Bispectrum and Nonlinear Biasing of Galaxies: Perturbation Analysis, Numerical Simulation, and SDSS Galaxy Clustering

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Abstract

We consider nonlinear biasing models of galaxies with particular attention to a correlation between the linear and quadratic biasing coefficients, $b_1$ and $b_2$. We first derive perturbative expressions for $b_1$ and $b_2$ in halo and peak biasing models. We then discuss our computations of the power spectra and bispectra of dark matter particles and halos using $N$-body simulation data and of volume-limited subsamples of Sloan Digital Sky Survey (SDSS) galaxies, and determine their $b_1$ and $b_2$. We find that the values of those coefficients at linear regimes ($k < 0.2 \, h \, Mpc^{-1}$) are fairly insensitive to the redshift-space distortion and the survey volume shape. The resulting normalized amplitudes of the bispectra, $Q$, for equilateral triangles are insensitive to the values of $b_1$, implying that $b_2$ indeed correlates with $b_1$. The present results explain the previous finding of Kayo et al. (2004, PASJ, 56, 415) for the hierarchical relation of three-point correlation functions of SDSS galaxies. While the relations between $b_1$ and $b_2$ are quantitatively different for specific biasing models, their approximately similar correlations indicate a fairly generic outcome of the biasing due to the gravity in primordial Gaussian density fields.

Key words: cosmology: large-scale structure of universe — methods: statistical — observations — theory

1. Introduction

One of the major uncertainties in precise cosmology is galaxy biasing relative to the underlying mass distribution. It hampers extracting cosmological information from the large-scale structure of the universe. In particular, the biasing is sensitive to the unknown physical conditions of galaxy formation. Thus, its phenomenological parametrization and understanding is crucial for advancing our knowledge of the evolution of the universe.

A reasonable approximation often adopted is a local linear biasing model, which assumes a simple scale-independent relation between the density contrast fields of a galaxy and the mass:

$$
\delta_g(x, z) = b(z) \delta(x, z).
$$

(1)

In the above, the density contrast of the mass is defined as

$$
\delta(x, z) \equiv \frac{\rho(x, z) - \bar{\rho}(z)}{\bar{\rho}(z)},
$$

(2)

where the over-bar indicates the mean over the entire universe. The number density contrast of the galaxy density field, $\delta_g$, is defined similarly. Because this model contains only a single time-dependent parameter, $b(z)$, it is not surprising that it cannot accurately describe the observed features of galaxy clustering. Furthermore, the three-point correlation function of SDSS galaxies (Kayo et al. 2004) indicates the importance of the higher order correction to equation (1); while the galaxy two-point correlation functions are well represented by the linear biasing model (e.g., Zehavi et al. 2005), the normalized amplitudes, $Q$, of the corresponding three-point functions for equilateral triangles do not show the expected scaling with respect to $b(z)$:

$$
Q \propto \frac{1}{b}.
$$

(3)

In reality, however, $Q$ for equilateral triangles calculated from SDSS galaxies in redshift space proves to be almost scale-independent, and approximately follows the hierarchical relation, $Q = 0.5–1.0$. Moreover its dependence on the morphology, color, and luminosity is not statistically significant. Given the robust morphological, color, and luminosity dependences of the two-point correlation function, Kayo et al. (2004) argued that galaxy biasing is complex and requires a contrived relation between...
the linear biasing and its higher order correction terms. Another possibility is that the observed value is significantly contaminated by the redshift-space distortion, and does not properly reflect the actual clustering information in real space. Indeed, previous N-body studies of the redshift-space distortion on three- and four-point correlation functions suggest that this is the case in nonlinear regimes (Suto 1993; Matsubara & Suto 1994; Suto & Matsubara 1994). The interpretation is further complicated by the fact that the higher-order statistics is sensitive to relatively rare large-scale structures in particular samples (Nichol et al. 2006).

The main purpose of the present paper is to provide a physical explanation for the approximate hierarchical relation for $Q$ of SDSS galaxy clustering on the basis of perturbation analysis and numerical simulations. This is a step toward understanding the nonlinear nature of biasing. Specifically, we employ a local nonlinear biasing model (Fry & Gaztañaga 1993),

$$\delta_b(x,z) = \sum_{n=0}^{\infty} \frac{b_n(z)}{n!} [\delta(x,z)]^n.\quad (4)$$

We consider density peaks, dark matter halos, simulated halos, and the SDSS galaxies as specific examples for the density fields, $\delta_b$ in the left-hand side.

An outline of the paper is as follows: section 2 briefly summarizes the halo and peak biasing models as analytically tractable examples. We computed the linear and quadratic biasing coefficients perturbatively and found that these models roughly agree with the observed hierarchical relation. To be more realistic, we analyze simulated halo catalogs and SDSS galaxies in section 3, and find that the perturbative result is valid even if we take account of a variety of selection effects and particularly redshift-space distortion. The quantitative agreement between the two indicates a generic correlation among the biasing coefficients in the SDSS galaxies. Finally, section 4 is devoted to a summary and conclusion of the paper.

2. Perturbative Predictions in Halo and Peak Biasing Models

2.1. Basic Statistics and Perturbation Biasing Models

Throughout the present analysis, we work in Fourier space. For definiteness, we adopt the definition of the Fourier transform of an arbitrary function, $g(x)$ as

$$g(k) = \int \frac{d^3x}{(2\pi)^3} g(x) e^{-ik\cdot x}.\quad (5)$$

Then, the power spectrum and the bispectrum of the mass density field are defined as

$$\langle \delta(k_1) \delta(k_2) \rangle \equiv P(k_1) \delta_\Omega(k_1 + k_2),\quad (6)$$

$$\langle \delta(k_1) \delta(k_2) \delta(k_3) \rangle \equiv B(k_1,k_2,k_3) \delta_\Omega(k_1 + k_2 + k_3),\quad (7)$$

where $\delta_\Omega$ denotes the Dirac delta, and we assume that the universe is isotropic and homogeneous. We introduce the normalized amplitude of the mass bispectrum as

$$Q_m(k_1,k_2,k_3) \equiv \frac{B(k_1,k_2,k_3)}{P(k_1)P(k_2) + P(k_2)P(k_3) + P(k_3)P(k_1)}.\quad (8)$$

Note that, strictly speaking, the above statistic is different from that defined in configuration space. Since both are expected to be nearly identical in a weakly nonlinear regime, we focus on the Fourier-space analysis throughout this paper (see also Verde et al. 2002; Hikage et al. 2005). An analysis in configuration space is now in progress.

