Weak lensing skew-spectrum

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ABSTRACT

We introduce the skew-spectrum statistic for weak lensing convergence maps and test it against state-of-the-art high-resolution all-sky numerical simulations. We perform the analysis as a function of source redshift and smoothing angular scale for individual tomographic bins. We also analyse the cross-correlation between different tomographic bins. We compare the numerical results to fitting-functions used to model the bispectrum of the underlying density field as a function of redshift and scale. We derive a closed form expression for the skew-spectrum for gravity-induced secondary non-Gaussianity. We also compute the skew-spectrum for the projected κ inferred from cosmic microwave background (CMB) studies. As opposed to the low redshift case, we find the post-Born corrections to be important in the modelling of the skew-spectrum for such studies. We show how the presence of a mask and noise can be incorporated in the estimation of a skew-spectrum.

Key words: Cosmology – Weak Lensing – Methods: analytical, statistical, numerical.

1 INTRODUCTION

Recently completed cosmic microwave background (CMB) experiments such as the Planck Surveyor1 (Planck Collaboration XVI 2014; Planck Collaboration VI 2018), have helped establishing a standard model of cosmology, with the baseline cosmological parameters now known with an unprecedented accuracy. However, many fundamental questions in cosmology remain open. These include the nature of dark matter and dark energy, a possible modification of General Relativity on cosmological scales (Clifton et al. 2012; Joyce et al. 2014) and the nature of neutrinos mass hierarchy (Lesgourgues & Pastor 2006). Next generation of large-scale surveys will provide a massive amount of high-precision data carrying complementary information that can help answer at least some of these questions. Indeed, observational programs of many ongoing as well as future surveys including the surveys e.g. Euclid2 (Laureijs et al 2006), CFHTLS,3 PAN-STARRS,4 Dark Energy Surveys5 (Allam et al. 2016), WiggleZ6 (Drinkwater et al. 2010), Rubin Observatory,7 (Tyson et al. 2003), BOSS8 (Eisenstein et al. 2011), KIDS (Kuijken et al. 2015), Roman Space Telescope (National Research Council 2010), lists weak lensing as their main science driver. From the early days of detection weak lensing (see e.g. Munshi et al. 2008 for a review) studies have now reached a level of maturity. Surveys such as Euclid will constrain the cosmological parameters with sub-per cent accuracy and answer many of the most challenging questions that cosmology is facing today.

Weak lensing at smaller angular scales probes scales that are in the highly non-linear regime and contains a wealth of cosmological information. This gravity-induced non-linearity (Bernardeau et al. 2002) introduces mode-coupling that is responsible for the resulting departure from Gaussianity (Bartolo et al. 2004). Higher order statistics beyond power spectrum estimation is typically used in exploitation of the information content of weak lensing maps. An accurate modelling of higher order statistics is important for modelling the covariance of the lower order estimators as well as to break cosmological parameter degeneracy. Early studies of higher order statistics concentrated on cumulants (Bernardeau 1994a, b) in real-space (Bernardeau, Mellier & van Waerbeke 2002; Bernardeau, van Waerbeke & Mellier 2003). Future surveys such as Euclid will have a near all-sky coverage and thus enable quantifying higher order statistics in the harmonic domain where measurements of individual modes will be less correlated (Amendola et al. 2013).

Most theoretical modelling in the highly non-linear regime were based on perturbative calculations or its extensions (Bernardeau et al. 2002), variants of the halo models (Cooray et al. 2002), effective field theory (EFT; Baumann 2012), or fitting-functions that are calibrated from simulations (Scoccimarro & Frieman 1999; Gil-Marín et al. 2012). Many different estimators are currently available for analysing departures from Gaussianity, including morphological estimators (Munshi et al. 2012), position-dependent power spectra (Munshi et al. 2020b), line-correlations (Eggemeier & Smith 2017), extreme value statistics (Harrison & Coles 2011), peak-statistics (Kacprzak et al. 2016; Shan et al. 2018), void statistics (Krause et al. 2013), and probability distribution functions (Codis et al. 2016; Uhlemann et al. 2016; Valageas 2016; Gruen et al. 2018).

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The measurements of angular correlations have the benefit of simplicity especially while the survey geometry (mask) is complicated (Munshi & Jain 2001, 2000; Munshi 2000). However, such measurements are highly correlated. In contrast, the measurements in the harmonic domain are less correlated and contain independent information especially when the sky coverage is high. One of the motivation of this study is to develop analytical predictions for one such proxy statistics to the bispectrum called skew-spectrum (Cooray 2001; Munshi & Heavens 2010) and test them against state-of-art numerical simulations. We will borrow the concepts developed for constructing skew-spectrum for the study of non-Gaussianity in the context of cosmic microwave background (CMB) observations by WMAP9 (Smidt et al. 2010; Calabrese et al. 2010) and Planck (Planck Collaboration XIII 2016a, b) satellites. However our aim here is also to include gravity-induced secondary non-Gaussianity. The skew-spectrum is the lowest order member in the family of higher order spectrum is the lowest order member in the family of higher order to include gravity-induced secondary non-Gaussianity. The skew-spectrum is the lowest order member in the family of higher order spectrum (Munshi et al. 2008). However, the treatment eventually breaks down when δ becomes non-linear (δ ≥ 1). Expanding δ in a Fourier series, and assuming δ ≤ 1, for the perturbative series to be convergent, we get
\[ \delta(k) = \delta^{(1)}(k) + \delta^{(2)}(k) + \delta^{(3)}(k) + \ldots; \]  
\[ \delta^{(2)}(k) = \int \frac{d^3k_1}{2\pi} \int \frac{d^3k_2}{2\pi} \delta \cdot (k_1 + k_2 - k) F_2(k_1, k_2) \delta^{(1)}(k) \]  
\[ \times \delta^{(1)}(k'); \]  
\[ F_2(k_1, k_2) = \frac{5}{7} + \frac{1}{2} \left( \frac{k_1}{k_2} + \frac{k_2}{k_1} \right) \left( \frac{k_1 \cdot k_2}{k_1 k_2} \right) + \frac{2}{7} \left( \frac{k_1 \cdot k_2}{k_1 k_2} \right)^2. \]  

The linearized solution for δ will be denoted by δ^{(1)}(k); higher order terms δ^{(2)}, δ^{(3)}, ... will denote the second- and third-order corrections to the linear solution. The 3D wave vectors are denoted as k, k_1, k_2, and their magnitudes as k = |k| and k_i = |k_i|. More details of our Fourier convention will be introduced in Section 3.2. Our Eulerian formalism is based on a perfect-fluid approach. This is valid at large scales (i.e. before shell crossing). We will be taking the fitting-function approach in the non-perturbative regime. In recent years many new methods have been developed to tackle the gravitational instability in the non-linear regime including EFT methods (see e.g. Munshi & Regan 2012 and references therein).

