The 2dF Galaxy Redshift Survey: The bias of galaxies and the density of the Universe

Licia Verde\textsuperscript{1,2}, Alan F. Heavens\textsuperscript{3}, Will J. Percival\textsuperscript{3}, Sabino Matarrese\textsuperscript{4}, Carlton M. Baugh\textsuperscript{5}, Joss Bland-Hawthorn\textsuperscript{6}, Terry Bridges\textsuperscript{6}, Russell Cannon\textsuperscript{6}, Shaun Cole\textsuperscript{5}, Matthew Colless\textsuperscript{7}, Chris Collins\textsuperscript{8}, Warrick Couch\textsuperscript{9}, Gavin Dalton\textsuperscript{10}, Roberto De Propris\textsuperscript{9}, Simon P. Driver\textsuperscript{11}, George Efstathiou\textsuperscript{12}, Richard S. Ellis\textsuperscript{13}, Carlos S. Frenk\textsuperscript{5}, Karl Glazebrook\textsuperscript{14}, Carole Jackson\textsuperscript{7}, Ofer Lahav\textsuperscript{12}, Ian Lewis\textsuperscript{10}, Stuart Lumsden\textsuperscript{15}, Steve Maddox\textsuperscript{16}, Darren Madgwick\textsuperscript{12}, Peder Norberg\textsuperscript{5}, John A. Peacock\textsuperscript{3}, Bruce A. Peterson\textsuperscript{7}, Will Sutherland\textsuperscript{3}, Keith Taylor\textsuperscript{6}

\textsuperscript{1} Department of Physics and Astronomy, Rutgers University, 136 Frelinghuysen Road, Piscataway, NJ 08854–8019 USA.
\textsuperscript{2} Princeton University Observatory, Princeton, NJ 08544, USA.
\textsuperscript{3} Institute for Astronomy, University of Edinburgh, Royal Observatory, Blackford Hill, Edinburgh EH9 3HJ, UK
\textsuperscript{4} Dipartimento di Fisica, ‘G. Galilei’, Università di Padova, and INFN, Sezione di Padova, via Marzolo 8, I-35131 Padova, Italy
\textsuperscript{5} Department of Physics, University of Durham, South Road, Durham DH1 3LE, UK
\textsuperscript{6} Anglo-Australian Observatory, P.O. Box 296, Epping, NSW 2121, Australia
\textsuperscript{7} Research School of Astronomy & Astrophysics, The Australian National University, Weston Creek, ACT 2611, Australia
\textsuperscript{8} Astrophysics Research Institute, Liverpool John Moores University, Twelve Quays House, Birkenhead, L14 1LD, UK
\textsuperscript{9} Department of Astrophysics, University of New South Wales, Sydney, NSW 2052, Australia
\textsuperscript{10} Department of Physics, University of Oxford, Keble Road, Oxford OX1 3RH, UK
\textsuperscript{11} School of Physics and Astronomy, University of St Andrews, North Haugh, St Andrews, Fife, KY6 9SS, UK
\textsuperscript{12} Institute of Astronomy, University of Cambridge, Madingley Road, Cambridge CB3 0HA, UK
\textsuperscript{13} Department of Astronomy, California Institute of Technology, Pasadena, CA 91125, USA
\textsuperscript{14} Department of Physics & Astronomy, Johns Hopkins University, Baltimore, MD 21218-2686, USA
\textsuperscript{15} Department of Physics, University of Leeds, Woodhouse Lane, Leeds, LS2 9JT, UK
\textsuperscript{16} School of Physics & Astronomy, University of Nottingham, Nottingham NG7 2RD, UK