The above statistics can be generalized in a straightforward manner to any biased field. By keeping the linear and quadratic terms in equation (4), one obtains

$$P_b(k) = b_1^2 P(k),\quad (9)$$

$$Q_b(k_1,k_2,k_3) = \left[ \frac{Q_m(k_1,k_2,k_3)}{b_1^4} + \frac{b_2}{b_1} \right].\quad (10)$$

Therefore, $Q_b$ for the biased field is directly dependent on the quadratic biasing coefficient, $b_2$, unlike in the case of the power spectrum. As we show below, this term plays an important role to approximately satisfy the hierarchical relation. In order to understand the qualitative relations among the biasing coefficients, we first consider analytically workable biasing models, i.e., halo and peak biasing.

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1 In this paper, we use the term “real space” in order to imply the analysis without any redshift-space distortion in $k$-space. Thus, we reserve the term “configuration space” in order to distinguish from $k$-space.
2.2. Halo Biasing

We follow the formalism of Mo & White (1996) for halo biasing. We define the (unconditional) mass function, \( n_{\text{halo}}(m, z) \), of halos with mass \( m \) at redshift \( z \). We rewrite the unconditional mass function as

\[
\frac{m^2 n_{\text{halo}}(m, z) \, dm}{\bar{\rho}} = v f(v) \frac{d v}{v}, \quad v(m, z) \equiv \frac{\delta_{\text{sc}}(z)}{\sigma(m)}. \tag{11}
\]

We denote by \( \sigma(m) \) the linearly extrapolated value of the rms of the initial density fluctuation field smoothed with a tophat filter of scale \( R = [3m/(4\pi \bar{\rho})]^{1/3} \), where \( \bar{\rho} \) is the mean comoving background density. The characteristic density contrast for spherical collapse is given by

\[
\delta_{\text{sc}}(z) \simeq \frac{3(12\pi)^{2/3}}{20} \left[ 1 + 0.0123 \log_{10} \Omega_m(z) \right] / D(z), \tag{12}
\]

where \( \Omega_m(z) \) is the matter density in units of the critical density at redshift \( z \), and \( D(z) \) is the linear growth rate (Kitayama & Suto 1996). In the above equation and throughout the paper, we assume a spatially-flat model with a non-vanishing cosmological constant. Specifically, we adopt the present value of the mass density \( \Omega_m = 0.3 \), the cosmological constant, \( \Omega_{\Lambda} = 0.7 \), the Hubble constant, \( h = H_0/(100 \, \text{km} \, \text{s}^{-1} \, \text{Mpc}^{-1}) = 0.7 \), and the amplitude of the density fluctuations smoothed over \( 8 \, h^{-1} \text{Mpc} \), \( \sigma_8 = 0.9 \).

There exist two popular models for \( f(v) \). One is based on a spherical collapse model (Press & Schechter 1974),

\[
v f_{\text{PS}}(v) = 2 \sqrt{\frac{v^2}{2\pi}} \exp(-v^2/2). \tag{13}
\]

The other is based on an ellipsoidal collapse model (Sheth & Tormen 1999; Sheth et al. 2001),

\[
v f_{\text{ST}}(v) = 2A(p) \left[ 1 + \frac{1}{(qv)^p} \right] \sqrt{\frac{q v^2}{2\pi}} \exp(-q v^2/2), \tag{14}
\]

where \( p \approx 0.3 \), \( A(p) = [1 + 2^{-p}(1/2 - p)/\sqrt{2}]^{-1} \approx 0.3222 \), and \( q \approx 0.75 \) (Cooray & Sheth 2002).

Consider a comoving volume \( V \) and its total mass \( M \) at redshift \( z_0 \). Let \( n_{\text{halo}}(m, z_1|M, V, z_0) \, dm \) denote the conditional number density of halos whose mass is between \( m \) and \( m + dm \) at \( z_1 \). A reasonable approximation for \( n_{\text{halo}}(m, z_1|M, V, z_0) \) is obtained by applying the extended Press–Schechter theory (e.g., Bower 1991; Bond et al. 1991):

\[
\frac{m^2 n_{\text{halo}}(m, z_1|M, V, z_0) \, dm}{\bar{\rho}} = v_{10} f(v_{10}) \frac{d v_{10}}{v_{10}}, \tag{15}
\]

\[
v_{10} = \frac{\delta_{\text{sc}}(z_1) - \delta_0(z_0)}{\sqrt{\sigma^2(m) - \sigma^2(M)}}, \tag{16}
\]

where \( \delta_0(z_0) \) is the linearly extrapolated mass density contrast at \( z_0 \). If we define the actual density contrast \( \delta = M/(\bar{\rho}V) - 1 \), \( \delta_0(z_0) \) is written as

\[
\frac{\delta_0}{1 + z_0} = \sum_{i=0}^{\infty} a_i \delta_i, \quad a_0 = 0; \quad a_1 = 1; \quad a_2 = -17/21; \quad a_3 = 341/567; \quad a_4 = -55805/130977; \ldots \tag{17}
\]

where the above coefficients are obtained assuming the spherical collapse model (Bernardeau 1994).