### 2.2 Phenomenological fitting-functions

Beyond the quasi-linear regime non-perturbative tools become necessary. One such approach was developed in Scoccimarro & Frieman (1999) who proposed the so-called Hyper Extended Perturbation Theory (HEPT) in the highly non-linear regime and a fitting-function that connects it with the tree-level perturbative calculation. The fitting function which interpolates these two regime is calibrated using numerical simulations. Over the years similar but more accurate fitting formula were developed by other authors (Gil-Marin et al. 2012), which essentially generalize the kernel F_2 defined in equation (1c) by introducing scale-dependent coefficients a(n_i, k_i), b(n_j, k_j), and c(n_i, k_j):
\[ F_2(k_1, k_2) = \frac{5}{7} a(n_i, k_i) a(n_j, k_j) + \frac{1}{2} \left( \frac{k_1 \cdot k_2}{k_1 k_2} \right) \left( \frac{k_1}{k_1} + \frac{k_2}{k_2} \right) \]  
\[ \times b(n_i, k_i) b(n_j, k_j) + \frac{2}{7} \left( \frac{k_1 \cdot k_2}{k_1 k_2} \right)^2 c(n_i, k_i) c(n_j, k_j). \]

Here n_e is the effective spectral slope associated with the linear power spectra n_e = d ln P(z)/d ln k. \( k_{nl} \) is the ratio of a given length scale to the non-linear length scale \( q = k / k_{nl} \), where \( k_{nl}^2 / 2\pi^2 D_z^2(z) P_{nl}(k_{nl}) = 1. \) Here \( D_z(z) \) represents the linear growth rate of perturbations at redshift z. At length scales where \( q < 1 \), the relevant length scales are well within the quasi-linear regime, \( a = b = c = 1 \), and we recover the tree-level perturbative results. In the regime where \( q \gg 1 \), and the length scales we are considering are well within the non-linear scale, we recover \( a \neq 1 \) but \( b = c = 0 \). In this limit, the bispectrum becomes independent of configuration. It was recently pointed out by Munshi et al. (2020a) that the fitting function of Gil-Marin et al. (2012) is not very accurate in describing the weak lensing bispectrum. A more accurate fitting function was developed recently in Takahasi et al. (2020) which we will be using in this study. The original fitting function by (Scoccimarro & Frieman 1999) involved

9https://map.gsfc.nasa.gov/
just six free parameters and was valid for \( k < 3 \text{h} \text{Mpc}^{-1} \) and \( 0 < z < 1 \). The improved (Gil-Marín et al. 2012) formula has a limited range of validity \( k < 0.4 \text{h} \text{Mpc}^{-1} \) and \( 0 < z < 1.5 \) and contains nine parameters. The fitting function by Takahasi et al. (2019) contains 52 free parameters. Introduction of such a large number of free parameters increases the validity range to \( k = 10 \text{h} \text{Mpc}^{-1} \) and \( z \approx 1–3 \). This will be important in modelling non-Gaussianity on arcminute scales probed by the future stage-IV experiments.

### 3 WEAK LENSING STATISTICS IN PROJECTION

In this section, we will relate the convergence bispectrum \( \kappa \) with its 3D density contrast \( \delta \) counterpart. Then this bispectrum will be used to construct the convergence skew-spectrum.

#### 3.1 Projected weak lensing bispectrum

The weak lensing convergence \( \kappa \) and 3D density contrast \( \delta \) is related by the following expression:

\[
\kappa(\theta) = \int_0^{r_{\text{sky}}} dr \, w(r) \delta(r, \theta); \\
\delta(r) = \frac{32\pi M}{2 \Omega_1 H_0^2 c^2} dA_r dA_{r-l} dA_{r-b};
\]

We have introduced a kernel \( w(r) \) above. In this expression, \( r = |r| \) is the comoving radial distance to the source, \( \theta \) describes the angular position on the sky, \( \Omega_1 \) is the cosmological matter density parameter (total matter density in units of the critical density), \( H_0 \) is the Hubble constant, \( c \) is the speed of light, \( \alpha = 1/(1 + z) \) is the scale factor at a redshift \( z \), \( dA_r \) represents the comoving angular diameter distance at a distance \( r \) and \( dA_{r-l} \) is the comoving radial distance to the source plane. The corresponding redshift will be represented by \( z_c \).

To keep the analysis simple, in our study we will ignore the source distribution and assume them to be localized on a single source plane defined by \( z_c \). We will study various statistics as a function of \( z_c \). To simplify the analysis, we will also ignore photometric redshift errors. Throughout we will use the flat-sky approximation (Kilbinger et al. 2017; Kitching et al. 2017). Needless to say, such complications are essential to link predictions to observational data, and will be presented in an accompanying study.

Fourier decomposing \( \delta \) along and perpendicular to the line-of-sight direction we obtain:

\[
\kappa(\theta) = \int_0^{r_{\text{sky}}} dr \omega(r) \left( \frac{dk}{2\pi} \int \frac{d^2k}{(2\pi)^2} \phi(k, \theta) \right) \]

Here, we have decomposed the 3D wavenumber \( k \) along and perpendicular to the radial direction, \( k = (k_1, k_2, k_3) \). We have used the following convention for the 3D Fourier Transform and its inverse:

\[
\delta(k) = \frac{1}{(2\pi)^3} \int d^3r \exp(-ik \cdot r) \delta(r); \\
\delta'(r) = \int d^3k \exp(i k \cdot r) \delta(k).
\]

The corresponding 3D power spectrum and bispectrum for \( \delta \) are:

\[
\langle \delta(k_1) \delta(k_2) \rangle = (2\pi)^3 \delta_{\text{BD}}(k_1 + k_2) P_\delta(k_1), \\
\langle \delta(k_1) \delta(k_2) \delta(k_3) \rangle = (3\pi)^3 \delta_{\text{BD}}(k_1 + k_2 + k_3) B_\delta(k_1, k_2, k_3).
\]

