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HAL Id: hal-03645917
https://hal.science/hal-03645917
Submitted on 6 Sep 2022

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Non-Gaussian statistics of critical sets in 2D and 3D: Peaks, voids, saddles, genus, and skeleton

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(Received 3 October 2011; published 26 January 2012)

The formalism to compute the geometrical and topological one-point statistics of mildly non-Gaussian two-dimensional and three-dimensional (3D) cosmological fields is developed. Leveraging the isotropy of the target statistics, the Gram-Charlier expansion is reformulated with rotation-invariant variables. This formulation allows us to track the geometrical statistics of the cosmic field to all orders. It then allows us to connect the one-point statistics of the critical sets to the growth factor through perturbation theory, which predicts the redshift evolution of higher-order cumulants. In particular, the cosmic nonlinear evolution of the skeleton’s length, together with the statistics of extrema and Euler characteristic are investigated in turn. In two dimensions, the corresponding differential densities are analytic as a function of the excursion set threshold and the shape parameter. In 3D, the Euler characteristics and the field isosurface area are also analytic to all orders in the expansion. Numerical integrations are performed and simple fits are provided whenever closed form expressions are not available. These statistics are compared to estimates from N-body simulations and are shown to match well the cosmic evolution up to root mean square of the density field of ~0.2. In 3D, gravitational perturbation theory is implemented to predict the cosmic evolution of all the relevant Gram-Charlier coefficients for universes with scale-invariant matter distribution. The one-point statistics of critical sets could be used to constrain primordial non-Gaussianities and the dark energy equation of state on upcoming cosmic surveys; this is illustrated on idealized experiments.

DOI: 10.1103/PhysRevD.85.023011

I. INTRODUCTION

Random fields are ubiquitous in physics. In cosmology, the large-scale distribution of matter (LSS) and the sky maps of the polarized CMB radiation are described as, respectively, three-dimensional (3D) and two-dimensional (2D) random fields. In primordial theories where initial seeds for cosmic structures have quantum origin, the fields of initial density fluctuations are close to Gaussian. Subsequent evolution develops further non-Gaussian signatures via nonlinear gravitational dynamics. Hence, these signatures might be used to constrain the dark energy equation of state via 3D galactic surveys, or shed light on the physics of the early Universe. The non-Gaussianities can be accessed through the higher-order moments of the field, which are generally difficult to estimate directly in real-life observations, due to their sensitivity to very rare events. The geometrical analysis of the critical sets of the field provides more robust measures of non-Gaussianity and has become an active field of investigation [1,2]. Another active area of research is centered around the use of the bi- or trispectrum for evaluation of the non-Gaussian effects [3–8]. This paper presents statistical tools that may help to detect unique signatures of the fundamental processes in our Universe.

Several statistics have been used to characterize the morphology of the density field of the large-scale structures. Peaks are one of the first components to have been studied in detail [9]. Maxima of the density fields are indeed considered as the location where gravitational collapse occurs to form the halos where galaxies form. Their distribution is thus a key ingredient in semianalytical models and for predictions of the so-called halo model [10,11]. In 2D, the statistics of peaks of the cosmic microwave background have been used to constrain its Gaussianity [12–15].

The Euler characteristic and the genus, which only differ by their normalization, have also been thoroughly studied, since they describe in part the topology of the field. They provide a different approach to the usual geometrical probes such as the power spectrum. The genus has been used in redshift surveys [2,16–19]. It has also been applied to the CMB to constrain the Gaussianity of primordial fluctuations, with results from COBE [20–22] and WMAP [1,23]. The genus can be defined in a probabilistic framework where its theory is more easily derived. Gaussian predictions have been known for a long time [24,25] and have been extended by Matsubara to non-Gaussian fields arising from gravitational nonlinear dynamics [26] and to proper accounting of redshift distortions [27,28]. More generally, the Euler characteristic is one of the Minkowski functionals (a complete set of morphological descriptors) [29] and has been combined to describe the shape of structures [30]. This paper is an extension of this approach to other characteristics and higher orders.
The skeleton has been introduced more recently [31,32]. It traces the filamentary structure of a field, extracting the “ridges” linking the maxima through saddle points. It has been applied in 2D for the CMB [33,34] and in 3D for the study of the large-scale structure [35–37]. Most of the efforts have focused up to now on improving the algorithm [38] and our theoretical knowledge: [38] gives a complete theoretical description of the properties of the skeleton for Gaussian fields. One of the motivations of the present paper is to extend this description of the skeleton beyond this specific Gaussian case. Indeed, mildly non-Gaussian random fields play an important role in cosmology: they can be present in the primordial fluctuations and open a window to the physics during the inflation phase [39], and they also arise, e.g., from the nonlinear dynamics of gravitational clustering [40].

In this paper, we develop the formalism for computing the geometrical and topological one-point statistics for mildly non-Gaussian 2D and 3D cosmological fields. Our formalism is based on the Gram-Charlier expansion of the joint probability distribution of the field and its derivatives using rotation-invariant variables which allows us to analyze the isotropic one-point statistics to all orders in cumulant expansion. In the following, we will predict the evolution of quantities such as the Euler characteristic, the density of extrema, and the differential length of the skeleton as a function of root mean square of the cosmic density field, $\sigma$. The evolution of these geometrical critical sets can thus be linked to the cosmic evolution of the underlying field. The next generation of redshift surveys, such as Big-BOSS or EUCLID and WFIRST,1 aim at constraining the equation of state of dark energy by mapping millions of galaxies. For example, [41] showed how using the genus in Big-BOSS could give a precision of 5% on a constant equation of state. Similarly, [42] argues that a joint power spectrum and extrema counts probe has applications to CMB non-Gaussianities and to the cosmic evolution of the large-scale structures. In particular, Sec. IV B 2 shows how to predict the cumulants that appear in these expansions using perturbation theory when cosmic gravitational clustering is responsible for their growth, Sec. IV B 3 compares the results to measurements on numerical realizations, and Sec. IV B 4 describes an idealized dark energy probe experiment. Appendix D details the link between the order in the Gram-Charlier expansion and that corresponding to gravitational perturbation theory, introducing the so-called Edgeworth expansion. It is then applied to the computation of the $\sigma^2$-order correction to the Euler characteristic in Appendix E. The results from perturbation theory to scale-invariant power spectra are given in Appendix F. Appendix G presents the algorithm implemented to measure the Euler characteristic.

II. ASYMPTOTIC EXPANSION OF THE JOINT PROBABILITY DISTRIBUTION FUNCTION OF THE FIELD AND ITS DERIVATIVES

The expectation of topological and geometrical statistics like the Euler characteristic, extrema counts, the differential length of skeleton requires the knowledge of the joint probability distribution function (JPDF) of the field and its derivatives up to second order [25]. In this paper, we advance the formalism that considers these statistics in non-Gaussian regime treated as the correction to the Gaussian limit. We start with developing the relevant version of non-Gaussian JPDF of the field and its derivatives, following closely the exposition in [43]. Starting with some notations, for a given field $\rho$, we define the moments $\sigma^2 = \langle \rho^2 \rangle$, $\sigma_1^2 = \langle (\nabla \rho)^2 \rangle$, $\sigma_2^2 = \langle (\Delta \rho)^2 \rangle$. Combining these moments, we can build two characteristic lengths, $R_0 = \sigma/\sigma_1$ and $R_s = \sigma_1/\sigma_2$, and the spectral parameter $\gamma = \sigma_1^2/(\sigma_2 \sigma)$. We choose to normalize the field $\rho$ and its derivatives to have unit variances: $x = \rho/\sigma$, $x_i = 1/\sigma_1 \nabla_i \rho$, and $x_{ij} = 1/\sigma_2 \nabla_i \nabla_j \rho$.

A. The Gram-Charlier expansion of the JPDF

A statistically homogeneous $N$-dimensional random field $x$ is fully characterized by the joint one-point distribution function of its value and its derivatives $P(x)$, $x = (x, x_i, x_{ij}, x_{ijk}, \ldots)$. The simplest case, frequently arising because of the central limit theorem (e.g., simple inflation models), is the Gaussian JPDF:

$$G(x) \equiv (2\pi)^{-N/2}|C|^{-1/2} \exp \left( -\frac{1}{2} x \cdot C^{-1} \cdot x \right).$$

---

where \( C = \langle \mathbf{x}^T \cdot \mathbf{x} \rangle \) is the covariance matrix of the \( \mathbf{x} \) variables and we have considered that all the variables are defined as having zero mean and unit variance \( (C_{ij} = 1 \text{ for each } i) \). For a general non-Gaussian field, the JPDF can be expanded around a Gaussian arranged to match the mean and the covariance of the field variables \( \mathbf{x} \), using a multivariate Gram-Charlier expansion [44–47]:

\[
P(\mathbf{x}) = G(\mathbf{x}) \left[ 1 + \sum_{n=3}^{\infty} \frac{1}{n!} \text{Tr}[\langle \mathbf{x}^n \rangle_{\text{GC}} \cdot \mathbf{h}_n(\mathbf{x})] \right].
\]

The correction to the Gaussian approximation is the series in Hermite tensors

\[
\mathbf{h}_n(\mathbf{x}) = (-1)^n G^{-1}(\mathbf{x}) \frac{\partial^n G(\mathbf{x})}{\partial \mathbf{x}^n},
\]
of rank \( n \). These tensors are the orthogonal polynomials related to the Gaussian kernel \( G(\mathbf{x}) \), which allows us to construct the so-called Gram-Charlier (GC) coefficients from the moments:

\[
\langle \mathbf{x}^n \rangle_{\text{GC}} = \langle \mathbf{h}_n(\mathbf{x}) \rangle.
\]

The statistics of the Euler characteristic (or equivalently the genus) and extremal points [25], and the basic description of the critical lines of the field in the stiff approximation [38], require to know the JPDF only up to the second derivatives. In the Gaussian limit, in this case, the only nontrivial covariance parameter is the cross correlation between the field and the trace of the Hessian

\[
\gamma = \langle \text{Tr}(\mathbf{x}_{ij}) \rangle [9, 38].
\]

B. Rotational invariants of the field and its derivatives

The JPDF in the form of Eq. (1) is not ideal to study critical sets statistics since the coordinate representation does not take advantage of the isotropic nature of the statistical descriptors. The pioneering works of [26,48], where the first correction to Euler characteristic was computed, demonstrate the arising complexities. Instead, we developed the equivalent of the Gram-Charlier expansion for the JPDF of the field variables that are invariant under coordinate rotation [15,43]. Such distribution can be computed via explicit integration of the series (1) over rotations. However, we obtain it here directly from general principles: the moment expansion of the non-Gaussian JPDF corresponds to the expansion in the set of polynomials which are orthogonal with respect to the weight provided by the JPDF in the Gaussian limit. Thus, the problem is reduced to finding such polynomials for a suitable set of invariant variables.

The rotational invariants that are present in the problem are the field value \( x \) itself, the square modulus of its gradient, \( q^2 = \sum x_i^2 \), and the invariants of the Hessian, the matrix of the second derivatives \( x_{ij} \). A rank \( N \) symmetric matrix has \( N \) invariants with respect to rotations. The eigenvalues \( \lambda_i \) provide one such representation of invariants, however they are complex algebraic functions of the matrix components. An alternative representation is a set of invariants that are polynomial in \( x_{ij} \), with one independent invariant polynomial per order, from one to \( N \). A familiar example is the set of coefficients, \( I_s \), of the characteristic equation for the eigenvalues, where the linear invariant is the trace, \( I_1 = \sum \lambda_i \), the quadratic one is \( I_2 = \sum_{i<j} \lambda_i \lambda_j \), and the \( N \)th order invariant is the determinant of the matrix, \( I_N = \prod \lambda_i \). Aiming at simplifying the JPDF in the Gaussian limit\(^2\) (e.g., [24,38,49]), we use their linear combinations, \( J_s \),

\[
J_1 = I_1, \quad J_{s \geq 2} = I_1 - \sum_{p=2}^{s} \frac{(-1)^p C_p^s}{N!} I_{1-p} I_p,
\]

where \( J_{s \geq 2} \) are (renormalized) coefficients of the characteristic equation of the traceless part of the Hessian and are independent in the Gaussian limit on the trace \( J_1 \).

Our choice of normalization of the field and its derivatives gives simple relations for the linear and quadratic in the field moments of the invariant variables:

\[
\langle \zeta \rangle = 0, \quad \langle J_1 \rangle = 0, \quad \langle \zeta J_1 \rangle = 0, \quad \langle \zeta^2 \rangle = 1,
\]

\[
\langle J_1^2 \rangle = 1, \quad \langle q^2 \rangle = 1, \quad \langle J_2 \rangle = 1,
\]

i.e., the linear invariant variables have zero means and unit variances and the quadratic ones have unit means. Since we expand JPDF around the Gaussian form that is arranged to match all these terms, the relations (4) are retained in non-Gaussian case.

C. The Gram-Charlier non-Gaussian joint probability distribution in two dimensions

The rotation-invariant formalism is straightforwardly adapted to 2D random fields, which are of interest for, e.g., CMB and weak lensing studies. There are just two invariants that describe the Hessian in 2D,

\[
I_1 = \text{Tr}(x_{ij}) = x_{11} + x_{22},
\]

\[
I_2 = \det[x_{ij}] = x_{11} x_{22} - x_{12}^2,
\]

which in the Hessian eigenframe have the form

\[
I_1 = \lambda_1 + \lambda_2, \quad I_2 = \lambda_1 \lambda_2.
\]

We shall also use the orthogonal set

\[
J_1 = I_1, \quad J_2 = I_1^2 - 4 I_2 = (x_{11} - x_{22})^2 + 4 x_{12}^2.
\]

Introducing \( \zeta = (x + \gamma J_1)/\sqrt{1 - \gamma^2} \) in place of the field value \( x \), we can build the 2D Gaussian JPDF \( G_{2D}(\zeta, q^2, J_1, J_2) \), normalized over \( d\zeta dq^2 dJ_1 dJ_2 \), which has a fully factorized form:

\(\text{By Gaussian JPDF, we mean the limit of the Gaussian field } x.\)

Note that even in this limit, the rotation-invariant variables that are nonlinear in field variables \( x \) are not Gaussian distributed.
The non-Gaussian rotation-invariant JPDF can then be
polynomials are the "probabilists" Hermite polynomials,
For the quadratic variables (q^2 and J_2), the corresponding polynomials are the Laguerre polynomials:

\[ \phi_n(x) = x^n \exp(-x/2) / \sqrt{2^n n!} \]

For the quadratic variables (q^2 and J_2), the corresponding polynomials are the Laguerre polynomials:

\[ \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) H_n(x) H_m(x) dx = n! \delta_{nm} \]  \hspace{1cm} (9)

The non-Gaussian rotation-invariant JPDF can then be written in the form of the direct series in the products of these polynomials:\

\[ P_{2D}(\zeta, q^2, J_1, J_2) = G_{2D} \left[ 1 + \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(-1)^{i+j+k+l}}{i!j!k!l!} \times \langle \xi^i q^j J_1^k J_2^l \rangle_{GC} H_i(\zeta) L_j(q^2) H_k(J_1) L_l(J_2) \right] \]  \hspace{1cm} (11)

where \( \sum_{i,j,k,l} \) stands for summation over all combinations of non-negative i, j, k, l such that i + 2j + k + 2l\nadds to the order of the expansion n, defined as the number of fields. This arrangement sorts the series in the order of increasing power of the field variables.\(^4\) When counting the order of a term, j and l enter multiplied by two, since each of q^2 and J_2 contributes a quadratic in the field quantity.

The orthogonality of the polynomials allows us to extract the coefficients by projection:

\[ \langle H_i(\zeta) L_j(q^2) H_k(J_1) L_l(J_2) \rangle = \int_{-\infty}^{\infty} \frac{d\xi}{(2\pi)^{1/2}} \int_{-\infty}^{\infty} dq^2 \int_{-\infty}^{\infty} dJ_1 \times \int_{0}^{\infty} dJ_2 P_{2D}(\zeta, q^2, J_1, J_2) H_i(\zeta) L_j(q^2) H_k(J_1) L_l(J_2) \]

\[ = \frac{(-1)^{i+j+k+l}}{j!k!} \langle \xi^i q^j J_1^k J_2^l \rangle_{GC}. \]  \hspace{1cm} (12)

We can thus express the Gram-Charlier coefficients as combinations of moments:

\[ \langle \xi^i q^j J_1^k J_2^l \rangle_{GC} = \frac{j!k!}{(i+j+k+l)!} \langle H_i(\zeta) L_j(q^2) H_k(J_1) L_l(J_2) \rangle. \]  \hspace{1cm} (13)

The chosen convention for the prefactor ensures that when the polynomial expressions are expanded, the highest power term is given directly by the corresponding moment:

\[ \langle \xi^i q^j J_1^k J_2^l \rangle_{GC} = \langle \xi^i q^j J_1^k J_2^l \rangle + \ldots \]  \hspace{1cm} (14)

For example,

\[ \langle \xi^3 \rangle_{GC} = \langle H_3(\zeta) \rangle = \langle \xi^3 - 3\xi \rangle = \langle \xi^3 \rangle - 3(\langle \xi \rangle)^3. \]

\[ \langle q^2 J_1 \rangle_{GC} = \langle L_1(q^2) H_1(J_1) \rangle = \langle 1 - q^2 \rangle J_1 \]

\[ = \langle q^2 J_1 \rangle - \langle J_1 \rangle = \langle q^2 J_1 \rangle. \]

\[ \langle J_1 J_2 \rangle_{GC} = \langle H_2(J_1) L_1(J_2) \rangle = -\langle (J_1^2 - 1)(1 - J_2) \rangle \]

\[ = \langle J_1 J_2 \rangle - \langle J_1^2 \rangle - \langle J_2 \rangle + 1 = \langle J_1 J_2 \rangle - 1, \]

where we used our normalization convention Eq. (8). Consequently, the \( n = 3 \) Gram-Charlier coefficients simply coincide with the third-order moments, but the higher-order ones are given by the combination of moments.

The combinations of moments that constitute the Gram-Charlier coefficients vanish in the Gaussian limit. Thus, they can be expressed via the cumulants of the field that have the same property. For the variables that are linear in the field, e.g., \( \zeta, J_1 \), the ordinary cumulants arise. However, the variables \( q^2 \) and \( J_2 \) are not Gaussian distributed in the limit of the Gaussian field due to their quadratic nature, and their own cumulants do not vanish in this limit, e.g., \( \langle J_1^2 \rangle_{cum} = \langle J_1^2 \rangle - \langle J_1 \rangle^2 \rightarrow 1 \). To take this effect into account, we introduce the notion of "field cumulants." A field cumulant (FC) of a polynomial variable is the cumulant of the expression in field variables, \( x, x_i, x_{ij} \), that constitutes it. By definition, it vanishes in the limit of the Gaussian field. For example, \( \langle J_1^2 \rangle_{FC} = \langle (x_{11} - 2x) + 4x_{11}^2 \rangle_c = \langle J_1^2 \rangle - \langle J_1 \rangle^2 \rightarrow 0 \). In the rest of the paper, we will always mean field cumulants and will use the subscript \( c \) to tag them. Expressing the Gram-Charlier coefficients via the field cumulants, we find that for the orders \( n = 3, 4, 5 \) they coincide, e.g.,

\[ \langle \xi^2 q^2 J_1 \rangle_{GC} = \langle H_2(\zeta) L_1(q^2) H_1(J_1) \rangle \]

\[ = \langle q^2 \rangle J_1 - \langle \xi \rangle \langle q^2 \rangle \phi(\zeta) \]

\[ = \langle \xi^2 q^2 J_1 \rangle - \langle \xi \rangle \langle q^2 \rangle \phi(\zeta) = \langle \xi^2 q^2 \rangle_c \phi(\zeta). \]

This property follows from the link between the Gram-Charlier and the Edgeworth expansions (see Appendix D). For the next orders \( n \geq 6 \), the Gram-Charlier coefficients are more complex combinations of cumulants, as explained in Appendix E.

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\(^3\)The orthogonality expressions and the choice of normalization for the Gram-Charlier coefficient [Eq. (12)] explain the quadruple factorial at the denominator together with the \( (-1)^{i+j+k+l} \) factor.