Now one can write the corresponding conditional density contrast of halos:

\[
\delta_{\text{halo}}(m, z_1|M, V, z_0) = \frac{n_{\text{halo}}(m, z_1|M, V, z_0)(1 + \delta)}{n_{\text{halo}}(m, z_1)} - 1. \tag{19}
\]

Expanding equation (19) in terms of \( \delta \) in the limit of \( \sigma(M) \to 0 \) yields the biasing coefficients of halos (Mo et al. 1997; Scoccimarro et al. 2001; Cooray & Sheth 2002):

\[
b_1(m, z) = 1 + \epsilon_1 + E_1, \tag{20}
\]

\[
b_2(m, z) = 2(1 + a_2)(\epsilon_1 + E_1) + \epsilon_2 + E_2, \tag{20}
\]

where

\[
\epsilon_1 = \frac{qv^2(m, z) - 1}{\delta_{\text{sc}}(z)}, \quad \epsilon_2 = \frac{qv^2(m, z)}{\delta_{\text{sc}}(z)} \left[ \frac{qv^2(m, z) - 3}{\delta_{\text{sc}}(z)} \right], \tag{21}
\]

\[
E_1 = \frac{2p}{\delta_{\text{sc}}(z)} \left[ 1 + [qv^2(m, z)]^p \right], \quad \frac{E_2}{E_1} = \frac{1 + 2p}{\delta_{\text{sc}}(z)} + 2\epsilon_1. \tag{22}
\]
The above results for ellipsoidal collapse reduce to those for spherical collapse, if one sets $p = 0$ and $q = 1$.

### 2.3. Peak Biasing

The biasing coefficients of the peak model can be obtained similarly as the halo model described in the above subsection. The conditional and unconditional number densities of peaks with peak height $v \equiv \delta / \sigma$ are derived in Bardeen et al. (1986). In this case, $\sigma$ denotes the rms value of the density fluctuation smoothed over $R \equiv [3m/(4\pi \bar{\rho})]^{1/3}$, and $\delta$ is the density contrast of the peaks in the smoothed field. Substituting those formulae into the right-hand side of equation (19), one similarly obtains the biasing coefficients for the peak model (Mo et al. 1997):

\begin{align}
  b_1(v, z) & = 1 + \frac{v^2 + g_1}{\delta_{sc}(z)}, \\
  b_2(v, z) & = 2(1 + a_2) \frac{v^2 + g_1}{\delta_{sc}(z)} + \left[ \frac{v}{\delta_{sc}(z)} \right]^2 \left( v^2 - 1 + 2g_1 + \frac{2g_2}{v^2} \right),
\end{align}

where the above functions, $g_1$ and $g_2$, are defined in equation (25) of Mo et al. (1997).

### 2.4. Results of Analytic Models

We consider the above three models (spherical halo, ellipsoidal halo, and peak) to explore the correlation between $b_1$ and $b_2$ analytically.

Figure 1 shows the biasing coefficients, $b_1$ (thick lines) and $b_2/b_1$ (thin lines) at $z = 0$ (solid) and $z = 2$ (dashed). We adopt the cold dark matter transfer function of Bardeen et al. (1986) as the initial mass power spectrum (with $\Omega_m = 0.3$, $\Omega_\Lambda = 0.7$, and the shape parameter $\Gamma = \Omega_m h = 0.21$). In the three biasing models, $b_1$ and $b_2/b_1$ behave similarly as functions of $M_{\text{halo}}$ or $v$. This implies the presence of a certain correlation between $b_1$ and $b_2/b_1$. To see this point more clearly, we plot $b_2/b_1$ in terms of $b_1$ for each model (figure 2). All of the three models exhibit very similar correlations between $b_1$ and $b_2/b_1$. The differences among the three models become even smaller at higher redshifts; compare $z = 0$ (thick lines) and $z = 2$ (thin lines).

In analyzing the simulation data below, the estimate of the biasing coefficients is made after averaging over a finite mass range.

We calculate the mass-averaged values of $b_1$ and $b_2$ for halo models at $z = 0$:

\begin{align}
  B_n(m_{\text{min}}, m_{\text{max}}) & = \frac{\int_{m_{\text{min}}}^{m_{\text{max}}} \int_{m_{\text{halo}}} dm n_{\text{halo}}(m, z = 0) b_n(m, z = 0)}{\int_{m_{\text{min}}}^{m_{\text{max}}} \int_{m_{\text{halo}}} dm n_{\text{halo}}(m, z = 0)}, \quad (n = 1, 2). \tag{25}
\end{align}

The difference between $b_n$ and $B_n$ is illustrated in figure 2, where spherical (open) and ellipsoidal (filled) halo models [equations (13), (14)] are assumed. We plot the results for two mass ranges, corresponding to S (triangle) and L (square) halo subsamples (see subsection 3.1 for detail).

Figure 3 plots $Q_b$, for biased fields as a function of $b_1$ in real space. The dotted, dashed, and solid lines represent the
Fig. 2. Correlation of $b_1$ and $b_2/b_1$ from halo and peak biasing. Different line types are results of halo biasing of spherical collapse (solid), ellipsoidal collapse (dashed), and peak biasing (dotted), evaluated at redshifts of 2 (thin) and 0 (thick). The open (filled) symbols represent the mass-averaged values, $B_1$ and $B_2/B_1$, defined in equation (25), assuming the spherical (ellipsoidal) halo model. The mass ranges correspond to L (square), and S (triangle), defined in subsection 3.1.