Using the small-angle approximation the projected power spectrum \( P^s(l) \) and bispectrum \( B^s(l_1, l_2, l_3) \) of the convergence field \( \kappa \) can be expressed respectively in terms of the 3D power spectrum \( P_\kappa(k) \) and bispectrum \( B_\kappa(k_1, k_2, k_3) \):  

\[
P^s(l) = \int_0^{r_{\text{sky}}} dr \alpha^3 \frac{dA_r}{dA_{r-l}} \frac{P_\kappa(k)}{dA_{r-b}} \frac{l}{dA_r}; \\
B^s(l_1, l_2, l_3) = \int_0^{r_{\text{sky}}} dr \frac{d^3r}{dA_r} B_\kappa \left( \frac{l_1}{dA_r}, \frac{l_2}{dA_r}, \frac{l_3}{dA_r} \right).
\]

Detailed derivations of these expressions can be found in Munshi et al. (2008). Cross-correlating tomographic bins can be used to define cross-spectra \( P_{\alpha \beta}^{s} \) and cross-skew spectra \( B_{\alpha \beta}^{s} \):  

\[
P^s_{\alpha \beta}(l) = \int_0^{r_{\text{sky}}} dr \frac{d^3r}{dA_r} \frac{P_\kappa(r \omega_{\alpha \beta})}{dA_{r-b}} \frac{l}{dA_r}; \\
B^s_{\alpha \beta}(l_1, l_2, l_3) = \int_0^{r_{\text{sky}}} dr \frac{d^3r}{dA_r} B_\kappa \left( \frac{l_1}{dA_r}, \frac{l_2}{dA_r}, \frac{l_3}{dA_r} \right)
\]

The integration takes contribution only from the overlapping redshift range of the two bins. Thus, the upper limit extends only to the source plane defined by the lower redshift \( r_{\text{min}} = \min(r_a, r_b) \). Notice that \( P_{\alpha \beta} = P_{\alpha \alpha} \) but \( B_{\alpha \beta} \neq B_{\alpha \alpha} \) and they carry independent information. It is possible to directly deal with shear bispectrum and relate them to density bispectrum thus avoiding the map making process. See Munshi et al. (2011d) for bispectra constructed for higher spin objects, i.e. shear as well as flexions.

#### 3.2 Skew-spectrum in all-sky and flat-sky

The skew-spectrum statistic for \( \kappa \) is constructed by cross-correlating the squared \( \kappa \) with itself. We start by introducing the spherical harmonic transform of a convergence map \( \kappa(\Omega) \).

\[
\kappa(\ell) = \int d\Omega Y_\ell^m \kappa(\Omega) \approx \frac{32\pi M}{2 \Omega_1 (H_0 c)^2} \int \frac{d^2k}{(2\pi)^2} Y_\ell^m(k) \phi(k, \theta)
\]

\[
The corresponding reduced bispectrum and skew-spectrum is identical to \( C^s_{\ell} \) at high \( \ell \) with the identification \( \ell \rightarrow \ell' \). The weak lensing \( \kappa \) maps are highly non-Gaussian. The bispectrum characterizes departure from Gaussianity and is defined as the three-point coupling of harmonic coefficients. The statistics beyond bispectra, e.g. the trispectra and its higher order analogues are increasingly noise dominated. By assuming isotropy and homogeneity the all-sky bispectrum \( B^s_{\ell_1 \ell_2 \ell_3} \) is defined as

\[
\langle \kappa_{\ell_1 \ell_2} \kappa_{\ell_2 \ell_3} \kappa_{\ell_3 \ell_1} \rangle_c = \sum_{\ell} B^s_{\ell_1 \ell_2 \ell_3} \left( \frac{\ell_1}{\ell} \right) \left( \frac{\ell_2}{\ell} \right) \left( \frac{\ell_3}{\ell} \right)
\]

The Wigner-3j symbol appearing above in parentheses ensures rotational invariance. It is only non-zero for the triplets \( (\ell_1, \ell_2, \ell_3) \) that satisfy the triangular condition and \( \ell_1 + \ell_2 + \ell_3 \) is even. The reduced bispectrum \( B^s_{\ell_1 \ell_2 \ell_3} \) is useful in directly linking the all-sky bispectrum and its flat-sky counterpart. For the convergence field \( \kappa \)
Here $k^2_{\ell m}$ is defined through the following expression:

$$B_{\ell_1 \ell_2 \ell_3}^{k^2_{\ell m}} := \sqrt{\frac{(2\ell_1 + 1)(2\ell_2 + 1)(2\ell_3 + 1)}{4\pi}} \left( \begin{array}{ccc} \ell_1 & \ell_2 & \ell_3 \\ 0 & 0 & 0 \end{array} \right) b_{\ell_1 \ell_2 \ell_3}. \quad (12)$$

Finally we are in a position to define the skew spectrum as the cross power-spectra formed by cross-correlating the squared $k^2(\Omega)$ maps against the original map $k(\Omega)$.

$$S_\ell := \frac{1}{2\ell + 1} \sum_m \text{Real} \left\{ \langle k^2_{\ell m} k_{\ell m} \rangle \right\} = \sum_{\ell_1 \ell_2 \ell_3} B_{\ell_1 \ell_2 \ell_3}^{k^2_{\ell m}} J_{\ell_1 \ell_2 \ell_3}; \quad (13a)$$

$$J_{\ell_1 \ell_2 \ell_3} := \sqrt{\frac{(2\ell_1 + 1)(2\ell_2 + 1)(2\ell_3 + 1)}{4\pi}} \left( \begin{array}{ccc} \ell_1 & \ell_2 & \ell_3 \\ 0 & 0 & 0 \end{array} \right). \quad (13b)$$

Here $[k^2_{\ell m}]$ represents the harmonic multiplies computed using a harmonic decomposition of $k^2$ and * denotes complex conjugation. The commonly used (normalized) one-point skewness parameter $S_\ell = \langle k^2_{\ell} \rangle / \langle k^2 \rangle^2$ can be recovered from the skew-spectrum. The third-order moment ($k^3$) is given by:

$$\langle k^3(\theta_1) \rangle = \frac{1}{4\pi} \sum_\ell \langle 2\ell + 1 \rangle S_\ell \beta^3_\ell(\theta_1); \quad (14)$$