\(^4\)This ordering should not be confused with the order given by perturbation theory. See Appendix D for further explanations.
D. The Gram-Charlier non-Gaussian joint probability distribution in three dimensions

In 3D, the polynomial invariants of the matrix $x_{ij}$, as written in the coordinate frame, are

\[ I_1 \equiv \text{Tr}(x_{ij}) = x_{11} + x_{22} + x_{33}, \]
\[ I_2 = x_{11}x_{22} + x_{22}x_{33} + x_{11}x_{33} - x_{12}^2 - x_{13}^2 - x_{23}^2, \]
\[ I_3 \equiv \det(x_{ij}) = x_{11}x_{22}x_{33} + 2x_{12}x_{23}x_{13} - x_{11}x_{23}^2 - x_{22}x_{13}^2 - x_{33}x_{12}^2. \] (15)

The same invariants expressed through eigenvalues are

\[ I_1 = \lambda_1 + \lambda_2 + \lambda_3, \quad I_2 = \lambda_1\lambda_2 + \lambda_2\lambda_3 + \lambda_1\lambda_3, \]
\[ I_3 = \lambda_1\lambda_2\lambda_3. \] (16)

These variables can be made partially independent using the linear combinations

\[ J_1 = I_1, \quad J_2 = I_2 - 3I_1, \]
\[ J_3 = I_3 - \frac{9}{2}I_1I_2 + \frac{27}{2}I_3. \] (17)

In $N$ dimensions (ND), the Gaussian JPDF, $G_{\text{NP}}(\xi, q^2, J_1, J_2, J_3)$, retains the complete factorization with respect to $\xi$, $q^2$, $J_1$, $J_2$,

\[ G_{\text{ND}} = \frac{(N/2)^{N/2}q^{-N/2}}{(2\pi)^{N/2}} \exp \left[ -\frac{\xi^2}{2} - \frac{Nq^2}{2} - \frac{J_1^2}{2} \right] G(J_{\geq2}). \] (18)

so in the moment expansion this sector always gives rise to Hermite $H_j(\xi)$, $H_j(J_1)$, and generalized Laguerre $L^{(N-2)/2}(Nq^2/2)$ polynomials. However, the distributions of the rest $J_{s=2}$ contained in $G(J_{\geq2})$ are coupled. Specifically, in 3D, $G(J_2, J_3) = -25\sqrt{3}/12\pi \exp[-\frac{5}{2}J_2]$, and $J_3$ is distributed uniformly between $-J_2^{3/2}$ and $J_2^{3/2}$.

\[ P_{3D} = G_{3D} \left[ 1 + \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(-1)^{i+j+k+l}3!}{i!(1+2j)!!k!(3+2l)!!} (\xi q^2)^i J_{1}^{(i/2)} H_{1}(\xi) L_{j}^{(1/2)} \left( \frac{3}{5} q^2 \right) H_k(J_1) L_l^{(3/2)} \left( \frac{5}{2} J_2 \right) \right] \]
\[ + \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(-1)^{i+j+k+l}25}{i!(1+2j)!!k!(3+2l)!!} (\xi q^2)^{i/2} J_{1}^{(i/2)} H_{1}(\xi) L_{j}^{(1/2)} \left( \frac{3}{5} q^2 \right) H_k(J_1) L_l^{(3/2)} \left( \frac{5}{2} J_2 \right) \]
\[ + \sum_{n=3}^{\infty} \sum_{i,j,k,l,m=1} (-1)^{i+j+k+l+m} \frac{c_{im}}{i!(1+2j)!!k!(3+2l)!!} (\xi q^2)^{i/2} J_{1}^{(i/2)} \left( \frac{3}{5} q^2 \right) H_k(J_1) L_m^{(3/2)} \left( \frac{5}{2} J_2 \right) \] (22)

The first non-Gaussian correction term contains $J_2$ but not $J_3$, the second is linear in $J_3$ and contains no $J_2$, while the last one contains all the remaining combinations of $J_2$ and $J_3$. In this last term, we leave the normalization coefficients $c_{im}$ undetermined, but $c_{01} = 0$, so the first nonzero contribution, $I = m = 1$, is of the fifth power of the field and is

\[ G_{3D}(\xi, q^2, J_1, J_2, J_3) \]
\[ = \frac{25\sqrt{15}}{8\pi^2} \sqrt{q^2} \exp \left[ -\frac{1}{2} \frac{1}{2} - \frac{3}{2} \frac{1}{2} - \frac{1}{2} J_1^2 - \frac{5}{2} J_2 \right] \] (19)

with the ranges that variables span summarized in the normalization integral

\[ \int_{-\infty}^{\infty} d\xi \int_{0}^{\infty} dq^2 \int_{-\infty}^{\infty} dJ_1 \int_{0}^{\infty} dJ_2 \]
\[ \times \int_{-J_2^{3/2}}^{J_2^{3/2}} dJ_3 G_{3D}(\xi, q^2, J_1, J_2, J_3) = 1. \] (20)

Let us denote the orthogonal polynomials in the variables $J_2$ and $J_3$, $F_{lm}(J_2, J_3)$, where $l$ is the power of $J_2$ and $m$ is the power of $J_3$. They obey the orthonormality condition

\[ \int_{0}^{\infty} dJ_2 \int_{-J_2^{3/2}}^{J_2^{3/2}} dJ_3 G_{3D}(J_2, J_3) F_{lm}(J_2, J_3) = \delta_l^0 \delta_m^0. \] (21)

We shall not give the full theory of these polynomials here, but note that one can construct them by a Gram-Schmidt orthogonalization procedure [say using the MATHEMATICA ORTHOGONALIZE routine using the dot product defined by Eq. (21)] \footnote{For instance, the orthogonal polynomial of order one in $J_2$ and $J_3$ is proportional to $(5J_2 - 11)J_3$} to any given order. Two special cases

\[ F_{00} = \sqrt{\frac{3 \times 2!}{(3+2)!}} \frac{25\sqrt{3}}{(2J_2)} \] \quad and \quad \[ F_{01} = \frac{5}{\sqrt{21}} J_3 \]

are sufficient to obtain the general expression for the Euler characteristic of the excursion sets of the field to arbitrary order, and to calculate the critical point and skeleton statistics to quartic order. Hence, we write the moment expansion for invariant non-Gaussian JPDF, $P_{3D}(\xi, q^2, J_1, J_2, J_3)$, as a series in the power order of the field, $n$, in the form

\[ \langle \xi^i q^j J^k \rangle_{\text{GC}} = \frac{(-1)^i j! (-1)^j l!}{(3/2)^i (5/2)^j} \times \left( H_i(\xi) L_j^{(1/2)} \left( \frac{3}{5} q^2 \right) H_k(J_1) L_l^{(3/2)} \left( \frac{5}{2} J_2 \right) \right) \] (23)

and for $l = 0$ and $m = 1$.
\[ \langle \xi^2 q^3 J^3 \rangle_{GC} = \frac{(-1)^j}{(3/2)^j} \langle H_i(\xi)L_j^{1/2}(3/2)H_k(\xi)J^3 \rangle \] (24)

Equations (11) and (22) are the main results of this section. They fully characterize the joint probability function of a field and its derivatives in two and three dimensions, up to second-order derivatives, to an arbitrary order in the moment expansion (see also [15,25]).

III. NON-GAUSSIAN CRITICAL SETS IN 2 AND 3D

Given formulas (11) and (22), one can easily compute any rotation-invariant statistics that depend exclusively on these descriptors of the field. In particular, one can compute the filling factor of high excursion set (Secs. III A), the length of field isocontours in 2D and the area of isosurfaces in 3D (Secs. III B and III C), the Euler characteristic of the field (Secs. II B and II C), the density of extrema (Secs. III B and III C), and the properties of the critical lines that describe the skeleton of the field (Secs. III B and III C).

A. The filling factor of high excursion set

The filling factor, which is the volume fraction occupied by the field with values exceeding the threshold \( \nu \), presents the simplest case when one considers only the field and not its derivatives. The JPDF, Eqs. (12) and (33), is then reduced both in 2D and 3D to the same expression:

\[ P(\nu) = G(\nu) \left[ 1 + \sum_{n=3}^{\infty} \frac{(\nu^3)^n}{n!} H_n(\nu) \right] \] (25)

This is the classical one-variable expression for the Gram-Charlier expansion. This expansion does not have good convergence properties [47], but it is very simple to implement. When the origin of non-Gaussianities is known and can be used to predict the amplitude of each coefficient, this expansion can be resummed to a more physically motivated one. For example, as shown in Appendix D, when the non-Gaussianities give rise to a hierarchy of cumulants such that \( \langle \chi^n \rangle \propto \sigma^{n-2} \) (notably in the case of gravitational evolution), the resummation to a series in terms of \( \sigma \) leads to the Edgeworth expansion [45]:

\[ P(\nu) = G(\nu) \left[ 1 + \frac{(\nu^3)^2}{6} H_2(\nu) + \frac{(\nu^3)^3}{24} H_4(\nu) + \frac{(\nu^3)^4}{72} H_6(\nu) + \ldots \right] \] (26)

From the non-Gaussian PDF given by Eq. (25), the filling factor, \( f(\nu) \) immediately follows

\[ f(\nu) = \int_{\nu}^{\infty} P(\nu)d\nu = \frac{1}{2} \text{Erfc} \left( \frac{\nu}{\sqrt{2}} \right) + G(\nu) \sum_{n=3}^{\infty} \frac{(\nu^3)^n}{n!} \frac{H_{n-1}(\nu)}{n} \] (27)

B. Non-Gaussian critical sets in two dimensions

I. The length of 2D isocontours

The average length of the isocontour \( x = \nu \) is one of the two Minkowski functionals [29,51,52] that characterize the level sets in 2D. In our formalism, this is the simplest statistics for which the invariant variables appear nontrivially. To compute the average length per unit volume \( L \) of a \( \nu \) isocontour,

\[ L(\nu) = \frac{1}{R_0} \int_{-\infty}^{\nu} dx \int_{0}^{\infty} dq^2 P_{2D}(x, q^2) \delta^D(x - \nu)q \]

\[ = \frac{1}{R_0} \int_{0}^{\infty} dq^2 P_{2D}(\nu, q^2)q \] (28)

one needs JPDF of the field and the magnitude of its gradient. From Eq. (11),

\[ P_{2D}(x, q^2) = \frac{1}{\sqrt{2\pi}} e^{-\langle x^2 \rangle / 2} q \int_{-\infty}^{\infty} e^{-\langle \chi^2 \rangle / 2} \left[ 1 + \sum_{n=3}^{\infty} \frac{i^{n-2} (q)^{n-2} (-1)^j}{i!/j!} \right. \]

\[ \times \left. \langle x^i q^j \rangle_{GC} H_i(\chi)L_j(q^2) \right]. \] (29)

Performing integration over \( q^2 \) with the help of the identity \( \int_{0}^{\infty} e^{-t} t^j L_i(t)dt = -\Gamma(j + \frac{1}{2})/(4\Gamma(j + 1)) \) allows us to write the length of the \( \nu \) isocontour to an arbitrary order in non-Gaussianity as

\[ L(\nu) = \frac{1}{2\sqrt{2\pi} R_0} e^{-(\nu^2)/2} \left[ 1 + \frac{1}{\sqrt{2}} \sum_{n=3}^{\infty} \frac{i^{n-2} (\nu)^{n-2} (-1)^j j!}{i!/j!} \right. \]

\[ \times \left. \frac{\Gamma(j - \frac{1}{2})}{\Gamma(j + 1)} \langle x^i q^j \rangle_{GC} H_i(\nu) \right]. \] (30)

This result is a generalization of the published results to first [48] and second [53] orders. Appendix E2 reorders the expression (30) as an Edgeworth expansion.

2. The Euler characteristic of non-Gaussian 2D fields

The Euler characteristic of a manifold is defined as the alternating sum of the Betti numbers: \( \chi = b_0 - b_1 + b_2 - \ldots \). The first Betti numbers have a simple interpretation: \( b_0 \) is the number of connected components, \( b_1 \) is the number of two-dimensional holes, and \( b_2 \) is the number of three-dimensional voids. It is also one of the Minkowski functionals [29] and, in particular, it is additive. For example, the Euler characteristic is 1 for a ball and 2 for a sphere (which is a region surrounding a void). The Euler characteristic is an important quantity in topology and plays a central role in many different theories. For example, it can be linked to the geometry of the surface via the Gauss-Bonnet theorem or the extrema of a function via the Morse theory [54].

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In the specific case of an orientable and closed two-dimensional surface, it can be shown to be topologically equivalent to a sphere with \( g \) handles. One can thus use \( g \), called the genus, as an equivalent to the Euler characteristic. There is a direct relation between the Euler characteristic of a closed surface and its genus: \( \chi = 2 - 2g \).

In cosmology, the genus has been used to characterize the topology of the isocontours. However, the definition generally used by cosmologists is that the genus is the number of holes minus the number of isolated regions, which leads to an offset of 1: for an isolated region, the number of holes minus the number of isolated regions, called the genus, as an equivalent to the Euler characteristic of a random field, we use a central result from Morse theory [54]. Morse theory makes the link between the topology of a manifold and the critical points of the functions defined upon it. It shows that the Euler characteristic is simply equal to the alternating sum of the number of critical points of index \( i \) (i.e., where the Hessian has \( i \) negative eigenvalues): \( \chi = \sum (-1)^i n_i \). For a two-dimensional random field, the Euler characteristic \( \chi(\nu) \) of the excursion set \( x > \nu \) is given by

\[
\chi(\nu) = n_{\text{max}}(\nu) - n_{\text{sad}}(\nu) + n_{\text{min}}(\nu),
\]

where \( n_{\text{max}}(\nu), n_{\text{sad}}(\nu), \) and \( n_{\text{min}}(\nu) \) are the number of maxima, saddle points, and minima above the excursion threshold, \( \nu \). The density of Euler characteristic can thus be computed by integrating the determinant of the Hessian, \( I_2 \):

\[
\chi(\nu) = \frac{1}{R^2} \int_{-\infty}^{\infty} dJ_1 \int_{(\nu + y_1)^2}^{\infty} \frac{d\zeta}{\sqrt{1 - \nu^2}} \times \int_0^{2\pi} dJ_2 P_{\text{ext}}(\zeta, J_1, J_2) I_2,
\]

where \( P_{\text{ext}} \) is the distribution function under the condition of zero gradient:

\[
P_{\text{ext}}(\zeta, J_1, J_2) = \frac{1}{\pi} \int_0^{\infty} dq^2 P_{2D}(\zeta, q^2, J_1, J_2) \delta_D(q^2).
\]

As demonstrated in Appendix A 1, this integration yields

\[
\chi(\nu) = \frac{1}{2} \text{Erfc}(\frac{\nu}{\sqrt{2}}) \chi(-\infty) + \frac{1}{4\pi \sqrt{2} \sqrt{2\pi} R^2} \exp\left(-\frac{\nu^2}{2}\right) \times \left[ \gamma^2 H_1(\nu) + \sum_{n=3}^{\infty} \sum_{i,j,k} \frac{i+j+k-n}{i!j!k!} (\zeta_i q^2 J_1^k) \gamma(1 - \gamma^2)^{n/2} \right]
\times \sum_{k=0}^{\text{min}(2k, k)} \frac{2! \kern-2pt \sqrt{2 \pi}}{2^{k+2} s!(2-s)!} (k-s)! H_{i+k+1-2s}(\nu) + \sum_{n=3}^{\infty} \sum_{i,j,k} \frac{i+j+k+2-n}{i!j!k!} (\zeta_i q^2 J_1^k J_2) \gamma(1 - \gamma)^{n/2} \gamma^k H_{i+k-1}(\nu) \right].
\]

Note that in the rare event limit, \( \nu \gg 1 \) or \( \nu \ll 1 \), Eq. (34) represents, respectively, the differential number of maxima or minima, as the contribution to Eq. (31) from other extremal points becomes negligible [56]. In this limit, Eq. (34) yields both quantities to all orders.

A comment is in order on the value of the Euler characteristic density evaluated at infinitely low threshold, \( \chi(-\infty) \). In a finite, topologically nontrivial space, \( \chi(-\infty) \) evaluates to the Euler characteristic of the manifold itself, divided by its volume. While the result of Morse theory in Eq. (31) is exact, the expression via the local estimate of the extrema densities, Eq. (32), does not contain this effect as \( I_2 = 0 \). The expansion in Eq. (34) shows the appearance of the boundary term \( \frac{1}{2} \text{Erfc}(\nu) \chi(-\infty) \) when the moments of the form \( \langle J_1^2 q^2 \rangle - \langle J_2 q^2 \rangle \) do not cancel exactly. We conjecture that such boundary effects are captured by the \( R_\gamma \) scale introduced in Ref. [56].
term accounts for the topology of the manifold on which the field is considered. It might perhaps be one of the most important applications for the genus statistics of masked data sets with the mask introducing nontrivial topology.

When the non-Gaussianities give a hierarchy of cumulants such that a cumulant of $n$ fields is of order $\sigma^{n-2}$ (notably for gravitational evolution), the Gram-Charlier series can be reordered as the series in powers of $\sigma$ (called an Edgeworth expansion, see Appendix D):

$$\chi(n) = \frac{1}{2} \text{Erfc} \left( \frac{\nu}{\sqrt{2}} \right) \chi(-\infty) + \frac{1}{4\pi\sqrt{2\pi}R^2} \exp \left( -\frac{\nu^2}{2} \right) \left[ \chi^2 H_1(n) + \sum_{n=1}^{\infty} \chi^{(n)}(n) \right], \quad (35)$$

with $\chi^{(n)} \sim \sigma^n$. The first two orders of this expansion in powers of $\sigma$ are (see Appendix E)

$$\chi^{(1)}(n) = \left( 2\gamma^2g^2I_1 + 4(g^2I_2) \right) H_0(n) - \left( \gamma^2x^2 + \gamma(x^2I_1) \right) H_2(n) + \frac{\gamma^2}{6} \left( x^3 \right) H_4(n), \quad (36)$$

$$\chi^{(2)}(n) = \left( \frac{\gamma^2}{2} \langle q^4 \rangle_c + 2\gamma \langle q^2x^2 \rangle_c + \langle x^2 \rangle_c \right) H_1(n) - \left( \frac{\gamma^2}{2} \langle x^2q^2 \rangle_c + \frac{\gamma}{3} \langle x^3J_1 \rangle \right) H_3(n) + \frac{\gamma^2}{24} \left( x^4 \right) H_5(n)$$

$$- \left( 4\gamma \langle x^2q^2J_1 \rangle + \langle q^2x^2J_1 \rangle + 4\langle x^2q^2 \rangle \right) H_1(n) + \left( \gamma^2 \langle x^2q^2 \rangle^2 + \frac{\gamma}{3} \langle x^3 \rangle \langle q^2J_1 \rangle + \gamma \langle x^2 \rangle \langle q^2J_1 \rangle \right)$$

$$+ \left( \frac{\gamma^2}{4} \langle x^2J_1 \rangle^2 + \frac{\gamma}{3} \langle x^3 \rangle \langle x^2 \rangle \right) H_3(n) - \left( \frac{\gamma}{6} \langle x^2q^2 \rangle \langle x^3 \rangle \right) H_5(n) + \frac{\gamma^2}{72} \langle x^3 \rangle^2 \right) H_7(n), \quad (37)$$

while the $\sigma^3$ correction, $\chi^{(3)}(n)$, is given in Appendix E by Eq. \((E15)\).

Figure 1 compares the first-order non-Gaussian prediction to simulations of non-Gaussian density field developed through nonlinear gravitational instability. As described in detail in Sec. IV B 3, we performed 25 scale-invariant $(n = -1)$ 256$^3$ N-body simulations which are then cut into 25 slices each to form 2D fields. The cubic cumulants required to form the non-Gaussian prediction [Eq. \((36)\)] are then measured from these slices and the 2D Euler characteristic is determined using the suite of algorithms described in Appendix G. The agreement between the measured Euler characteristic and the theory is remarkable. Perhaps the most striking fact is that the non-Gaussian correction indeed arises as a combination of low-order Hermite polynomials. Alternatively, in this model the cumulants can be predicted by perturbation theory (PT) (see section IV B 1). The right panel in Fig. 1 demonstrates that both approaches give consistent results.\(^7\)

Figure 2 demonstrates how non-Gaussian deviations in the Euler characteristic develop with increase in nonlinearities as tracked by $\sigma$. At small $\sigma = 0.06$, the deviations are small in magnitude and are symmetric with respect to the threshold. Such signal is perfectly described by the first-order non-Gaussian correction to $\chi$, Eq. \((36)\), which contains only the even Hermite polynomials in $\nu$. As $\sigma$
increases, the magnitude of the non-Gaussian signal increases as well. The symmetric (even in parity) first-order approximation of Eq. (36) continues to describe the bulk of the effect until $\sigma \approx 0.1$; starting with $\sigma \approx 0.15$ the appearance of an antisymmetric, odd contribution (a hint of which can be already seen at $\sigma = 0.1$) becomes also apparent. This is exactly the contribution that is described by the next $\sim \sigma^2$ order in the perturbation expansion, which, given in Eq. (37), is constructed from the odd $H_i(\nu)$. We stress that the measurements in a simulated evolving field indeed demonstrate this transition in $\chi(\nu)$ parity properties as non-Gaussianity increases, thus strongly supporting our theoretical understanding of non-Gaussianity in this topological measure.

Another effect, revealed in Fig. 2, is a decrease of the relative uncertainty of the measurements as $\sigma$ increases: as expected, the stronger non-Gaussianities are easier to measure. Thus, when using non-Gaussian measures for the purpose of determining parameters of cosmological models, one is faced with a balance between the need for a strong signal, and the necessity to use higher-order terms in theoretical predictions if the effects are too large. Our study suggests $\sigma \sim 0.1$ as the best compromise regime if the theory is limited to the first-order perturbative corrections. This “rule-of-thumb” prescription finds confirmation in Sec. IV.

3. Non-Gaussian total and differential extrema counts in 2D

Extrema counts, especially that of the maxima of the field, also have long application to cosmology (e.g., [9], BBKS), however their theoretical development has been mostly restricted to the Gaussian fields. Given the knowledge of the JPDF [Eq. (11)], non-Gaussian extrema can be computed in the similar way as the Euler characteristic [15]. For example, the density number of all critical points above the threshold $\nu$ is

$$n(\nu) = n_{\max}(\nu) + n_{\text{rad}}(\nu) + n_{\min}(\nu)$$

$$= \frac{1}{R^2} \int_{-\infty}^{\infty} dJ_1 \int_{\nu + \gamma J_1/\sqrt{1-\gamma^2}}^{\infty} d\zeta$$

$$\times \int_0^\infty dJ_2 P_{\text{ext}}(\zeta, J_1, J_2) |I_2|, \quad (38)$$
the difference from Eq. (32) only involves taking the absolute value of the Hessian determinant $I_2$. Different types of critical points can be considered individually by restraining the integration domain in the $J_1 - J_2$ plane to ensure the appropriate signs for the eigenvalues. Specifically, the integration ranges for maxima are $J_1 \in [-\infty, 0]$, $J_2 \in [0, J^*_2]$; for minima $J_1 \in [0, \infty]$, $J_2 \in [0, J^*_2]$; and for saddle points $J_1 \in [-\infty, \infty]$, $J_2 \in [J^*_2, \infty]$.

The total number density of extrema is obtained by integrating over threshold $\nu$. The first (cubic) non-Gaussian corrections in 2D are

$$n_{\text{max}} = \frac{1}{8\sqrt{3}\pi R^2} + \frac{18(\nu^2 J_1) - 5(J^*_1) + 6(I_1 I_2)}{54\pi\sqrt{2}\pi R^2}, \quad n_{\text{min}} = \frac{1}{8\sqrt{3}\pi R^2} - \frac{18(\nu^2 J_1) - 5(J^*_1) + 6(I_1 I_2)}{54\pi\sqrt{2}\pi R^2}, \quad n_{\text{sad}} = \frac{1}{4\sqrt{3}\pi R^2} + 0. \quad (39)$$

The total number of saddles, as well as the total number of all the extremal points, is preserved to the first order (the latter follows for the former, since $n_{\text{max}} - n_{\text{sad}} + n_{\text{min}}$ is the Euler characteristic of the manifold and is thus fixed).

The differential distribution of the number of extrema with the threshold can be also found analytically in 2D:

$$\frac{\partial n_{\text{max}}}{\partial \nu} = \frac{1}{\sqrt{2\pi}R^2} \exp\left(-\frac{\nu^2}{2}\right) \left[ 1 + \text{erf}\left(\frac{\gamma\nu}{\sqrt{2(1 - \gamma^2)}}\right) \right] K_1(\nu, \gamma) + \frac{1}{\sqrt{2\pi}(1 - \gamma^2) R^2} \exp\left(-\frac{\nu^2}{2(1 - \gamma^2)}\right) K_3(\nu, \gamma),$$

$$\frac{\partial n_{\text{min}}}{\partial \nu} = \frac{1}{\sqrt{2\pi}R^2} \exp\left(-\frac{\nu^2}{2}\right) \left[ 1 - \text{erf}\left(\frac{\gamma\nu}{\sqrt{2(1 - \gamma^2)}}\right) \right] K_1(\nu, \gamma) - \frac{1}{\sqrt{2\pi}(1 - \gamma^2) R^2} \exp\left(-\frac{\nu^2}{2(1 - \gamma^2)}\right) K_3(\nu, \gamma),$$

$$\frac{\partial n_{\text{sad}}}{\partial \nu} = \frac{2}{\sqrt{2\pi}(1 - 2/3\gamma^2) R^2} \exp\left(-\frac{3\nu^2}{6 - 4\gamma^2}\right) K_2(\nu, \gamma). \quad (40a, 40b, 40c)$$

where $K_1$, $K_2$, $K_3$ are polynomials in $\nu$ with coefficients given by the cumulants and $\gamma$. Since

$$\frac{\partial n_{\text{max}}}{\partial \nu} + \frac{\partial n_{\text{min}}}{\partial \nu} - \frac{\partial n_{\text{sad}}}{\partial \nu} = -\frac{\partial}{\partial \nu} \chi(\nu)$$

we immediately find that $K_1$ is related to the series that appear in the expression for the Euler characteristic, Eqs. (34) and (36). Specifically, $K_1$ reads

$$K_1(\nu, \gamma) = \gamma^2 \left[ \frac{2}{8\pi} H_2(\nu) + \frac{2}{\gamma} \left(\frac{\nu^2}{4} (q^2 J_1) + \frac{1}{\gamma} (x J_2^2) \right) - \frac{1}{\gamma} \left(\frac{\nu^2}{4} (x J_2) \right) \right]$$

$$+ \frac{1}{\gamma} \left(\frac{\nu^2}{4} (x J_1) \right) H_3(\nu)$$

$$+ \frac{1}{6} \left(\frac{\nu^2}{4} (x^3) \right) H_3(\nu). \quad (42)$$

In the limit $\nu \gg 1$, the 3 terms are dominated by their respective exponentials. Since $2(1 - \gamma^2)$ and $(6 - 4\gamma^2)/3$ are smaller than 1, the $K_2$ and $K_3$ terms decrease more quickly than the $K_1$ term. Thus, in the limit of large $\nu$ the expression of the density of maxima is dominated by the $K_1$ term and coincides with the expression of the Euler characteristic, as expected.

