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Abundance of peaks and dips in 3D mass and halo density fields: a test for cosmology

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ABSTRACT

Using cosmological N-body simulations, we study the abundance of local maxima (peaks) and minima (dips) identified in the smoothed distribution of haloes and dark matter (DM) on scales of 10–100 s Mpcs. The simulations include Gaussian and local-type $f_{\rm NL}$ non-Gaussian initial conditions. The expression derived in the literature for the abundance (irrespective of height) of peaks for Gaussian fields is surprisingly accurate for the evolved halo and DM density fields for all initial conditions considered. Furthermore, the height distribution is very well fitted by a lognormal on quasi-linear scales. The abundance as a function of scale depends on the cosmological parameters (H_0 and background matter densities) through the shape of the power spectrum, but it is insensitive to the clustering amplitude. Further, the abundance in the smoothed halo distribution is substantially different in the non-Gaussian from the Gaussian simulations. The interpretation of this effect is straightforward in terms of the scale dependence of halo bias in non-Gaussian models. The abundance of extrema extracted from 3D large galaxy redshift surveys could be a competitive probe of the clustering amplitude, making it complementary to counts of galaxy clusters and peaks in weak-lensing maps.

Key words: galaxies: haloes - cosmology: theory - dark matter.

1 INTRODUCTION

Peaks in the underlying mass density field are the most likely sites for the formation of haloes, where gas is expected to accrete and form galaxies (White & Rees 1978). In the classical picture of Press & Schechter (1974), matter in regions with linear density contrast above a threshold, δ_c , is assigned to haloes of mass larger than *M*, where *M* defines the smoothing of the density field. This implies that haloes of mass *M* form at peaks with $\delta = \delta_c$ in the smoothed density contrast δ .

Naturally, most studies have focused on peaks associated with haloes. Indeed, statistical properties of local extrema (e.g. Adler 1981) have gained a great deal of attention in cosmology (Bardeen et al. 1986, hereafter BBKS). Correlations of haloes and their distribution in relation to the mass density field of the gravitationally dominant dark matter (DM), i.e. biasing (Kaiser & N. 1984), have been studied extensively with analytic methods and numerical simulations. For Gaussian initial conditions and on sufficiently large scales, haloes follow a linear-biasing relation, $\delta_h = b\delta$ between the halo number density contrast, δ_h , and the mass density contrast δ . The bias factor, *b*, depends on the height of the peaks associated

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with haloes and on their mass. An important result obtained in simulations (Dalal et al. 2008), and confirmed by analytic techniques (Grinstein & Wise 1986; Dalal et al. 2008; Matarrese & Verde 2008; Slosar et al. 2008), is that the presence of initial local-type non-Gaussianity introduces a peculiar scale dependence in the bias factor dubbed 'non-Gaussian bias'. The specific form of b(k) (k is the wavenumber of a given scale) opens the window for probing initial non-Gaussianity based on the clustering properties of galaxies in planned large redshift surveys, e.g. *Euclid* (Laureijs et al. 2011) and DESI (DESI Collaboration 2016).

It is well known that non-Gaussianity strongly affects the tails of density probability distributions (Adler 1981; Catelan, Lucchin & Matarrese 1988a). Several authors have further specialized these results to local density maxima of non-Gaussian density fields, where the non-Gaussianity is either of a generic form (e.g. Catelan, Lucchin & Matarrese 1988b; Gay, Pichon & Pogosyan 2012; Codis et al. 2013; Uhlemann et al. 2018) or developed via non-linear gravitational evolution of initial Gaussian conditions (e.g. Suginohara 1991; Matsubara 1994). In particular, Gay et al. (2012) and Codis et al. (2013) considered the effect of a generic non-Gaussianity on extrema counts and Minkowski functionals of the *DM* density field. However, these authors did not explicitly discuss the case of density fields constructed out of observed galaxy catalogues. Therefore, they did not take into account non-Gaussian bias, which is one of

the main effects discussed here. In this work, we consider peaks and dips in cosmological density field smoothed on scales much larger than those of galactic and galaxy cluster haloes (≤ 10 Mpcs). Using N-body simulations in large cosmological boxes, we focus on the total number of local extrema for density fields constructed from the *halo* distribution, as a proxy for a galaxy catalogue. Earlier analyses (Croft & Gaztanaga 1998; De & Croft 2007, 2010) have used this type of statistics for constraining parameters related to the linear matter power spectrum on smaller scales (≤ 10 Mpcs), such as the running of the spectral index and the neutrino mass. Our goal is to assess the extent to which the abundance of extrema in 3D fields inferred from current and forthcoming large galaxy redshift surveys can be used as a cosmological tool and, more specifically, a probe of local primordial non-Gaussianity. As we shall see, our main findings have a straightforward interpretation in terms of the non-Gaussian bias.