The smoothing beam (window) is denoted as $P_\ell(\cos \theta_1)$, and the skew-spectrum $S_\ell$ is related to the two-to-one correlation function $\xi_{12}^2$ in the real space. They are related by the following expression:

$$\xi_{12}^2(\theta_1) := \langle k^2(\theta_1) k(\theta_2) \rangle = \frac{1}{4\pi} \sum_\ell \langle 2\ell + 1 \rangle S_\ell \beta^3_\ell(\theta_1). \quad (15)$$

The flat-sky bispectrum is similarly defined through:

$$\langle (l_1, m_1) (l_2, m_2) (l_3, m_3) \rangle = \int d\Omega Y_{l_1 m_1}(\hat{\Omega}) Y_{l_2 m_2}(\hat{\Omega}) Y_{l_3 m_3}(\hat{\Omega}) B_{l_1 l_2 l_3}.$$  

The flat-sky bispectrum $B_{l_1 l_2 l_3}$ is identical to the reduced bispectrum $b_{l_1 l_2 l_3}$ for high multipole (Bartolo et al. 2004). This can be shown by using the following asymptotic relationship:

$$G_{l_1 m_1, l_2 m_2, l_3 m_3} = \int \frac{d\Omega Y_{l_1 m_1}(\hat{\Omega}) Y_{l_2 m_2}(\hat{\Omega}) Y_{l_3 m_3}(\hat{\Omega})}{4\pi} \frac{(2\ell_1 + 1)(2\ell_2 + 1)(2\ell_3 + 1)}{4\pi} \left( \begin{array}{ccc} \ell_1 & \ell_2 & \ell_3 \\ 0 & 0 & 0 \end{array} \right) \approx (2\pi)^2 \delta_{2D}(l_1 + l_2 + l_3). \quad (17a)$$

The skew-spectrum in the flat-sky is given by (Pratten & Munshi 2012):

$$S(l_2) = \int_0^\infty \frac{d\mu}{2\pi \sqrt{1 - \mu^2}} \left( \frac{l_1}{l_2} \right) P_b \left( \frac{l_1}{l_2} \right) T(l_1 l_2) \approx \beta^2(\ell_1 \theta_1) \beta^2(\ell_2 \theta_2) T(l_1 l_2) \quad (19a)$$

$$T(l_1 l_2) = \left[ \frac{36}{7} \int_0^1 \frac{d\ell_3}{l_3} \left( \frac{l_1 l_2 l_3}{l_1^2 + l_2^2} \right) - 3 \left( \frac{l_1}{l_2} \right)_l + l_2 \right] \int I_0(l_1 l_2 \theta_1^2) + 6 I_2(l_1 l_2 \theta_1^2). \quad (19b)$$

The angular integral in equation (18) can be done analytically using the modified Bessel functions represented in $I_m$. To simplify the notation we adopt a parametrization in terms of the variables $C_m^{\ell\ell}$. Here a few comments about the skew-spectrum are in order. The skewness has a high signal to noise. However, it lacks distinguishing power. Therefore, such statistics cannot distinguish various contributions, e.g., from primordial non-Gaussianity or non-Gaussianity from intrinsic alignment of source galaxies from the gravity-induced secondary non-Gaussianity. The skew-spectrum, on the other hand, retains some of the information regarding the shape of the spectrum, thus it can, in principle, allow us to separate various contributions or remove possible source of contamination from systematics.

We consider a direct estimator in the framework. Optimality can be achieved by using suitable weights (Komatsu, Spergel & Wandelt 2005; Munshi & Heavens 2010). However, optimality is not of crucial importance for analysing weak lensing maps as the secondary non-Gaussianity is expected to be detected with a very high signal to noise. A direct estimator, which is simpler to implement, will thus be useful for studying non-Gaussianity in weak lensing maps.

A few comments in order. The skewness, skew-spectrum, and bispectrum are all third-order statistics. However, their information contents are not the same. The skewness is an one-point statistics which can be constructed from the skew-spectrum. The skew-spectrum being a two-point statistics cannot however be constructed from the skewness. Similarly bispectrum is a Fourier (or harmonic) analogue of the three-point correlation function. Being a three-point statistics it cannot be constructed from the skew-spectrum and contains more information. The use of skew-spectrum can be treated as a data compression step as computation of the entire covariance matrix of the bispectrum can be a challenging task. Indeed the bispectrum, skew-spectrum, and skewness are equally sensitive to the second-order non-linearities. However, the skewness cannot distinguish among different contributions from various sources of non-linearities that the skew-spectrum is able to do. Physically skew-spectrum is the Fourier (or harmonic) transform of the two-to-one correlation function as defined in equation (15).

Next we consider skew-spectrum for specific models of bispectrum considered in Section 2.

### 3.3 The skew-spectrum in the tree-level standard perturbation theory (SPT)

Analytical predictions for the skew-spectrum for large smoothing angular scales can be obtained using perturbative calculations. However, for any arbitrary $\ell$ an order-by-order calculation is needed. We will obtain these results using a Gaussian smoothing beam where complete analytical results in closed form can be derived. We will consider the gravity-induced (secondary) non-Gaussianity.

$$S(l_2) = \int \frac{d\omega}{2\pi} \frac{\omega^3(r)}{d^3(r)} \int \frac{d\ell_1}{(2\pi)^2} P_b \left( \frac{l_1}{d_\ell(r)} \right) P_b \left( \frac{l_2}{d_\ell(r)} \right) \times \beta^2(l_1 \theta_1) \beta^2(l_2 \theta_2) T(l_1 l_2) \quad (19a)$$

$$T(l_1 l_2) = \left[ \frac{36}{7} \int_0^1 \frac{d\ell_3}{l_3} \left( \frac{l_1 l_2 l_3}{l_1^2 + l_2^2} \right) - 3 \left( \frac{l_1}{l_2} \right)_l + l_2 \right] \int I_0(l_1 l_2 \theta_1^2) + 6 I_2(l_1 l_2 \theta_1^2). \quad (19b)$$

The angular integral in equation (18) can be done analytically using the modified Bessel functions represented in $I_m$. To simplify the notation we adopt a parametrization in terms of the variables $C_m^{\ell\ell}$. The skewness has a high signal to noise. However, it lacks distinguishing power. Therefore, such statistics cannot distinguish various contributions, e.g., from primordial non-Gaussianity or non-Gaussianity from intrinsic alignment of source galaxies from the gravity-induced secondary non-Gaussianity. The skew-spectrum, on the other hand, retains some of the information regarding the shape of the spectrum, thus it can, in principle, allow us to separate various contributions or remove possible source of contamination from systematics.

