m}$$

$$\approx \left( 1 - \frac{m R_S^2}{2 R^2} \right) \epsilon_{n,m}^D,$$

(38)

for all $n \neq 0$.

So we can expect that $m = 4$ eccentricities, such as $\epsilon_{2,4}$, are approximately twice as sensitive to the size of the source as $m = 2$ eccentricities, such as the traditional $\epsilon_2 = \epsilon_{2,2}$ that determines the leading elliptic flow. Further, with this formula we can quantify what length scales can be probed. E.g., if the granular structure were different by a fraction of the system size $R_S/R = X \ll 1$, the fractional change in $\epsilon_{2,4}$ would be roughly $2X^2$. For

---

3 See, for instance, Ref. [63].
\[ \rho^{(2)}(p_T) \], being strongly sensitive to \( \sqrt{|\epsilon_{2,4}|^2} \), provides a unique probe of the initial conditions at sub-fermi scales.

\[ \epsilon_{n,m}(w) \approx \sqrt{R_s(w) / \langle \epsilon_{2,2}^2 \rangle} \left( 1 + 4 \left( \frac{w^2 - w_s^2}{w_s^2} \right) \right)^{m/2} \epsilon_{n,m}^{D}. \] (43)

In Fig. 5, Eq. 13 is represented, for 40–50% centrality, by the faint magenta lines. A reasonable agreement is found between the isotropic ISM and TR\(\text{ENTo} for small departures from \( w = 0.956 \), but significant deviations are seen for \( w \) below 0.6 fm. This might be explained by anisotropies at the granular scale, which should be expected from the generalized average involved in the TR\(\text{ENTo} reduced thickness function — a combination of projectile and target thicknesses which is nonlinear for parameter \( p \neq 1 \). Nonetheless, we note that a much better agreement with the isotropic ISM can be found by employing \( \epsilon_{2,4} \) — with no subtraction of correlations with \( \epsilon_{2,2} \) — instead of \( \epsilon_{2,4} \). We also stress that even the results for Eq. 43 exceed the twofold increase in sensitivity from \( m = 2 \) to \( m = 4 \) expected from Eq. 41, due to departures from linear behavior.

Above, we have presented analytical and numerical evidence for the higher sensitivity of the higher-order eccentricity \( \epsilon_{2,4} \) to the granularity of initial-state fluctuations. Even under the very conservative assumption of isotropic fluctuations at small scales, this eccentricity is found to be at least two times more sensitive to the granular scale \( w \) than the usual eccentricity \( \epsilon_{2,2} \). We thus conclude that the subleading principal component of elliptic flow \( \epsilon_{2,4}(p_T) \), being strongly sensitive to \( \sqrt{|\epsilon_{2,4}|^2} \), provides a unique probe of the initial conditions at sub-fermi scales.

\[ \epsilon_{n,m}(w) \approx \sqrt{R_s(w) / \langle \epsilon_{2,2}^2 \rangle} \left( 1 + 4 \left( \frac{w^2 - w_s^2}{w_s^2} \right) \right)^{m/2} \epsilon_{n,m}^{D}. \] (43)
V. CONCLUSIONS

In this paper, we show that the PCA of anisotropic flow is a promising tool for studying the hydrodynamic response of the QGP to fluctuations of the initial state. Once undesired contributions from radial flow fluctuations are properly removed, as proposed in [18], this analysis uncovers details of the hydrodynamic response to anisotropies of the initial energy distribution of the system. More specifically, it reveals corrections to the familiar scaling relation between flow harmonics and spatial eccentricities $V_n = k_n \epsilon_n [24] [23, 27]$. A more complete mapping between initial geometry and flow harmonics can be generalized by means of a cumulant expansion of the initial transverse energy density profile [24]. In this work, by extending this mapping to account for transverse-momentum dependence, we were able to successfully predict anisotropic flow harmonics on a differential basis. This allowed us to systematically gauge, for the first time, the relative importance of linear and nonlinear hydrodynamic response at different centralities and transverse-momentum ranges. Higher-order cumulants of the initial transverse profile of the system proved to be more important at lower transverse momentum and in more central collisions, while nonlinear hydrodynamic response was found to provide the most important corrections at higher transverse momentum and in more peripheral collisions. We also found that two nonlinear terms, proportional to $|\epsilon_{2,2}|^2 \epsilon_{2,2}$ and $\epsilon_{1,3}^2$, provide relevant corrections to $V_2(p_T)$. Surprisingly, the latter was found to surpass the former in importance on a wide transverse-momentum range.