We adopt standard notation. The mean total and baryonic mass densities (in units of the critical density) are denoted by $\Omega_{\rm m}$ and $\Omega_{\rm b}$, respectively. The Hubble constant is H_0 and $h = H_0/[100 \,{\rm km \, s^{-1} \, Mpc^{-1}}]$. The linear growth factor (normalized to unity at this time) at redshift, *z*, is *D* (*z*). The outline of the paper is as follows. In Section 2, we lay out known relations between the number of extrema and the underlying power spectrum for Gaussian fields. A description of the N-body simulations is provided in Section 3 and the corresponding results for the abundance of local extrema identified in smoothed density fields derived from the DM and halo distributions are in Section 4. In Section 5, we discuss the prospects for the application of the number of extrema as a test of cosmological parameters and conclude with a summary in Section 6.

2 DEFINITIONS AND THEORETICAL EXPECTATIONS

We define local maxima (peaks) in a smoothed random field, f, as points in space where the spatial gradient is $\partial_{\alpha} f = 0$ and the Hessian $\partial_{\alpha} \partial_{\beta} f$ is negative definite. Local minima (dips) are defined similarly but with a positive definite Hessian. For a random Gaussian field, peaks and dips have an equal total number *per unit volume*, which was computed by **BBKS** to be

$$n_0 \approx 0.016 R_*^{-3},$$
 (1)

where

$$R_* = \sqrt{3} \frac{\sigma_1}{\sigma_2},\tag{2}$$

and the spectral moments

$$\sigma_j^2 = \int \frac{k^2 dk}{2\pi^2} P(k) W_R^2(k) k^{2j}.$$
 (3)

The expression for n_0 is independent of the clustering amplitude, and it depends only on the shape of the power spectrum, P(k) of the field, and the smoothing Kernel $W_R(k)$. For $P(k) \sim k^n$ and a Gaussian smoothing window $W_R^2(k) = \exp(-k^2 R^2)$, it is easy to see that

$$\frac{R}{R_*} = \left(\frac{n+5}{6}\right)^{1/2}.\tag{4}$$

The total number of peaks is preserved under a local monotonous one-to-one mapping, $F(\delta)$, of the density field. Thus, we expect this quantity to be independent of time in the quasi-linear scales. On smaller scales, local extrema tend to merge and diffuse, leading to deviations from expression equation (1).

In addition to the DM density field, we also examine peaks and dips in the smoothed distribution of haloes. The corresponding spectral moments $\sigma_j \equiv \sigma_{j,h}$ are given by

$$\sigma_{j,h}^2 = \int \frac{k^2 dk}{2\pi^2} k^{2j} W_R^2(k) \Big[b^2(k) P(k) + \frac{1}{\bar{n}_h} \Big].$$
(5)

The expression in square brackets is a model for the power spectrum of the halo distribution, where P(k) here refers to the underlying density field and b(k) describes the scale-dependent halo bias. The term $1/\bar{n}_h$ is due to the finite number of haloes and approximated as a Poisson discreteness noise. Using the simulations described next, we have found that the added discreteness variance is strongly suppressed for large smoothing and is actually sub-Poissonian, in agreement with the findings of Casas-Miranda et al. (2002) and Hamaus et al. (2010). On linear scales, the halo bias b(k) is constant for Gaussian initial conditions but depends on the halo mass, i.e. $b(k) = b^{G}(M)$.