Fig. 3. $Q$ for biased fields as a function of $b_1$ in real space. We estimate $Q$ using equation (10) on the basis of perturbative predictions for peaks (dotted), ellipsoidal halos (dashed), and spherical halos (solid). We assume three different values for $Q_m$: $Q_m = 1.58$, $Q_m = 0.83$, and $Q_m = 4/7$ from top to bottom. The first two values were computed from N-body results at $k = 0.4 h \text{Mpc}^{-1}$, and $k = 0.18 h \text{Mpc}^{-1}$ (see section 3 below), and the last value corresponds to perturbation theory [equations (39) and (40)], where equilateral triangles ($k_1 = k_2 = k_3 = k$) are assumed. Left: $z = 0$, right: $z = 2$. Results of tree-level perturbation with respect to biasing, equation (10), for peak, ellipsoidal, and spherical halo models. For simplicity, we consider equilateral triangles and adopt three different values for $Q_m$: $Q_m = 4/7$ from the leading term of the gravitational nonlinearity [see equation (39) below], and $Q_m = 0.83$ and 1.58 from N-body results at $k = 0.18 h \text{Mpc}^{-1}$.
3. Comparison with N-Body Simulations and SDSS Galaxies

3.1. N-Body Simulations and SDSS Galaxies

The N-body simulations that we employ consist of 512^3 dark matter particles in a cubic box of 300 h^{-1} Mpc (comoving) on a side with the periodic boundary condition. The mass of each dark matter particle is 1.68 \times 10^{10} h^{-1} M_{\odot}. As mentioned in section 2, we use the cold dark matter transfer function of Bardeen et al. (1986) as the initial mass power spectrum (with \Omega_m = 0.3, \Omega_\Lambda = 0.7, and \Gamma = \Omega_m h = 0.21), and generate three independent Gaussian realizations. The initial density field at z = 36 is evolved up to z = 0 using the P^3M code of Jing and Suto (1998, 2002) with a gravitational softening length of \epsilon \sim 58 h^{-1} \text{kpc}. We assume the amplitude of the density fluctuations smoothed over 8 h^{-1} \text{Mpc}, \sigma_8 = 0.9.

In order to test the dependence of \Omega_b, on actual galaxy properties, we construct volume-limited subsamples of different colors from New York University Value-Added Galaxy Catalog (Blanton et al. 2005) based on the SDSS Data Release 4 (Adelman-McCarthy et al. 2006). The angular selection function of the survey is written in terms of spherical polygons (Hamilton & Tegmark 2004). Details of the SDSS can be found in the following papers: York et al. (2000) described an overview. Technical articles providing details include descriptions of the telescope design (Gunn et al. 2006), the photometric camera (Gunn et al. 1998), photometric analysis (Stoughton et al. 2002), the photometric system and calibration (Fukugita et al. 1996; Hogg et al. 2001; Ivezić et al. 2004; Smith et al. 2002; Tucker et al. 2006), the photometric pipeline (Lupton et al. 2001), astrometric calibration (Pier et al. 2003), selection of the galaxy spectroscopic samples (Eisenstein et al. 2001; Strauss et al. 2002), and spectroscopic tiling (Blanton et al. 2003). We only consider galaxies with redshifts of 0.05 \leq z \leq 0.1 and magnitudes \text{M_r} \leq -19.8 (68337 galaxies in total, over \sim 4259 deg^2). We divide those galaxies according to their colors so that each subsample would contain roughly the same number: red subsample has g-r > 0.86 (34351 in total), and blue subsample has g-r < 0.86 (33986). In table 1, N_g and n_g denote the total number of galaxies and the galaxy number density in each subsample, and the wavenumber corresponding to the mean galaxy separation is 2\pi n_g^{1/3}. The linear biasing coefficient, \bi, is evaluated at \hat{k} = 0.126 h^{-1} Mpc^{-1}, assuming the same set of cosmological parameters used in N-body simulations. In particular, the estimated value of \bi is sensitive to \sigma_8. The value of \sigma_8 is still controversial between different observations. For example, WMAP3 gives \sigma_8 = 0.742 \pm 0.051 (Spergel et al. 2006), while 2dFGRS gives 0.88^{+0.12}_{-0.08} (Gaztañaga et al. 2005). Thus, in principle our estimated values of \bi would increase by about 20\% if we had adopted the WMAP3 result.

In order to compare with analytical predictions for halo biasing and also with the SDSS subsamples, we identify dark matter
halos using a friends-of-friends algorithm. Specifically, we use the public code “FOF”, and adopt a linking length of 0.164 in units of the mean particle separation. We construct nine halo subsamples from three realizations (table 2). The indices (S, L, and LL) represent the three different halo mass ranges corresponding to $1.68 \times 10^{12} h^{-1} M_\odot < M < 1.18 \times 10^{13} h^{-1} M_\odot$, $1.18 \times 10^{13} h^{-1} M_\odot < M$, and $6.72 \times 10^{13} h^{-1} M_\odot < M$. The mass ranges of the first two, S and L, are determined so that they have approximately the same values of $b_1$ for the SDSS blue and red galaxies, respectively. The last one, LL, is constructed so as to check the $b_2 - b_1$ correlation around $b_1 \sim 1$, and is discussed in figure 8 below. We further divide the three different mass-selected samples into three subsamples: r-cube (s-cube) measures positions of halos in real (redshift) space using the line-of-sight component of the center-of-mass velocity of each halo. Wedge subsamples are constructed so as to have the same survey geometry as the SDSS sample. They measure the positions of halos in redshift space, using the line-of-sight component of the center-of-mass velocity unlike s-cube. In table 2, $\langle N_h \rangle$ and $\langle n_h \rangle$ denote the total number of halos and the halo number density in each subsample (averaged over three realizations). The wavenumber corresponding to the mean halo separation is $2\pi \langle n_h \rangle^{1/3}$, and the linear biasing coefficient, $\langle b_1 \rangle$ was evaluated at $k = 0.126 h \, \text{Mpc}^{-1}$.

### 3.2. Power Spectrum and $b_1$

We first estimate the linear biasing parameter according to equation (9):

$$ b_{1,i}(k) = \sqrt{\frac{P_i(k)}{P(k)}}, \quad (28) $$

where the subscript $i$ denotes the different subsamples of simulated halos and SDSS galaxies.