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ABSTRACT

We compute the bispectrum of the 2dF Galaxy Redshift Survey (2dFGRS) and use it to measure the bias parameter of the galaxies. This parameter quantifies the strength of clustering of the galaxies relative to the mass in the Universe. By analysing 80 million triangle configurations in the wavenumber range $0.1 < k < 0.5 \, h^{-1} \text{Mpc}$ (i.e. on scales roughly between 5 and 30 $h^{-1} \text{Mpc}$) we find that the linear bias parameter is consistent with unity: $b_1 = 1.04 \pm 0.11$, and the quadratic (nonlinear) bias is consistent with zero: $b_2 = -0.054 \pm 0.08$. Thus, at least on large scales, optically-selected galaxies do indeed trace the underlying mass distribution. The bias parameter can be combined with the 2dFGRS measurement of the redshift distortion parameter $\beta \simeq \Omega_m^{0.6} / b_1$, to yield $\Omega_m = 0.27 \pm 0.06$ for the matter density of the Universe, a result which is determined entirely from this survey, independently of other datasets.

Key words: galaxies: clustering– cosmology: large-scale structure of Universe, cosmological parameters

1 INTRODUCTION

Clustering of mass in the Universe is believed to be a result of amplification by gravitational instability of small perturbations generated in the early Universe. Comparison with theoretical predictions offers the chance to test models of the generation of the perturbations, as well as putting important constraints on cosmological parameters, which control the growth rate of the perturbations. A fundamental limitation on such a comparison has been that theoretical models predict the clustering properties of the mass in the Universe, and yet we have few direct measures of mass observationally. More readily observable is the distribution of luminous objects such
as galaxies, so to compare with theory one has to determine, or assume, the relationship between the clustering of mass and the clustering of galaxies. In general one will expect these to differ, because the efficiency of galaxy formation may depend in some nontrivial way on the underlying mass distribution. The idea that structures may be ‘biased’ tracers of the mass distribution goes back to Kaiser (1984), who explained the high clustering strength of Abell clusters as due to their forming in high-density regions of the Universe. In addition, observations indicating that different types of galaxy cluster differently (e.g. Dressler 1980; Postman & Geller 1984; White, Tully & Davis 1988; Hamilton 1988; Lahav, Nemiroff & Piran 1990; Lahav & Saslaw 1992) show that they cannot all be unbiased tracers of the mass. Bias became an attractive way to reconcile the low velocities of galaxies with the high-density Einstein-de Sitter model favoured in the 1980s (e.g. Davis et al. 1985), but after the Cosmic Background Explorer (COBE) determined the amplitude of primordial fluctuations on large scales (Smoot et al. 1992), the ‘standard’ biased Cold Dark Matter (CDM) model became less popular. With the advent of more detailed datasets in the cosmic microwave background (CMB) and large-scale structure, it is possible to investigate and constrain a wider range of galaxy formation models, and an unknown bias relation adds uncertainty to the process.

Since the efficiency of galaxy formation is not well understood theoretically, it makes sense to try to measure it empirically from observations. When the perturbations are small (or on large, linear scales), it is difficult to do this: there is a degeneracy between the unknown amplitude of the matter power spectrum $P(k)$ and the degree of bias, $b$, defined such that the galaxy power spectrum is $P_b(k) \equiv b^2 P(k)$. In principle, $b$ may be a function of scale, through the wavenumber $k$. At later times (or on smaller scales), however, the degeneracy is lifted by nonlinear effects. One feature of nonlinear gravitational evolution is that the overdensity field $\delta(x) \equiv (\rho(x)/\bar{\rho}) - 1$ becomes progressively more skewed towards high density. In principle skewness could also arise from non-Gaussian initial conditions; in practice this can be neglected (Verde et al. 2000), since CMB fluctuations are consistent with Gaussian initial conditions (Komatsu et al. 2001; Santos et al. 2001).