We also consider local-type non-Gaussianity (Salopek & Bond 1990; Gangui et al. 1993; Komatsu & Spergel 2001) for which the Bardeen potential Φ deep in matter domination is expanded around a random Gaussian field ϕ as

$$\Phi(\mathbf{x}) = \phi(\mathbf{x}) + f_{\rm NL} \left([\phi(\mathbf{x})]^2 - \langle \phi^2 \rangle \right).$$
(6)

The bispectrum of Φ induces the following scale dependence in the bias factor:

$$b(k) = b^{\mathrm{G}}(M) + \frac{\alpha(f_{\mathrm{NL}})}{k^2 T(k)},\tag{7}$$

where

$$\alpha(f_{\rm NL}) \equiv 3 f_{\rm NL} \frac{\partial \ln \bar{n}_h}{\partial \ln \sigma_8} \frac{\Omega_m H_0^2}{D(z)c^2},\tag{8}$$

and $\bar{n}_h(M)$ is the abundance of haloes (per unit *M*) computed for the Gaussian field without the $f_{\rm NL}$ terms. When implementing equation (5) to compare it to data (see Section 4), we use the following approximation:

$$\frac{\partial \ln \bar{n}_h}{\partial \ln \sigma_8} \approx \delta_c (b^{\rm G}(M) - 1), \tag{9}$$

with $\delta_c = 1.687$, which is valid for universal mass functions and the spherical collapse model.¹ We do not include expressions (e.g. Gay et al. 2012) for the theoretical corrections to equation (1) due to $f_{\rm NL}$ non-Gaussianity. Indeed, we will see next that the expression of **BBKS** remains accurate provided that the appropriate σ_i is used.

3 SIMULATIONS

Two sets of simulations, respectively, in a $2 h^{-1}$ Gpc and a $3 h^{-1}$ Gpc box, are available for initial conditions generated from ACDM initial power spectra with slightly different cosmological parameters, as described in the Table 1. The simulations were run with the GADGET2 (Springel 2005) N-body code on the Baobab cluster at the University of Geneva. The initial particle displacements were implemented at $z_i = 99$ using the public code 2LPTIC (Crocce, Pueblas & Scoccimarro 2006) for realizations with Gaussian initial conditions and its modified version (Scoccimarro et al. 2012) for non-Gaussian initial conditions of the local type. The transfer function for the smaller box (simulations 1, see Table 1) was obtained using the CLASS code (Blas, Lesgourgues & Tram 2011). This set contains runs for Gaussian initial conditions and two for local-type non-Gaussianity, respectively, with $f_{\rm NL} = 250$ and $f_{\rm NL} = -250$. For each of these initial conditions, we obtain eight random realizations corresponding to different random seeds.

¹See Biagetti et al. (2017) for a quantitative analysis on this approximation on the same set of simulations, sim 1, used here.

Table 1. Simulation parameters, where *L* is the box size (in unit of h^{-1} Gpc), N_p , number of simulation particles, and M_{halo} is the minimum halo mass identified in the simulation (in unit of $10^{12} h^{-1} M_{\odot}$). Both, simulations 1 and 2, include Gaussian and two choices for non-Gaussian initial conditions. Outputs of simulations 1 are available at z = 0 and z = 1, while only the output at z = 0 is available for simulations 2. In all simulations, the Hubble parameter is h = 0.7, and the spectral index of the initial power spectrum at large scales is $n_s = 0.967$.

	L	Np	Mhalo	σ_8	Ω_{m}	$\Omega_{\rm b}$
sim 1	2	1536^3	3.67	0.85	0.3	0.0455
sim 2	3	1024^3	37.9	0.81	0.272	0.0455



Figure 1. Total number of maxima versus the smoothing length, from the DM distribution in simulations 1 and 2 for Gaussian initial conditions at redshift z=0 The lines represent the corresponding theoretical prediction using equation (1), and the shaded area represents the 1σ shot-noise for the larger simulation.

The transfer function of the second set, simulations 2, was obtained using the CAMB code (Lewis, Challinor & Lasenby 2000). This set includes three types of models: Gaussian initial conditions ($f_{\rm NL} = 0$) and non-Gaussian initial conditions, respectively, with $f_{\rm NL} = 100$ and $f_{\rm NL} = -100$. For each type of models, we have three simulations corresponding to different random realizations of the initial conditions. The ROCKSTAR (Behroozi, Wechsler & Wu 2013) algorithm is employed to identify haloes, with linking length $\lambda = 0.28$.