In real space (r-cube in table 2), we first construct a density field of halos using cloud-in-cell pixelization, Fourier transform it, angularly average $P_i(k)$ over the direction of $k$, and finally remove the shot noise contribution, $1/n_h$, to obtain $P_i(k)$. We choose logarithmically equal bins for $k$, $\Delta(\log_{10} k) = 1/6$. The mean values, $\langle n_h \rangle$, of the number density of halos in each realization ($n_h$) are listed in table 2. The wavenumbers corresponding to the mean halo separation $2\pi \langle n_h \rangle^{1/3}$ roughly provide the limit of the reliability for estimating $P_i(k)$. The dark matter power spectrum $P(k)$ is estimated in two different methods. One is a direct estimate using the dark matter particles, $P^\text{sim,}\Gamma(k)$. The other employs an analytical prescription by Peacock and Dodds (1996), $P^\text{PD,}\Gamma(k)$. The upper-left panel of figure 5 shows these results. Symbols and curves represent the results using $P^\text{PD,}\Gamma(k)$ and $P^\text{sim,}\Gamma(k)$. Their agreement ensures the validity of the Peacock–Dodds prescription.

The wavenumber dependence of $1/b_1$ is weak for $k < 0.2 h \, \text{Mpc}^{-1}$. The increase of $1/b_1$ for $k > 0.2 h \, \text{Mpc}^{-1}$ is consistent with that expected from the halo finite-volume exclusion effect (e.g., Taruya et al. 2001). When the wavenumber becomes comparable or larger than the mean separation of halos ($\sim 0.5 h \, \text{Mpc}^{-1}$; c.f. table 2), $P_i(k)$ does not represent the intrinsic clustering signal properly. Therefore, the increase of $1/b_1$ against $k$ may not be real for $k > 0.5 h \, \text{Mpc}^{-1}$. In this paper we are interested in linear regimes where the $k$ dependence of $b_1$ is negligible. Thus, we do not consider the region $k > 0.2 h \, \text{Mpc}^{-1}$, where the scale-independent expansion, like equation (4), breaks down.

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2 Available at the website (http://www-hpcc.astro.washington.edu/).
Fig. 5. Inverse of the linear biasing parameters of simulated halos (table 2) and SDSS galaxies (table 1). Crosses and filled circles for simulated halos (SDSS galaxies) correspond to S and L samples (blue and red), respectively. Dashed and solid lines indicate results based on the direct estimation of \( P(k) \), while symbols in all panels use \( P^{PD}(k) \) or \( P^{DS}(k) \), equation (29). The quoted error bars for simulated halos are computed from three different realizations. We simply use the error bars for wedge subsamples just for reference in the case of SDSS galaxies.

In redshift space (s-cube in table 2), we compute \( P_r(k) \) using the center-of-mass peculiar velocity of each halo adopting the distant observer approximation. The dark matter power spectrum in redshift space was computed using two different methods. One was a direct estimate using the dark matter particles, \( P^{sim}(k) \). The other, \( P^{PD}(k) \), was an analytical prediction combining \( P^{PD}(k) \) and redshift distortion effects empirically. For the latter we considered the Kaiser effect (Kaiser 1987) and the finger-of-God effect, assuming an exponential peculiar velocity distribution (Cole et al. 1994, 1995; Kang et al. 2002). In this case the redshift space power spectrum is given as

\[
P^{PD_s}(k) = \left[ A(k_m) + \frac{2}{3} \beta_m B(k_m) + \frac{2}{3} \beta_m^2 C(k_m) \right] P^{PD_r}(k),
\]

where \( \beta_m \) denotes the value of equation (27) for dark matter particles, \( \kappa_m = k \sigma_p / (\sqrt{2} H_0) \), and the rms pairwise velocity dispersion of dark matter particles, \( \sigma_p = \sqrt{\sigma_v^2} \approx 520 \text{ km s}^{-1} \), \( \sigma_v \) is the one-dimensional velocity dispersion) is computed directly from \( N \)-body simulations. The upper-right panel in figure 5 displays the results for s-cube. Again, \( b_1 \) is fairly scale-independent in redshift space. Although the finger-of-God effect partially compensates the wavenumber dependence in r-cube for \( k > 0.2 \text{ h Mpc}^{-1} \), \( b_1(k) \) there is not reliable. The results of r-cube and s-cube imply that \( b_1(k) \) is not sensitive to the redshift distortion, i.e., \( b^2_1 \approx b^2_1 \) for \( k < 0.2 \text{ h Mpc}^{-1} \). In fact, the agreement between \( P^{sim}(k) \) and \( P^{PD}(k) \) suggests that the above feature can be explained using equation (29):

\[
A(k_m) = \frac{1}{\kappa_m} \arctan(\kappa_m),
\]

\[
B(k_m) = \frac{3}{\kappa_m^2} \left[ 1 - A(k_m) \right],
\]

\[
C(k_m) = \frac{5}{3 \kappa_m^2} \left[ 1 - B(k_m) \right].
\]
where subscripts $i$ and $m$ refer to different subsamples and dark matter particles, respectively, and $\beta_i$ indicates the value of equation (27) for each subsample, and $k_i = k_\sigma P_i / (\sqrt{2} H_0)$. The rms pairwise velocity dispersion of halos, $\sigma_\text{P.V.}$, is calculated from the simulated halo subsamples through $\sigma_\text{P.V.} = \sqrt{2}\sigma_\text{v.e.}$ for S and 397 km s$^{-1}$ for L. Equation (33) yields 1.04 (0.97) at $k = 0.05 h$ Mpc$^{-1}$ and 1.08 (1.01) at $k = 0.2 h$ Mpc$^{-1}$ for the S (L) subsample. These values explain the behavior in figure 5, and suggest that $b_{1,i}$ and $b_{1,m}$ in the linear regime agree within 10% level.