By predicting $V_n(p_T)$ exclusively from features of the initial transverse geometry, we were also able to reproduce, to a reasonable accuracy, the principal components of elliptic and triangular flow calculated in full event-by-event hydrodynamic simulations. By employing different eccentricities of the initial geometry, we found subleading principal components to be sensitive both to higher-order cumulants of the initial transverse geometry and to nonlinear hydrodynamic response. In the case of triangular flow, the most important contribution to the subleading component comes from a nonlinear term proportional to $\epsilon_{2,2} \epsilon_{1,3}$. The leading linear correction, proportional to $\epsilon_{3,5}$, on the other hand, was found to provide distinct contributions to the first subleading component — qualitatively different from the full simulation results.

In the case of elliptic flow fluctuations, the leading linear correction to $V_2(p_T)$, proportional to $\epsilon_{2,4}$, was found to provide an excellent prediction of the first subleading principal component in central collisions. In more peripheral collisions, this term was found to compete with the nonlinear terms $\propto |\epsilon_{2,2}|^2 \epsilon_{2,2}$ and $\propto \epsilon_{3,5}^2$ in importance. By employing both analytical arguments and numerical results, we showed this eccentricity to be sensitive to the initial granularity of the system. This provides compelling evidence that the measurement of the PCA observables proposed in Ref. [18] will shed light on the details and granular structure of the initial stages of high-energy nucleus-nucleus collisions.

In short, measurements of the principal components of different flow harmonics at different centralities can be employed to study different aspects of the hydrodynamic response of the QGP. Besides fluctuations of the energy density at smaller scales, principal component analysis can prove a useful tool to uncover fluctuations originating from initial flow [64]. This makes PCA an especially interesting tool for investigating the physics of smaller collision systems — although a study of this possibility is left to future work [64].

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Appendix A: Moments and cumulants of the initial density profile

In this appendix, we provide a complementary, more detailed description of the cumulants and moments of the initial transverse energy-density profile, used to define the eccentricities in Eq. (24).

1. Definition

The two-dimensional density (here, energy-density) profile of the initial state $\rho(\vec{x})$ can be characterized by its moments and cumulants. The corresponding generating functions are, respectively,

$$\rho(\vec{k}) = \int d^2 x \rho(\vec{x}) e^{i\vec{k} \cdot \vec{x}}$$

(A1)

and

$$W(\vec{k}) = \log \left( \frac{\rho(\vec{k})}{\bar{\rho}} \right),$$

(A2)

where $\bar{\rho}$ is an arbitrary scale and $\vec{x} = r (\cos \phi_x, \sin \phi_x)$ and $\vec{k} = k (\cos \phi_k, \sin \phi_k)$ are vectors on the transverse plane. The moments and cumulants can be defined by taking the Taylor expansion of both $\rho(\vec{k})$ and $W(\vec{k})$ around $k \equiv |\vec{k}| = 0$, followed by a Fourier expansion in $\phi_k$:

$$\rho(\vec{k}) = \sum_{m=0}^{\infty} \sum_{n=-m}^{m} \varrho_{n,m} k^m e^{-in\phi_k},$$

(A3)

$$W(\vec{k}) = \sum_{m=0}^{\infty} \sum_{n=-m}^{m} W_{n,m} k^m e^{-in\phi_k},$$

(A4)

where, from Eqs. (17) and (18), we can identify

$$\varrho_{n,m} = \frac{(i/2)^m \rho_0 \rho_{n,m}}{(m+n/2)! (m-n/2)!}, \quad W_{n,m} = \frac{(i/2)^m W_{n,m}}{(m+n/2)! (m-n/2)!}.$$  