One can thus hope to exploit the gravitational skewness, but skewness could equally well arise from biasing, e.g. from a galaxy formation efficiency that increased at dense points in the mass field. It is nevertheless possible to distinguish these two effects by considering the *shapes* of isodensity regions. If the field is unbiased, then the shapes of isodensity contours become flattened, as gravitational instability accelerates collapse along the short axis of structures, leading to sheet-like and filamentary structures (e.g. Zeldovich 1970). If the galaxy field is highly biased with the same power spectrum, however, the underlying mass field is of low amplitude, and thus expected to be close to the initial field, which is assumed to be Gaussian. These fields do not have highly-flattened isodensity contours, as bias does not flatten the contours; for example, Eulerian bias preserves the contour shape. Thus there is a difference which could be detected, for example, by studying the three-point correlation function. In this paper, we exploit this effect in Fourier space rather than real space, by analysing the bispectrum: $\langle \delta_{k_1} \delta_{k_2} \delta_{k_3} \rangle$, where $\delta_k$ is the Fourier transform of the galaxy overdensity field. The theory for the bispectrum is set out in Fry (1994), Hivon et al. (1995), Matarrese, Verde & Heavens (1997), Verde et al. (1998), Scoccimarro et al. (1998), Scoccimarro, Couchman & Frieman (1999) and Scoccimarro (2000).

The galaxy survey we use is the Anglo-Australian Telescope 2 degree field Galaxy Redshift Survey (Colless et al. 2001), as compiled in February 2001. It was created with the 2dF multi-fibre spectrograph on the Anglo-Australian Telescope (Lewis et al. 2001), and currently consists of over 200,000 galaxies with redshifts up to about $z = 0.3$, broadly in two regions centred near the south and north galactic poles. See http://www.mso.anu.edu.au/2dFGRS/ for further details. It is the first survey which is large enough to put tight constraints on the bias parameter, as previous surveys are too shallow or too sparse. In this paper, we use 127,000 galaxies from the February 2001 compilation of the catalogue, truncated at $0.03 < z < 0.25$.

The outline of the paper is as follows: in Section 2 we review the theory of growth of the bispectrum through gravitational instability, and discuss briefly the effects of redshift-space distortions; in Section 3 we illustrate our method of measuring the bias parameter. This method uses a new estimator of the bias parameter, which allows us to analyse many millions of $k$-vector triplets, thus drastically improving the signal to noise. In Section 4 we test the performance of the method. Finally in Section 5 we present our results and in Section 6 we discuss the implications of these results. An Appendix presents and describes in detail our new estimator of the bias parameter.

## 2 THE BISPECTRUM IN REAL AND REDSHIFT SPACE

The statistic we use to measure the bias of the galaxies is the bispectrum $B$, which is related to the three-point correlation function in Fourier space. For the mass, this is defined by

$$\langle \delta_{k_1} \delta_{k_2} \delta_{k_3} \rangle \equiv \langle (2\pi)^3 B(k_1, k_2, k_3) \delta^3(k_1 + k_2 + k_3) \rangle,$$

where $\delta_k \equiv \int d^3x \delta(x) \exp(-i \mathbf{k} \cdot \mathbf{x})$ is the Fourier transform of the mass overdensity $\delta(x) \equiv \rho(x)/\bar{\rho} - 1$ and $\delta^3$ is the Dirac delta function, which shows that the bispectrum can be non-zero only if the $k$-vectors close to form a triangle.

The power spectrum $P$ is similarly defined by

$$\langle \delta_{k_1} \delta_{k_2} \rangle \equiv \langle (2\pi)^3 P(k) \delta^3(k + k') \rangle.$$

Analogous relations hold for the power spectrum and bispectrum of the galaxy distribution. We assume that the mass overdensity is a Gaussian random field initially, as closely predicted by inflationary early universe models. Thus, at asymptotically early times the bispectrum is zero by symmetry. As gravitational instability develops, the field becomes asymmetric, because nonlinear effects skew the density field to high densities. In this way, a non-zero bispectrum develops. In the mildly nonlinear regime, we use second-order perturbation theory to compute the expected bispectrum. To second order (in the overdensity $\delta$) the Fourier coefficients develop a nonlinear component which is proportional to $\delta^2$, so the leading order term in the bispectrum grows like $\delta^4$. Since the 2dFGRS is not a survey of mass density, to interpret the bispectrum measured from the survey we must make some assumption about the distribution of mass relative to the distribution of galaxies. To date, this uncertainty in the relationship between the mass and the galaxy distribution has placed a limitation on the usefulness of galaxy catalogues as a probe of cosmology. We make the assumption that the galaxy overdensity field $\delta_g$ is the underlying mass overdensity by some deterministic function, which we expand in a Taylor series as (cf. Fry & Gaztanaga 1993)