To write the remaining $K_2$ and $K_3$ more compactly, we introduce two types of the scaled Hermite polynomials $H^*_n(\nu/k) = k^n H_n(\nu/k)$ and $H^*_{\text{sad}}(\nu, k) = k^{-n} H\text{sad}(\nu/k)$, where the choice of $k$ will be dictated by the exponential prefactor in the correspondent term. The polynomial $K_2$ is given by the straightforward expansion in $H^*_{\text{sad}}(\nu, \sqrt{1 - 2/3\gamma^2})$,

$$K_2(\nu, \gamma) = \frac{1}{8\pi \sqrt{3}} \left[ 1 - \left(\frac{\nu^2}{3} \gamma (x J_2) + \frac{2}{3} \gamma (x^2 J_1) - \frac{4}{3} (x J_2) \right) \right.$$

$$+ \frac{2}{9} \gamma (J^*_1 - 2J_1 J_2) H^*_1(\nu, \sqrt{1 - 2/3\gamma^2})$$

$$+ \frac{1}{6} \left(\frac{\nu^2}{4} (x^3) + 2\gamma (x^2 J_1) + \frac{4}{3} \gamma^2 (x J_1^2) + \frac{2}{3} \gamma^2 (x J_2) \right)$$

$$\left. + \frac{8}{27} \gamma^3 (J^*_1 + \frac{4}{9} \gamma^3 (J_1 J_2) \right) \right) H^*_3(\nu, \sqrt{1 - 2/3\gamma^2}). \quad (43)$$

The knowledge of $K_1$ and $K_2$ is sufficient to describe the differential number density of the sum of extrema of all the types.
The remaining term, \( K_3(\nu, \gamma) \), is the most complicated. It is simplest to express it as the expansion in \( \mathcal{H}^+(\nu, \sqrt{1 - \gamma^2}) \) but the expression is still quite long:

\[
K_3(\nu, \gamma) = \frac{\gamma(1 - \gamma^2)}{2(2\pi)^{3/2}} \left[ \mathcal{H}_1(\nu, \sqrt{1 - \gamma^2}) + \left( \frac{\gamma^2(1 + \gamma^2 - 26 \gamma^4 + 28 \gamma^6 - 8 \gamma^8)}{2(3 - 2\gamma^2)^3} \langle x^2 \rangle - \frac{2\gamma^3(13 - 14 \gamma^2 + 4 \gamma^4)}{(3 - 2\gamma^2)^3} \langle x^2 J_1 \rangle \right.
\]
\[
+ \frac{(1 - \gamma^2)(1 + 2 \gamma^2)}{3 - 2\gamma^2} \langle x^2 q^2 \rangle - \frac{24 - 26 \gamma^2 + 8 \gamma^4}{(3 - 2\gamma^2)^3} \langle x J_1^2 \rangle + \left. \frac{15 - 23 \gamma^2 + 8 \gamma^4}{(3 - 2\gamma^2)^3} \langle x J_2 \rangle + \frac{4(1 - \gamma^2)}{\gamma(3 - 2\gamma^2)} \langle q^2 J_1 \rangle \right]
\]
\[
- \frac{2(5 - 6 \gamma^2 + 2 \gamma^4)}{\gamma(3 - 2\gamma^2)^3} (J_1^3) + \frac{6(1 - \gamma^2)(2 - \gamma^2)}{\gamma(3 - 2\gamma^2)^3} (J_1 J_2) \mathcal{H}_0^+(\nu, \sqrt{1 - \gamma^2}) + \left. \frac{27 - 36 \gamma^2 + 224 \gamma^4 - 192 \gamma^6 + 48 \gamma^8}{6(3 - 2\gamma^2)^3} \langle x^3 \rangle \right]
\]
\[
+ \frac{2(-9 + 27 \gamma^2 - 18 \gamma^4 + 4 \gamma^6)}{\gamma(3 - 2\gamma^2)^3} \langle x^2 J_1 \rangle - \langle x^2 J_1 \rangle + \frac{6}{(3 - 2\gamma^2)^3} (x J_1^2) + \left. \frac{3}{(3 - 2\gamma^2)^3} (x J_2) \right]
\]
\[
+ \frac{4\gamma}{3(3 - 2\gamma^2)^3} (J_1^3) + \frac{2\gamma}{(3 - 2\gamma^2)^3} (J_1 J_2) \right] \mathcal{H}_1^+(\nu, \sqrt{1 - \gamma^2}) \right].
\]

(45)

Figure 3 (left panel) shows samples of this distribution for several values of the spectral parameter \( \gamma \) according to perturbation theory (see Sec. IV B 2). Figure 3 (right panel) shows the very good agreement between these predictions and the measurements of the corresponding number densities in simulations (see Sec. IV B 3). In 2D, small non-Gaussianities have little effect on the saddle points, slightly shifting the differential distribution toward the lower field thresholds, while keeping the total number of the saddles constant to first order. In contrast, there are more significant effects on the maxima and minima. The maxima distribution becomes broader, developing high \( \nu \) excess relative to the Gaussian formula. The minima distribution behaves oppositely. It sharpens around its peak at \( \nu \sim -1.5 \), while at the same time losing the lowest minima. This reflects the intuitive picture of non-linear gravitational clustering developing high-density peaks, while stretching and smoothing low \( \nu \) voids.

4. The differential length of the skeleton for non-Gaussian fields in 2D

The skeleton is a tool used to extract the filamentary structure of a field [31]. The theory of the skeleton for Gaussian fields has been presented in [38], which demonstrated that its differential length can be approximately (in the so-called stiff approximation) computed as

\[
\frac{dL_{\text{skeleton}}}{d\nu} = \frac{1}{R_e} \int d^2\vec{x}_i d^2x_{kl} P(\nu, \vec{x}_i, x_{kl}) \delta^D(\vec{x}_2) |\lambda_2|.
\]

(46)

where the \( \vec{x}_i \) are the components of the gradient in the Hessian eigenframe. To separate the part of the skeleton that maps closely the high-field value filamentary

![Graphs showing the differential distribution of maxima in 2D slices of 3D density field evolved to \( \sigma = 0.1 \) for several values of the initial spectral index (from left to right, red: \( n_s = -1.8 \), green: \( n_s = -1.5 \), blue: \( n_s = -1 \), purple: \( n_s = 0 \). Dotted lines are the Gaussian approximation, solid lines are the first-order non-Gaussian prediction. Right: Measured maxima (blue, right), saddle point (green, middle), and minima (red, left) distributions for \( n_s = -1 \) versus Gaussian (dashed lines) and the first-order non-Gaussian (solid lines) predictions. The shaded region corresponds to the 3-sigma dispersion within the simulation set.](#)
structures from the other critical lines, in [38] the notion of “primary” skeleton was introduced, defined by the constraint

$$\lambda_1 + \lambda_2 \leq 0. \quad (47)$$

The skeleton presents an insightful case study of how our invariant formalism can be extended to the cases in which the independent rotation symmetry of the gradient and the Hessian is broken. The skeleton definition couples the directions of the gradient and the Hessian: it singles out the direction associated with the highest eigenvalue of the Hessian ($\lambda_1$) which is the local direction of the skeleton. While the components of the gradient orthogonal to the skeleton’s direction are zero, their parallel component should be integrated over. This does not break the isotropy for the Hessian-related part of the expansion since the integrand only contains $\lambda_2$ which is a rotation invariant of the Hessian. However, it means that we cannot independently average over the directions of the gradient and the expansion has to treat its components separately.

We will first consider that the statistical properties of the components of the gradient are the same in the Hessian eigenframe as in any frame. Appendix C discusses this assumption and shows how to avoid it and that it only has a secondary impact on the results. In practice, the gradient part of the Gaussian PDF is

$$G(x_1, x_2) = \frac{1}{\pi} \exp(-x_1^2 - x_2^2). \quad (48)$$

The corresponding orthogonal polynomials needed to build an expansion around the Gaussian kernel $1/\sqrt{\pi} e^{-x^2}$ are the “physicists” Hermite polynomials $H_n^{\text{phys}}(x) = 2^{n/2} H_n(\sqrt{2}x)$. Their orthogonality relation is

$$\frac{2^{(n+1)/2}}{\sqrt{\pi}} \int_{-\infty}^{\infty} e^{-x^2} H_n(\sqrt{2}x) H_m(\sqrt{2}x) dx = \delta_{nm}. \quad (49)$$

The Gram–Charlier expansion is then

$$P(\xi, x_1, x_2, J_1, J_2) = G\left(1 + \sum_{n=3}^{\infty} \sum_{i,a,b,k,l} \frac{(-1)^i 2^{n/2} b^{/2}}{i!a!b!k!l!} \langle \xi | x_1^i x_2^a J_1^b J_2^k \rangle_{\text{GC}} H_i(\xi) H_a(\sqrt{2}x_1) 
\times H_b(\sqrt{2}x_2) H_k(J_1) L_i(J_2),
\right)$$

with the normalization of the Gram–Charlier coefficients again such as to have the highest-degree term to coincide with the corresponding moment:

$$\langle x_1^n \rangle_{\text{GC}} = 2^{-n/2} \langle H_n(\sqrt{2}x_1) \rangle. \quad (50)$$

Using the variable $x$ instead of $\xi$ allows us to introduce a threshold more easily. By definition of $\xi = (x + \gamma J_1)/\sqrt{1 - \gamma^2}$, the Jacobian of the transformation is $d\xi dJ_1 = 1/(1 - \gamma^2) dx dJ_1$ and the Gaussian PDF becomes

$$G(x, x_1, x_2, J_1, J_2) = \frac{1}{2\pi \sqrt{1 - \gamma^2}} \exp\left(-\frac{(x + \gamma J_1)^2}{2(1 - \gamma^2)} - x_1^2 - x_2^2 - \frac{J_1^2}{2} - J_2^2\right). \quad (51)$$

The expression of the non-Gaussian PDF in these variables is then

$$P(x, x_1, x_2, J_1, J_2) = G(x, x_1, x_2, J_1, J_2) \left(1 + \sum_{n=3}^{\infty} \sum_{i,a,b,k,l} \frac{(-1)^i 2^{n/2} b^{/2}}{i!a!b!k!l!} \langle \xi | x_1^i x_2^a J_1^b J_2^k \rangle_{\text{GC}} 
\times H_i(\sqrt{2}x_1) H_a(\sqrt{2}x_2) H_b(J_1) L_i(J_2)\right). \quad (52)$$

To express the integrand $|\lambda_2|$ via the Gaussian invariants and determine the appropriate limits of integration, we note that the skeleton constraint, $J_1 = \lambda_1 + \lambda_2 \leq 0$, implies that $\lambda_2 \leq 0$, since eigenvalues are sorted, $\lambda_1 \geq \lambda_2$. The relation between the integrand and the invariants is thus

$$|\lambda_2| = \frac{1}{2} (\sqrt{J_2} - J_1), \quad (53)$$

and the differential length of the skeleton is given by

$$\frac{\partial L_{\text{ske}}}{\partial \nu} = \frac{1}{2R_s} \int_0^{\infty} dJ_2 \int_{-\infty}^{0} dJ_1 \int_{-\infty}^{+\infty} dx_1 
\times \int_{-\infty}^{+\infty} dx_2 \delta^D(x_2)(\sqrt{J_2} - J_1) \left(1 + \sum_{i,a,b,k,l} \frac{(-1)^i 2^{n/2} b^{/2}}{i!a!b!k!l!} \langle \xi | x_1^i x_2^a J_1^b J_2^k \rangle_{\text{GC}} 
\times H_i(\sqrt{2}x_1) H_a(\sqrt{2}x_2) H_b(J_1) L_i(J_2)\right). \quad (54)$$

The integration over $x_2$ is trivial while the integration over $x_1$ corresponds to the projection on $H_0(\sqrt{2}x_1)$ that selects only the $a = 0$ term:

$$\frac{\partial L_{\text{ske}}}{\partial \nu} = \frac{1}{2R_s} \int_0^{\infty} dJ_2 \int_{-\infty}^{0} dJ_1 \left(\sqrt{J_2} - J_1\right) G(x = \nu, 0, J_1, J_2) 
\times \left(1 + \sum_{n=3}^{\infty} \sum_{i,a,b,k,l} \frac{(-1)^i 2^{n/2} b^{/2}}{i!a!b!k!l!} \langle \xi | x_1^i x_2^a J_1^b J_2^k \rangle_{\text{GC}} 
\times H_i(\sqrt{2}x_1) H_a(\sqrt{2}x_2) H_b(J_1) L_i(J_2)\right). \quad (55)$$

The result of the remaining integrations takes the form

$$\frac{\partial L_{\text{ske}}}{\partial \nu} = \frac{1}{2R_s} \int_0^{\infty} dJ_2 \int_{-\infty}^{0} dJ_1 \left(\sqrt{J_2} - J_1\right) G(x = \nu, 0, J_1, J_2) 
\times \left(1 + \sum_{n=3}^{\infty} \sum_{i,a,b,k,l} \frac{(-1)^i 2^{n/2} b^{/2}}{i!a!b!k!l!} \langle \xi | x_1^i x_2^a J_1^b J_2^k \rangle_{\text{GC}} 
\times H_i(\sqrt{2}x_1) H_a(\sqrt{2}x_2) H_b(J_1) L_i(J_2)\right). \quad (55)$$
In the Gaussian case, the coefficients $A$ and $B$ were given in \cite{38}: $A^{(0)} = \sqrt{1 - \gamma^2/(2\sqrt{2}\pi)}$ and $B^{(0)} = 1/(8\sqrt{\pi}) \times (\sqrt{\pi} + 2\gamma \nu)$. At the $\sigma$ order, the non-Gaussian corrections are

$$A^{(1)} = \sqrt{\frac{1 - \gamma^2}{2\sqrt{2}\pi}} \left[ \frac{\sqrt{\pi}(-2\gamma^3 + \gamma^5) + 2(3\gamma^2 - 5\gamma^4 + 2\gamma^6)H_1(\nu) - \sqrt{\pi}(3\gamma - 3\gamma^3 + \gamma^5)H_2(\nu) + 2(1 - \gamma^2)^3H_3(\nu)}{12(1 - \gamma^3)} \right]$$

$$+ \frac{-\sqrt{\pi}\gamma^2 + 2(2\gamma - 3\gamma^3 + \gamma^5)H_1(\nu) - \sqrt{\pi}H_2(\nu)}{4(1 - \gamma^3)} (x^2J_1) + \frac{-\sqrt{\pi}\gamma^3 - 2H_1(\nu) - \sqrt{\pi}\gamma H_2(\nu)}{4(1 - \gamma^3)} (xJ_1^2)$$

$$+ \frac{\sqrt{\pi}(1 - 2\gamma^2) - 2\gamma H_1(\nu) - \sqrt{\pi}\gamma^2 H_2(\nu)}{12(1 - \gamma^3)} (J_1^3) - \frac{\sqrt{\pi}\gamma}{4(1 - \gamma^3)} (x^2J_2) - \frac{\sqrt{\pi}}{4(1 - \gamma^3)} (J_1J_2)$$

$$+ \left( \frac{\sqrt{\pi}}{2(1 - \gamma^2)} - H_1(\nu) \right) (xx_2^2) + \frac{\sqrt{\pi}}{2(1 - \gamma^2)} (x^2J_1).$$

$$B^{(1)} = \frac{1}{8\sqrt{\pi}} \left[ \left( \frac{\sqrt{\pi}}{6} H_3(\nu) + \frac{\gamma}{3} H_4(\nu) \right) (x^3) - H_2(\nu) (x^2J_1) + \frac{\sqrt{\pi}}{2} H_1(\nu) (xJ_2) - (\sqrt{\pi}H_1(\nu) + 2\gamma H_2(\nu)) (xx_2^2) + 2(x^2J_1) \right].$$

Figure 4 (left panel) displays the variation of this differential length with respect to (w.r.t.) the underlying power index, $n_s$. The dynamical evolution tends to lower the total length of the skeleton. This corresponds to filaments merging. At the same time, at the very high density tail $\nu > 2$, notice an increase in the length of the skeleton per unit volume, which reflects the compression of the high-density regions. These predictions are easily adapted to the anti-skeleton that maps the troughs of the field by replacing both fields and cumulants with their opposite, i.e., replacing $\nu$ with $-\nu$ and the odd moments by their negative.

Figure 4 (right panel) shows a reasonable agreement between these predictions and the measured differential length. The discrepancies between the predictions and the measurements could have come from differences between the analytical definition of the skeleton and its numerical implementation \cite{57} and, possibly, from the approximation used to predict the differential length of the skeleton (the “stiff approximation,” as explained in \cite{38}). Both issues are more of a concern for low value thresholds $-1 < \nu < 1$, where both the skeleton and antiskeleton are the most complicated, while the differential behavior at high threshold values is much more robust, as was pointed out in \cite{38}. This complexity in interpreting the skeleton length make us stress that one should focus on differential lengths statistics at high thresholds, $\nu > 1$ (and $\nu < -1$ for the antiskeleton), when comparing theoretical predictions for this statistic to the data. When comparing Eq. (56) to the
data in the right panel of Fig. 4, we have adjusted the overall amplitude by a factor $L_{\text{tot,measured}}/L_{\text{tot,predicted}}$ which does not prevent us from estimating $\sigma$ from the shape of the differential length. Incidentally, in 3D, we found that no such correction is required (see below Sec. III C 4).

C. Non-Gaussian critical sets in three dimensions

In 3D, the basis of our consideration is Eq. (22). The expressions for the critical sets are somewhat more complex, while cosmology provides a context in which we can directly predict the value of the relevant cumulants within the framework of gravitational perturbation theory (see Sec. IV B 1).

1. The area of 3D isosurfaces

Let us again start with a 3D Minkowski functional, the surface, $A(\nu)$, of the $x = \nu$ isocontour, which is given by

$$A(\nu) = \int_{-\infty}^{\infty} dx \int_{0}^{\infty} dq^2 P(x, q^2) \delta^3(x - \nu)q = \int_{-\infty}^{\infty} dq^2 P(\nu, q^2)q,$$

where the joint PDF of the field and its gradient, $P(\nu, q^2)$, follows from Eq. (22) after integration over the $J_i$'s:

$$P(x, q^2) = \frac{3\sqrt{3}}{2\pi} \sqrt{q^2} \exp\left(-\frac{x^2}{2} - \frac{3}{2} q^2\right) \times \left(1 + \sum_{n=3}^{\infty} \sum_{i,j} (-3)^j \frac{(3j+1)!}{i!(1+2j)!!}\right) \times \langle x^i q^j \rangle_{\text{GC}} H_j(x) L_j^{(1/2)}(\frac{3}{2} q^2).$$

Performing integration using $\int_{0}^{\infty} \exp(-3/2q^2)q^2 L_j^{(1/2)}(q^2) dq^2 = -2\Gamma(j - \frac{1}{2})/(\sqrt{\pi}\Gamma(j + 1))$ yields

$$\mathcal{A}(\nu) = \frac{2}{\sqrt{3\pi R_0}} e^{-\nu^2/2} \left(1 + \frac{1}{2\sqrt{\pi}} \sum_{n=3}^{\infty} \sum_{i,j} (-3)^j \frac{(3j+1)!}{i!(1+2j)!!}\right) \times \frac{\Gamma(j - \frac{1}{2})}{\Gamma(j + 1)} \langle x^i q^j \rangle_{\text{GC}} H_j(\nu).$$

For instance, to first order in the non-Gaussian expansion, we have

$$\mathcal{A}(\nu) = \frac{2}{\sqrt{3\pi R_0}} e^{-\nu^2/2} \left(1 + \frac{1}{6}(x^3)_{\text{H}}(\nu) + \frac{1}{2}(xq^2)_{\text{H}}(\nu)\right)$$

(62)

Appendix E 2 demonstrates how to reorder expansion (61) as an Edgeworth expansion.

2. The Euler characteristic of non-Gaussian 3D fields

To compute the Euler characteristic in 3D, following the 2D case, one averages $-I_3 = -(J_1^3 - 3J_1 J_2 + 2J_3)/27$ over the full range of the second derivatives:

$$\chi(\nu) = -\frac{1}{R_0^3} \int_{-\infty}^{\infty} dJ_1 \int_{-(\nu + \gamma J_1)/(\sqrt{1 - \gamma^2})}^{\infty} d\xi \int_{0}^{\infty} dJ_2 \int_{-\xi J_2}^{\infty} dJ_3 P_{\text{H}}(\xi, J_1, J_2, J_3)I_3.$$

As shown in Appendix B 2, the resulting expression is

$$\chi(\nu) = \frac{1}{2} \text{Erfc}\left(\frac{\nu}{\sqrt{2}}\right) \chi(-\infty) + \frac{1}{27R_0^3} \left(\frac{3}{2\pi}\right)^{3/2} \exp\left(-\frac{\nu^2}{2}\right) \left[y^3 H_2(\nu) + \sum_{i,k} \frac{(-1)^i}{i!j!} \left(\frac{3}{2}\right)^j \langle x^i q^j J_1^k \rangle_{\text{GC}} (1 - \gamma^2)^j/2\right] \times \sum_{s=0}^{\min(k,k)} \frac{3!j!^{k+3-2s}}{s!(3-s)!(k-s)!} H_{i+k+2-2s}(\nu) \times \sum_{n=3}^{\infty} \sum_{i,j} \frac{(-1)^{k+1}}{i!j!} \left(\frac{3}{2}\right)^j \langle x^i q^j J_1^k J_3^{k+1} \rangle_{\text{GC}} (1 - \gamma^2)^j/2 \times \sum_{s=0}^{\min(k,k)} \frac{3!j!^{k+3-2s}}{s!(3-s)!(k-s)!} H_{i+k+2-2s}(\nu) \times \sum_{n=3}^{\infty} \sum_{i,j,k} \frac{(-1)^{k+1}}{i!j!k!} \left(\frac{3}{2}\right)^j \langle x^i q^j J_1^k J_3^{k+1} \rangle_{\text{GC}} (1 - \gamma^2)^j/2 \times \sum_{s=0}^{\min(k,k)} \frac{3!j!^{k+3-2s}}{s!(3-s)!(k-s)!} H_{i+k+2-2s}(\nu).$$

As in 2D, in the rare event limit, $\nu \gg 1$ or $\nu \ll 1$, Eq. (64) also represents, respectively, the general differential number of maxima or minima. When Eq. (64) is reordered in $\sigma$, the first corrections are (see Appendix E)

$$\chi^{(1)} = \left(\frac{9}{2} \gamma^2 (q^2 J_1) + 9\gamma(x I_2)\right) H_1(\nu) - \left(\frac{3}{2} \gamma^3 (x^2) + \frac{3}{2} \gamma^2 (x^2 J_1)\right) H_3(\nu) + \frac{1}{6} \gamma^3 (x^3) H_3(\nu),$$

(65)

and
\[
\chi^{(2)} = -\left(\frac{27}{2} \gamma(q^2 I_2) + 27 (x I_3)\right) H_0(\nu) + \left(\frac{9}{8} \gamma(x^2 J_1) + \frac{9}{2} \gamma^2 (x q^2 J_1) + \frac{9}{2} \gamma x I_3\right) H_2(\nu)
- \left(\frac{3}{4} \gamma^3 (x^2 q^2) + \frac{1}{2} \gamma^2 (x^2 J_1)\right) H_4(\nu) + \frac{1}{24} \gamma^3 (x^2) H_6(\nu) + \left(\frac{135}{16} \gamma^3 (J_1)^2 + \frac{27}{2} \gamma (q^2 J_1) (x I_2) + \frac{81}{2} \gamma (x^2 J_1) I_3\right) H_8(\nu)
- \left(\frac{45}{4} \gamma^3 (x q^2 J_1) + \frac{9}{2} \gamma (q^2 J_1) (x^2 J_1) + \frac{27}{2} \gamma (x q^2) (x I_2) + \frac{9}{2} \gamma (x^2 J_1) (x I_2) + \frac{9}{2} \gamma (x^3) I_3\right) H_2(\nu)
+ \left(\frac{15}{8} \gamma^3 (x q^2)^2 + \frac{3}{4} \gamma^2 (x^3) (q^2 J_1) + \frac{9}{4} \gamma^2 (x q^2) (x^2 J_1) + \frac{3}{4} \gamma (x^2 J_1)^2 + \frac{3}{2} \gamma (x^3) (x I_2)\right) H_4(\nu)
- \left(\frac{1}{4} \gamma^3 (x q^2 (x^3) + \frac{1}{4} \gamma^2 (x^3) (x^2 J_1)\right) H_6(\nu) + \frac{1}{72} \gamma^3 (x^3)^2 H_8(\nu).
\tag{66}
\]

Figure 5 shows that the first-order expansion provides a good match to the measured 3D Euler characteristic, matching closely the functional form of the non-Gaussian correction, antisymmetric in 3D. Measurements are performed by computing the alternating sum of the counts of critical points.