(A5)

Note that

$$\varrho_{n,m} = \frac{1}{m!} \int_{-\pi}^{\pi} \frac{d\phi_k}{2\pi} \frac{d^m k}{dk^m} \rho(\vec{k}) e^{in\phi_k} \bigg|_{k=0}$$

$$= \frac{i^m}{m!} \int d^2 x \rho(\vec{x}) r^m \int_{-\pi}^{\pi} \frac{d\phi}{2\pi} [\cos(\phi_k - \phi_x)]^m e^{in\phi_k},$$

(A6)

Using the result

$$I(\phi_x) \equiv \int \frac{d\phi_k}{2\pi} (\cos(\phi_k - \phi_x))^m e^{in\phi_k}$$

$$= \frac{1}{2^m} \int_{-\pi}^{\pi} \frac{d\phi}{2\pi} (e^{i\phi} + e^{-i\phi})^m e^{in\phi}$$

$$= \frac{1}{2^m} \int_{-\pi}^{\pi} \frac{d\phi}{2\pi} e^{-i(2l-m)\phi} e^{in\phi}$$

$$= \frac{1}{2^m} \frac{m!}{(m-n/2)!} \int_{-\pi}^{\pi} \frac{d\phi}{2\pi} e^{in\phi} e^{i\phi_k},$$

(A7)

we get

$$\rho_{n,m} = \frac{1}{\rho_0} \int d^2 x \rho(\vec{x}) r^m e^{in\phi_x}$$

(A8)

provided $m - |n|$ is a positive multiple of 2. For negative or odd values of $m - |n|$, the integral in Eq. (A7) vanishes.

2. Relation between cumulants and moments

The cumulants can be written in terms of the moments by separating the zero mode and using the Taylor expansion of $\log h$ around $h = 1$:

$$W(\vec{k}) - W(\vec{0}) = \sum_{l=1}^{\infty} \frac{(-1)^{l-1}}{l!} \left( \sum_{m=1}^{\infty} \sum_{n=-m}^{m} \frac{\varrho_{n,m}}{\rho_0} k^m e^{-in\phi_k} \right)^l$$

$$= \sum_{l=1}^{\infty} \frac{(-1)^{l-1}}{l! (\rho_0)^l} \sum_{m=1}^{\infty} \sum_{n=-m}^{m} l! \times$$

$$\times \prod_{m=0}^{\infty} \prod_{n=-m}^{m} \frac{1}{p_{n,m}!} (\varphi_{m,n} k^m e^{-in\phi_k})^{p_{n,m}}$$

$$= \sum_{l=1}^{\infty} \frac{(-1)^{l-1} (\ell - 1)!}{l! (\rho_0)^l} \times$$

$$\times \sum_{k=1}^{\infty} k^\ell \sum_{m=1}^{\infty} \sum_{n=-m}^{m} e^{i \sum_{n p_{n,m} \phi_k x} \sum_{\{p_{n,m}\}} \times$$

$$\times \prod_{m=0}^{\infty} \prod_{n=-m}^{m} \frac{1}{p_{n,m}!} (\varphi_{m,n})^{p_{n,m}},$$

(A9)

where the multinomial theorem was employed to rewrite the $\ell$th power of a sum as a sum over the powers $p_{n,m}$ of each term, under the condition that $\sum_{m,n} p_{n,m} = \ell$.

Equating the powers of $k$ and $e^{i\phi_k}$ in Eqs. (A4) and
allows us to identify the cumulants,

\[
\mathcal{W}_{n,m} = \sum_{\{p_{n,m}\}} (-1)^{\frac{1}{2}(\sum_{n,m} p_{n,m} - 1)} \times \sum_{m, n} \left( \prod_{m=0}^{\infty} m \right) \prod_{n=-m}^{m} \frac{1}{p_{n,m}!} \left( \frac{\rho_{m,n}}{\rho_0} \right)^{p_{n,m}},
\]