$$\delta_g = \sum_{i=0}^{\infty} \frac{b_i \delta^i}{i!}.
$$
We must keep terms up to \( i = 2 \), since these enter in the bispectrum at the same level as second-order perturbation theory growth terms, and we ignore higher-order terms. In order for \( \delta_0 \) to have zero mean, there is a (calculable) \( b_0 \) term, but we ignore it as it contributes only to \( k = 0 \); \( b_1 \) is the linear bias parameter, and \( b_2 \) is the quadratic bias parameter. A non-zero \( b_2 \) would indicate nonlinear biasing of galaxies with respect to mass. Both of these parameters are estimated in this paper from the 2dFGRS.

In real space, the two effects of nonlinear growth and nonlinear bias contribute terms to a non-zero bispectrum:

\[
B(k_1, k_2, k_3) = P_g(k_1)P_g(k_2) \left( \frac{J(k_1, k_2)}{b_1} + \frac{b_2}{b_1^2} \right) + \text{cyc.} \tag{4}
\]

where there are two additional cyclic terms \((2,3)\) and \((3,1)\). Details of the theory leading to (4) may be found in e.g. Matarrese, Verde & Heavens (1997). We assume here that the galaxy power spectrum is \( P_g(k) = b_1^2 P(k) \) (see Heavens, Matarrese & Verde 1998 for discussion of this point); \( J \) is a function that depends on the shape of the triangle in \( k \)-space, but only very weakly on cosmology (e.g. Bouchet et al. 1992; Bouchet et al. 1995; Catelan et al. 1995). Note that we assume a deterministic bias; other authors (e.g. Scherrer & Weinberg 1998; Dekel & Lahav 1999; Taruya, Koyama & Soda 1999) have investigated stochastic bias, where there is a random component to the relationship between \( \delta \) and \( \delta_0 \). In the case of stochastic bias the bispectrum has still the form of equation (3) but the function \( J \) is modified into \( J' \) in such a way that when the correlation coefficient of the stochastic bias \( r \) goes to unity (i.e. bias becomes deterministic), \( J' \rightarrow J \). Theoretical considerations suggest that \( r \sim 1 \) on scales of interest (e.g. Tegmark & Peebles 1998; Blanton et al. 2000; Seljak 2000).

In redshift space, both the power spectrum and the bispectrum are modified by redshift-space distortions, arising because the distance estimator (the redshift) is perturbed by peculiar velocities. These distortions are radial in nature, and can be analysed, at some expense in complexity, in radial and angular basis functions (e.g. Fisher, Scharf & Lahav 1994; Heavens & Taylor 1995; Ballinger, Heavens & Taylor 1995; Hamilton 1998; Tadros et al. 1999). Here, we adopt the distant-observer approximation (Kaiser 1987), and assume that nonlinear effects can be modelled by an incoherent small-scale velocity field, characterised by the pairwise velocity dispersion \( \sigma_p \). The large-scale in-fall leads to distortions which depend on the redshift distortion parameter \( \beta = \Omega_m^{1/6}/b_1 \), where \( \Omega_m \) is the matter density parameter. Assuming in addition an exponential distribution for the pairwise velocity the combined effect gives the power spectrum in redshift space (denoted by subscript \( s \))

\[
P_s(k) = \frac{P(k)(1 + \beta \mu^2)^2}{1 + k^2 \sigma_p^2 \sigma_p^2/2} \tag{5}
\]