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Next we consider skew-spectrum for specific models of bispectrum considered in Section 2.
Figure 1. Examples of realizations of all-sky weak lensing convergence or κ maps used for our study. The left- and right-panel panels correspond, respectively, source redshifts \( z_s = 2.0 \) and 0.5. The κ maps, we have used, were generated at a HEALPIX resolution \( N_{\text{side}} = 4096 \), and we have degraded them to \( N_{\text{side}} = 2048 \) for our study.

\[
S(l_2) = \sigma^2 \mathcal{P}_r(l_2) R_2 \int \frac{l_1 dl_1}{2\pi} \int \frac{l_2 dl_2}{2\pi} \beta^2(l_1 \theta_1) \beta^2(l_2 \theta_2) T(l_1 l_2);
\]

\[
R_2 = \int_0^{\ell_{\text{max}}} dr \frac{w^3(r)}{d^{3+m}(r)} D^2_r(z) / \left[ \int_0^{\ell_{\text{max}}} dr \frac{w^3(r)}{d^{3+m}(r)} D^2_r(z) \right]^2 .
\]

To separate the temporal \( r \) and angular \( l \) parts of the integral, we replaced the linear power spectrum \( \mathcal{P}_r(k) \) with a power-law form, i.e. \( P_r(k) = AD^2_r(z)k^n \). Due to the choice of normalization here the skew-spectrum is independent of the power spectrum amplitude \( A \). The resulting skewness \( S_1 \) can then be written as

\[
S_1 := \int l_1 dl_1 S_2(l_2) .
\]

The \( F_2 \) kernel for many modified gravity theories have a similar structural form and can be treated analytically. Similarly, the EFT based approaches introduces corrective terms to \( F_2 \) that too have a very similar form (Munshi & Regan 2012). The kernel describing the primordial non-Gaussianity can also be treated in a similar manner. We will focus on certain well-known cases of modified gravity theories. The analytical results for these models are important as there are no established numerical fitting-function available in these scenarios.

4 NUMERICAL SIMULATIONS

The simulated all-sky maps we have used in our study are available for public use (Takahashi et al. 2017).\(^{10}\) Ray-tracing through numerical (N-body) simulation were used to generate these maps. The underlying simulations followed the gravitational clustering of 2048\(^3 \) particles. The lensing maps were generated using multiple lens planes. A range of source redshifts were used \( z_s = 0.05-5.30 \) with \( \Delta z_s = 0.05 \). In our numerical study, we have used the maps with source redshifts \( z_s = 0.5, 1.0, 1.5, 2.0 \). For CMB maps, the lensing potentials were constructed using the deflection angles which were used to construct the lensing potentials and eventually the κ maps. In recent studies inclusion of post-Born terms in lensing statistics were outlined (Pratten & Lewis 2016). The maps we use include post-Born corrections. In this study, we will see that at a low source redshift such corrections do not play any significant role although they do play an important role at higher redshifts, e.g. in case of lensing of CMB. The healpix\(^{11}\) (Gorski et al. 2005) equal area pixelization scheme was used in generating these maps. The number of pixels denoted by \( N_{\text{pix}} \) scale as \( N_{\text{pix}} = 12N_{\text{side}}^2 \) where \( N_{\text{side}} \) is the parameter that defines the resolution which take values \( N_{\text{side}} = 2^N \) with \( N = 1, 2, \ldots \). The set of maps we use in this study are generated at \( N_{\text{side}} = 4096 \) and were cross-checked against higher resolution maps. These maps constructed at different \( N_{\text{side}} \) were found to be consistent with each other up to the angular harmonics \( \ell \leq 2000 \). Various additional tests were performed including the Electric/Magnetic (E/B) decomposition of the shear maps for the construction of κ maps (Takahashi et al. 2017). We have used high resolution maps generated at \( N_{\text{side}} = 4096 \) and degraded them to \( N_{\text{side}} = 2048 \) to analyse them for harmonic modes satisfying \( \ell < 2N_{\text{side}} \). The background cosmological parameters used for these simulations are: \( \Omega_{\text{CDM}} = 0.233, \Omega_b = 0.046, \Omega_M = \Omega_{\text{CDM}} + \Omega_b, \Omega_{\Lambda} = 1 - \Omega_M + h = 0.7 \). The amplitude of density fluctuation \( \sigma_8 = 0.82 \) and the spectral index \( n_s = 0.97 \). Fig. 1 shows examples of maps used in numerical study. Several authors have used these maps in the context of CMB lensing (Namikawa et al. 2019) and for weak lensing of galaxies at low redshifts (Munshi et al. 2020a; Munshi & McEwen 2020).

5 THE PSEUDO SKEW-SPECTRUM ESTIMATOR

An inverse covariance weighting is needed in the construction of an optimal maximum likelihood (ML) estimator or a quadratic maximum likelihood (QML) estimator (Efsthathiou 2004). For high resolution maps the size of the data vector, however, makes these estimators difficult to implement (Oh, Spergel & Hinshaw 1990). As a result many suboptimal estimators, which use heuristic weighting schemes, have been developed. The so-called pseudo-\( C_\ell \) (PCL) technique was introduced in Hivon et al. (2002) in the harmonic domain. Later a related correlation function based approach was introduced in Szapudi et al. (2001). These estimators are unbiased but are suboptimal. Typically various heuristic weighting depending on sky coverage, as well as noise characteristics can improve the optimality of these estimators typically in noise dominated high-\( \ell \) (or smaller angular scales) regime. The maximum likelihood estimators, on the other hand, can be efficiently used for larger smoothing scales. Different hybridization schemes can be used to combine the large

\(^{10}\)http://cosmo.phys.hirosaki-u.ac.jp/takahasi/alsky_raytracing/

\(^{11}\)https://healpix.jpl.nasa.gov/
angular scale (equivalently the low $\ell$) estimates using QML with small angular scale (high $\ell$) PCL estimates (Efthathiou 2004). In our study, we will use a direct pseudo-$C_\ell$ estimator for the skew-spectrum. The direct estimator from the masked sky $\hat{S}_\ell$ is related to the underlying all-sky $S_\ell$ skew-spectrum through a mode-mixing matrix $M_{\ell\ell}$ that depends on the mask.