In cosmological applications, we do not expect our Universe to have complex topology at the large-scale structure scales, if at all, and, thus \(\chi(-\infty) = 0\). This matches exactly \(\langle I_3 \rangle = 0\) from Eq. \((63)\), i.e., the following relation between the moments of the 3D Hessian invariants holds

\[
\langle J_1^3 \rangle - 3 \langle J_1 J_2 \rangle + 2 \langle J_3 \rangle = 0.
\tag{67}
\]

### 3. Non-Gaussian total and differential extrema counts in 3D

The total number of critical points in 3D is given by

\[
n_{\text{ext}} = \frac{1}{R^3} \int_{-\infty}^{\infty} dJ_1 \int_{-\infty}^{\infty} dJ_2 \int_{-\infty}^{\infty} dJ_3 P_{\text{ext}}(\xi, J_1, J_2, J_3) |I_3|.
\tag{68}
\]

For a specific extrema type, sign constraints must be imposed on the sorted eigenvalues \(\lambda_1 \geq \lambda_2 \geq \lambda_3\) which establish the nonoverlapping ranges of integration in \((J_1, J_2, J_3)\) space, defining the maxima \((- - -\)), the minima \((+++)\), and the saddle points of “filamentary” \((+--\) and “pancake” \((++-)\) types. While \(J_2\)
Gaussian maxima and filamentary saddles of the density.

Fig. 7 shows the results of the numerical integration for non-Gaussian fields in 3D for several values of the spectral index (from left to right, red: \( n_s = -2.5 \), green: \( n_s = -2 \), blue: \( n_s = -1 \), purple: \( n_s = 0 \)). Dotted lines are the Gaussian approximation, solid lines are the non-Gaussian prediction for \( \sigma = 0.1 \) computed via Monte Carlo integration. Right: Measured differential distribution of critical points in 3D for a mildly non-Gaussian field from \( n_s = -1 \) simulations at \( \sigma = 0.1 \) versus the first-order non-Gaussian prediction (solid lines) and the Gaussian approximation (dashed lines).

The total number of extrema can be found analytically at all orders of the non-Gaussian expansion. Here, we give the required sign for the eigenvalues. The left panel of Fig. 6, the ranges of integration are shown for each extrema type.

The total number of extrema has changed very insignificantly at \( \sigma = 0.1 \) for maxima and minima and \( \sim 2\% \) for pancake saddles. We note that according to Eq. (69) and the values of the cumulants from Table VI, the total number of extrema has changed very insignificantly at \( \sigma = 0.1 \), by just over \( 2\% \) for maxima and minima and \( \sim 0.07\% \) for the saddle points. This shows that the differential distribution of extrema counts is a much more sensitive tool to probe non-Gaussianities.

4. The differential length of the skeleton for non-Gaussian fields in 3D

In 3D, the length of the skeleton is given in the stiff approximation by

\[
\frac{\partial L_{\text{ske}}}{\partial \nu} = \frac{1}{R_s^2} \int d^3 \tilde{x}_j d^3 x_{kl} P(\nu, \tilde{x}_j, x_{kl}) \delta_3(\tilde{x}_j) \delta_3(\tilde{x}_k) |\Lambda_2 |.
\]
with the constraint \( \lambda_1 + \lambda_2 + \lambda_3 \leq 0 \). As in 2D, in order to compute the differential length of the skeleton we need to separate the component of the gradient aligned with the highest eigenvalue of the Hessian from the others. Let \( Q \) be the normalized norm of the projection of the gradient in the hyperplane orthogonal to this direction:

\[
Q^2 = \frac{q^2 - x_1^2}{(q^2 - x_1^2)}. 
\]  

(71)

In the Gaussian case, we have \( Q^2 = \frac{q^2}{(x_1^2 + x_2^2)} \). The Gaussian PDF for \( q^2 \) is \( \mathcal{G}(q^2) = \frac{3}{2^{3/2} \pi} e^{-3/2 q^2} \) which leads to the PDF for \( x_1 \) and \( Q^2 \):

\[
G(x_1, Q^2) dQ^2 = \sqrt{\frac{3}{2 \pi}} e^{-(3/2)q^2} e^{-Q^2} dQ^2. 
\]  

(72)

After integration over \( x_1 \), the Gram-Charlier expansion of the PDF is thus

\[
P(x, Q^2, J_1, J_2, J_3) = \mathcal{G}(x, Q^2, J_1, J_2, J_3) \left[ 1 + \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(-1)^{i+j} 25}{i! j! k! l!} \left( \xi^i \eta^j \lambda^k \mu^l \right) \mathcal{H}_i \mathcal{H}_j \mathcal{L}_k \mathcal{L}_l \mathcal{H}_{k+1} \mathcal{H}_{j+1} \mathcal{L}_{k+1} \mathcal{L}_{l+1} (Q^2) \right]. 
\]  

(73)

The differential length of the skeleton is given by

\[
\frac{\partial L^{\text{skel}}}{\partial \nu} = \frac{1}{R^2} \int_{-\infty}^{\infty} dJ_1 \int_{-\infty}^{\infty} dJ_2 \int_{-\infty}^{\infty} dJ_3 \frac{1}{\pi} P(\xi, Q^2 = 0, J_1, J_2, J_3) |\lambda_2 \lambda_3|. 
\]  

(74)

The integrand \( |\lambda_2 \lambda_3| \) is a nontrivial combination of the invariant variables: one would have to find the sorted roots of the third-degree characteristic polynomial \( \lambda^3 - \lambda_1 \lambda^2 + \lambda_2 \lambda - \lambda_3 = 0 \). Furthermore, [38] has shown that the differential length of the skeleton does not have an analytical expression in 3D. It is thus easier to use a numerical integration, as was described for the extrema. (Note that we use \( \lambda_1 + \lambda_2 + \lambda_3 \leq 0 \) instead of \( \lambda_1 + \lambda_2 \leq 0 \) that was adopted in [38] as the constraint for defining the skeleton; this choice yields a perfect match to the total length of the skeleton as measured in simulations.)

For \( n_s \in [-2.5, 0] \), the Gaussian and first-order prediction for the differential skeleton length are given in Appendix B. Figure 8 shows the variation of a non-Gaussian 3D skeleton with the spectral index, and illustrates the match between the predicted and the measured skeleton. The match is found to be significantly improved as compared to the 2D case.

FIG. 8 (color online). Left: differential length of the skeleton in 3D for several values of the spectral index (from left to right, red: \( n_s = -2.5 \), green: \( n_s = -2 \), blue: \( n_s = -1 \), purple: \( n_s = 0 \)). The dotted lines are the Gaussian case, the solid lines are the non-Gaussian prediction for \( \sigma = 0.1 \). Right: measured differential length of the 3D skeleton for a mildly non-Gaussian, \( \sigma = 0.1 \), field (blue band), versus first-order non-Gaussian prediction (solid line) and Gaussian approximation (dashed line).
IV. APPLICATIONS: CMB PRIMORDIAL NON-GAUSSIANITIES, DARK ENERGY IN THE LSS

Having developed description of non-Gaussian geometrical statistics [expressed in 2D by Eq. (34) for the Euler characteristic, (40) for extrema, (56) for the skeleton, and in 3D by, respectively, Eqs. (64), (69), and (74)], let us now illustrate how these statistics can be used to address problems arising in cosmology. The first example uses 2D extrema counts to estimate primordial non-Gaussianities in synthetic CMB maps, while the second example estimates the density dispersion, $\sigma(z)$ of dark matter simulations from differential extrema counts, the Euler characteristic, and skeleton length as a function of redshift. Note that alternative illustrations can be found in [15,43], including models involving second-order corrections.

A. Primordial non-Gaussianities in the cosmic microwave background

Let us generate sets of parameterized non-Gaussian maps using the package SKY-NG-SIM [59] of HEALPIX. In this so-called harmonic model, the PDF, $f_T(T)$ of the pixels’ temperature, $T$, is given by

$$f_T(T) = \exp\left(-\frac{T^2}{2\sigma_0^2}\right)\left|\sum_{i=0}^{n} \alpha_i c_i H\left(\frac{T}{\sigma_0}\right)\right|^2,$$

where the $c_i$ are normalization constants. In this paper, we use $\text{nside} = 2048$, $\ell_{\text{max}} = 4096$, $n = 2$, $\sigma_0 = 1$, and vary $\alpha_1$ and $\alpha_2$ on a $50 \times 50$ polar grid in the range $[0.1, 0.8]$ subject to $\alpha_1^2 + \alpha_2^2 < 1$. For each set of maps, we compute its derivatives, and arithmetically average the corresponding cumulants, using the code MAP2CUM [15] for which the invariant variables $J_1$ and $J_2$ on a sphere are defined via the mixed tensor of covariant derivatives $J_1 = x_{ij}j^i$ and $J_2 = J_1^2 - 4|x_{ij}j^i|^2$. The differential counts are then evaluated for a range of threshold, $\nu \in [-5, 5]$. For our reference map, $\alpha_1 = 0.6$, the number of extrema is computed by the procedure HOTSPOT [60]. Figure 9 illustrates our ability to recover input model from a given set of maxima and minima extracted from a given realization of the sky. Here, the residual departure from the Gaussian number count is shown. Indeed, the maximum likelihood solution, solid lines, corresponds to a very good match to the measured residual differential number counts. Note that this class of model has an interesting feature that $\langle x^3 \rangle$ is significantly smaller than say, $\langle J_1 \rangle$. Indeed, for this model, $\langle x^3 \rangle = -0.076, \langle x^2 J_1 \rangle = 0.072, \langle q^2 x \rangle = -0.091, \langle x J_1^3 \rangle = -0.15, \langle x J_2 \rangle = -0.087, \langle q^2 J_1 \rangle = 0.054, \langle J_1^3 \rangle = 0.31, \langle J_1 J_2 \rangle = 0.023$. This simple-minded experiment should be improved to account for, e.g., more realistic and physically motivated non-Gaussian maps (such as $f_{\text{gal}}$ models), realistic galactic cuts and hit maps, variable instrumental response. The main challenge is to model incompleteness in observed samples in order to weight accordingly the range of relevant thresholds.

B. Constraining the dark energy equation of state via the geometry of the LSS

At supercluster scales where nonlinearity is mild, we can expect the non-Gaussianity of the matter density to be also mild, but still essential for a quantitative understanding of the filamentary cosmic web in between the galaxy clusters. Indeed, e.g., Fig. 7 shows that the interplay of gravity and (possibly accelerated) expansion will induce some level of distortion in the PDF of the differential counts.

![Figure 9](image_url)

FIG. 9 (color online). Middle panel: the predicted (solid line) excess number of maxima (bottom), and minima (top) in $\Delta \nu = 0.25$ bins as a function of the threshold, $\nu$, on top of the measured count (dashed line) from a single realization full sky NSIDE = 1024 HEALPIX map (histogram). The temperature field is smoothed with the Gaussian filter of 10 arcmin FWHM, resulting in $R_c \approx 11$ arcmin $\approx 3$ pixels. Left and right panels: likelihood contours in the parameter space of the Harmonic Oscillator model of non-Gaussianity given by Eq. (83); the input model corresponds to $\alpha_1 = 0.6, \alpha_2 = 0.6$, represented by a green dot in the likelihood map. The five contours correspond to 1,2,3,4, and 5 sigma likelihood contours, while the blue and red dots mark the most likely estimate for maxima and minima, respectively. Hence, different models of non-Gaussianity can be well distinguished by their effects on differential extrema counts.
Importantly, this distortion is not sensitive to the overall number count as the PDF are normalized. Fitting measured counts to the distorted theoretical expectation should therefore allow us to robustly estimate some features of the underlying dark energy equation of state.

1. Extension of the gravitational perturbation theory to critical set cumulants

When the field is the 3D density field of the large-scale structure of the Universe and for Gaussian initial conditions, the non-Gaussianities appear because of the nonlinearities of gravitational clustering. As mentioned in the introduction, Eqs. (64), (69), and (74) have in common that they can be rearranged as

$$
E_R(\nu, z) = \sum_n C_n(\nu, R) \sigma^2_0(R) D(z)^n,
$$

(76)

where $\sigma^2_0$ is today’s variance of the cosmic density field at the corresponding smoothing scale, $R$, and $D(z)$ is the growth factor as a function of redshift. For instance, for the Euler characteristic, $C_1 \sigma \sigma D$ is given by Eq. (65) while $C_2 \sigma^2 \sigma D^2$ is given by Eq. (66). When gravity is generating mildly non-Gaussianities, we will show that it is possible to predict the emerging statistical properties of the field using a perturbative theory. As we will demonstrate below, all the cumulants involved in the Edgeworth expansion entering the $C_n$’s may then be predicted from first principle. While the geometrical analysis of catalogs, say using codes like DISPERSE [61], allows us to measure $E_R(\nu, \nu_f)$, and provided that (i) we consider large enough scales so that Eq. (76) converges, and (ii) that gravity is the driving cause of non-Gaussianity, Eq. (76) can be inverted for $D(z)$. In turn, measuring $D(z)$ below $z \sim 1$ probes the dark energy equation of state. One of the difficulties in implementing our statistics to observed surveys is that $\sigma$ is not known, hence we cannot normalize $\nu \equiv \rho/\sigma$. Following [62], a work-around is to introduce instead of $\nu$ the density contrast $\nu_f$, corresponding to a given filling factor, $f$, and related by

$$
f = \int_0^\infty d\nu P(x) = \int_{\nu_f}^\infty d\nu G(x).
$$

(77)

Here, $f(\nu)$ is given by Eq. (27) and $G$ stands for the corresponding Gaussian PDF, so that $E_R(\nu, \nu_f, z)$ is a measurable quantity encoding the departure from Gaussianity which is not already encoded in $P(\nu)$. In practice, we will not make use of Eq. (77) below and we will assume perfect knowledge of $\nu$ in these experiments.

2. Prediction of the 3rd-order cumulants in perturbation theory

In the formalism of perturbation theory [40], it is easy to see that the $p$th cumulant of the field is of order $\sigma^{p-2}$, if the field is normalized to a constant variance. Using this prescription, one can resum the Gram-Charlier expansion to the so-called Edgeworth expansion. The Edgeworth expansion is an expansion in $\sigma$ and its order is thus physically motivated. In practice, it means its convergence properties can be justified on physical grounds, that it can thus be truncated more easily [47]. Appendix D shows that the $\sigma$ term from the Edgeworth expansion only comes from the third-order term of the Gram-Charlier expansion, while the $\sigma^2$ term comes from both the fourth- and part of the sixth-order terms. We will therefore restrict our predictions of the involved cumulants to third order. At this order, moments, cumulants, and Gram-Charlier coefficient coincide. For the next order in $\sigma$, one would have to extend the predictions to quartic cumulants and to decompose sixth-order Gram-Charlier coefficients in cumulants (see Appendix E). For third-order terms, the calculation of all the cumulants is similar to the well-known prediction of the skewness [40,63–65]. We will only consider an Einstein-de Sitter (EdS) universe since other cosmologies can be described as corrections to this prediction.

For a three-dimensional field, the PT cumulants involving the field and its derivatives can be written as

$$
C_R = \left\langle \left( \partial^a_{\nu_1} \partial^a_{\nu_2} \partial^a_{\gamma_{\nu_3}} \delta_R \right) \left( \partial^b_{\nu_1} \partial^b_{\nu_2} \partial^b_{\gamma_{\nu_3}} \delta_R \right) \left( \partial^c_{\gamma_{\nu_1}} \partial^c_{\gamma_{\nu_2}} \partial^c_{\gamma_{\nu_3}} \delta_R \right) \right\rangle,
$$

where $\partial^a_{\nu}$ is the $n$th derivative with respect to the $n$th coordinate, and $\delta_R$ is the field smoothed on scale $R$. Using gravitational perturbation theory, we show in Appendix F 1 that $C_R$ obeys

$$
C_R = 2 \int d^3 k_1 d^3 k_2 [F_{a, b, \gamma}(k_1, k_2) + F_{b, a, \gamma}(k_1, k_2)] P(k_1) P(k_2) W(k_1 R) \times \frac{W(k_2 R)}{W(k_1 + k_2 R)},
$$

(78)

where the so-called generalized geometric factors $F_{a, b, \gamma}(k_1, k_2)$ [here $\alpha$ is the shorthand for the triplet $\alpha_1, \alpha_2, \alpha_3$, etc.], read

$$
F_{a, b, \gamma}(k_1, k_2) = F_2(k_1, k_2) G(k_1, k_2) \gamma_{\alpha_1 + \alpha_2 + \alpha_3} \beta_1 + \beta_2 + \beta_3 \times (-1)^{\gamma_1 + \gamma_2 + \gamma_3},
$$

(79)

with

$$
F_2(k_1, k_2) = \frac{5}{7} + \frac{k_1 \cdot k_2}{k_1^2} + 2 \left( \frac{k_1 \cdot k_2}{k_1^2} \right)^2,
$$

$$
G(k_1, k_2) = \left( k_1^{(1)} \right)^{\alpha_1} \left( k_1^{(2)} \right)^{\alpha_2} \left( k_1^{(3)} \right)^{\alpha_3} \left( k_2^{(1)} \right)^{\beta_1} \left( k_2^{(2)} \right)^{\beta_2} \times \left( k_2^{(3)} \right)^{\beta_3} \left( k_1^{(1)} + k_1^{(2)} \right)^{\gamma_1} \left( k_1^{(2)} + k_2^{(2)} \right)^{\gamma_2} \times \left( k_1^{(3)} + k_2^{(3)} \right)^{\gamma_3}.
$$

(80)

The smoothing of the field over the scale $R$ is taken care of via the window function, $W(k R)$. As explained in Appendix F 1 for a Gaussian filtering and a scale-invariant power spectrum, these integrals can be completed and the resulting cumulants are given explicitly in terms of the smoothing length and the power index in Appendix F 2.
instance, for the \( n_s = -3 \) power-law index, we extend the classical result

\[
\langle x^3 \rangle = \frac{34}{7},
\]

with

\[
\langle x x_2^2 \rangle = \frac{34}{7} \frac{2}{3 \cdot 5!!}, \quad \langle x x_4 x_2 \rangle = \frac{34}{7} \frac{2}{3 \cdot 5!!},
\]

\[
\langle x x_4^2 \rangle = \frac{34}{7} \frac{2}{3 \cdot 5!!}. \quad \langle x x_6 \rangle = \frac{34}{7} \frac{2}{3^2}.
\]

In Sec. III B, we also use the perturbation theory to predict 2D statistics. Since our 2D fields are the slices of the gravitationally evolved 3D density field, the cumulants in field variables that involve only the derivatives restricted to the slice are the same as predicted by the 3D perturbations theory and are given in Table VI. However, to apply to 2D slices, one needs to combine these cumulants of the field variables into 2D invariants.

3. Comparison to measurements of the critical set cumulants in Monte Carlo simulations

Measurements are carried on sets of simulations produced using GADGET-2 [66] code and scale-invariant initial conditions generated with MPGRAFIC [67]. We use 256 \(^3\) particles and 10, 25, and 10 realizations for, respectively, \( n_s = 0, -1, \) and \( -2 \). To be consistent with the perturbation theory used, we set an Einstein-de Sitter cosmology \((\Omega_m = 1, \Omega_{\Lambda} = 0)\). The 2D results are obtained by cutting these 3D simulations into 25 slices which are averaged along the shortest dimension to give 2D fields. Following [68], we benefit from the scale invariance of our simulations to combine them. For each simulation, we compute the density field with a cloud-in-cell procedure and we smooth on lengths of \( R = 5, 10, 15, 20, \) and 25 pixels.

To eliminate the snapshots contaminated by initial transients or incorrect large-scale mode couplings, we compute the correlation length \( l_0 \), defined by \( \sigma^2(l_0) = 1 \). We only keep the snapshots verifying [69]

\[
\frac{L}{N^{1/3}} \ll l_0 \ll L,
\]

where \( L \) is the size of the box and \( N \) the number of particles. The first condition ensures that the grid effects have relaxed, while the second condition limits the coupling of the modes modified by the finite size of the box. In practice, we choose

\[
2 \frac{L}{N^{1/3}} < l_0 < \frac{1}{20} L.
\]

Even within a selected snapshot, all the scales cannot be trusted. The smoothing length has to be large to probe scales for which the discreteness of the simulation is not important, but small enough compared to the size of the box. A detailed study shows that the smallest scale containing information is not given by the initial interparticle distance, as one could naïvely think. Indeed, the resolution is improved in clustered regions, where the density is higher than the density of the initial conditions.