Density fields are interpolated from the DM and halo distributions in the simulation box on a 512³ cubic grid using the Clouds-in-Cells scheme. The grid spacing is thus $3.9 h^{-1}$ Mpc and $5.85 h^{-1}$ Mpc, for simulations 1 and 2, respectively. The density fields were additionally smoothed with a Gaussian window of eight different widths in the range $20 h^{-1}$ Mpc to $500 h^{-1}$ Mpc. For each smoothed field, local maxima (minima) were identified as grid points surrounded by grid points with lower (higher) density values. Fig. 1 shows the total number of maxima in the smoothed DM density field in the full boxes of simulations 1 and 2 at z = 0. The theoretical predictions obtained from the BBKS expression (equations 1 and 2) using the linear power spectrum $P(k) = P_L(k)$ for the two models are also shown, as indicated in the figure.² The shaded area encompasses the expected range of (1σ) shot-noise for simulations 2. The number drops like R^{-3} , consistently with equation (1) since $R_* \propto R$ upto a factor of $\mathcal{O}(1)$, which depends on the shape of the power spectrum at scale R (cf. equation 4). The figure refers to the Gaussian simulations only. A similar figure can be found in Ludlow & Porciani (2011) but for comparison of the theoretical expression with peaks identified in the initial conditions of their simulations.

4 RESULTS

4.1 Total number of minima and maxima

Gaussian initial conditions imply equal probability of producing peaks and dips, upto fluctuations due to the finite box size. However, on scales ≤ 10 s of Mpcs, non-linear gravitational evolution breaks the initial symmetry through the merging and smearing of dips and peaks. For non-Gaussian initial conditions, the statistical symmetry between maxima and minima is already broken initially.

We choose to first analyse the (total) number, n_1 , per unit volume of minima, n_{\min} , and maxima, n_{\max} , in the simulations. The differences between the abundance of minima and maxima will be discussed at a later stage. More precisely, we consider

$$n_1 = \frac{1}{2} \left(n_{\min} + n_{\max} \right), \tag{10}$$

which is computed from the smoothed density fields for the various simulations. An advantage of n_1 is that it boosts the statistical significance of the measured abundance. For a Gaussian field, $n_1 = n_0$ given in equation (1). Inclusion of non-Gaussian terms modify the abundance of either the minima or maxima by a leading-order correction proportional to the skewness of the density field and its derivatives (Gay et al. 2012). The combined leading-order correction for both minima and maxima cancel out in the expression of n_1 . Consequently, the BBKS prediction equation (1) remains valid up to a small correction of order f_{NL}^2 .

According to Fig. 1, differences in n_1 between the simulations are visually hard to examine directly. Thus, we consider the statistic

$$\Upsilon \equiv \frac{n_1 R^3}{0.016},\tag{11}$$

where R is the width of the smoothing window. According to equation (1), for a Gaussian field, $\Upsilon = (R/R_*)^3$. The three-panel Fig. 2 summarizes the main results. The top panel plots Υ , averaged over the eight random realizations in simulations 1, against the smoothing length, R, for the DM density field. The shaded area represents the 1σ shot-noise in n_1 corresponding to the finite number of peaks and dips in the simulation box. It is estimated as $\sqrt{n_0 L^3/2}$, where n_0 is the theoretical value according to equation (1), and the factor of 1/2 arises from the definition of n_1 that involves both minima and maxima. We have checked that the scatter from the eight individual runs (not shown for clarity) is consistent with this estimate of the shot-noise. For our Gaussian as well as non-Gaussian simulations, the results in the top panel for z = 1 and z = 0 are almost identical. The dotted line shows $(R/R_*)^3$ computed according to the theoretical expression equation (2) derived for Gaussian fields, where σ_i are computed using equation (3) with the initial power spectrum $P_{\rm L}(k)$. There is a reasonable match between the dotted curve and $(R/R_*)^3$ derived from n_1 for the Gaussian simulations (black and red solid

 $^{^{2}}$ In performing the integration in equation (3), it is important to impose a low-*k* cut-off corresponding to the finite box size of the simulations.



Figure 2. The quantity Υ as estimated from equation (11). *Top:* from the number of peaks and troughs in the DM distribution of simulations 1. *Middle:* The same as the *Top* but for the halo distribution. *Bottom:* For DM and haloes for simulations 2, at z = 0 only. In all panels, the grey area represents the shot-noise estimated from the expression using the theoretical linear power spectrum.