$$\hat{S}_\ell = M_{\ell\ell} S_\ell; \quad \hat{S}_\ell = M_{\ell\ell}^{-1} \hat{S}_\ell; \quad \langle \hat{S}_\ell \rangle = S_\ell. \quad (22a)$$

Here $S_\ell^{(21)}$ denotes the skew-spectrum computed from a map in the presence of a mask $w(\Omega)$, $S_\ell^{(21)}$ is the all-sky estimate. The mode-coupling matrix $M_{\ell\ell}$ is given in terms of the power spectrum of the mask $w(\Omega)$ as follows:

$$M_{\ell\ell} = (2\ell + 1) \sum_{\ell'} \left( \begin{array}{ccc} \ell & \ell' & \ell'' \\ 0 & 0 & 0 \end{array} \right)^2 \left( \frac{2\ell'' + 1}{4\pi} \right) |w_{\ell''}|. \quad (23)$$

Here $w_\ell = 1/(2\ell + 1) \sum_m w_{\ell m}^2$ is the power spectrum of the mask constructed from the harmonic-coefficient $w_{\ell m}$ of the map. The coupling matrix $M_{\ell\ell}$ encodes the mode-mixing due to the presence of a mask. We have used this estimator for estimation of skew-spectrum from individual tomographic bins as well as cross-correlating two different bins. In case of cross-correlation we have used the same mask for the two different bins. Generalization of the PCL method were developed in Munshi et al. (2011a,b,d,e).

In our study, we have used the mask which is shown in Fig. 2. To construct this mask all pixels (shown in maroon) lying within 22 deg of either the galactic or ecliptic planes are discarded. The remaining unmasked pixels (shown in yellow) cover 14,490 deg$^2$ of the sky, making fraction of the sky covered $f_{sky} \approx 0.35$. Various aspects of noise involved in cross-correlating CMB lensing maps and galaxy lensing maps are discussed in Fabbian, Lewis & Beck (2019).

Typically, to construct an unbiased PCL estimator the noise contribution is subtracted from the total estimates. This however is not necessary for the construction of the skew-spectrum estimator as the bispectrum of a Gaussian noise is zero. However, presence of noise in the data does increase the variance of the estimator. We will not attempt to construct the covariance matrix of our estimator. Such a generalization will be presented in a future publication.

6 RESULTS

In this section, we discuss the numerical results presented in this paper. We have used all-sky simulations generated at $N_{side} = 4096$ for validating the theoretical predictions. We have used the simulations generated at lower redshifts for weak lensing studies along with the lensing maps generated at $z_s = 1100$ (last scattering surface). Examples of the maps and mask used is presented respectively in Figs 1 and 2. We have used these maps for constructing the skew-spectra at individual redshift as well as computing the skew-spectrum by cross-correlating two different redshifts. Below we list our findings.

(i) Skew-spectra from individual tomographic bins: First, we compute the theoretical skew-spectra $S_\ell$ using equation (21) as a function of harmonics $\ell$ for various smoothing angular scales as well as redshifts. The results for lower redshift bins are plotted in Fig. 3 and the results for the last scattering surface (LSS) is plotted in Fig. 4. In Fig. 3 the panels from left to right correspond to redshifts $z_s = 0.5, 1.0, 1.5,$ and 2.0. In each panel three different smoothing angular scales are considered, from top to bottom the curves correspond to FWHM of the Gaussian beam $\theta_s = 0.0, 5.0, \text{and} 10$ arcmin as indicated. For the CMB sky shown in Fig. 4 the

![Figure 3](https://example.com/fig3.png)

**Figure 3.** The skew-spectrum $S_\ell$ defined in equation (21) is being plotted for various smoothing angles and redshifts. The noisy purple lines correspond to results from the simulations. The panels from left to right correspond to redshifts $z_s = 0.5, 1.0, 1.5,$ and 2.0. In each panel, we show skew-spectra corresponding to Gaussian beams with full width at half maxima of $\theta_s = 0.0, 5.0, \text{and} 10$ arcmin as indicated. We have considered all-sky simulations and no noise was included. The theoretical predictions, shown in black, are computed using the expressions equations (13a)–(13b). The fitting function of Takahasi et al. (2019) was used throughout this study to model the gravity-induced bispectrum $B_\ell$. We use all modes below $\ell_{max} = N_{side}$ in the computation.
panels from left to right correspond to four different Gaussian beams \( \theta_s = 0.0, 5.0, \) and 10 arcmin. We have computed the skew-spectra using the Born approximation as well as including the post-Born correction terms (Pratten & Lewis 2016). We found that the post-Born corrections will be important in modelling the skew-spectra at high redshifts. However, for the lower redshifts we found this corrections to be negligible as expected. We have considered all-sky simulations and no noise was included. The theoretical predictions are computed using the expressions in equations (13a)–(13b). The fitting functions of Takahasi et al. (2019) was used throughout in this study to model the gravity-induced bispectrum \( B \). We use all modes below \( \ell_{\text{max}} = 2N_{\text{side}} \) in our computation. We have also removed all modes \( \ell_{\text{max}} < 100 \) from our computation. These fitting functions are found to be an excellent description of the simulated data.

(ii) Cross-correlating two tomographic bins: The skew-spectra computed by cross-correlating \( \kappa^2(\hat{\Omega}) \) and \( \kappa(\hat{\Omega}) \) from two different redshift bins is being plotted in Figs 5 and 6. In particular, squared \( \kappa_1 = \kappa(z_1) \) defined for a source redshift \( z_1 \) and \( \kappa_2 = \kappa(z_2) \) at redshift \( z_2 \) is being cross-correlated in the harmonic domain. For this plot we restrict ourselves to \( z_1 > z_2 \). We use the expression of mixed bispectrum given in equations (9b)–(9c) for computing the theoretical predictions. The expression for the estimator for the skew-spectrum is given in equation (21). From left to right-hand panels correspond to \( z_1 = 2.0, 1.5 \) and 1.0 respectively and various curves in each panel correspond to \( z_2 \) as indicated. The maps used were constructed at HEALPix resolution \( N_{\text{side}} = 4096 \). We have filtered all \( \ell \) \( > 2048 \) modes out before analysing them. No additional smoothing was considered. We do not include any noise due to intrinsic ellipticity distribution of galaxies. We have used one single all-sky realization to compute the skew-spectra and no mask was included. The skew-spectra constructed by cross-correlating \( \kappa^2 \) at the Last Scattering Surface (LSS, \( z_s = 1100 \)) and \( \kappa \) at lower redshift is presented in Fig. 7. Similarly, the skew-spectrum constructed using \( \kappa \) at LSS and \( \kappa^2 \) at lower redshift is presented in Fig. 8. We found that the post-Born correction is negligible in modelling the skew-spectrum constructed cross-correlating maps from two redshifts.