The pertinent characteristic length can be computed using the quantity \( N_c(R) = \bar{n} \xi(R) \), where \( R \) is the smoothing length, \( \bar{n} \) the average number of particles per cell, and \( \xi \) the correlation function. \( N_c \) characterizes the average number of neighbors [70]. The associated characteristic length is then \( l_c \), such as \( N_c(l_c) = 1 \). We keep the smoothing length for which [69]

\[
l_c \ll R \ll L.
\]

In practice, we choose \( 2l_c < R < L/20 \). The rotational invariants of the field are computed for each field (using Fourier transforms to estimate the derivatives), and are averaged over realizations once the relevant cumulants are generated. The results for the 3rd-order moments of the field are given in Table VI and are compared to predictions from first-order perturbation theory. Table II reexpresses them in terms of invariants and Table III gives the 4th-order cumulants.

![Table II](image)

**TABLE II.** Predicted and measured 3D third-order invariant cumulants for three values of the power-spectrum index. The method to estimate these cumulants in simulations is described in Sec. IV B 3. The predictions follow from the coordinate cumulants computed in Table V1 given in Appendix F 2. According to perturbation theory, the dimensionless third-order cumulants are proportional to \( \sigma \).

<table>
<thead>
<tr>
<th>( n_s = 0 )</th>
<th>( n_s = -1 )</th>
<th>( n_s = -2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>prediction</td>
<td>measurement</td>
<td>prediction</td>
</tr>
<tr>
<td>( \langle x^3 \rangle / \sigma )</td>
<td>3.144</td>
<td>3.08 ± 0.08</td>
</tr>
<tr>
<td>( \langle x q^2 \rangle / \sigma )</td>
<td>2.096</td>
<td>2.05 ± 0.03</td>
</tr>
<tr>
<td>( \langle x^2 J_1 \rangle / \sigma )</td>
<td>-3.248</td>
<td>-3.15 ± 0.06</td>
</tr>
<tr>
<td>( \langle x J_1 \rangle / \sigma )</td>
<td>3.871</td>
<td>3.75 ± 0.06</td>
</tr>
<tr>
<td>( \langle x J_2 \rangle / \sigma )</td>
<td>1.545</td>
<td>1.54 ± 0.02</td>
</tr>
<tr>
<td>( \langle q^2 J_1 \rangle / \sigma )</td>
<td>-1.335</td>
<td>-1.28 ± 0.02</td>
</tr>
<tr>
<td>( \langle J_1 \rangle / \sigma )</td>
<td>-4.644</td>
<td>-4.50 ± 0.08</td>
</tr>
<tr>
<td>( \langle J_1 J_2 \rangle / \sigma )</td>
<td>-0.679</td>
<td>-0.65 ± 0.01</td>
</tr>
<tr>
<td>( \langle J_3 \rangle / \sigma )</td>
<td>1.304</td>
<td>1.28 ± 0.03</td>
</tr>
</tbody>
</table>
4. Growth factor measurement experiment: Toward a possible dark energy probe?

To demonstrate the potential of our non-Gaussian critical set statistical estimates to map the evolution of the growth factor $D(z)$ [Eqs. (64), (69), and (74)], let us discuss a very idealized topological dark energy probe experiment. For this purpose, we generate a cosmic sequence of a set of 25 $n_s = -1$ scale-invariant simulations with 256$^3$ particle smoothed over $R = 10$ and 20 pixels and measure for each redshift the corresponding differential number of extrema, $\xi_k^\mathrm{ext}(\nu, z)$ (respectively, min, max, filament-saddle, and pancake saddle), Euler characteristic $\xi_k^\mathrm{Euler}(\nu, z)$, and skeleton length, $\xi_k^\mathrm{ske}(\nu, z)$. We denote the set of such measurements as $\xi_k^\mathrm{Euler}(\nu, z)\rangle_k$, where $k$ stands for the type of statistics and $R$ the smoothing scale [cf. Eq. (76)]. We then match these histograms to the predictions of the first-order Gram-Charlier PT model and deduce the best fit $\hat{\sigma}$ for a range of redshifts, using for now a least-squares fit with uniform weights. This estimated “geometric” $\hat{\sigma}$ is then plotted as a function of underlying $\sigma$ for the six critical set statistics (4 types of extrema, Euler characteristic, and skeleton length$^9$) and shown in Fig. 10. On average, $\hat{\sigma}$ is a good match to $\sigma$ (when $\sigma \approx 0.2$), with a dispersion $\sigma^2$ which scales with the smoothing length $R$ and the size of the survey $L$ as

\[ \sigma^2 = \sigma^2_k[(L_c/R_c)(R/L)]^2, \]

where the values $\sigma^2$ that were measured in our numerical survey ($L_c = 256$ pixels) are given in Table IV for two values of the smoothing scale, $R_c = 20$ and $R_c = 10$ pixels. The deviation at larger $\sigma$ seen on Fig. 10 could be explained as an effect of the higher-order non-Gaussianities and should be reduced by going beyond a first-order description. This is corroborated by the observations that the critical sets which are least sensitive to the higher and, presumably, more nonlinear thresholds (e.g., minima and saddle points) tend to depart from the non-Gaussian linear correction later. On the other hand, our statistics are less sensitive at very low $\sigma$’s due to the universal behavior of the geometrical descriptors in the Gaussian limit. Hence, the optimal level for the recovery of the variance $\hat{\sigma}$ is around $\sigma = 0.1$, according to Fig. 10. Thus, varying the smoothing scale with redshift, $R(z)$, in such a way as to ensure a level of nonlinearity near $\sigma(z) = 0.1$ may help the reconstruction of $D(z)$. Note finally and importantly that the overall spread in the measured $\hat{\sigma}$ reflects cosmic variance and could be reduced in proportion to the square root of the volume of the survey, following Eq. (81). For instance, a naive extrapolation of the cluster CFHT-LS wide counts$^{10}$ [62] to a full sky experiment with 40 bins of redshift below $z = 1.15$ would yield a noise-to-signal ratio of 1% per bin for $D(z)$ using the Euler characteristic alone.

Let us now briefly discuss how these very idealized measurements can be extrapolated to anticipate the accuracy of a dark energy experiment based on critical sets. A dark energy probe may directly attempt to estimate the so-called equation of state parameters $(\omega, w_0)$ from these critical sets while relying on the cosmic model for the growth rate$^{72}$,

\[ D(z|w_0, w_a) = \frac{5\Omega_m H_0^2}{2} \frac{H(a)}{a^3 H^3(a')} \int_0^a \frac{da'}{a^3} H^3(a'), \]

\[ H^2(a) = H_0^2 \left[ \frac{\Omega_m}{a^3} + \Omega_\Lambda \exp \left( 3 \int_0^a \frac{1 + w(z')} {1 + z'} dz' \right) \right], \]

with $\Omega_m$, $\Omega_\Lambda$, and $H_0$, respectively, the dark matter and dark energy densities and the Hubble constant at redshift $z = 0$, $a = 1/(1 + z)$, the expansion factor, and the equation of state $w(z) = w_0 + w_za/(1 + z)$. Given Eqs. (76) and (82), optimizing the probability of observing all counts at all redshifts with respect to $(\omega, w_0)$ yields a maximum likelihood estimate for the dark energy equation of state parameters$^{11}$:

$^{9}$The skeleton length is only fitted on the range $\nu < 0$ or $\nu > 2$ where the agreement between predictions and measurements is the best.

$^{10}$http://www.astromatic.net/iip/w1.html

$^{11}$Here, we assume for simplicity that the different counts are uncorrelated; an improvement would be to use multinomial statistics.
\[
(\hat{w}_0, \hat{w}_a) = \arg\max_{w_0, w_a} \left\{ \prod_{z, i} \text{Poisson}(E_k^i) \times [\nu, D(z|w_a, w_0), E^k_{z,i}] \right\},
\]

where the predicted “number counts,” \(E^k_{\nu}(\nu, D(z|w_a, w_0))\), are given by Eq. (76), given Eq. (82).\(^{12}\)

Here, \(E^k_{z,i}\) is the measured \(K\) statistics in the threshold bin \(i\) at redshift \(z\), and Poisson \((\mu, x)\) is the PDF of a Poisson process, \(x\), of mean \(\mu\). A joint analysis of all critical sets, \(\{E^k(\nu, z)\}_k\), is achieved via the product on \(k\) (assuming somewhat optimistically that all estimators are independent).

\(^{12}\)As the Euler characteristic is cumulative, the model needs to be differentiated since Eq. (83) assumes that the counts are independent. In other words, \(\kappa_{Euler}(\nu, z) = \partial_\nu \partial_\lambda\) given by Eq. (41).

We shall not carry such a maximization in full here. Instead, Fig. 11 displays (very naive) likelihood contours for an experiment corresponding to the CFHT-LS extrapolation \([62]\) of the cluster counts to a \(1/4\) sky experiment \([\text{2\% error on } D(a)\text{ for 40 bins, which corresponds to the } 256^3, n_s = -1\text{ scale-invariant simulation smoothed over } R = 20\text{ pixels (which corresponds to approximately 230 maxima) for the larger and lighter contours and } R = 10\text{ pixels (1800 maxima) for the small and darker ones. The skeleton is smoothed only for } R = 20\text{ to ensure sufficient smoothness. The shaded area corresponds to 1 standard deviation (computed from 25 simulations). Note that the critical sets which are most sensitive to the higher thresholds tend to depart from the non-Gaussian linear correction earlier, which is expected in the context of gravitational clustering.}\]

<table>
<thead>
<tr>
<th>(R_c = 20)</th>
<th>(R_c = 10)</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\sigma_{max}^c)</td>
<td>(-0.030)</td>
</tr>
<tr>
<td>(\sigma_{min}^c)</td>
<td>(-0.034)</td>
</tr>
<tr>
<td>(\sigma_{sad-fil}^c)</td>
<td>(-0.049)</td>
</tr>
<tr>
<td>(\sigma_{sad-pan}^c)</td>
<td>(-0.064)</td>
</tr>
<tr>
<td>(\sigma_{skel}^c)</td>
<td>(-0.033)</td>
</tr>
<tr>
<td>(\sigma_{Euler}^c)</td>
<td>(-0.027)</td>
</tr>
</tbody>
</table>
appropriate rescaling of Eq. (81) given the prefactors of Table IV]. Here, we simply assume that the fit to the critical sets yields access to a noisy \( D(a | w_0, w_a) \) via Eq. (82) (for a Gaussian noise with a signal-to-noise ratio of 50 = 1/0.02); for a grid of models obeying Eq. (82) we therefore compute the likelihood of a given draw. The corresponding 3-sigma contours suggest that critical sets could constrain \( w_0 \) to \( \sim 5\% \) and \( w_a \) to \( \sim 10\% \).

Beyond the obvious oversimplifications of this model, here are a few limitations/caveats which need to be discussed: (i) The accuracy of the predictions, \( E^k(v, z) \), is as good as PT; it therefore requires a galaxy sample large enough to probe the larger scales, so that \( \sigma = 0.2 \). On scales of the order of 30 Mpc/h, given that \( \sigma(R, z) = \sigma_0(R) D(z) \), with \( \sigma_0(R) \) the rms of the cosmic density field at redshift zero smoothed over \( R \), if we rely on the estimates for \( \sigma \) given by Eq. (81), we may expect to probe \( D(z) \) near \( z \sim 0.5^{13} \); (ii) For scale-invariant power spectra, the cumulants entering Eqs. (64), (69), and (74) and the shape parameter, \( \gamma \), depend on the power-law index \( n_s \) (see Appendix F 2). For realistic \( \Lambda \)CDM cosmologies, following [40] one strategy could be to let the index \( n_s \) be a (redshift-dependent) parameter to be also estimated while fitting the counts\(^ {14} \); (iii) One could also estimate the optimal weighting strategy to best constrain \( (w_0, w_a) \) using combinations of critical set estimators, varying threshold and redshifts weights. This is achieved by replacing Poisson \( (\xi_k^R(v, z), \xi_k(v, z)) \) in Eq. (83) by Poisson \( (w_{i,k}^R \xi_k^R(v, z), w_{i,k}^R \xi_k^R(v, z)) \) and minimizing the error on the estimated dark energy parameters w.r.t. the positive weights \( w_{i,k}^R \); (iv) In this paper, we ignored all issues regarding completeness, masking, etc. (see Sec. V). In closing, note that the 2D statistics, Eqs. (34), (40), and (56), could also be implemented on nonlinear projected data, given some dynamical model for the projected cumulants, e.g., in the context of weak lensing (see, e.g., [42,48,73,74]).

Carrying out the road map sketched in this section while addressing these issues could be one of the targets of the upcoming surveys that have been planned specifically to probe dark energy, either from ground-based facilities (e.g., VST-KIDS, DES, Pan-STARRS, LSST\(^ {15} \)) or space-based observatories (EUCLID, SNAP, and JDEM\(^ {16} \)).

In this paper, we have chosen to focus mainly on the non-Gaussianities arising from gravitational instability. It should be noted however that our formalism is more general. The Gram-Charlier expansion is valid for any mildly non-Gaussian field, while perturbation theory only enters the prediction of the cumulants. An important example of a situation where the cumulants can be computed independently is the \( f_{\text{NL}} \) parameterization of non-Gaussianities in the CMB. In this parameterization, cumulants can be computed and our formalism can be adapted to give predictions of the evolution of the properties of the field with the value of \( f_{\text{NL}} \). For example, [75] recently used the skeleton to constrain \( f_{\text{NL}} \) on two-dimensional

\[^{13}\text{Note that the invariants moments } q^2, J_1, \text{ or } J_2 \text{ involve higher-order derivatives, and are therefore sensitive to smaller, less linear scales.}

\[^{14}\text{Alternatively, cumulants like Eq. (F11) can be computed numerically for realistic power spectra.}


weak-lensing maps. Their approach was numerical and our formalism could provide the theoretical framework for this type of investigation.

V. CONCLUSION AND DISCUSSION

We have introduced the Gram-Charlier expansion to describe the point probability distribution of a field and its derivatives [Eqs. (11) and (22)] in the mildly non-Gaussian regime in two and three dimensions. We have derived from these 2D and 3D PDF the non-Gaussian expressions for the Euler characteristic, Eqs. (34) and (64), for differential extrema counts, Eqs. (40) and (69), and for the skeleton differential length Eqs. (56) and (74). For the Minkowski functionals, including the Euler characteristic (and therefore the rare event tail of the differential number counts) the expansion is valid to arbitrary order in Gram-Charlier and is also given explicitly to third order in rms of the density field. All differential distributions were checked against Monte Carlo realizations of the field and its critical sets: excellent agreement was found for extrema counts (and therefore for the Euler characteristic) for dispersion up to $\sim 0.2$. For the skeleton, some level of discrepancies still arises for some intermediate contrasts (see Figs. 4 and 8). The origin of this disagreement may originate either from the shift approximation behind the theory, or residual biases in the measurements. As argued in Sec. IV B 4, the range of contrasts corresponding to this mismatch may in practice be masked out when fitting observed differential lengths.

Furthermore, using gravitational clustering perturbation theory, we have shown how to predict all the 3D cumulants that appear in the Gram-Charlier expansion from nonlinear gravitational dynamics. We can therefore predict the evolution of quantities such as the Euler characteristic, the density of extrema, and the differential length of the skeleton as a function of rms of the cosmic density field. The evolution of these critical sets can thus be related to the evolution of the underlying dark matter field. This connection is achieved via the gravitational distortion of quantities which can be measured independently of any monotonic bias. We have sketched a possible dark energy experiment based on the statistical analysis of these critical sets in order to constrain its equation of state, which in the very idealized regime we investigated, seems promising.

Possible extension of this work beyond the scope of this paper involves (i) implementing the dark energy estimation of Sec. IV B 4 first on realistic mock catalogs extracted from large-scale simulations such as [76], and eventually on the above-cited surveys, while accounting for incomplete, magnitude-limited samples and modeling the corresponding biases; one remaining task would be to demonstrate that the statistical analysis of geometrical critical sets can be competitive in practice to constrain this equation of state, compared to, e.g., weak lensing [77–80] or Baryonic acoustic oscillations probes [41]; (ii) deriving the invariant JPDF for higher-order derivatives and/or antiderivatives (e.g., to link geometry to the kinematic of the flow along filaments), or for fields of higher dimensions (e.g., to study the saddle points of high dimensions landscapes), following the tracks presented in [38]; (iii) extending the theory of critical sets to walls and manifolds of higher dimensions; (iv) exploring the implication of the departure from a Gaussian JPDF on other statistics, such as the mean connectivity of the peaks of the cosmic web, following [81]; (v) implementing second-order PT, or the so-called renormalized perturbation theory [82] in order to extend the domain of applicability of these asymptotic expansions, and possibly build a generalized stable clustering model for our cumulants; (vi) extending the formalism to N-points statistics, e.g., to investigate the non-Gaussian correlation of peaks, or non-Gaussian count-in-cells.

ACKNOWLEDGMENTS

We thank S. Colombi, V. Desjacques, D. Weinberg, F. Bernardeau, S. Prunet, and K. Benabed for comments, T. Soubie for providing us with his skeleton code, and D. Munro for freely distributing his Yorick programming language and OpenGL interface (available at http://yorick.sourceforge.net/). C. G. and C. P. thank University of Alberta for hospitality in the spring of 2010 and in the spring–summer of 2011. Special thanks to L. Vigroux and the Institut d’Astrophysique for funding some critical stages of this work. D.P. thanks the University of Oxford, the France-Canada Mobility fund, and the CNRS (France) for support when this investigation was initiated within the framework of the Horizon project, http://www.projet-horizon.fr. We warmly thank S. Prunet and S. Rouberol for running the Horizon cluster for us. C. P. acknowledges support from the physics department of the University of Oxford, and thanks Merton College, Oxford.

APPENDIX A: DERIVATIONS OF THE EULER CHARACTERISTIC IN 2 AND 3D

1. Euler characteristic in 2D

For a two-dimensional random field, the Euler characteristic $\chi(\nu)$ of the excursion set $x > \nu$ is given by

$$
\chi(\nu) = \frac{1}{R^2} \int_{-\infty}^{\infty} dJ_1 \int_{(\nu + \gamma J_1)/\sqrt{1-\gamma^2}}^{\infty} d\xi \times \int_0^{\infty} dJ_2 P_{\text{ext}}(\xi, J_1, J_2) I_2,
$$

where $P_{\text{ext}}$ is the distribution function under the condition of zero gradient:

$$
P_{\text{ext}}(\xi, J_1, J_2) = \frac{1}{\pi} \int_0^{\infty} dq^2 P(\xi, q^2, J_1, J_2) \delta_D(q^2).$$

Using polar coordinates $P_{\text{ext}}(\xi, J_1, J_2) = \frac{1}{2} P(\xi, q^2 = 0, J_1, J_2)$, and since $L_j(0) = 1$ for any $j$, we have

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\[ P_{\text{ex}}(\zeta, J_1, J_2) = \frac{1}{\pi} G_{2D}(\zeta, J_1, J_2) \left[ 1 + \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1) L_l(J_2)}{i! j! k! l!} \right]. \]

(A3)

Remarkably, the Euler characteristic can be evaluated completely for this general distribution function. For this, we note that \( J_2 = \frac{1}{2}(J_1^2 - J_2) = \frac{1}{2}(H_1(J_1) + L_1(J_2)) \), thus, the calculations are reduced to the integration of products of Hermite and Laguerre polynomials. As the first step, the integral over \( J_2 \) leaves only the terms with \( L_0(J_2) \) (\( l = 0 \)) or \( L_1(J_2) \) (\( l = 1 \)) in the Gram-Charlier expansion:

\[ \chi(\nu) = \frac{1}{4\pi R^2} \int_{-\infty}^{\infty} dJ_1 \int_{(\nu + J_1)/\sqrt{1 - \gamma^2}}^{\infty} d\zeta G(\zeta, J_1) \]
\[ \times \left[ H_2(J_1) + H_2(J_1) \sum_{n=3}^{\infty} \sum_{i,j,k,l} \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1) L_k(J_2)}{i! j! k! l!} \right. \]
\[ \times \left[ \langle \zeta^2 J_1^2 \rangle_{\zeta} H_i(\zeta) H_k(J_1) L_l(J_2) \right] \]
\[ \times \frac{\langle \zeta^2 J_1^2 \rangle_{\zeta} H_i(\zeta) H_k(J_1) L_l(J_2)}{l! j! k!} \left. \right] \]

(A4)

Let us first establish the important relation between some coefficients of the expansion for an isotropic field by evaluating \( \chi(-\infty) \). Integrating over full range of \( \zeta \) selects \( i = 0 \) terms,

\[ \chi(-\infty) = \frac{1}{4\pi R^2} \int_{-\infty}^{\infty} dJ_1 G(J_1) \left[ H_2(J_1) + H_2(J_1) \right. \]
\[ \times \left. \sum_{n=3}^{\infty} \sum_{j,k} \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j! k!} \right] \]
\[ \times \sum_{n=3}^{\infty} \sum_{j,k} \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j! k!} \left. \right] \]

(A5)

and subsequent evaluation retains only \( j = n/2 - 1 \) in each even order of the expansion

\[ \chi(-\infty) = \frac{1}{4\pi R^2} \sum_{j=1}^{\infty} \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j!} \]
\[ \times \frac{(\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j!} \left. \right] \]

(A6)

The limiting Euler number \( \chi(-\infty) \) is determined by the topology of the manifold which the sum of the cumulants in Eqs. (A6) must obey. We note that in this derivation we have obtained only the non-Gaussian contribution to \( \chi(-\infty) \) which comes via hierarchy of the terms \( \langle q^2 J_2 \rangle_{GC} \),

\[ j \geq 1. \]

The Gaussian term of the same structure corresponds to \( j = 0 \) and was set to zero by our normalization \( \langle J_2 \rangle = \langle J_1^2 \rangle - J_2 = 0 \). It is an open question whether this is strictly consistent only with \( \chi(-\infty) = 0 \) as, e.g., in an infinite Euclidean space, but requires more accurate treatment on the 2D sphere where \( \chi(-\infty) = 2 \). For now, we will leave the boundary term \( \chi(-\infty) \) included in our expressions but unspecified. This is also justified by the practical case of the boundary topology introduced by a mask on the data. In this case, the moments of the field remain the same as on unmasked manifold, but \( \chi(-\infty) \) measured in realizations will nevertheless reflect the topology of the mask, being now disentangled from moment statistics.