Figure 3. A test of the analytic prediction for the non-Gaussian model. The dashed and dash–dotted curves are taken from the middle panel in the previous figure. The curves with the circles plot $(R/R_*)^3$ computed with the approximate σ_i given in equations (5)–(7).

curves). Overall, the impact of $f_{\rm NL}$ is very small, in agreement with the fact that, for DM, n_1 depends on $f_{\rm NL}$ only at order $f_{\rm NL}^2$.

The middle panel refers to results obtained from the halo distribution in simulations 1. The solid curves corresponding to the Gaussian simulations at z = 1 and z = 0 are similar. In great contrast to the upper panel, both $f_{\rm NL} = 250$ and $f_{\rm NL} = -250$ models (dashed and dash-dotted line) at the two redshifts are substantially different. It is interesting to check how well the BBKS expression in equation (2) fits the Υ computed from the haloes in the non-Gaussian simulations. To do that we compute $(R/R_*)^3$ using equation (2) for σ_1 and σ_2 computed directly from the halo density fields. The results are plotted as the plus signs and circles, respectively, for the $f_{\rm NL} = 250$ and $f_{\rm NL} = -250$ simulations. We present the z = 1 case only, but the excellent agreement of Υ with $(R/R_*)^3$ computed from n_1 also holds at z = 0. The bottom panel summarizes results for simulations 2 (z=0) of the larger box. The haloes in these simulations have a larger mass and therefore follow a different biasing relation than haloes in simulations 1, yielding different quantitative results. For these simulations also, the BBKS expression (computed with σ_i measured in the simulations), shown as the plus signs and circles, furnishes an excellent match. Therefore, despite the fact that relations equations (1) and (2) are formally obtained for Gaussian fields, they remain accurate for the non-Gaussian fields considered here, provided the actual σ_i are used.

In Fig. 3, we compare the theoretical expectation of equation (5) against Υ measured from the non-Gaussian simulations. The theoretical curve fits good the data on scales $R \leq 100 \text{ Mpc h}^{-1}$ and provides a qualitatively good description at all scales. Deviations may be due in part to our approximation equation (9) and, especially at large scales, to the finite box size of the simulations.

To conclude this section, we note that the effect of $f_{\rm NL}$ on Υ is only weakly degenerate with that of σ_8 because n_1 primarily depends on the ratio of spectral moments $\sigma_{1,h}/\sigma_{2,h}$.



Figure 4. The relative difference between the total number of maxima and minima in simulations 2 versus the smoothing width, at redshift z = 0.

4.2 Asymmetry and height distribution

So far we have considered n_1 , without distinguishing between minima and maxima. In Fig. 4, we examine the asymmetry between the abundances of minima and maxima as a function of the smoothing width for simulations 2 at redshift z = 0. There is a clear excess of N_{max} , which is significantly above the level of the shot-noise (grey area). The trend is reversed at larger scales for both DM non-Gaussian models, but it becomes immersed in the shot-noise. Results of the three individual runs for the Gaussian DM simulation are also shown. It is clear that the shot-noise estimated theoretically as described above (grey area) is consistent with the scatter in the individual runs.

We explore the probability density distribution (PDF) of the value of the densities at the minima and maxima. We define

$$\nu = \frac{\delta}{\sigma_0}$$
 and $\nu_{\ln} = \frac{\ln(1+\delta)-\mu}{\sigma_{\ln}}$, (12)

where σ_0 is the rms of density field all over space, while μ and σ_{\ln} are the mean and rms of the values of $\ln(1+\delta)$ at either the minima or maxima. The quantity, v_{ln} is motivated by the result that the PDF of the density field is well approximated by a lognormal distribution (e.g. Coles & Jones 1991; Kofman et al. 1994). In Figs 5 and 6, we plot the PDF of ν (top) and ν_{ln} (bottom) for a smoothing of $R = 20 h^{-1}$ Mpc and $80 h^{-1}$ Mpc for simulations 2. For $R = 20 h^{-1}$ Mpc, mean values $\langle v \rangle$ measured in the simulations are in the range [-1.45, 1.38] for dips and [1.98, 2.02] for peaks. The corresponding numbers for $R = 80 h^{-1}$ Mpc are [-1.83, -1.74]and [1.90 - 1.98]. Each colour band corresponds to the Gaussian and 2 non-Gaussian simulations, respectively, for haloes and DM, as indicated in the figure. The band in each case is bounded from below and above by the minimum and maximum value of the corresponding PDF. The tight widths of the bands imply that PDF of densities at either maxima or minima is weakly sensitive to whether the initial conditions were Gaussian or not. This is expected given that corrections arise at order $f_{\rm NL}^2$ as noted above. Given the tight width of each band and for the sake of clarity, the plot does not indicate which of the individual simulations are shown (i.e. whether Gaussian or not).