(iii) Accuracy of predictions: To quantify the difference of predicted skew-spectra and the one estimated from numerical simulation, we have used the following statistics:

\[
\Delta_b = \frac{1}{\sigma_b} \left[ \hat{S}_b - S^b \right].
\]

Here \( \hat{S}_b \) represents the binned theoretical skew-spectrum and \( S^b \) is the estimated binned skew-spectrum from numerical simulation and \( \sigma_b \) is the standard deviation of the fluctuations in individual \( \ell \) modes within a bin. We have chosen a bin-size of \( \delta_b \). The results are shown in Fig. 9. The left-hand panel shows the errors in skew-spectra obtained by cross-correlating \( \kappa^2_{\text{SS}} \) and low redshift \( \kappa \) (upper curves) and their symmetric counterparts (lower curves). The fitting functions underpredict the simulation results for \( \kappa^2_{\text{LSS}} \kappa^2 \). The difference between theory and simulation is lowest for \( z_s = 0.5 \) and increases with the redshift. For \( z_s = 2.0 \) it can be as high as \( 1.5\sigma_b \). The results are more pronounced for the intermediate bins. The middle- and right-hand panels of Fig. 9 depicts \( \Delta_b \) for \( \kappa^2 \) (middle-panel) and \( \kappa^2 \kappa, \) respectively. The difference is highest for skew-spectra involving \( z_s = 0.5 \) and lower for higher redshift \( z_s = 2.0 \). The \( \Delta_b \) can reach a value of 2.5 for lower redshifts. The theory typically underpredicts the data.

The individual skew-spectral bins are correlated as the skew-spectrum is an integrated measure, i.e. individual \( \ell \) modes (bins) depend on the entire range of \( \ell \) modes (bins). So a straightforward \( \chi^2 \) analysis (using a diagonal covariance matrix) is not possible. Nevertheless, notice that we have considered noise-free simulations in characterization of errors. Inclusions of noise will increase \( \sigma_b \) and decrease \( \Delta_b \). We have considered full-sky maps but, inclusion of the masks will increase the scatter and thus further reduce the value of \( \Delta_b \). Hence, the deviations seen here should be seen as a maximum possible deviation for the chosen \( N_{\text{side}} \).

(iv) Mask: We have examined the impact of an Euclid-type mask on skew-spectrum in a Pseudo-\( C_L \) based approach introduced in Section 5. The results are presented in Fig. 10. The upper solid-lines in each panel correspond to all-sky theoretical predictions of \( S_1 \). The upper lines with scatter correspond to the estimates from one realization of the simulated maps. The left-hand panel corresponds to the source redshift \( z_s = 1.0 \) and the right-hand panel corresponds to the source redshift \( z_s = 0.5 \) and 0.0, respectively.
Figure 5. The skew-spectrum computed by cross-correlating $\kappa^2$ and $\kappa$ from two different redshift bins is plotted. In particular, the squared $\kappa_1 = \kappa(z_s^1)$ defined for a source redshift $z_s^1$ and $\kappa_2 = \kappa(z_s^2)$ at redshift $z_s^2$ are being cross-correlated in the harmonic domain. For this plot we restricted ourselves to $z_s^1 > z_s^2$. The smooth lines correspond to the theoretical predictions and the lines with error bars correspond to results from numerical simulations. We use the expression of mixed bispectrum given in equations (9b)–(9c) for computing the theoretical predictions. The expression for the estimator for the skew-spectrum is given in equation (21). The panels from left to right correspond to $z_s^1 = 2.0, 1.5$ and $1.0$ respectively and various curves in each panel correspond to $z_s^2$ as indicated.

Figure 6. Same as Fig. 5 but the skew-spectrum is being computed cross-correlating $\kappa_1$ and $\kappa_2^2$ instead of $\kappa_1^2$ and $\kappa_2$ for $z_s^1 > z_s^2$. to $z_s = 2.0$. The lower dashed-curves in each panel correspond to the PCL based theoretical predictions $\tilde{S}_l$ computed using equation (22a). The corresponding (lower) lines with scatter are estimates from one realization of partial sky with the Euclid-type mask, shown in Fig. 2, applied.

(v) Noise: The impact of noise which we assume to be Gaussian on estimation of skew-spectrum is shown in Fig. 11. In both panels the source plane is fixed at $z_s = 1$. The solid lines in each panel represent the theoretical skew-spectrum for $z_s = 1$. The dashed lines represent the pseudo skew-spectrum represented as $\tilde{S}_l$ with an Euclid-type mask being included. If we compare the scatter with corresponding plots in Fig. 10 we can see how the inclusion of noise increases the scatter though the estimator remains unbiased. Skew-spectrum for a Gaussian noise alone is zero so the only effect the noise has on the estimator is to increase its scatter.

The noise was generated at each pixel using a Gaussian deviate with variance $\sigma = \sigma_e / \sqrt{n}$. Where we take $\sigma_e$ represents the variance of the observed ellipticity $\sigma_e = 0.3$, and $\bar{n}$ is the average number density of source galaxies per pixel computed using total number of observed galaxies, fraction of sky covered and number of pixel at a specific HEALPix resolution. We have used two different values of $\bar{n}$. The left-hand panel of Fig. 11 corresponds to a source density of $n_s = 30 \text{ arcmin}^{-2}$ and the right-hand panel corresponds to $n_s = 10 \text{ arcmin}^{-2}$. The fraction of the sky covered by the mask was taken to be $f_{\text{sky}} = 0.35$.

7 Conclusions

In this paper, we have introduced the skew-spectrum statistic as a probe for weak lensing bispectrum. While we found an excellent agreement of numerical simulations and fitting-function based theoretical predictions for the autocorrelation we have studied, we also found significant deviation in many other situations and found that the current analytical uncertainty is not sufficient for high accuracy work. We have primarily focused on gravity-induced secondary bispectrum in a $\Lambda$CDM cosmology. However, several extensions of our study are possible.
Figure 7. The skew-spectra computed by cross-correlating $\kappa_{LSS}^2$ at the last scattering surface of $z_{LSS} = 1100$ and low redshift weak lensing convergence maps at $z_s$ is being plotted. As before the solid smooth lines in different panels correspond to theoretical results and the lines with error bars correspond to the results from numerical simulations. The panels from left to right correspond to $z_s = 2.0, 1.5, 1.0$, and $0.5$, respectively. No mask or smoothing was considered. We use the expression of mixed bispectrum given in equations (9b)–(9c) for computing the theoretical predictions.