(A11) gives us the cumulants \( \mathcal{W}_{n,m} \) in terms of the moments \( \rho_{m,n} \). The sum is over all possible partitions with weights given by

\[
\hat{\Omega}_{\rho,n,m} = (-1)^{\frac{1}{2}(\sum_{n,m} p_{n,m} - 1)} \left( \sum_{m, n} p_{n,m} - 1 \right)! \prod_{n,m} (p_{n,m}!)^{-1}.
\]

Using Eqs. (A5) and (A8), we can finally find \( W_{\tilde{n},\tilde{m}} \) in terms of \( \rho_{\tilde{n},\tilde{m}} \):

\[
W_{\tilde{n},\tilde{m}} = (\frac{\tilde{m}+\tilde{n}}{2})! (\frac{\tilde{m}-\tilde{n}}{2})! \sum_{\{p_{\tilde{n},\tilde{m}}\}} (-1)^{\frac{1}{2}(\sum_{\tilde{n},\tilde{m}} p_{\tilde{n},\tilde{m}} - 1)} \times \left( \prod_{\tilde{m}, \tilde{n}} m \right) \prod_{\tilde{m}, \tilde{n}} \frac{1}{p_{\tilde{n},\tilde{m}}!} \left( \frac{\rho_{\tilde{n},\tilde{m}}}{\rho_0} \right)^{p_{\tilde{n},\tilde{m}}},
\]

where only even values of \( m - |n| \) are considered. Once again, the sum is over all possible partitions, with \( \sum m p_{n,m} = \tilde{m} \) and \( \sum n p_{n,m} = \tilde{n} \). The weight of each term is of the form:

\[
\tilde{\Omega}_{\rho,n,m} = \tilde{\Sigma}_{\rho,n,m} \times \left( \frac{\tilde{m}+\tilde{n}}{2} \right)! \left( \frac{\tilde{m}-\tilde{n}}{2} \right)! \left( \frac{\tilde{m}+\tilde{n}}{2} \right)! \left( \frac{\tilde{m}-\tilde{n}}{2} \right)! p_{n,m}!.
\]

From Eq. (A13) — or, equivalently, from Eqs. (A12) and (A14) and all the relevant partitions — one can recover Eqs. (19), (20), (21) and (22).

**Appendix B: Other Harmonics and Corrections**

For the sake of clarity and simplicity, we have opted to omit a few results from the main text of this article. This had the advantage of making the text paper clearer and the plots less polluted. For completeness, we present a few extra results in this appendix.

Here, we show more comprehensive results for the mapping of the anisotropic flow from eccentricities of the initial geometry. In all these results, the dashed blue curve with hollow squares represents predictions from the most complete estimates available. For \( n = 2, 3 \), these estimates can be found on Eqs. (29) and (30). For \( n = 4, 5 \), they are given by

\[
V_4 \simeq \kappa_4^{(4)} \epsilon_{4,4} + \kappa_4^{(4)} \epsilon_{4,6} + \kappa_4^{(4)} \epsilon_{4,8} + \mathcal{O}(m = 10)
\]

\[
+ \kappa_2^{(4)} \epsilon_{2,2} + \kappa_1^{(4)} \epsilon_{1,3} \epsilon_{3,3} + \mathcal{O}(\epsilon^3),
\]

(B1)

Wherever present, the solid black curve with hollow squares represents the results from full hydrodynamic simulations. Dot-dashed curves represent results from predictors combining the leading eccentricity \( \epsilon_{n,n} \) and a single linear correction \( \propto \epsilon_{n,m>n} \). Dotted curves, on the other hand, exhibit results from predictors combining the leading eccentricity \( \epsilon_{n,n} \) and a single nonlinear subleading term.

In Section [IV] we have shown results for the quality of different predictors of the elliptic and triangular flow harmonics. Figure [3] in particular, displays the Pearson correlation coefficient between the actual flow harmonics from hydrodynamic simulation events and predictors of these harmonics from the initial-state eccentricities. In Fig. [8], we repeat the same results, but, for completeness, include other predictors of \( V_2(p_T) \) and \( V_3(p_T) \), each of them built from a pair of terms containing the leading term \( \propto \epsilon_{n,n} \). Figure [9] exhibits the same kind of analysis for \( V_4(p_T) \) and \( V_5(p_T) \). It is noteworthy that our predictions for higher harmonics are not quite as good as the ones for elliptic and triangular flow. Note also that the linear estimator is better here than in some previous analyses because of our choice to define eccentricities via cumulants rather than moments.