Figure 5. *Top:* The PDF of v at minima and maxima in simulations 2, as indicated in the figure. The black curve is the theoretical prediction for P(v) given in BBKS. *Bottom:* The PDF of v_{ln} at minima and maxima for simulations 2. Here, the black line is a Gaussian with zero mean and unit variance.

For $R = 20 h^{-1}$ Mpc, the BBKS theoretical prediction for P(v) (expression 4.3 in their paper) shown as the black in the top panel, is a poor fit to any of the PDFs measured in the simulations. However, $P(v_{ln})$ for the DM density field (dotted), exhibit only minor differences at the tails, where the PDF for maxima is slightly skewed to positive values relative to the Gaussian (black in the bottom panel), the distribution at minima is negatively skewed. The rather small differences between the PDF from the haloes and the corresponding DM are due to the deviations from linear biasing. Fig. 6 shows the same results but for $R = 80 h^{-1}$ Mpc. This large smoothing greatly reduces the effect of non-linear evolution, bringing the BBKS theoretical PDF (black curve, top panels) closer to the measured PDF than it is for $R = 20 h^{-1}$ Mpc. We have checked the shape of DM PDF(v) approaches the BBKS prediction as the smoothing length is increased beyond $R = 80 h^{-1}$ Mpc. The lognormal curve (black,



Figure 6. The same as the previous figure but for $R = 80 h^{-1}$ Mpc.

bottom) remains a good fit to $P(v_{\rm in})$ for the DM, although not as good as in the smaller smoothing. It is interesting that the lognormal describes the halo PDF fairly well for this smoothing. At $R = 20 h^{-1}$ Mpc and $80 h^{-1}$ Mpc, non-Gaussian $f_{\rm NL}$ signatures in the halo bias are small (cf. equation 7). This explains the similarity between the halo PDFs in the simulations irrespective of the initial statistic.

5 ABUNDANCE OF EXTREMA AS A COSMOLOGICAL TEST

We offer a preliminary assessment of using total number of peaks and dips as a test of the initial conditions. Therefore, this complements the study of Croft & Gaztanaga (1998) and De & Croft (2007, 2010), who focused on the shape of the linear power spectrum and that of Codis et al. (2013), who did not consider halo density field and therefore the non-Gaussian bias. A proper analysis should take into account the covariance between the abundances corresponding to different smoothing scales. However, this task is beyond the



Figure 7. Abundance of dips and peaks as a cosmological test for estimating $\Omega_{\rm m}$ from a survey such as *Euclid*. Values of $\Delta \chi^2 = 1$ correspond to 1σ limits from the fiducial value of $\Omega_{\rm m}$.

scope of the current paper. Instead, we will focus on the expected discriminatory power of extrema abundance at distinct scales. As an example, we consider the *Euclid* mission (Laureijs et al. 2011), which will target emission line galaxies in the redshift range 0.9 < z < 1.8 across ~35 per cent of the sky. For Planck's cosmological parameters, the corresponding survey volume is $48(h^{-1} \text{ Gpc})^3$. Furthermore, the typical host halo mass is ~ $10^{11-12} h^{-1} M_{\odot}$, in broad agreement with the minimum halo mass resolved in simulations 1.

We wish to assess the ability that a measured total number N of extrema in a survey can reject a certain model given the hypothesis of an assumed fiducial underlying model. For this purpose, we assume that N follows a Poisson distribution

$$P_{\bar{N}}(N) = \frac{\bar{N}^N}{N!} e^{-\bar{N}},\tag{13}$$

where \bar{N} is the mean number expected in a particular given model. Given an observed N, the preferred of two competing models with expected mean numbers \bar{N}_1 and \bar{N}_2 , respectively, is determined by

$$D_{\bar{N}_1\bar{N}_2} = -2\ln\frac{P_{\bar{N}_1}}{P_{\bar{N}_2}} = 2N\ln\frac{\bar{N}_2}{\bar{N}_1} + 2(\bar{N}_1 - \bar{N}_2).$$
(14)