Returning to the differential case, we use the following property of the Hermite polynomials:

\[ \int_{-\infty}^{\infty} d\zeta G(\zeta) H_k(\zeta) = \begin{cases} \frac{1}{2} \text{Erfc} \left( \frac{\nu}{\sqrt{2}} \right), & i = 0, \\ \frac{1}{\sqrt{2\pi}} \exp \left( -\frac{\nu^2}{2} \right) H_{i-1}(a), & i \geq 1. \end{cases} \]

(A7)

Let us first treat the \( i = 0 \) case. Expanding the appearing product of Hermite polynomials,

\[ H_p(J_1) H_k(J_1) = \sum_{s=0}^{\min(p,k)} \frac{p! k!}{s! (p-s)! (k-s)!} H_{k+p-2s}(J_1), \]

the contribution to the Euler characteristic reads

\[ \chi(\nu)_{i=0} = \frac{1}{8\pi R^2} \int_{-\infty}^{\infty} dJ_1 G(J_1) \text{Erfc} \left( \frac{\nu + \gamma J_1}{\sqrt{2(1 - \gamma^2)}} \right) \]
\[ \times \left[ H_2(J_1) + \sum_{s=0}^{\min(2,k)} \frac{2^{j+k-s-n} (\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j! k!} \right] \]
\[ \times \sum_{s=0}^{\min(2,k)} \frac{2^{j+k-s-n} (\zeta)^i (J_1)^j (J_2)^k H_i(\zeta) H_k(J_1)}{j! k!} \left. \right] \]

(A8)

We now use

\[ \frac{1}{2} \int_{-\infty}^{\infty} dJ_1 G(J_1) \text{Erfc} \left( \frac{\nu + \gamma J_1}{\sqrt{2(1 - \gamma^2)}} \right) H_k(J_1) \]
\[ = \begin{cases} \frac{1}{2} \text{Erfc} \left( \frac{\nu}{\sqrt{2}} \right), & k = 0, \\ \frac{1}{\sqrt{2\pi}} \exp \left( -\frac{\nu^2}{2} \right) (\gamma)^k H_{k-1}(\nu), & k > 0, \end{cases} \]

(A9)

to obtain
\[
\chi(n_{i0}) = \frac{1}{8\pi R_c^2} \text{Erfc}\left(\frac{\nu}{\sqrt{2}}\right) \sum_{j=0}^{\infty} \left[ (q_j^2 J_j^2)^{\text{GC}} - (q_j^2 J_j^2)^{\text{AC}} \right] + \frac{1}{4\pi \sqrt{2} \pi R_c} \exp\left(-\frac{\nu^2}{2}\right) \chi^2 H_1(\nu) + \sum_{n=3}^{\infty} \frac{i+j+k-n}{j!} \left(\frac{2}{j!}\right) H_{k+1-2n}(\nu) + \sum_{n=3}^{\infty} \sum_{j=0}^{k-12} \frac{(-1)^{j+k+1}}{j!} \left(\frac{2}{j!}\right)^2 (q_j^2 J_j^2)^{\text{GC}} H_{k-1}(\nu).
\]

(A10)

where in the latter summations only terms that lead to non-negative index of Hermite polynomials should be included. The remaining \(i > 0\) terms of the Euler number calculation are

\[
\chi(n_{i0}) = \frac{1}{4\pi \sqrt{2} \pi R_c^2} \int_{-\infty}^{\infty} dJ_1 G(J_1) \exp\left(-\frac{(\nu + \gamma J_1)^2}{2(1-\gamma^2)}\right) \times H_i\left(\frac{\nu + \gamma J_1}{\sqrt{1-\gamma^2}}\right) H_k(J_1) + \sum_{n=3}^{\infty} \frac{i+j+k-2n}{j!} \left(\frac{2}{j!}\right) H_{k+1-2n}(\nu)
\]

(A11)

We see that it has the same form as the second part of the \(\chi_{n=0}\) contribution which can be joined just by extending the range of \(i\) to start from zero. Thus, together with the result for \(\chi(\infty)\) we find the full Gram-Charlier expansion for the Euler characteristic, Eq. (34).

### 2. Euler characteristic in 3D

Following the 2D calculation, one just averages \(-I_3 = -(J_3^2 - 3J_1J_2 + 2J_3)/27\) over the full range of the second derivatives

\[
\chi(\nu) = -\frac{1}{R_c^2} \int_{-\infty}^{\infty} dJ_1 \int_{-\infty}^{\infty} dJ_2 \int_{-\infty}^{\infty} dJ_3 \frac{d\zeta}{(\nu + \gamma J_1)/(\sqrt{1-\gamma^2})} \times P_{\text{ext}}(\zeta, J_1, J_2, J_3).
\]

(A14)

In 3D, we have

\[
P_{\text{ext}}(\zeta, J_1, J_2, J_3) = \left(\frac{3}{2\pi}\right)^{3/2} P(\zeta, J_1, J_2, J_3).
\]

(A15)

Writing \(I_3 = (H_3(J_1) + \frac{2}{3} H_1(J_1) L_1^{(3/2)}(\zeta J_2) + 2J_3)/27\), one can perform all the integration explicitly using the orthogonality properties of the polynomial expansion of \(P_{\text{ext}}\) given by Eq. (22). First, integration over \(J_3\) and \(J_2\) gives the intermediate formula

\[
\chi(\nu) = -\frac{1}{27 R_c^2} \left\{ \frac{3}{2\pi} \right\}^{3/2} \int_{-\infty}^{\infty} dJ_1 \int_{-\infty}^{\infty} dJ_2 \int_{-\infty}^{\infty} d\zeta \frac{d\zeta}{(\nu + \gamma J_1)/(\sqrt{1-\gamma^2})} \times H_3(J_1) + \sum_{n=3}^{\infty} \frac{i+j+k-2n}{j!} \left(\frac{2}{j!}\right) H_{k+1-2n}(\nu)
\]

(A16)
where we have used $L_j^{(1/2)}(0) = (2j + 1)!/(2^j j!)$. The Euler number at the infinitely low threshold that encompass all the manifold

$$
\chi(-\infty) = -\frac{1}{27R_s^4 \frac{3}{2\pi}} \sum_{j=0}^{\infty} \left( -\frac{3}{2j} \right) \left[ (q^2 J_1)^{i<j} - 3(q^2 J_1 J_2)^{i<j} + 2(q^2 J_3)^{i<j} \right]
$$

(A17)

$$
= -\frac{1}{R_s \frac{3}{2\pi}} \sum_{j=0}^{\infty} \left( -\frac{3}{2j} \right) \left[ (q^2 J_1)^{i<j} \right].
$$

(A18)

As a function of the threshold, $\nu$, the integration is performed using the properties of the Hermite polynomials, first to get [given Eqs. (A7), (A9), and (A12)]

$$
\chi(\nu)|_{-\infty} = -\frac{1}{54R_s^4 \frac{3}{4\pi}} \int_{-\infty}^{\nu} dJ_1 G(J_1) \operatorname{Erfc} \left( \sqrt{\frac{\nu + \gamma J_1}{2(1 - \gamma^2)}} \right)
\left[ H_3(J_1) + \sum_{n=3}^{\infty} \sum_{j,k} \frac{1}{j!k!} \left( -\frac{3}{2} \right) \left( q^2 J_1^k \right) H_k(J_1) H_3(J_1) \right]
$$

(A19)

and

$$
\chi(\nu)|_{0} = -\frac{1}{27R_s^4 \frac{3}{4\pi}} \int_{0}^{\nu} dJ_1 G(J_1) \operatorname{Erfc} \left( \sqrt{\frac{\nu + \gamma J_1}{2(1 - \gamma^2)}} \right)
\left[ \sum_{n=3}^{\infty} \sum_{j,k} \frac{1}{j!k!} \left( -\frac{3}{2} \right) \left( q^2 J_1^k \right) H_k(J_1) H_3(J_1) \right]
$$

(A20)

and, finally, Eq. (64).

**APPENDIX B: FITS TO EXTREMA COUNTS AND SKELETON IN 3D**

1. Differential extrema counts

The Gaussian distribution of maxima is a classical result and fitting formulas have been proposed, e.g., in [9]. For the sake of consistency, we propose a formula using the same functional form as the next fits. For a scale-invariant power spectrum of spectral index $n_s$, the Gaussian distribution of maxima is accurately fitted by

$$
\frac{\partial n_s^{(0)}}{\partial \nu} = \frac{1}{10^3 \sqrt{2\pi}} \int_{0}^{\nu} dJ_1 \left[ \frac{97}{30} + \frac{n_s}{23} - \frac{n_s^2}{85} \right] \left[ \frac{182}{29} - \frac{4n_s}{25} - \frac{15n_s^2}{31} + \frac{n_s^3}{78} \right] \left[ \frac{368}{65} - \frac{2n_s}{19} - \frac{27n_s^2}{38} - \frac{n_s^3}{23} \right] \left[ \frac{61}{43} - \frac{n_s}{24} - \frac{n_s^2}{50} \right] \left[ \frac{29}{52} - \frac{n_s}{32} - \frac{2n_s^2}{27} \right] \left[ \frac{3}{31} - \frac{n_s}{314} - \frac{n_s^2}{85} \right] \left[ \frac{1}{37} - \frac{n_s}{383} - \frac{n_s^2}{288} \right] \left[ \frac{5}{17} - \frac{n_s}{58} - \frac{n_s^2}{19} - \frac{n_s^3}{377} \right].
$$

(B1)

The first-order non-Gaussian correction is fitted by

$$
\frac{\partial n_s^{(1)}}{\partial \nu} = \frac{\sigma}{10^3 \sqrt{2\pi}} \int_{0}^{\nu} dJ_1 \left[ \frac{449}{64} - \frac{27n_s}{80} - \frac{84n_s^2}{67} + \frac{5n_s^3}{37} \right] \left[ \frac{47}{4} - \frac{15n_s}{28} - \frac{35n_s^2}{38} + \frac{2n_s^3}{27} \right] \left[ \frac{245}{16} - \frac{19n_s}{22} - \frac{9n_s^2}{7} + \frac{4n_s^3}{29} \right] \left[ \frac{233}{28} - \frac{8n_s}{19} - \frac{17n_s^2}{36} + \frac{7n_s^3}{45} \right] \left[ \frac{81}{17} - \frac{7n_s}{24} - \frac{25n_s^2}{46} + \frac{n_s^3}{26} \right] \left[ \frac{13}{14} - \frac{n_s}{24} - \frac{4n_s^2}{43} + \frac{n_s^3}{90} \right] \left[ \frac{5}{17} - \frac{n_s}{58} - \frac{n_s^2}{19} - \frac{n_s^3}{377} \right].
$$

(B2)
For the saddle points, we have
\[
\frac{\partial n^{(0)}_{+}}{\partial \nu} = -\frac{1}{10^{3}\sqrt{2\pi}}e^{-(\nu^{2}/2)}\left[\left(\frac{454}{47} + \frac{n_{s}}{298} - \frac{n^{2}_{s}}{25} - \frac{n^{3}_{s}}{119}\right) + \left(1 - \frac{4n^{3}_{s}}{27} - \frac{5n^{3}_{s}}{117} + \frac{n^{3}_{s}}{21}\right)H_{1}(\nu) + \left(3 - \frac{n_{s}}{34} + \frac{6n^{2}_{s}}{35} + \frac{n^{3}_{s}}{22}\right)H_{2}(\nu) + \left(-\frac{37}{2} - \frac{10n_{s}}{39} + \frac{n^{3}_{s}}{24}\right)H_{3}(\nu) + \left(-\frac{11}{45} - \frac{n_{s}}{20} + \frac{n^{3}_{s}}{112}\right)H_{4}(\nu) + \left(-\frac{1}{42} - \frac{n^{2}_{s}}{268} - \frac{n^{3}_{s}}{308}\right)H_{5}(\nu)\right]
\]
and
\[
\frac{\partial n^{(1)}_{+}}{\partial \nu} = -\frac{\sigma}{10^{3}\sqrt{2\pi}}e^{-(\nu^{2}/2)}\left[\left(\frac{33}{28} - \frac{n_{s}}{223} - \frac{5n^{2}_{s}}{14} - \frac{2n^{3}_{s}}{31}\right) + \left(-\frac{30}{19} + \frac{25n_{s}}{88} + \frac{23n^{2}_{s}}{27} + 9n^{3}_{s}\right)H_{1}(\nu) + \left(\frac{31}{40} - \frac{n_{s}}{9} + \frac{6n^{2}_{s}}{38} + \frac{31n^{3}_{s}}{61}\right)H_{2}(\nu) + \left(\frac{43}{16} - \frac{30n^{2}_{s}}{31} + \frac{6n^{3}_{s}}{37}\right)H_{3}(\nu) + \left(-\frac{12}{23} - \frac{6n_{s}}{37} + \frac{32n^{2}_{s}}{51} + \frac{n^{3}_{s}}{7}\right)H_{4}(\nu) + \left(-\frac{6}{7} - \frac{n_{s}}{32} + \frac{3n^{2}_{s}}{20} + \frac{n^{3}_{s}}{52}\right)H_{5}(\nu)\right].
\]
Replacing the field by its opposite (i.e., replacing \(\nu\) with \(-\nu\) and the odd Gram-Charlier coefficients by their opposite) yields the expressions for minima and the second kind of saddle points:
\[
\frac{\partial n^{(0)}_{\min}}{\partial \nu}(\nu) + \frac{\partial n^{(1)}_{\min}}{\partial \nu}(\nu) = -\frac{\partial n^{(0)}_{\max}}{\partial \nu}(-\nu) - \frac{\partial n^{(1)}_{\max}}{\partial \nu}(-\nu),
\]
\[
\frac{\partial n^{(0)}_{++}}{\partial \nu}(-\nu) + \frac{\partial n^{(1)}_{++}}{\partial \nu}(-\nu) = -\frac{\partial n^{(0)}_{-}}{\partial \nu}(-\nu) - \frac{\partial n^{(1)}_{-}}{\partial \nu}(-\nu).
\]

The corresponding fits for the first-order corrections are thus obtained by changing the signs of the even Hermite polynomials. Figure 12 demonstrates the performance of the first-order theoretical correction fit to the measurements of extrema counts in \(n = 0\) simulations. The level of non-Gaussianity increases with \(\sigma\) and some deviations of the first-order fit become noticeable at \(\sigma \sim 0.15\). Most susceptible are number counts of \(n_{++}\) “pancake-like” saddle points, with \(n_{--}\) for “filament-like” saddle points still being fit very well. The difference of the peak distribution between the two types of the saddle points is correctly represented. The maxima \(n_{\max}\) fit continues to track the increasing shoulder of the high \(\nu\) tail, at the expense of some accuracy at the low \(\nu\) section.

At \(\sigma = 0.2\), some aspects of the extrema counts are still fitted remarkably well, i.e., the filament-like saddle counts, the high threshold tail of the maxima distribution, relative peak values for all types of extrema. But, in other places the fitting formulas start showing artifacts not present in the measurements. Here, we see artificial oscillations in the maxima distribution at intermediate thresholds. Similarly, high \(\nu\) tail of pancake-like saddles is grossly incorrect. This demonstrates exactly how the first-order predictions start to fail—due to the limited set of the Hermite polynomials that the first-order correction is expanded in, the fits start to show oscillations, even allowing for negative values, whenever the data is poorly represented by such set. As a rule of thumb, \(\sigma = 0.2\) is probably the limit for usefulness of the first-order description of non-Gaussian corrections.

![FIG. 12](color online). Development of non-Gaussian deviations in the distribution of extrema in \(n = 0\) simulations versus the first-order theoretical approximation.
2. Skeleton length

For $n_s \in [-2.5, 0]$, the Gaussian prediction for the differential skeleton length is well fitted by

$$
\frac{\partial L^{\text{skel}(0)}}{\partial \nu} = \frac{1}{10^3 \sqrt{2\pi}} e^{-(\nu^2/2)} \left[ \left( \begin{array}{c} 3041/23 + 4n_s - 3n_s^2 - n_s^3 \\ 111/25 + 22n_s^2/23 + n_s^3/4 \end{array} \right) H_1(\nu) + \left( \begin{array}{c} 2401/44 - 23n_s - 89n_s^2/11 - 32n_s^3/41 \\ 38/63 - 64n_s^2/33 - 40n_s^3/93 \end{array} \right) H_2(\nu) \right].
$$

while the first-order correction is fitted by

$$
\frac{\partial L^{\text{skel}(1)}}{\partial \nu} = \frac{\sigma}{10^3 \sqrt{2\pi}} e^{-(\nu^2/2)} \left[ \left( \begin{array}{c} -2450/33 - 719n_s^2/41 + 2n_s^3/19 + 67n_s^3/24 \\ -2143/57 - 1332n_s^2/19 + 910n_s^3/64 \end{array} \right) H_1(\nu) + \left( \begin{array}{c} 4479/38 - 1952n_s^2/23 + 5912n_s^3/37 + 1909n_s^3/31 \\ 5239/32 - 14842n_s^2/291 + 3788n_s^3/35 + 715n_s^3/18 \end{array} \right) H_2(\nu) + \left( \begin{array}{c} 4926/59 - 388n_s^2/23 + 2979n_s^3/76 + 178n_s^3/13 \\ 834/35 - 118n_s^2/35 + 240n_s^3/29 + 74n_s^3/27 \end{array} \right) H_3(\nu) \right].
$$

APPENDIX C: COMPUTING THE SKELETON LENGTH IN THE HESSIAN EIGENFRAME

The calculations made in Sec. III B 4 used the assumption that the properties of the components of the gradients are the same in the Hessian eigenframe as in any frame. This assumption translates into $\langle \hat{x}_2^2 \rangle = \langle x_2^2 \rangle = 1/2$, $\langle \hat{x}_2^3 \rangle = \langle x_2^3 \rangle = 1/2$, and $\langle \hat{x}_2^2 \hat{x}_1^2 \rangle = \langle x_2^2 x_1^2 \rangle = \langle x_2^2 J_1 \rangle$. These equalities are true in the Gaussian case: the first and second derivatives are independent so that the statistical properties of the gradient do not depend on the Hessian eigen directions. However, it is no longer true when non-Gaussianities give rise to correlations between the gradient and the Hessian. To avoid this assumption, one can measure the values of $\langle \hat{x}_2^2 \rangle$ and $\langle x_2^2 J_1 \rangle$. To do so, the Hessian needs to be diagonalized at each point, which makes the measurements a bit more difficult. Furthermore, the cumulants measured in the Hessian eigenframe seem difficult to predict via the perturbation theory described in the next section.

The assumption on the variances $\langle \hat{x}_2^2 \rangle$ and $\langle x_2^2 \rangle$ is more subtle. These terms are not explicitly present in the expansion. But, they are actually implicitly assumed when the Gaussian kernel for Gram-Charlier expansion is constructed. If these variances are taken in a fixed frame, the isotropy and the normalization of the gradient $\langle q^2 \rangle = 1$ fix these variances to 1/2. But, if the Hessian eigenframe is used, these variables are still Gaussian but their variance can change. The non-Gaussianities then affect the Gaussian kernel and the expression of the expansion. However, this problem can be easily solved by using these variances to normalize the variables. We can use the variable $Q = \sqrt{2(\hat{x}_2^2)}$ instead of $x_2$ to make sure that the variance is fixed and that the expansion using this new variable will not change. To summarize, if the properties of the gradient are not assumed to be the same in the Hessian eigenframe as in a fixed frame, one needs to diagonalize the Hessian, compute the components of the gradient in the eigenframe, normalize them to a fixed variance, measure the cumulants $\langle \zeta Q^2 \rangle$ and $\langle Q^2 J_1 \rangle$, and use the previous analytical predictions with the new normalized variable $Q$ instead of the original $x_2$. Figure 13 shows the results of such a procedure. A small improvement is seen near the maximum of the skeleton curve, but the difference with the previous results is almost negligible: the statistical properties of the components of the gradient are almost the same in the Hessian eigenframe as in a fixed frame and considering the apparition of a difference with non-Gaussianities is just a higher-order correction.

FIG. 13 (color online). Prediction for the 2D skeleton length using the components of the gradient in a fixed coordinate frame (dashed line) versus the Hessian eigenframe (solid line).
APPENDIX D: CONNECTING THE EDGOWORTH TO THE GRAM-CHARLIER EXPANSION

In this section, we give a formal derivation of the Edgeworth series \[47\]. For convenience, we restrict it to the case of a monovariate distribution, but it can be generalized directly \[83\]. Understanding the link between the Gram-Charlier and the Edgeworth expansions will be the key to a physically motivated resummation of the expansion. For a given PDF \( P(x) \), we define the moment-generating function

\[
M(t) = \int_{-\infty}^{\infty} P(x)e^{tx}dx = \sum_{p=0}^\infty \frac{\langle x^p \rangle}{p!}t^p, \tag{D1}
\]

and the cumulant-generating function

\[
C(t) = \sum_{p=2}^\infty \frac{\langle x^p \rangle}{p!}t^p. \tag{D2}
\]

By definition of the cumulants, we have

\[
M(t) = \exp(C(t)). \tag{D3}
\]

Now, let us take the Fourier transform of the PDF:

\[
\tilde{P}(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} P(x)e^{itx}dx. \tag{D4}
\]

We have

\[
\tilde{P}(t) = M(it) = \exp(C(it)). \tag{D5}
\]

We can get the PDF by taking the inverse Fourier transform:

\[
P(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-itx} \exp\left(\sum_{p=2}^\infty \frac{\langle x^p \rangle c}{p!}it^p\right) dt. \tag{D6}
\]

In order to have an expansion around the Gaussian case, we extract the \( p = 2 \) term which corresponds to the purely Gaussian contribution:

\[
P(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-itx} e^{-(\sigma^2/2)t^2} \exp\left(\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}it^p\right) dt, \tag{D7}
\]

where \( \sigma^2 = \langle x^2 \rangle \). If we expand the last exponential, we will thus have the Gaussian term

\[
\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-itx} e^{-(\sigma^2/2)t^2} dt = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x^2/2\sigma^2)}, \tag{D8}
\]

and the other terms will involve

\[
\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-itx} e^{-(\sigma^2/2)t^2} i^pt^p dt
= (-1)^p \frac{d^p}{dx^p} \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x^2/2\sigma^2)}\right). \tag{D9}
\]

The expansion can therefore be formally written as

\[
P(x) = \exp\left(\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}\right) \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x^2/2\sigma^2)}. \tag{D10}
\]

Let us expand the exponential in Taylor series:

\[
P(x) = \left(1 + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}}{2} + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^2}{2} + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^3}{6} + \ldots \right)G(x), \tag{D11}
\]

where \( G(x) = \exp(-x^2/2\sigma^2)/\sqrt{2\pi\sigma^2} \) is the Gaussian PDF. Developing powers of the operator allows us to rewrite Eq. (D11) as

\[
P(x) = \left(1 + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}}{2} \right.
+ \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^2}{2}
+ \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^3}{6}
\times \left(1 + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^3}{2}\right) + \ldots \right)G(x). \tag{D12}
\]

By definition of the Hermite polynomials, \((-1)^k d^k G/\ dx^k = H_k(x/\sigma)G(x)/\sigma^2\), it follows that

\[
P(x) = \left(1 + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}}{2} H_{p+1}(x/\sigma) \right)
+ \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^2}{2} H_{p+1}(x/\sigma) \right)
+ \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^3}{6}
\times \left(1 + \frac{\sum_{p=3}^\infty \frac{\langle x^p \rangle c}{p!}(-1)^p \frac{d^p}{dx^p}^3}{2}\right) + \ldots \right)G(x). \tag{D13}
\]