(e.g. Ballinger, Peacock & Heavens 1996; Hatton & Cole 1998), where \( \mu \) is the cosine of the wavevector to the line of sight, which is a fixed direction in the distant observer approximation. Other modifications have been suggested, such as a Gaussian, or exponential one-particle velocity dispersion, which yield different functional forms for the redshift distortion. Note that \( \sigma_p \) is usually written, as here, implicitly divided by the Hubble constant. The bispectrum is modified similarly, and again various modifications have been proposed. We use the form

\[
B_s(k_1, k_2, k_3) = (B_{12} + B_{23} + B_{31}) \times \left[ \left( 1 + \frac{\alpha_1^2 k_1^2 \sigma_p^2}{2} \right) \left( 1 + \frac{\alpha_2^2 k_2^2 \sigma_p^2}{2} \right) \left( 1 + \frac{\alpha_3^2 k_3^2 \sigma_p^2}{2} \right) \right]^{-1/2} \tag{6}
\]

where

\[
B_{12} = (1 + \beta \mu_1^2)(1 + \beta \mu_2^2) \frac{\text{Ker}(k_1, k_2)}{b_1} + \frac{b_2}{b_1^2} P_g(k_1)P_g(k_2)
\]

and the kernel function \( \text{Ker} \) is \( J \) modified for redshift space (see Verde et al. 1998 equation (13) for the formula); \( \alpha_V \) is an adjustable parameter which is shape-dependent, and must be calibrated from simulations. Scoccimarro, Couchman & Frieman (1999) propose an alternative* for the denominator of (6), namely

\[
(1 + \alpha_S^2 k_1^2 \sigma_p^2 + k_2^2 \sigma_p^2 + k_3^2 \sigma_p^2/2)^2
\]

where the one-particle dispersion \( \sigma_p = \sigma_0/\sqrt{2} \) if the small-scale velocities are incoherent, and again the parameter \( \alpha_S \) needs to be calibrated for different triangle shapes and cosmologies. We find that the formula (6) recovers the true bias parameter in an ensemble of simulated biased 2dFGRS catalogues with smaller scatter than (8). We see from (6) and (7) how the bispectrum can allow us to measure the bias parameters. The left-hand side is potentially observable, and we can hope to constrain \( b_1 \) and \( b_2 \) by considering triangles of different shape (and hence different \text{Ker}).

3 METHOD

The previous Section shows the theoretical model for the bispectrum, and its dependence on the two parameters \( b_1 \) and \( b_2 \); note that, apart from these two parameters, the bispectrum depends on observable quantities such as \( \beta \), \( \sigma_p \), and \( P_s \). The real-space galaxy power spectrum is obtained from the angle-averaged redshift-space power (cf. Verde et al. 1998):

\[
P(k) = P_s(k) \left[ \frac{4(\sigma_p^2 k^2 - \beta \mu_1^2 \beta_2^2)}{\sigma_p^2 k^4} + \frac{2\beta^2}{3\sigma_p^4 k^2} + \frac{\sqrt{2}(\sigma_p^2 k^2 - 2\beta \mu_1^2 \beta_2^2)}{k^2 \sigma_p^4} \tan^{-1} \left( \frac{k \sigma_p}{\sqrt{2}} \right) \right]^{-1}
\]

The real-space power spectrum obtained this way agrees well with the APM power estimated by Baugh & Efstathiou (1994) and Efstathiou & Moody (2001). The input catalogue is based on a revised and extended version of the APM galaxy catalogue (Maddox et al. 1990a).

The bispectrum and power spectrum data come from transforming the galaxy distribution as follows. The galaxies are weighted with the optimum weight for measuring the power spectrum (Feldman, Kaiser & Peacock 1994), which also minimises the variance of higher-order correlation functions (Scoccimarro 2000). The weight is \( w(r) = 1/[1 + P_0 \overline{\mu}(r)] \), where \( \overline{\mu}(r) \) is the average number density of galaxies at position \( r \), and \( P_0 \) is the power spectrum to be estimated. For speed reasons, \( P_0 \) was fixed at 5000 \( h^{-1} \)Mpc\(^3\) so that a Fast Fourier transform could be employed. This is optimal for minimising the variance at \( k \sim 0.1 \)h Mpc\(^{-1}\), whereas our signal comes from wavenumbers with a smaller power, but in fact altering \( P_0 \) hardly changes the results.