The mean value of *D* over all measurements, which we loosely denote by $\Delta \chi^2$ is

$$\Delta \chi^2 = \sum_{N} P_{\bar{N}} D_{\bar{N}_1 \bar{N}_2} = 2\bar{N} \ln \frac{\bar{N}_2}{\bar{N}_1} + 2(\bar{N}_1 - \bar{N}_2), \tag{15}$$

where we have used $\sum_{N} P_{\bar{N}}(N) = 1$ and $\sum_{N} N P_{\bar{N}}(N) = \bar{N}$. For $\bar{N}_2 = \bar{N}$, the quantity $\Delta \chi^2$ yields the confidence level with which a model with \bar{N}_1 can be rejected if the underlying model is \bar{N} . We use this statistic to assess whether the abundance of dips and peaks can be used to reject certain models given a Gaussian cosmological model with fiducial cosmological parameter. We focus on $\Omega_{\rm m}$ and $f_{\rm NL}$, separately.

Fig. 7 examines $\Delta \chi^2$ as a function of the matter density $\Omega_{\rm m}$. Here, \bar{N} is computed using equations (1)–(3) for fiducial DM linear power spectrum with the cosmological parameters corresponding to simulations 1. The same parameters with the exception of $\Omega_{\rm m}$ are used in the same expression to derive \bar{N}_1 . This figure therefore refers



Figure 8. The same as the previous figure but for $f_{\rm NL}$ instead of $\Omega_{\rm m}$.

to a Gaussian model ($f_{\rm NL} = 0$) and, in addition to DM density fields, it is also relevant for haloes with linear constant bias with respect to the DM. Only two filtering scales are considered, as indicated in the figure. It is remarkable that for $R = 50 h^{-1}$ Mpc, the 1σ level ($\Delta \chi^2 = 1$) is at $\Delta \Omega \approx \pm 0.01$ from the fiducial $\Omega_{\rm m} = 0.3$. It should be pointed out that for the Λ CDM linear power spectrum the abundance on a filtering scale given in h^{-1} Mpc is degenerate with respect to $\Omega_{\rm m}h$ and $\Omega_{\rm b}h$. This sensitivity to $\Omega_{\rm m}$ declines rapidly at $R = 100 h^{-1}$ Mpc due to the $1/R^3$ dependence of the number of dips and peaks.

The sensitivity to $f_{\rm NL}$ is demonstrated in Fig. 8 plotting $\Delta \chi^2$ with \bar{N} from the fiducial model and \bar{N}_1 for $f_{\rm NL} \neq 0$ but with all other parameters fixed at the fiducial values. These curves refer to filtered halo distribution where the theoretical expressions in equations (5)–(7) are used in equation (1) to derive the mean number of dips and peaks \bar{N} in a *Euclid* volume survey at z = 1. In these calculations, we consider a halo mass distribution consistent with simulations 1, with a minimum mass of $3.67 \times 10^{12} h^{-1} M_{\odot}$. For this mass threshold, we have seen in the previous section that the theoretical predictions are in reasonable agreement with the simulations. The sensitivity to $f_{\rm NL}$ is improved for the larger filtering widths, R, because of the stronger $f_{\rm NL}$ -dependence of halo bias on larger scales. For $R = 300 h^{-1}$ Mpc, we find $\Delta \chi^2 = 1$ for deviations $\Delta f_{\rm NL} \approx \pm 25$. This is encouraging especially if combined with measurements as a function of filtering scales and different halo masses.

6 DISCUSSION AND CONCLUSIONS

Locating points of maxima and minima is straightforward even for 3D density fields estimated from realistic galaxy redshift surveys. Since the total abundance is computed irrespective of height, it should be robust against the details of how the density field is estimated from the data. The total abundance is also insensitive to redshift space distortions that, in any case, can be modelled with standard perturbation theory for smoothing widths $R \gtrsim 50 h^{-1}$ Mpc (along the lines of Lam, Desjacques & Sheth 2010; Codis et al. 2013). Further, it depends explicitly only on the shape of the power