For wedge subsamples, we determine the positions of halos in redshift space properly using the line-of-sight velocity component unlike s-cube where we employ the distant-observer approximation. In estimating $P_1(k)$, we follow Feldman, Kaiser, and Peacock (1994), Matarrese et al. (1997), and Verde et al. (2002), and define the field as

$$ F_i(r) = \lambda w(r)[n_i(r) - \alpha n_i(r)], $$

(34)

where $w(r)$ is the weight, $\lambda$ is a constant to be determined, $n_i(r)$ is the number density of random particles, $n_i(r)$ is the number density of each subsamples, and $\alpha$ is the ratio of particle numbers of actual and random catalogs. In this paper, $w(r)$ is unity inside the survey volume, and zero otherwise. If we set $\lambda = I_{22}^{-1/2}$, where

$$ I_{jk} = \int d^3 r w^j(r) n_i^j(r) $$

(35)

with the mean number density $n_i$ for each subsample (Matarrese et al. 1997), then the power spectrum is

$$ \langle |F_i(k)|^2 \rangle = P_i(k) + \frac{I_{22}}{I_{33}}(1 + \alpha) $$

(36)

(Verde et al. 2002). The dark matter power spectrum for the wedge subsamples is calculated by the two methods again: the first is an analytical calculation using equation (29) and the second is a direct estimation based on equation (36). The results are plotted in the lower-left panel of figure 5. The symbols correspond to the estimate using equation (28) with $P(k)$ evaluated from the Peacock–Dodds prescription, i.e., $P_{\text{PD}}(k)$. The solid and dashed lines use the direct estimate for $P(k)$, instead, i.e., $P_{\text{sim}}(k)$. The symbols and the lines deviate for $k < 0.1 h$ Mpc$^{-1}$ where the effect of complicated survey volume shape cannot be ignored. For $k > 0.1 h$ Mpc$^{-1}$ the estimate of $b_1$ in wedge configuration well reproduces that in s-cube, upper-right panel of figure 5.

Finally, $b_1(k)$ for SDSS galaxies is computed similarly as wedge subsamples. Again we assume the same cosmological parameters used in the $N$-body simulations. As in the wedge case, we used $P(k) = P_{\text{PD}}(k)$ and $P_{\text{sim}}(k)$ for the symbols and lines, respectively. The lower-right panel of figure 5 is the result. The red (blue) sample is almost the same as L (S) in the lower-left panel. The lower-left panel of figure 5 is indeed in good agreement with the upper panels for $k < 0.2 h$ Mpc$^{-1}$, which we may interpret as being an indication that the SDSS results in redshift space are directly related to their real-space property.

We find very similar deviations between the lines and the symbols at $k < 0.1$, which ensures our interpretation due to the effect of the complicated survey boundary shape.

### 3.3. Bispectrum and Q

Finally, we are in a position to consider the three-point statistics. We calculate the bispectra $B_i(k_1, k_2, k_3)$ using the same methods as $P_i(k)$ for r-cube and s-cube. For wedge and SDSS subsamples we use the formula

$$ \langle F_i(k_1) F_i(k_2) F_i(k_3) \rangle = \frac{I_{33}}{I_{22}^{1/2}} \left( B_i(k_1, k_2, k_3) + \frac{I_{32}^2}{I_{33}} [P_i(k_1) + P_i(k_2) + P_i(k_3)] + (1 - \alpha^2) \frac{I_{31}^2}{I_{33}} \right) $$

(37)

(Verde et al. 2002). Figure 6 plots $Q_i(k)$, the reduced amplitude of the bispectrum for equilateral triangles,

$$ Q_i(k) \equiv \frac{B_i(k, k, k)}{3P_i^2(k)}, $$

(38)

which should be compared with figure 5. The difference of $b_1$ between the two different subsamples in each panel of figure 5 does not show up in figure 6. This is fully consistent with the finding of Kayo et al. (2004) for three-point correlation functions. Furthermore, comparisons among simulated halo subsamples indicate that this feature is not a simple outcome of the redshift distortion effect, but reflects the intrinsic correlation between $b_1$ and $b_2/b_1$ in real space, as we discussed in section 2.

In order to proceed further, we attempt to estimate $b_2(k)/b_1(k)$ by combining $Q_i(k)$ and $Q_{\text{m}}(k)$. We directly compute $Q_{\text{m}}(k)$ from simulation particles for r-cube, s-cube, and wedge (filled triangles in figure 6). For comparison, tree-level perturbation theory predicts that

$$ Q(k_1, k_2, k_3) = 2 \times \frac{F_3(k_1, k_2) P(k_1) P(k_2) + F_2(k_2, k_3) P(k_2) P(k_3) + F_2(k_3, k_1) P(k_3) P(k_1)}{P(k_1) P(k_2) + P(k_2) P(k_3) + P(k_3) P(k_1)}. $$

(39)
where we follow Jain and Bertschinger (1994) and define $F_2(k_1, k_2)$ as

$$F_2(k_1, k_2) = \frac{5}{7} + \frac{1}{2} \frac{k_1 \cdot k_2}{k_1 k_2} \left( \frac{k_1}{k_2} + \frac{k_2}{k_1} \right) + 2 \frac{(k_1 \cdot k_2)^2}{7 k_1 k_2}.$$  

Strictly speaking, this expression of $F_2(k_1, k_2)$ is valid only for the Einstein–de Sitter universe, but its dependence on cosmology has proved to be very weak (Matsubara 1995; Scoccimarro et al. 1998). For equilateral triangles, equation (39) reduces to a constant value, $4/7$, which is plotted as a dashed line (PT) in the upper-left panel of figure 6 for real space. In redshift space, one obtains $Q = 0.533$ by setting $\gamma = 0$ and $\beta = \beta_m$ in the perturbation expression, equation (26), which is also plotted as a dashed line (PT) in the upper-right and lower-left panels. For $k < 0.2 \, h \, \text{Mpc}^{-1}$, the direct estimates (filled triangles; DM) agree with the perturbation values. In directly evaluating $Q$ from $N$-body simulations, we use equation (8) for $Q_m$, and a similar expression for $Q_i$.