In Section [VII] and, more specifically, in Fig. [4], we have presented results for the second principal component of elliptic and triangular flow harmonics. There, we plotted results from full event-by-event hydrodynamic simulations, from the predictors in Eqs. (29) and (30), and from a few predictors containing a pair of terms.
FIG. 6. Pearson correlation coefficient between the flow harmonics $V_2(p_T)$ (upper panel) and $V_3(p_T)$ (lower panel) and predictions of their event-by-event fluctuations from eccentricities of the initial transverse geometry. Different curves correspond to different predictors, with the blue dashed curve corresponding to the full expressions in Eqs. (29) and (30). Events are simulated for $Pb+Pb$ collisions at $\sqrt{s_{NN}} = 2.76$ TeV, within a hybrid event-by-event hydrodynamic model (TRENTo+Music+UrQMD).

Each. In Fig. 8 we, once again, plot the first subleading principal component of elliptic and triangular flow. However, in this figure, we include other predictors as well. In Fig. 9 we show the first subleading principal component of $V_4(p_T)$ and $V_5(p_T)$. For these higher harmonics, results are less impressive as higher harmonics likely require a larger number of terms to describe.

Finally, we have also calculated principal components beyond the first subleading one. Figure 10 shows the third principal components of triangular and elliptic flow fluctuations. Results are shown only for full hydrodynamic simulations and for the full predictors on Eqs. (29) and (30). The agreement between the two curves is quite striking, considering the level of detail captured by the third principal component.
FIG. 7. Pearson correlation coefficient between the flow harmonics \( V_4(p_T) \) (upper panel) and \( V_5(p_T) \) (lower panel) and predictions of their event-by-event fluctuations from eccentricities of the initial transverse geometry. Different curves correspond to different predictors, with the blue dashed curve corresponding to the full expressions in Eqs. (29) and (30). Events are simulated for \( Pb + Pb \) collisions at \( \sqrt{s_{NN}} = 2.76 \) TeV, within a hybrid event-by-event hydrodynamic model (TRENTo+MUSIC+UrQMD).

FIG. 8. Second principal component of elliptic (upper panel) and triangular (lower panel) flow, both from full hydrodynamic simulations and event-by-event predictions from eccentricities of the initial geometry. Different curves correspond to different predictors, with the black solid curve corresponding to the full hydrodynamic results and the blue dashed curve corresponding to the full expressions in Eqs. (29) and (30). Events are simulated for \( Pb + Pb \) collisions at \( \sqrt{s_{NN}} = 2.76 \) TeV, within a hybrid event-by-event hydrodynamic model (TRENTo+MUSIC+UrQMD).
FIG. 9. Second principal component of $V_2(p_T)$ (upper panel) and $V_5(p_T)$ (lower panel), both from full hydrodynamic simulations and event-by-event predictions from eccentricities of the initial geometry. Different curves correspond to different predictors, with the black solid curve corresponding to the full hydrodynamic results and the blue dashed curve corresponding to the full expressions in Eqs. (B1) and (B2). Events are simulated for $Pb+Pb$ collisions at $\sqrt{s_{NN}} = 2.76$ TeV, within a hybrid event-by-event hydrodynamic model (TrENTo+Music+UrQMD).

FIG. 10. Third principal component of elliptic (upper panel) and triangular (lower panel) flow, both from full hydrodynamic simulations and event-by-event predictions from eccentricities of the initial geometry. The black solid curve corresponds to the full hydrodynamic results, while the blue dashed curve corresponds to the full expressions in Eqs. (29) and (30). Events are simulated for $Pb+Pb$ collisions at $\sqrt{s_{NN}} = 2.76$ TeV, within a hybrid event-by-event hydrodynamic model (TrENTo+Music+UrQMD).