spectrum. Any dependence on the amplitude (e.g. σ_8) is indirectly encoded in non-linear corrections to the shape of the gravitationally evolved power spectrum. The lack of sensitivity of this abundance statistics on the amplitude thereby implies that it breaks most of the degeneracy between $f_{\rm NL}$ and the primordial amplitude of scalar perturbations, which arises in measurements of galaxy clusters counts and shear peaks in weak lensing maps for instance (this degeneracy can also be broken by combining clusters and voids; see Kamionkowski, Verde & Jimenez 2009). We have demonstrated that a primordial non-Gaussianity of the local- $f_{\rm NL}$ type imprints a strong signal in the abundance of peaks and dips of the halo density field owing to the non-Gaussian bias. An important result of the current paper is that the **BBKS** prediction derived for Gaussian density field can account for this effect reasonably well, provided that the matter power spectrum is replaced by the halo power spectrum. Therefore, the abundance of peaks and dips (a one-point statistics) is sensitive to the scale-dependent bias in the halo power spectrum (a two-point statistics), such as the covariance of cluster counts (Cunha, Huterer & Doré 2010). This effect disappears when the density field perfectly traces the matter distribution as is the case for shear peaks for instance.

We have made a preliminary assessment of the applicability of the total abundance statistics as a test of $f_{\rm NL}$ for a survey with specifications similar to those of the Euclid mission (Laureijs et al. 2011). From a measurement at a single smoothing scale R, we obtain an uncertainty of $\Delta f_{\rm NL} = 25$ (for $R = 300 \, h^{-1}$ Mpc) and 40 (for $R = 50 h^{-1}$ Mpc). This suggests that a measurement combining different smoothing scales and halo masses should be able to achieve a sensitivity of $\Delta f_{\rm NL} \leq 10$. While the sensitivity of this approach will likely be worse than the limits set by the latest cosmic microwave background measurements from Planck, $f_{\rm NL} = 0.8 \pm 5$ (Planck Collaboration XVII 2016), this approach should be competitive with galaxy clusters and shear peak counts in weak-lensing maps, for which the forecasted uncertainty is $\Delta f_{\rm NL} \sim 9$ (e.g. Pillepich, Porciani & Reiprich 2012, for a galaxy survey like EROSITA) and $\Delta f_{\rm NL} \sim$ 13 (e.g. Marian et al. 2011, for a weak-lensing survey with Euclid specifications), respectively. However, our approach may also be affected by the Eddington bias that plagues galaxy cluster counts or shear peaks. Namely, additive noise in the data will presumably increase the number of peaks while reducing the number of dips, which would mimic a small positive $f_{\rm NL}$. We will defer a more detailed study of this effect to future work.

The abundance of extrema depends on the cosmological parameters of the background cosmology. Here, we explored the dependence on $\Omega_{\rm m}$ alone with very encouraging results of an accuracy at the level of $\Delta \Omega_{\rm m} \sim 0.01$. For a given filtering scale given in h^{-1} Mpc, the abundance dependence nearly degenerates with the combination $\Omega_{\rm m}h$. Thus, this result regarding $\Omega_{\rm m}$ could alternatively by expressed as an accuracy of 0.7 km s⁻¹ on H_0 if all other parameters are fixed.

In principle, the moments can be estimated directly from the density field. Their ration could then be used to extract cosmological information. We point out that computing the moments directly requires taking derivatives of the (smoothed) density fields, a procedure associated with enhancement of the noise. In any case, extrema counts and direct moment estimation are two independent measurements, linked only via theory. It would be of interest to assess this link by comparing the outcome of the two measurements from the observations. After the preparation of this manuscript, we have realized that according to the expressions in equations (1) and (2), the number of extreme is formally infinite for a Top-Hat smoothing of a density field with the power spectrum, p(k), of the standard Λ CDM.

The reason is the divergent contribution of high-*k* modes to spectral moment, σ_2 , for *any value* of the width, R_{TH} , of the Top-Hat window function. In contrast, the moment σ_1 is finite containing negligible contribution from modes $k \gg R_{\text{TH}}$. This motivates a mixed smoothing strategy where the density field is smoothed with a Gaussian window of width R_G and additional Top-Hat smoothing of width $R_{\text{TH}} > R_G$. Therefore, σ_2 would be sensitive to R_G while σ_2 to R_{TH} . This smoothing protocol yields greatly higher extreme counts than a single Gaussian smoothing and will consequently yield much tighter constraints on the f_{NL} and other parameters of interest than the preliminary tests given here. We will explore all this in a subsequent paper, where we will present application to available data.

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