Figure 7 plots $b_2/b_1$ computed from

$$(b_2/b_1)_i(k) \equiv b_{1,i}(k) Q_i(k) - Q_m(k).$$

For SDSS subsamples, we use the value evaluated from wedge for $Q_m(k)$. Strictly speaking, equation (41) ignores redshift-space distortion effects. In reality, however, we made sure that almost the same values were derived even when we used equation (26) instead. Therefore we use equation (41) both in real- and redshift-space. A comparison between figure 5 and figure 7 indicates that subsamples with larger $b_1$ tend to have a larger $b_2/b_1$. This is qualitatively consistent with the analytical results shown in figure 2.

To compare more quantitatively the simulation results with analytic models, we averaged $b_{1,i}(k)$ and $(b_2/b_1)_i(k)$ over the range of $0.08 \, h \, \text{Mpc}^{-1} < k < 0.2 \, h \, \text{Mpc}^{-1}$. The results are plotted in the left panel of figure 8. The three curves show the analytic models plotted in figure 2 at $z = 0$, while the squares correspond to the mass-averaged values for ellipsoidal halo model, again in figure 2. The triangles, crosses, and circles represent r-cube, s-cube, and wedge subsamples, respectively. If the analytic
ellipsoidal halo model is exact, the triangles and squares should agree within the error bars (we estimate from variance for three different realizations). The small differences between them may be ascribed to: i) the inaccuracy of the higher order biasing coefficients in the halo model. The parameters ($p$ and $q$) in equation (14) are empirically determined so as to reproduce the mass function, but they do not guarantee any reliability for $b_2$; ii) a weak scale dependence, which is seen in figures 5 and 7. This is not expected in the analytic model, implying its practical limitation; iii) inaccuracy of the estimators of $b_1$ and $b_2 = b_1$; and iv) stochasticity of biasing (Dekel & Lahav 1999; Taruya & Suto 2000; Taruya et al. 2001; Yoshikawa et al. 2001). The present halo model assumes deterministic biasing, which may affect the mean values of the biasing coefficients, especially in the higher order. Given those realistic issues, we would interpret that the triangles and squares are in reasonable agreement.

The right panel of figure 8 plots $b_2/b_1$ against $b_1$ for SDSS galaxy subsamples, which should be compared with the left panel for simulated halos. In addition to the color-selected subsamples in table 1, we construct nine luminosity threshold subsamples (table 3) following Zehavi et al. (2005). We construct realistic galaxy mock samples in order to see if the observational data can be understood from simple theoretical modeling. We employ a halo occupation distribution approach (HOD), which assigns galaxies within simulated halos. According to the simplest version of HOD, the mean number of galaxies within halos of mass $M$ is set as

$$N(M) = \begin{cases} 1 + \left( \frac{M}{M_1} \right)^\alpha, & M > M_{\min}, \\ 0, & \text{otherwise}. \end{cases}$$  \hspace{1cm} (42)$$

In the above expression, the first term represents a central galaxy, while the second corresponds to satellite galaxies. The three parameters ($M_{\min}$, $M_1$, and $\alpha$) are determined to reproduce the observed sample of galaxies. In practice, we adopt the values (table 3) fitted by Zehavi et al. (2005) for the SDSS luminosity galaxies (see their table 2). Their fit assumes the density profile proposed by Navarro, Frenk, and White (1996) for the galaxy distribution in each halo, and satellite galaxies within each halo are assigned following the Poisson distribution of the mean value of equation (42). We construct mock galaxy samples from our simulated halo catalogs using the routine of Skibba et al. (2006), which simultaneously takes care of the redshift-space distortion.

Fig. 7. $b_2/b_1$ for simulated halos (table 2) and SDSS galaxies (table 1). The crosses and the filled circles for simulated halos (SDSS galaxies) correspond to the S and L samples (blue and red), respectively. The quoted error bars for simulated halos are computed from three different realizations. We simply use the error bars for wedge subsamples just for reference in the case of SDSS galaxies.
Correlation between $b_1$ and $b_2/b_1$. Left: analytic models and simulated halos (table 2). The lines are analytical predictions for spherical (solid), ellipsoidal (dashed) halo models, and peak model (dotted) as in figure 2 at $z = 0$. The squares are the mass-averaged values corresponding to S and L subsamples for the ellipsoidal halo model. Other symbols are for halo subsamples. The quoted error bars are computed from three different realizations. Right: SDSS color-selected samples (table 1; crosses), SDSS luminosity threshold samples (table 3; triangles), HOD mock galaxies (table 3; squares), and HOD analytic prediction (circles). Note that we do not distinguish $b_n$ and $B_n$ here.

Table 3. HOD parameters corresponding to the SDSS luminosity threshold galaxies.

| $M_{r, \text{max}}$ | $z_{\text{min}}$ | $z_{\text{max}}$ | $N_{\text{SDSS, gal}}^{\text{SDSS, gal}}$ | $\log_{10} M_{\text{min}}$ | $\log_{10} M_1$ | $\alpha$ | $(N_{\text{mock, gal}}^{\text{mock, gal}})$ |
|---------------------|------------------|------------------|--------------------------------------|------------------|----------------|--------|--------------------------------------|
| -22.0 ............ | 0.02             | 0.22             | 7704                                 | 13.91            | 14.92          | 1.43   | 1417                                  |
| -21.5 ............ | 0.02             | 0.19             | 24711                                | 13.27            | 14.60          | 1.94   | 7235                                  |
| -21.0 ............ | 0.02             | 0.15             | 41969                                | 12.72            | 14.09          | 1.39   | 28151                                 |
| -20.5 ............ | 0.02             | 0.13             | 69217                                | 12.30            | 13.67          | 1.21   | 75292                                 |
| -20.0 ............ | 0.02             | 0.10             | 63770                                | 12.01            | 13.42          | 1.16   | 141424                                |
| -19.5 ............ | 0.02             | 0.08             | 56384                                | 11.76            | 13.15          | 1.13   | 246421                                |
| -19.0 ............ | 0.02             | 0.06             | 30532                                | 11.59            | 12.94          | 1.08   | 365719                                 |
| -18.5 ............ | 0.02             | 0.05             | 24636                                | 11.44            | 12.77          | 1.01   | 514269                                 |
| -18.0 ............ | 0.02             | 0.04             | 16123                                | 11.27            | 12.57          | 0.92   | 743540                                 |