Joining the terms having the same Hermite polynomial would give the Gram-Charlier expansion. The Gram-Charlier coefficient \( \langle x^n \rangle_{GC} \) can therefore be expressed in terms of cumulants:

\[
\langle x^n \rangle_{GC} = \langle x^n \rangle_c + \frac{n!}{2!} \sum_{p_1+p_2=n} \frac{\langle x^{p_1} \rangle c \langle x^{p_2} \rangle c}{p_1! p_2!}
+ \frac{n!}{3!} \sum_{p_1+p_2+p_3=n} \frac{\langle x^{p_1} \rangle c \langle x^{p_2} \rangle c \langle x^{p_3} \rangle c}{p_1! p_2! p_3!} + \ldots. \tag{D14}
\]

where the notation \( \sum_{p_1+p_2+n} \) means a summation on \( p_1 \) and \( p_2 \) from 3 to \( \infty \) under the condition \( p_1 + p_2 = n \). In the context of gravitational instability, we have \( \langle x^n \rangle \propto \sigma^{2n-2} \). The expansion can be reordered according to the
order of the cumulants, introducing the coefficients $S_p = \langle x^p \rangle_c / \sigma^{p-2}$:

$$P(x) = \left(1 + \sum_{p=3}^{\infty} \frac{S_p \sigma^{p-2}}{p!} H_p \left(\frac{x}{\sigma}\right) \right)$$

$$\quad + \frac{1}{2} \sum_{p_1=3}^{\infty} \sum_{p_2=3}^{\infty} \frac{S_{p_1} S_{p_2} \sigma^{p_1+p_2-4}}{p_1! p_2!} H_{p_1+p_2} \left(\frac{x}{\sigma}\right)$$

$$\quad + \frac{1}{6} \left(\sum_{p_1=3}^{\infty} \sum_{p_2=3}^{\infty} \sum_{p_3=3}^{\infty} S_{p_1} S_{p_2} S_{p_3} \sigma^{p_1+p_2+p_3-6} \right)$$

$$\quad \times H_{p_1+p_2+p_3} \left(\frac{x}{\sigma}\right) + \ldots \right) G(x).$$ (D15)

The first terms are

$$P(x) = \left(1 + \frac{S_3}{3!} H_3 \left(\frac{x}{\sigma}\right) + \frac{S_4 \sigma^2}{4!} H_4 \left(\frac{x}{\sigma}\right) \right)$$

$$\quad + \frac{S_5 \sigma^3}{5!} H_5 \left(\frac{x}{\sigma}\right) + \ldots + \frac{1}{2} \left(\frac{S_3 S_3 \sigma^2}{3! 3!} H_6 \left(\frac{x}{\sigma}\right) \right)$$

$$\quad + \frac{S_3 S_4 \sigma^3}{3! 4!} H_7 \left(\frac{x}{\sigma}\right) + \frac{S_4 S_3 \sigma^3}{4! 3!} H_7 \left(\frac{x}{\sigma}\right) + \ldots$$

$$\quad + \frac{1}{6} \left(\frac{S_3 S_3 S_3 \sigma^3}{3! 3! 3!} H_9 \left(\frac{x}{\sigma}\right) + \ldots \right) G(x).$$ (D16)

Ordering the terms w.r.t their order in $\sigma$ gives the Edgeworth expansion:

$$P(x) = \left(1 + \frac{S_3}{6} H_3 \left(\frac{x}{\sigma}\right) + \frac{S_4 \sigma^2}{24} H_4 \left(\frac{x}{\sigma}\right) \right)$$

$$\quad + \frac{S_5 \sigma^3}{72} H_6 \left(\frac{x}{\sigma}\right) + \frac{S_4 S_3 \sigma^3}{120} H_6 \left(\frac{x}{\sigma}\right)$$

$$\quad + \frac{S_3 S_3 S_3 \sigma^3}{144} H_9 \left(\frac{x}{\sigma}\right) + \frac{S_3 S_3 S_3 \sigma^3}{1296} H_9 \left(\frac{x}{\sigma}\right) + \ldots \right) G(x),$$

which corresponds to the expression found in the literature [40]. Equation (D14) allows us to determine the orders to which a given coefficient contributes to the summation. Indeed, a coefficient with $n$ fields can be decomposed as a sum of

(i) a linear (in cumulants) term, which is of order $\sigma^{n-2}$, 
(ii) if $n > 3$, a quadratic term, of order $\sigma^{p_1-2} + \sigma^{p_2-2} = \sigma^{n-4}$, 
(iii) if $n > 6$, a cubic term, of order $\sigma^{p_1-2} + \sigma^{p_2-2} + \sigma^{p_3-2} = \sigma^{n-6}$, 
(iv) . . .

Thus, the orders in $\sigma$ can be obtained as follows:

(i) the $\sigma$ order comes from the $n = 3$ terms, 
(ii) the $\sigma^2$ order comes from the $n = 4$ term and part of the $n = 6$ one, 
(iii) the $\sigma^3$ order comes from the $n = 5$ term and part of the $n = 7$ and $n = 9$ ones,

(iv) the $\sigma^4$ order comes from part of the $n = 6$, $n = 8$, $n = 10$, and $n = 12$ terms, 
(v) . . .

In the next section, we will see how to compute explicitly the required elements of the expansion up to some order in $\sigma$.

**APPENDIX E: CRITICAL SETS IN THE EDGEWORTH PT EXPANSION**

1. **Truncating the Gram-Charlier expansion to order $\sigma^2$**

According to Appendix D, the $\sigma^2$ term arises from the 4th-order cumulants and from part of the 6th-order Gram-Charlier coefficients. Indeed, the 6th-order coefficients contain quadratic $S_3^2$-like terms. For example, Eq. (D14) gives $\langle x^6 \rangle_{GC} = \langle x \rangle_c^6 + 10 \langle x^3 \rangle_c^2$. In general, these contributions can also be computed by expressing the Gram-Charlier coefficients in terms of moments using the expressions of the orthogonal polynomials [see Eqs. (13), (23), and (24)], and then expressing moments in terms of cumulants. Perturbation theory ensures us that a term involving a cumulant with $p$ fields will be of order $\sigma^{p-2}$.

For example, the 6th-order Hermite polynomial is $H_6(\xi) = 6^6 - 15 \xi^4 + 45 \xi^2 - 15$. So,

$$\langle \xi^6 \rangle_{GC} = \langle H_6(\xi) \rangle = \langle \xi^6 \rangle - 15 \langle \xi^4 \rangle + 30,$$

where we have used the fact that $\langle \xi^2 \rangle = 1$. We can then use the decomposition of moments into cumulants:

$$\langle \xi^4 \rangle = \langle \xi \rangle_c^4 + 3 \langle \xi^2 \rangle^2,$$

$$\langle \xi^6 \rangle = \langle \xi \rangle_c^6 + 15 \langle \xi^4 \rangle_c \langle \xi^2 \rangle + 10 \langle \xi^3 \rangle^2 + 15 \langle \xi^2 \rangle^3.$$

In the end,

$$\langle \xi^6 \rangle_{GC} = \langle \xi \rangle_c^6 + 10 \langle \xi^2 \rangle^2.$$

The first term is of order $\sigma^4$ and the second one is of order $\sigma^2$. Table V gives all $\sigma^2$ terms coming from the 6t-order Gram-Charlier coefficients used in the computation of the Euler characteristic. The expansion to 9th order is available online (http://www.iap.fr/users/pichon/Gram).

2. **Minkowski functionals**

**A. 2D length of isocontours**

In order to illustrate how to use Table V, let us consider the reordering of the length of the isocontours in 2 dimensions presented in Sec. III B 1. To obtain an expression up to $\sigma^2$ order, one has to compute the $n = 3$, 4, and 6 term from the general expression (30):

17 Or $\sigma^{2p-2}$ if not normalized.
Only terms contributing to the Euler characteristic are shown. Terms involving \( J_3 \) do not exist in 2D, and those involving \( q^4 \) differ in 2 and 3 dimensions.

<table>
<thead>
<tr>
<th>Coefficient</th>
<th>( \sigma^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( (\varepsilon^0)_{GC} )</td>
<td>10(( \varepsilon^0 ))^2</td>
</tr>
<tr>
<td>( (\varepsilon^3)<em>{J_1}^{\prime}</em>{GC} )</td>
<td>10(( \varepsilon^3 ))( (\varepsilon^3)_{J_1} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_1}^{\prime}</em>{GC} )</td>
<td>6(( \varepsilon^4 ))( (\varepsilon^4)_{J_1} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_1}^{\prime}</em>{GC} )</td>
<td>9(( \varepsilon^4 ))( (\varepsilon^4)_{J_1} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_2}^{\prime}</em>{GC} )</td>
<td>6(( \varepsilon^4 ))( (\varepsilon^4)_{J_2} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_3}^{\prime}</em>{GC} )</td>
<td>10(( \varepsilon^4 ))( (\varepsilon^4)_{J_3} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_4}^{\prime}</em>{GC} )</td>
<td>4(( \varepsilon^4 ))( (\varepsilon^4)_{J_4} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_5}^{\prime}</em>{GC} )</td>
<td>(( \varepsilon^4 ))( (\varepsilon^4)_{J_5} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_6}^{\prime}</em>{GC} )</td>
<td>(( \varepsilon^4 ))( (\varepsilon^4)_{J_6} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_7}^{\prime}</em>{GC} )</td>
<td>(( \varepsilon^4 ))( (\varepsilon^4)_{J_7} )</td>
</tr>
<tr>
<td>( (\varepsilon^4)<em>{J_8}^{\prime}</em>{GC} )</td>
<td>(( \varepsilon^4 ))( (\varepsilon^4)_{J_8} )</td>
</tr>
</tbody>
</table>

Using the results from Table V, this expression can be truncated at the \( \sigma^2 \) order:

\[
\mathcal{L}(\nu) = \frac{1}{2\sqrt{2}} e^{-(\nu^2/2)} \left( 1 + \frac{1}{6} (\varepsilon^3)_{H_3}(\nu) + \frac{1}{2} (\varepsilon^4)_{H_4}(\nu) \right) + \frac{1}{24} (\varepsilon^4)_{H_4}(\nu)
\]

\[
- \frac{1}{16} (\varepsilon^4)_{H_4}(\nu) - \frac{1}{720} (\varepsilon^6)_{GC} H_6(\nu)
\]

\[
+ \frac{1}{48} (\varepsilon^4)_{GC} H_4(\nu) - \frac{1}{128} (\varepsilon^2 q^2)_{GC} H_2(\nu)
\]

\[
+ \frac{1}{96} (\varepsilon^6)_{GC} H_6(\nu).
\]

Using the expression of the \( \sigma^2 \) contribution of the \( n = 6 \) terms, we get the \( \sigma^2 \) expression, and its truncation to \( \sigma^2 \) order reads

\[
\mathcal{A}(\nu) = \frac{2}{\sqrt{3\pi}} e^{-(\nu^2/2)} \left( 1 + \frac{3}{2} (\varepsilon^3)_{H_3}(\nu) + \frac{1}{2} (\varepsilon^3)_{H_3}(\nu) \right)
\]

\[
- \frac{9}{40} (\varepsilon^4)_{H_4}(\nu) + \frac{3}{4} (\varepsilon^2 q^2)_{H_2}(\nu)
\]

\[
+ \frac{1}{8} (\varepsilon^4)_{H_4}(\nu) + \frac{27}{560} (\varepsilon^6)_{GC} H_6(\nu)
\]

\[
- \frac{9}{80} (\varepsilon^2 q^2)_{GC} H_2(\nu) + \frac{1}{16} (\varepsilon^2 q^2)_{GC} H_4(\nu)
\]

\[
+ \frac{1}{240} (\varepsilon^6)_{GC} H_6(\nu).
\]

\[
\text{B. 3D surface of the isocontours}
\]

Following the same routine, Eq. (61) in Sec. III C 1 becomes to the required order

\[
\mathcal{A}(\nu) = \frac{2}{\sqrt{3\pi}} e^{-(\nu^2/2)} \left( 1 + \frac{3}{2} (\varepsilon^3)_{H_3}(\nu) + \frac{1}{2} (\varepsilon^3)_{H_3}(\nu) \right)
\]

\[
- \frac{9}{40} (\varepsilon^4)_{H_4}(\nu) + \frac{3}{4} (\varepsilon^2 q^2)_{H_2}(\nu)
\]

\[
+ \frac{1}{8} (\varepsilon^4)_{H_4}(\nu) + \frac{27}{560} (\varepsilon^6)_{GC} H_6(\nu)
\]

\[
- \frac{9}{80} (\varepsilon^2 q^2)_{GC} H_2(\nu) + \frac{1}{16} (\varepsilon^2 q^2)_{GC} H_4(\nu)
\]

\[
+ \frac{1}{240} (\varepsilon^6)_{GC} H_6(\nu).
\]

\[
18\text{Since no second derivative is involved in this simple example, } x \text{ can be used instead of } \varepsilon.
\]
All other terms in Eq. (E3) contribute to orders in $\sigma$ higher than 2.

### 3. 2D Euler characteristic

It is interesting to look at the first correction to the Euler characteristic since it has been predicted in other papers using a somewhat different approach [26,48]. The $\sigma$-order term is given by the $n = 3$ term of the Gram-Charlier expansion. In 2 dimensions, Eq. (34) yields

$$
\chi(\nu) = \frac{1}{4\pi \sqrt{2\pi R_c^2}} e^{-(\nu^2/2)} \left[ \nu^2 H_1(\nu) + (2\gamma q^2 J_1) + \sqrt{1 - \nu^2} \right] - \nu^2 (q^2 J_1^2) - \nu^2 (q^2 J_2^2) + \gamma (J_1^3) H_0(\nu)
$$

This result can be simplified using the variable $x$ and $I_2$:

$$
\chi(\nu) = \frac{1}{4\pi \sqrt{2\pi R_c^2}} e^{-(\nu^2/2)} \left[ \nu^2 H_1(\nu) + (2\gamma q^2 I_1) + 4(x I_2) H_0(\nu) - (\gamma^2 x^2 q^2) + \gamma (x^2 I_1) H_2(\nu) + \frac{\gamma^2}{6} (x^3) H_4(\nu) \right].
$$

The expression given by [48] for the two-dimensional genus is

$$
G_2(\nu) = \frac{1}{2\pi^{3/2}} \left( \frac{\sigma R_\gamma}{\sqrt{2 \sigma}} \right)^2 e^{-(\nu^2/2)} \left[ H_1(\nu) + \frac{S^{(0)}}{6} H_2(\nu) + \frac{S^{(1)}}{3} H_2(\nu) + \frac{S^{(2)}}{3} \right],
$$

where in terms of the non-normalized field, $\rho$,

$$
S^{(0)} = \frac{\langle \rho^3 \rangle}{\sigma^3}, \quad S^{(1)} = \frac{3}{4} \frac{\langle \rho^2 \nabla^2 \rho \rangle}{\sigma^3}, \quad S^{(2)} = \frac{3d}{2(d-1)} \frac{\langle (\nabla \rho)^2 \nabla^2 \rho \rangle}{\sigma^3}.
$$

With our normalization and for $d = 2$, these quantities can be expressed in terms of the rotational invariants:

$$
S^{(0)} = \frac{\langle x^3 \rangle}{\sigma}, \quad S^{(1)} = -\frac{3}{4} \frac{\langle x^2 I_1 \rangle}{\gamma \sigma}, \quad S^{(2)} = -\frac{3}{\gamma} \langle q^2 I_1 \rangle.
$$

Factoring $\gamma^2$ from our expression for the Euler characteristic gives the same prefactor since $\gamma/R_\gamma = \sigma_1/\sigma$. The $H_0(\nu)$ term is obviously the same:

$$
\sigma \frac{S^{(0)}}{6} = \frac{1}{6} \langle x^3 \rangle.
$$

Using the isotropy relations given in [48], one can check the identities:

$$
\langle x q^2 \rangle = -\frac{1}{2\gamma} \langle x^2 I_1 \rangle, \quad \langle x I_2 \rangle = -\frac{3}{4} \gamma \langle q^2 I_1 \rangle,
$$

which are needed to prove that the coefficients of $H_2(\nu)$ and $H_4(\nu)$ agree. Similarly, one can express the Euler characteristic up to $\sigma^2$ order. The result is more compact when expressed in terms of $x$ and $I_2$:

$$
\chi(\nu) = \frac{1}{4\pi \sqrt{2\pi R_c^2}} e^{-(\nu^2/2)} \left[ \nu^2 H_1(\nu) + (2\gamma q^2 I_1) + 4(x I_2) H_0(\nu) - (\gamma^2 x^2 q^2) + \gamma (x^2 I_1) H_2(\nu) + \frac{\gamma^2}{6} (x^3) H_4(\nu) \right]
$$

$$
+ \left( \frac{\gamma^2}{2} \langle q^4 \rangle \right) H_1(\nu) + 2 \gamma \langle x q^2 J_1 \rangle + 2 \langle x^2 I_2 \rangle \right] H_1(\nu) - \left( \frac{\gamma^2}{2} \langle x^2 q^2 \rangle \right) \frac{\gamma}{3} \langle x_3 J_1 \rangle H_3(\nu) - \frac{\gamma^2}{24} \langle x_3 I_1 \rangle H_5(\nu)
$$

$$
- (4\gamma \langle x q^2 \rangle \langle q^2 J_1 \rangle + \langle q^2 J_1 \rangle \langle x^2 J_1 \rangle + 4 \langle x q^2 \rangle \langle x I_2 \rangle) \right] H_1(\nu) + \left( \gamma^2 \langle x q^2 \rangle \right)^2 \frac{\gamma}{3} \langle x^3 q^2 \rangle + \gamma \langle x q^2 \rangle \langle x^2 J_1 \rangle
$$

$$
+ \frac{1}{4} \langle x^2 J_1 \rangle^2 + \frac{2}{3} \langle x^3 \rangle \langle x I_2 \rangle \right] H_3(\nu) - \left( \frac{\gamma^2}{6} \langle x q^2 \rangle \langle x^3 \rangle + \frac{\gamma}{6} \langle x^3 \rangle \langle x^2 J_1 \rangle \right) H_5(\nu) + \frac{\gamma^2}{72} \langle x^3 \rangle H_7(\nu).
$$

This result can be simplified using the isotropy relations (E11):
\[ \chi(\nu) = \frac{1}{4\pi^2/2R^2} e^{-(\nu^2/2)} \left( \gamma^2 H_3(\nu) - \gamma^2 H_0(\nu) - \frac{1}{2} \gamma x^2 H_1(\nu) + \frac{\gamma^2}{6} \langle x^3 \rangle H_4(\nu) + \left( \frac{\gamma^2}{2} \langle q^4 \rangle_c + 2 \gamma \langle x q^2 \rangle_c + 2 \langle x^2 I_2 \rangle_c \right) H_1(\nu) - \left( \frac{\gamma^2}{2} \langle x^2 q^2 \rangle_c + \frac{\gamma}{3} \langle x^3 \rangle_1 \right) H_3(\nu) + \frac{\gamma^2}{24} \langle x^4 \rangle_c H_5(\nu) \right) + \left( \frac{\gamma^2}{2} \langle q^4 \rangle_c + 2 \gamma \langle x q^2 \rangle_c + 2 \langle x^2 I_2 \rangle_c \right) H_1(\nu) - \frac{\gamma}{6} \langle x^3 \rangle H_3(\nu) - \frac{\gamma}{12} \langle x^3 \rangle H_5(\nu) + \frac{\gamma^2}{72} \langle x^3 \rangle^2 H_7(\nu) \right]. \] (E13)

This expression agrees with [53] given the following identities:

\[ \langle x^2 q^2 \rangle_c = -\frac{1}{3\gamma} \langle x^3 \rangle_1, \quad \langle q^4 \rangle_c + \frac{3}{\gamma} \langle x q^2 \rangle_c + \frac{2}{\gamma^3} \langle x^2 I_2 \rangle_c = 0. \] (E14)

These relations are consistent with the measurements. The compactness of the expansion in invariants allows us to compute the \( \sigma^3 \) correction:

\[ \chi^{(3)}(\nu) = \left( \frac{3}{2} \gamma \langle q^4 \rangle_c \langle x^2 q^2 \rangle_1 - \gamma \langle x q^2 \rangle_c \right) + \frac{3}{2} \gamma \langle q^2 \rangle_c \langle x q^2 \rangle_1 - \frac{1}{2} \gamma \langle x^2 q^2 \rangle_c \right) H_0(\nu) + \left( \frac{1}{3} \gamma \langle x^3 \rangle q^2 \langle x q^2 \rangle_1 - \frac{1}{12} \gamma \langle x^3 \rangle q^2 \langle x^2 q^2 \rangle_1 + \frac{1}{12} \gamma \langle x^3 \rangle q^2 \langle x^3 \rangle_1 - \frac{1}{6} \gamma \langle x^3 \rangle q^2 \langle x^3 \rangle_1 \right) H_0(\nu) + \left( \frac{1}{12} \gamma^2 \langle x^3 \rangle q^2 \langle x^3 \rangle_1 \right) H_0(\nu) + \left( \frac{1}{144} \gamma^2 \langle x^3 \rangle q^2 \langle x^3 \rangle_1 - \gamma \langle x^3 \rangle q^2 \langle x^3 \rangle_1 \right) H_0(\nu) + \left( \frac{1}{1296} \gamma^2 \langle x^3 \rangle q^2 \langle x^3 \rangle_1 \right) H_0(\nu) + \left( \frac{1}{27R^4} \left( \frac{3}{2\pi} \right)^{3/2} \frac{1}{\sqrt{2\pi}} e^{-(\nu^2/2)} \left[ \gamma^3 H_3(\nu) + \left( \frac{9}{2} \gamma \langle q^2 \rangle c + 3 \gamma \sqrt{1 - \gamma^2 \langle \xi^2 \rangle} - 3 \gamma \langle \xi \rangle \right) \right) - \left( \frac{3}{2} \gamma \sqrt{1 - \gamma^2 \langle \xi^2 \rangle} + \frac{3}{2} \gamma \langle q^2 \rangle c - \frac{3}{2} \gamma \langle \xi \rangle \right) \right) H_0(\nu) + \left( \frac{1}{6} \gamma^3 \langle 1 - \gamma^2 \rangle^{3/2} (\xi^3) - \frac{1}{2} \gamma^2 \langle 1 - \gamma^2 \rangle (\xi^2) \right) H_0(\nu) \right]. \] (E15)

4. 3D Euler characteristic

In 3D, the first-order term from Eq. (64) is

\[ \chi(\nu) = \frac{1}{27R^4} \left( \frac{3}{2\pi} \right)^{3/2} \frac{1}{\sqrt{2\pi}} e^{-(\nu^2/2)} \left[ \gamma^3 H_3(\nu) + \left( \frac{9}{2} \gamma \langle q^2 \rangle c + 3 \gamma \sqrt{1 - \gamma^2 \langle \xi^2 \rangle} - 3 \gamma \langle \xi \rangle \right) \right) - \left( \frac{3}{2} \gamma \sqrt{1 - \gamma^2 \langle \xi^2 \rangle} + \frac{3}{2} \gamma \langle q^2 \rangle c - \frac{3}{2} \gamma \langle \xi \rangle \right) \right) H_0(\nu) + \left( \frac{1}{6} \gamma^3 \langle 1 - \gamma^2 \rangle^{3/2} (\xi^3) - \frac{1}{2} \gamma^2 \langle 1 - \gamma^2 \rangle (\xi^2) \right) H_0(\nu) \right]. \] (E16)

As in 2D, it can be simplified by using the variables \( x, I_2 \), and \( I_3 \):

\[ \chi(\nu) = \frac{1}{27R^4} \left( \frac{3}{2\pi} \right)^{3/2} \frac{1}{\sqrt{2\pi}} e^{-(\nu^2/2)} \left[ \gamma^3 H_3(\nu) + \left( \frac{9}{2} \gamma \langle q^2 \rangle c + 9 \gamma \langle x I_2 \rangle \right) H_1(\nu) - \left( \frac{3}{2} \gamma^2 \langle x^2 q^2 \rangle \right) H_1(\nu) + \left( \frac{3}{2} \gamma \langle x^2 I_2 \rangle \right) H_3(\nu) + \frac{1}{6} \gamma^3 \langle x^3 \rangle H_5(\nu) \right]. \] (E17)