Figure 8. Same as Fig. 7 but are constructed by cross-correlating $\kappa_{LSS}$ against $\kappa_s^2$.

7.1 Skew-spectrum in beyond \Lambda CDM scenarios

In most modified gravity theories and dark energy models, the bispectrum is currently known only in the perturbative regime. We have provided analytical expressions for the skew-spectrum in such scenarios. To go beyond perturbative regime a non-linear model for the bispectrum is required. It is expected that a fitting-function based description in such scenarios will eventually be available as more accurate simulations are performed. Similarly, the modelling of bispectrum based on effective field theories will also be extended to modified gravity theories. Once such results are available, they can readily be used to compute the skew-spectrum in these models.

7.2 Higher order corrections

The theoretical expressions of the skew-spectrum are derived using many simplifying assumptions. We have ignored the corrections due to magnification bias as well as reduced shear which should be included in more accurate theoretical predictions. In addition, the skew-spectrum here is computed using the Limber approximation (Kitching et al. 2017).

7.3 Skew-spectrum from shear maps

We have computed the skew-spectrum from a convergence map. However, for many practical purposes a skew-spectrum estimated directly from shear maps can bypass many of complications of the map making process.

7.4 Intrinsic alignment

The intrinsic alignment (IA) of galaxies (see Vlah, Chisari & Schmidt 1910, and the references therein) are caused by the tidal interaction and is a source of contamination to gravity-induced (extrinsic) weak lensing. The lensing bispectrum induced by IA is typically at the level of 10 per cent of the lensing induced bispectrum. Several methods have been proposed to mitigate or remove such contamination using joint analysis of power spectrum and bispectrum. The skew-spectrum retains some of the shape information of the original bispectrum. A joint analysis of power spectrum and skew-spectrum can thus be useful in separation of these two different contributions. The skew-spectrum introduced in this study can further be optimized by introducing weights to judge the level of cross-contamination from the intrinsic alignment much in the same way as was achieved...
Figure 9. In this figure we present the accuracy of the fitting function we have used in our study. We have used the binned skew-spectra for our comparison. A bin-size of $\delta_b = 100$ is being used. Here $\Delta r$ represents the normalized deviation from predictions and results from simulations (see the text for details). The left-hand panel corresponds to the skew-spectrum computed by cross-correlating the CMB sky and the low redshift weak lensing. The middle- and the right-hand panel correspond to the skew-spectra obtained by cross-correlating two tomographic lensing maps. The plots in the left-hand panel are obtained using the results presented in the Figs 7 and 8. The upper set of curves correspond to the cross-correlation of tomographic bins against convergence map at $z_{\text{LSS}}$ i.e., $\langle \kappa^2(z_s)\kappa_{\text{LSS}}^2 \rangle$, whereas the lower curves correspond to their symmetric counterparts $\langle \kappa(z_s)\kappa_{\text{LSS}}^2 \rangle$. The error in skew-spectra computed using lower redshift maps that are associated with $\langle \kappa^2(z_s)\kappa_{\text{LSS}}^2 \rangle$ are depicted in the middle panel. Their symmetric counterparts are shown in the right-hand panel. The line-styles used in the middle and right-hand panels are identical. For the corresponding skew-spectra see Figs 5 and 6, respectively.

Figure 10. The pseudo-$S_\ell$ for two different redshifts are presented. A Euclid-type mask was used in our study. The regions which are within 22 deg of the galactic or ecliptic plane are removed from our study. The fraction of the sky left unmasked is $f_{\text{sky}} = 0.35$ (roughly 14 490 deg$^2$ of the sky). The left-hand (right) panel corresponds to $z_s = 1.0$ ($z_s = 2.0$). The upper smooth solid curves represents the theoretical $S_\ell$. The two upper curves represent estimated skew-spectrum from a single realization. The curve that shows more scatter represents skew-spectrum estimated from an all-sky map. The curve which shows more scatter correspond to $\hat{S}_\ell$ using equation (22a).

Figure 11. The impact of noise (assumed Gaussian) on estimation of skew-spectrum is presented. In both panels the source plane is fixed at $z_s = 1$. The solid lines in each panel represent the theoretical skew-spectrum for $z_s = 1$. The dashed line represent the pseudo skew-spectrum or $\hat{S}_\ell$ for the Euclid-type mask being considered. Inclusion of Gaussian noise increases the scatter but the estimator remains unbiased. The left-hand panel corresponds to source density of $n_x = 30$ arcmin$^{-2}$ and the right-hand panel corresponds to $n_x = 10$ arcmin$^{-2}$. 

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in case of point source contamination of CMB studies designed to
detect primordial non-Gaussianity from Planck data.

7.5 Primordial non-Gaussianity and active perturbations

We have considered the gravity-induced non-Gaussianity in our study
as it is the most dominant source of non-Gaussianity in weak lensing
maps. However, similar results can also be obtained for computing the
subdominant contributions from primordial non-Gaussianity as well as
secondary sources of non-Gaussianity induced by active sources, e.g. cosmic
strings.

7.6 Baryonic feedback

We have not included any baryonic feedback in our modelling of the
skew-spectrum but such corrections can be incorporated in the
skew-spectrum for direct comparison with any realistic data.

7.7 Covariance and likelihood

We have not discussed the covariance of the skew-spectrum in this study. An accurate
description of the covariance will be an important ingredient of cosmological likelihood analysis involving
skew-spectrum. A simple form of covariance can be derived under the assumption of Gaussianity and thus ignoring all higher order
correlation contributing to the covariance. Such an estimation will be useful in the noise-dominated regime but will not be sufficient in the
highly non-linear scales characterized by high signal-to-noise probed by the future surveys such as Euclid. The methods developed so far
in computing the covariance include the ones based on perturbative analysis, halo model, or simulated mocks (Rizzato et al. 2019). These
methods can be adapted to compute the skew-spectrum covariance.

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DATA AVAILABILITY

No new data were generated in support of this research. The simulations used in this work are publicly available.

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