According to the current HOD approach, we evaluate the biasing coefficients in two ways. One is an analytical estimate (HOD analytic prediction), which is based on equation (25):

$$B_n = \frac{\int_{M_{\text{min}}}^{M_{\text{max}}} dM n_{\text{halo}}(M, z = 0)(N(M))b_n(M, z = 0)}{\int_{M_{\text{min}}}^{M_{\text{max}}} dM n_{\text{halo}}(M, z = 0)(N(M))}, \quad (n = 1, 2),$$

where we use the Sheth–Tormen mass function, equation (14), for $n_{\text{halo}}(M, z = 0)$, and equation (20) for $b_n(M, z = 0)$, and $M_{\text{max}}$ is set as the maximum mass of our simulated halos. The other is a direct evaluation of the biasing coefficients from the HOD mock galaxy samples using equations (28) and (41). The right panel of figure 8 plots these results for SDSS color selected samples (crosses), for SDSS luminosity threshold samples (triangles), for HOD analytic prediction (circles), and finally for HOD mock
galaxies (squares with error bars). We note a few interesting features in the plot: i) The similarity between the color-selected and luminosity threshold samples. While they are based on different selection criteria, the resulting $b_2/b_1 - b_1$ correlation seems to be roughly the same. ii) The HOD analytic prediction reproduces the values of $b_2/b_1$ and $b_1$ for each luminosity threshold galaxy samples. This is surprising since the fitting procedure is designed to reproduce their two-point correlation functions alone. iii) The discrepancy between the HOD mock and analytical results may come from the difference between Sheth–Tormen model and our halo samples in $n_{\text{hala}}$ and/or $b_1$. We made sure that $n_{\text{hala}}$ of our halo samples is in good agreement with equation (14). On the other hand as the left panel of figure 8 indicates, $b_2$ for our simulated halos is systematically lower than the analytic prediction, equation (25). Thus, the agreement between the SDSS data and the HOD models needs to be interpreted with caution. We also computed the values of $b_2/b_1$ and $b_1$ of HOD mock galaxy samples in real space. We, however, do not plot the results, since they are almost the same as those in redshift space.

Even with the above subtlety, we find a fairly generic correlation between $b_2/b_1$ and $b_1$, which is the most important result of the current study. We note here that Gaztañaga et al. (2005) performed a related analysis using 2dF galaxy data. They found that $b_1 = 0.93^{+0.10}_{-0.08}$ and $b_2/b_1 = -0.34^{+0.11}_{-0.08}$. These values are in good agreement with the right panel of figure 8. Incidentally, they speculated a crude correlation of $b_2/b_1 \sim b_1 - 1.2$, which is indeed consistent with our finding on the basis of systematic results from analytic, and numerical simulations and SDSS galaxy data.

4. Summary and Discussion

We have found a fairly generic correlation between linear and quadratic biasing coefficients, $b_1$ and $b_2/b_1$, using a variety of different methodologies: perturbative expansions of peak and halo biasing models, $N$-body simulations of halos, SDSS galaxy data analysis, and the corresponding halo occupation distribution predictions (analytic and mock). The presence of such correlations was suggested earlier by a previous finding that the normalized three-point correlation functions of SDSS galaxies in redshift space follow the hierarchical relation approximately, $Q = 0.5\text{–}1.0$, despite the robust morphological, color and luminosity dependences of the corresponding two-point correlation functions (Kayo et al. 2004). We derived the $b_1$–$b_2$ correlation explicitly and showed for the first time that it indeed explains the observed behavior of $Q$ for equilateral triangles calculated from SDSS galaxies in linear regimes.

The major findings of the present paper are summarized as follows:

- Even with the presence of redshift distortion and complicated survey shape effects, $b_1$ can be accurately estimated in $k < 0.2\ Mpc^{-1}$. Thus, $b_1$ estimated from SDSS galaxies is expected to reflect the true value in real space.
- The values of $Q_0$ for equilateral triangles from simulated halos, mock galaxies based on HOD model, and SDSS galaxies do not show any clear dependence on $b_1$. The dependence on $b_1$ is not an artifact from the redshift-space distortion contamination, but is a consequence of the intrinsic correlation between $b_2/b_1$ and $b_1$.
- The HOD model, equation (42), whose parameters are determined so as to reproduce the observed two-point statistics alone, seems to also be successful in predicting the value of $b_2$.

There are a few remaining tasks that should be done following the present result. First, it is interesting to compare the current analysis in Fourier space with that in configuration space. As long as the nonlinear effect is negligible, $Q_{m}$ and $Q_{b}$ in $k$-space are expected to be the same as those in configuration space. Nevertheless, we would like to make sure of it, and to further explore the connection between bispectra and the three-point correlation functions of halos and galaxies. Second, the success of HOD to reproduce the correlation between $b_2/b_1$ and $b_1$ at the current level is very encouraging. Naturally, it is important to further improve the HOD model. Third, the present paper considers only equilateral configuration of Fourier space triangles for $Q$. In reality, however, the different shapes of triangles has a wealth of information. While the $b_1$ dependence of $Q$ barely vanishes for equilateral triangles due to the correlation between $b_1$ and $b_2/b_1$, this is not the case for arbitrary triangle shapes. Therefore, it is interesting to explore whether the $b_1$ dependence of $Q$ for the other triangle shapes and to see if the results are consistent with the present analysis and/or to constrain higher order biasing coefficients more tightly. Work along the line is now in progress.

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