Matsubara’s expression for the three-dimensional genus is

\[ G_3(\nu) = \frac{1}{(2\pi)^3} \left( \frac{\sigma_1}{\sqrt{3}\sigma_i} \right)^3 e^{-(\nu^2/2)} \left[ H_2(\nu) + \left( \frac{S(0)}{6} H_3(\nu) + S(1) H_4(\nu) + S(2) H_6(\nu) \right) \right]. \] (E18)

with

\[ S(0) = \frac{1}{\sigma_i} \langle x^3 \rangle, \quad S(1) = -\frac{3}{4} \frac{\gamma}{\gamma \sigma_i}, \quad S(2) = \frac{9}{4} \frac{\gamma^2}{\gamma \sigma_i}. \] (E19)

Factorizing the expression for \( \chi(\nu) \) with \( \gamma^3 \) leads to the same prefactor except for the sign. This sign difference comes from the fact that \( \chi(\nu) \) is the Euler characteristic of the excursion set while \( G_3(\nu) \) is the genus of the isocontour.
and thus $\chi(\nu) = -G_3(\nu)$. Using the isotropy identities given in Eq. (E11), it is then easy to show that each term coincides. Finally, the 3D Euler characteristic up to $q^2$ order expressed in terms of $x_1, x_2$, and $x_3$ is

$$\chi(\nu) = \frac{1}{27\pi^2} \left[ \frac{3}{2} \right]^{3/2} \frac{1}{\sqrt{2\pi}} \int e^{-q^2/2} \left[ \gamma^2 H_2(\nu) + \left( \frac{9}{2} \gamma^2 q^2 J_1 \right) H_1(\nu) + \left( \frac{3}{2} \gamma^2 xq^2 \right) H_3(\nu) \right].$$

In Eq. (F3), we assumed zero smoothing of the field for now. Using the notations of [40], given the Euler and continuity equation for an EdS universe, we can express $\delta^{(2)}$ in terms of $\delta^{(1)}$ as

$$\delta^{(2)}(k) = \int d^3 k_1 d^3 k_2 \delta^D(k - k_1 - k_2) F_2(k_1, k_2) \times \delta^{(1)}(k_1) \delta^{(1)}(k_2).$$

Thus, Eq. (F4) becomes

$$I = \int d^3 k_1 d^3 k_2 d^3 k_3 d\varepsilon \left( k_1 + k_2 + k_3 \right) \left( \frac{5}{7} + \frac{k_1 \cdot k_2}{k_1^2 k_2^2} + 2 \frac{(k_1 \cdot k_2)^2}{k_1^2 k_2^2} \right).$$

To simplify, we can combine the derivatives and $F_2$ into a generalized shape factor:

$$\mathcal{F}_{\alpha \beta \gamma}(k_1, k_2, k_3, k_4) = \left( i k_1 \right)^{\alpha_1} \left( i k_2 \right)^{\alpha_2} \left( i k_3 \right)^{\alpha_3} \left( i k_4 \right)^{\alpha_4} \left( i k_5 \right)^{\alpha_5} \mathcal{F}_2(k_4, k_5).$$

which gives

$$I = \int d^3 k_1 d^3 k_2 d^3 k_3 d\varepsilon \left( k_1 + k_2 + k_3 + k_4 \right) \left( \frac{5}{7} + \frac{k_1 \cdot k_2}{k_1^2 k_2^2} + 2 \frac{(k_1 \cdot k_2)^2}{k_1^2 k_2^2} \right) \times \mathcal{F}_{\alpha \beta \gamma}(k_1, k_2, k_3, k_4, k_5) \times \delta^{(1)}(k_1) \delta^{(1)}(k_2) \delta^{(1)}(k_3) \delta^{(1)}(k_4) \delta^{(1)}(k_5).$$

**APPENDIX F: GEOMETRIC CUMULANTS FROM PERTURBATION THEORY**

In this section, we will compute the 3rd-order cumulants involved in the non-Gaussian expansion in the case where non-Gaussianities arise from the nonlinear gravitational evolution.

1. Derivation for an EdS universe

For a three-dimensional field, the cumulants involving the field and its derivatives can be written as

$$\langle \left( \partial^{a_1} \partial^{a_2} \partial^{a_3} \delta \right) \left( \partial^{b_1} \partial^{b_2} \partial^{b_3} \delta \right) \left( \partial^{c_1} \partial^{c_2} \partial^{c_3} \delta \right) \rangle = \langle \left( \partial^{a_1} \partial^{a_2} \partial^{a_3} \delta \right) \langle \partial^{b_1} \partial^{b_2} \partial^{b_3} \delta \rangle \langle \partial^{c_1} \partial^{c_2} \partial^{c_3} \delta \rangle \rangle$$

where $\partial^n_i$ is the nth derivative with respect to the $i$th coordinate. Using gravitational perturbation theory [40], we expand the field as $\delta = \delta^{(1)} + \delta^{(2)} + \ldots$. Inserting this expansion in Eq. (F1) gives the lowest nonvanishing order contribution to the cumulant:

$$\langle \left( \partial^{a_1} \partial^{a_2} \partial^{a_3} \delta \right) \left( \partial^{b_1} \partial^{b_2} \partial^{b_3} \delta \right) \left( \partial^{c_1} \partial^{c_2} \partial^{c_3} \delta \right) \rangle = \langle \left( \partial^{a_1} \partial^{a_2} \partial^{a_3} \delta \right) \langle \partial^{b_1} \partial^{b_2} \partial^{b_3} \delta \rangle \langle \partial^{c_1} \partial^{c_2} \partial^{c_3} \delta \rangle \rangle$$

Let us compute the first term in Eq. (F2), $I = \langle \left( \partial^{a_1} \partial^{a_2} \partial^{a_3} \delta \right) \langle \partial^{b_1} \partial^{b_2} \partial^{b_3} \delta \rangle \langle \partial^{c_1} \partial^{c_2} \partial^{c_3} \delta \rangle \rangle$, which is best expressed in Fourier space as (with $k^{[j]}$ the jth component of vector $k$)

$$I = \int d^3 k_1 d^3 k_2 d^3 k_3 e^{i(\varepsilon_{k_1} + \varepsilon_{k_2} + \varepsilon_{k_3})} \langle \left( i k_1 \right)^{\alpha_1} \left( i k_2 \right)^{\alpha_2} \times \left( i k_3 \right)^{\alpha_3} \rangle \times \left( i k_1 \right)^{\beta_1} \left( i k_2 \right)^{\beta_2} \left( i k_3 \right)^{\beta_3} \times \left( i k_4 \right)^{\gamma_1} \left( i k_5 \right)^{\gamma_2} \times F_2(k_4, k_5).$$

Finally, the 3D Euler characteristic up to $q^2$ order expressed in terms of $x_1, x_2$, and $x_3$ is

$$\chi(\nu) = 1/27\pi^2 (3/2)^{3/2} 1/\sqrt{2\pi} \int e^{-q^2/2} \left( \gamma^2 H_2(\nu) + \left( \frac{9}{2} \gamma^2 q^2 J_1 \right) H_1(\nu) + \left( \frac{3}{2} \gamma^2 xq^2 \right) H_3(\nu) \right].$$
The integrated moment is then expanded using Wick’s theorem:

\[
I = \int d^3k_1 d^3k_2 d^3k_4 d^3k_5 e^{i(k_1 + k_2 + k_4 + k_5) \cdot x} \\
\times \mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2, k_4, k_5) \\
\times \langle \delta^{(1)}(k_1) \delta^{(1)}(k_2) \delta^{(1)}(k_4) \delta^{(1)}(k_5) \rangle.
\]  

(8)

The first term leads to \( F_2(k_4, -k_4) \) which vanishes, while the second and third terms are equivalent:

\[
I = 2 \int d^3k_1 d^3k_2 d^3k_4 d^3k_5 \mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2, k_4, k_5) \\
\times \delta^{(1)}(k_1 + k_4) P(k_1) \delta^{(1)}(k_2 + k_5) P(k_2).
\]  

(F9)

Integrating over \( k_4 \) and \( k_5 \) gives

\[
I = 2 \int d^3k_1 d^3k_2 \mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2) P(k_1) P(k_2).
\]  

(F10)

where we use the notation \( \mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2) = \mathcal{F}_{\beta, \alpha, \gamma}(k_1, k_2, -k_1, -k_2) \) [see Eq. (80) in the main text]. Finally, combining the three terms in Eq. (F2) gives us the cumulant:

\[
\langle (\partial_\alpha)^n \partial_\beta^m \partial_\gamma^p \delta \rangle (\partial_\alpha)^1 \partial_\beta^2 \partial_\gamma^3 \delta) (\partial_\alpha^1 \partial_\beta^2 \partial_\gamma^3 \delta)
\]

\[
= 2 \int d^3k_1 d^3k_2 \mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2) + \mathcal{F}_{\beta, \gamma, \alpha}(k_1, k_2)
\]

+ \( \mathcal{F}_{\gamma, \alpha, \beta}(k_1, k_2) P(k_1) P(k_2).
\]

Let us now take into account the smoothing of the field over a scale, \( R \); the cumulants become

\[
2 \int d^3k_1 d^3k_2 [\mathcal{F}_{\alpha, \beta, \gamma}(k_1, k_2) + \mathcal{F}_{\beta, \gamma, \alpha}(k_1, k_2)
\]

+ \( \mathcal{F}_{\gamma, \alpha, \beta}(k_1, k_2) P(k_1) P(k_2) W(k_1 R)
\]

\[
\times W(k_2 R) W(|k_1 + k_2| R).
\]

(F11)

Equation (F11) is the general expression for the cumulants used in the main text. The appearance of the dependence on the relative orientation of the wave vectors \( k_1 \) and \( k_2 \) in the filter factor \( W(|k_1 + k_2| R) \) is the source of most of the complexity of the theory and, in some sense, the essence of perturbation theory. It reflects the fact that the nonlinear field, smoothed at radius \( R \), is not determined solely by the average quantities at this radius, but also by what happens at shorter scales. For Gaussian filtering, the filter function can be expanded in Legendre series with respect to \( k_1 \cdot k_2 \) [71]:

\[
W(|k_1 + k_2| R) = \exp \left(-\frac{1}{2}(k_1^2 + k_2^2)R^2\right) \sum_{\ell=0}^{\infty} (-1)^\ell (2\ell + 1)
\]

\[
\times P_\ell (k_1 \cdot k_2) \frac{\pi}{\sqrt{2k_1 k_2 R^2}}.
\]

(F12)

Then, \( \mathcal{F}_{\alpha, \beta, \gamma} \) can be decomposed on the basis of the Legendre polynomials and the integration over the angles then just requires the orthogonality relation of the Legendre polynomials:

\[
\int_1^1 dx P_m(x) P_n(x) = \frac{2}{2m + 1} \delta_{m,n}.
\]

(F13)

It also means that in practice, one does not need Legendre polynomials of order higher than the degree of the integrated term (here, 2 from \( F_2 \) plus the number of derivatives) and can truncate the expansion. The result is an integral of Bessel functions, which can be expressed using hypergeometric functions. The results for all the independent third-order cumulants are given in Appendix B 2.

2. Gaussian filtered scale-invariant geometric cumulants

Using the above-defined procedure, the cumulants can be computed for a Gaussian filter. For a scale-invariant power spectrum of index \( n \) (called \( n_s \) in the main text), the results can be analytically expressed using the hypergeometric function \( 2F_1(a, b, c, x) \). For example, the result for the skewness is already known [71]:

\[
\frac{1}{\sigma} \langle x^3 \rangle = 3_2 F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, \frac{3}{2}, \frac{1}{2} \right) - \frac{6(3 + n)(8 + 7n)}{21n^2} \frac{2}{2} F_1 \left( \frac{3 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}, \frac{3}{2}, \frac{1}{2} \right).
\]

(F14)

Proceeding the same way with the cumulants involving derivatives of the field yields

\[
\frac{1}{\sigma} \langle xx_1^2 \rangle = -\frac{4(37020 + 51456n + 37092n^2 + 13828n^3 + 2579n^4 + 189n^5)}{35n^2(2 + n)^2(4 + n)^2(5 + n)^2} \frac{2}{2} F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, \frac{3}{2}, \frac{1}{2} \right)
\]

\[
+ \frac{3}{2} \left( 3(30720 + 51456n + 37092n^2 + 13828n^3 + 2579n^4 + 189n^5) \right) \frac{2}{2} F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, \frac{3}{2}, \frac{1}{2} \right)
\]

\[
+ \frac{n}{2} \left( 3(30720 + 51456n + 37092n^2 + 13828n^3 + 2579n^4 + 189n^5) \right) \frac{2}{2} F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, \frac{3}{2}, \frac{1}{2} \right)
\]

(F15)
\[ \frac{1}{\sigma} \langle x_{i1}x_{i2} \rangle = -\frac{4(-46080 - 37984n - 2848n^2 + 5718n^3 + 1949n^4 + 189n^5)}{105n^2(2 + n)^2(4 + n)^2(5 + n)^2} \left( (4 + 2n)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \\
+ 32 F_1 \left( \frac{3 + n}{2}, \frac{5 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \left) + \frac{3(-5760 - 2624n + 608n^2 + 476n^3 + 63n^4)}{105n^2(2 + n)(4 + n)(5 + n)(6 + n)} \left( 32 F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \\
+ n_2 F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \right), \tag{F17}
\]

\[ \frac{1}{\sigma} \langle x_{i2}^2 \rangle = -\frac{4(9120 + 96176n + 57062n^2 + 17883n^3 + 2894n^4 + 189n^5)}{105n^2(2 + n)^2(4 + n)^2(5 + n)^2} \left( (4 + 2n)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{3 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \\
+ 32 F_1 \left( \frac{3 + n}{2}, \frac{5 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \left) + \frac{3(8640 + 8836n + 3818n^2 + 791n^3 + 63n^4)}{105n^2(2 + n)(4 + n)(5 + n)(6 + n)} \left( 32 F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \\
+ n_2 F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, -\frac{1}{2}; \frac{1}{4} \right) \right. \right), \tag{F18}
\]

The last ones also depend on the spectral parameter \( \gamma = \sqrt{(n + 3)/(n + 5)}:

\[ \frac{1}{\sigma} \langle x_{i1}x_{i2} \rangle = -\frac{\gamma}{42n^2(2 + n)} \left( (448 + 260n + 42n^2)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (96 + 50n + 7n^2)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F19}
\]

\[ \frac{1}{\sigma} \langle x_{i1}x_{i2}x_{i3} \rangle = -\frac{8\gamma}{245n^2(2 + n)(4 + n)} \left( (7680 + 6384n + 2262n^2 + 395n^3 + 28n^4)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (4320 + 3438n + 1180n^2 + 201n^3 + 14n^4)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F20}
\]

\[ \frac{1}{\sigma} \langle x_{i1}x_{i2}^2 \rangle = -\frac{4\gamma}{735n^2(2 + n)(4 + n)} \left( (425600 + 21280n + 8002n^2 + 1557n^3 + 126n^4)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (31920 + 15280n + 5582n^2 + 1059n^3 + 54n^4)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F21}
\]

\[ \frac{1}{\sigma} \langle x_{i1}^2x_{i2} \rangle = -\frac{4\gamma}{735n^2(2 + n)(4 + n)} \left( (8(1280 + 1064n + 608n^2 + 186n^3 + 21n^4)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (31920 + 1528n + 862n^2 + 255n^3 + 78n^4)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F22}
\]

\[ \frac{1}{\sigma} \langle x_{i1}x_{i2}x_{i3} \rangle = -\frac{4\gamma}{245n^2(2 + n)(4 + n)} \left( (4(-5120 - 4256n - 1046n^2 - 23n^314n^4)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
+ (11520 + 9168n + 2134n^2 + 39n^3 - 28n^4)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F23}
\]

\[ \frac{1}{\sigma} \langle x_{i1}^2x_{i2} \rangle = -\frac{4\gamma}{735n^2(2 + n)(4 + n)} \left( (4(-1024 - 8512n - 706n^2 + 675n^3 + 126n^4)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{7 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (3(7680 + 6112n + 410n^2 - 471n^3 - 84n^4)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right), \tag{F24}
\]

\[ \frac{1}{\sigma} \langle x_{i1}^2 \rangle = -\frac{\gamma}{21n^2(2 + n)} \left( (448 + 260n + 42n^2)_{2,1} F_1 \left( \frac{3 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right. \\
- (96 + 50n + 7n^2)_{2,1} F_1 \left( \frac{5 + n}{2}, \frac{5 + n}{2}, \frac{3}{2}; \frac{1}{4} \right) \right). \tag{F25}
\]
TABLE VI. Predicted and measured (Sec. IV B 3) third-order 3D coordinate cumulants. According to perturbation theory, the dimensionless third-order cumulants are proportional to $\sigma$.

<table>
<thead>
<tr>
<th>$n_x = 0$</th>
<th>$n_x = -1$</th>
<th>$n_x = -2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\langle x^3 \rangle / \sigma$</td>
<td>3.144</td>
<td>3.08 ± 0.09</td>
</tr>
<tr>
<td>$\langle xx_1^2 \rangle / \sigma$</td>
<td>0.699</td>
<td>0.68 ± 0.01</td>
</tr>
<tr>
<td>$\langle xx_1^2 \rangle / \sigma$</td>
<td>0.567</td>
<td>0.554 ± 0.009</td>
</tr>
<tr>
<td>$\langle xx_1 x_2 \rangle / \sigma$</td>
<td>0.361</td>
<td>0.348 ± 0.006</td>
</tr>
<tr>
<td>$\langle xx_1 x_2 \rangle / \sigma$</td>
<td>0.103</td>
<td>0.102 ± 0.002</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>0.111</td>
<td>0.107 ± 0.002</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 x_4 \rangle / \sigma$</td>
<td>0.0124</td>
<td>0.0122 ± 0.0003</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-0.0068</td>
<td>-0.0061 ± 0.0006</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-0.0316</td>
<td>-0.0306 ± 0.0006</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-0.120</td>
<td>-0.116 ± 0.002</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-0.254</td>
<td>-0.214 ± 0.005</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-0.210</td>
<td>-0.203 ± 0.004</td>
</tr>
<tr>
<td>$\langle x_1 x_2 x_3 \rangle / \sigma$</td>
<td>-1.08</td>
<td>-1.05 ± 0.02</td>
</tr>
</tbody>
</table>

\[
\frac{1}{\sigma} \langle x_1^3 \rangle = -\frac{12\gamma}{245n^2(2+n)(4+n)} \left( 4(1024 + 8512n + 3478n^2 + 767n^3 + 70n^4)F_1 \left( \frac{3 + n}{2} , \frac{7 + n}{2} , \frac{3}{2} \right) \right) \\
- \left( 23040 + 18336n + 7306n^2 + 1569n^3 + 140n^4 \right)F_1 \left( \frac{5 + n}{2} , \frac{7 + n}{2} , \frac{3}{2} , \frac{1}{2} \right). \tag{F26}
\]

\[
\frac{1}{\sigma} \langle x^2 x_1 \rangle = -\frac{\gamma}{21n^2} \left( 96 + 124n + 42n^2 \right)F_1 \left( \frac{3 + n}{2} , \frac{3 + n}{2} , \frac{3}{2} , \frac{1}{4} \right) - 3(3 + n)(8 + 7n)F_1 \left( \frac{3 + n}{2} , \frac{5 + n}{2} , \frac{3}{2} , \frac{1}{4} \right). \tag{F27}
\]

All other cumulants are either identical to one of these because of isotropy, or null (mostly because of the parity of the derivatives). These formulas are thus sufficient to build any combination (e.g., the cumulants of the rotational invariants). In Sec. IV B 3, Tables VI and III give numerical values and show they agree with measurements.

**APPENDIX G: EULER CHARACTERISTIC ALGORITHM**

The very fast, linear in the number of pixels, EULER2D algorithm that we use to measure the Euler characteristic in 2D is based on the Gauss-Bonnet theorem. According to this theorem, the Euler characteristic of a region can be obtained by integrating the curvature over the boundary surface. We however note that explicit integration of the curvature is not needed, since any continuous deformation of the regions conserve the topology. Hence, on a grid the curvature can be thought of as concentrated in the corners of the cells at the boundary between the cells above and below the threshold and just counted by addition. What is needed is to assign the appropriate curvature weight for each boundary grid vertex, depending on which cells that form it are below and which are above the threshold. The sum of the weights over all the vertices will give us the

![Fig. 14](color online). Simple configurations used to determine the number of the geometrical configurations. The Euler characteristic of the example is written below the drawings. The red numbers are the new weights that are determined at the current step of the bootstrap, while the blue ones are the previously determined ones.
Euler characteristic of the excursion set. Our implementation of the algorithm is complemented by the cluster analysis which allows us to count the Euler characteristics of the individual isolated regions.

To determine the weights, we use a bootstrap method: we consider elementary configurations for which the Euler characteristic is known in order to determine the number to associate with a new geometrical configuration.

For a 2D regular grid, the geometric configurations can be classified in six categories, represented in Fig. 15. The number associated with the vertex in the first configuration (when all four surrounding pixels are not part of a region) is trivial: it is zero, since adding some empty pixels around a region does not change its topology. Similarly, a vertex surrounded by four pixels has also the weight of zero since vertices interior to the excursion set do not change its Euler characteristic. The second configuration can be studied by looking at the case of a region made of only one isolated pixel (Fig. 14(a)): its Euler characteristic is unity and all four corners of the pixel are equivalent and should thus carry the weight 1/4. These vertices are of a type when three neighbors are outside the region and one is inside. Recording this value, we can now look at the next category of pixel. Figure 14(b) shows that if the region is made of two adjacent pixels, the Euler characteristic is still unity and the intermediate corners of pixels must contribute the weight of zero since four corner vertices are of the type considered before and carry the weight of 1/4. We can then consider an L-shaped region made of three pixels (Fig. 14(c)) to find that a vertex surrounded by three pixels in a region has a weight of −1/4. The last remaining configuration is more problematic: when two pixels touch only by a corner (Fig. 14(d)), one has to decide whether they are part of the same region or not. Depending on this choice, the weight is 1/2 or −1/2. In case we do track the topology of individual regions, one of the decisions has to be made, which will bias the statistical result (but will be exact for a given definition of a connected region). If we are interested in statistical analysis of the total Euler characteristic of the complete excursion set, we note that, statistically in 2D, the two pixels touching by just the vertex are as likely to be connected as they are not. Thus, we make a statistical choice of assigning the weight zero to this last vertex type.

The final assignment of curvature weights is given in Fig. 15. One can easily repeat the procedure in 3D, or generalize it onto more complicated grids. It can be noticed that for the simple case of a 2D regular grid, the result is simply the number of turns the boundary contour rotates at a given corner [84]. Furthermore, with our statistical prescription, the weights associated with different geometrical configurations only depend on the number of surrounding pixels in the region, not on their configuration, which makes coding the algorithm very simple.

In complete implementation, the genus algorithm is complimented by the cluster analysis, so that the Euler characteristic of each individual isolated cluster can be tracked. This approach has also been implemented for 3D grids and more complicated HEALPix pixelization in EULER3D and EULERHEALPIX families of codes. The codes are available from the authors.

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PHYSICAL REVIEW D 85, 023011 (2012)