* Note that in equation 38 of Scoccimarro, Couchman & Frieman (1999) there is an extra power of 2 outside the square brackets, which we omit in (8).
3.1 Estimating the bispectrum

In this Section we describe how we estimate the bispectrum from the data, taking into account the survey shape, selection function and shot noise.

We follow Feldman, Kaiser & Peacock (1994) and Matarrese, Verde & Heavens (1997), and transform the field

\[ F(r) \equiv \lambda w(r) [n(r) - \alpha n_v(r)] \]

where \( \lambda \) is a constant to be determined, \( n_v(r) \) is the number density of a random catalogue with the same selection function as the real catalogue, but with \( 1/\alpha \) times as many particles. If we set \( \lambda = I_{22}^{-1/2} \), where

\[ I_{ij} \equiv \int d^3 r \, w^i(r) w^j(r) \]

(Matarrese, Verde & Heavens 1997), then the power spectrum may be estimated from

\[ \langle |F_k|^2 \rangle = P_g(k) + \frac{I_{21}}{I_{22}} (1 + \alpha) \]

and the bispectrum from

\[ \langle F_{k_1} F_{k_2} F_{k_3} \rangle = \frac{I_{33}}{I_{22}} \left[ B_g(k_1, k_2, k_3) + \frac{I_{32}}{I_{33}} \left[ P_g(k_1) + P_g(k_2) + P_g(k_3) \right] + (1 - \alpha^2) \frac{I_{41}}{I_{33}} \right] . \]

An underlying assumption is that the power spectrum is roughly constant over the width of the survey window function in \( k \) space. Because of the rather flat geometry of the survey regions, and the holes due to star drills, this criterion is not satisfied in detail. In practice, we have used multiple mock catalogues (Cole et al. 1998) with the same selection criteria as the survey regions to check that this assumption does not feed through into a biased estimate of the bias parameters. Also, the power spectrum and bispectrum estimates are convolved with the window function. This can lead to changes in shape from convolution, and erroneous interpretation of correlated noise as real features in the power spectrum. These effects are, however, not important in the wavenumber range \( 0.1 \leq k \leq 0.5 \, h \, \text{Mpc}^{-1} \) which we use for the bispectrum analysis (Percival et al. 2001).

To compute the \( F_k \), we generated random catalogues with approximately 5 times as many particles as the real catalogue, and analysed the North and South Galactic Pole regions (NGP & SGP) separately. We ignored the random fields present in the 2dFGRS as these complicate the window function and add very little information for the current analysis. Fast Fourier transforms were performed on a \( 512 \times 512 \times 256 \) grid which encompassed all the particles, leading to a grid spacing of about \( 1 \, h^{-1} \, \text{Mpc} \).

3.2 Choice of triangles

We use the real parts of \( F_{k_1} F_{k_2} F_{k_3} \) as our data, for triangles in \( k \) space (i.e. where \( k_1 + k_2 + k_3 = 0 \)). Each triangle allows us to estimate a linear combination of the parameters natural to this analysis: \( c_1 \equiv 1/b_1 \) and \( c_2 \equiv b_2/b_1 \), through equations (6) and (7). Note that we must use triangles of different shape (and hence different \( K_{\sigma} \{ k_1, k_2 \} \)) to lift the degeneracy between nonlinear gravity and nonlinear bias. As explained in the introduction, this is equivalent to analysing the shapes of structures, which are different in the two cases.

There is clearly a huge number of possible triangles to investigate (many millions), and we are faced with a problem of how to choose the triangles to analyse. The problem is that triangles which have a wavevector in common will be correlated, through cross-terms in the 6-point function. A likelihood analysis as originally suggested in Matarrese, Verde & Heavens (1997) and Verde et al. (1998) with millions of correlated data is unfeasible, so we take a different approach. We use two sets of triangles of different configurations: one set with one wavevector twice the length of another, and another set with two wavevectors of common length. For the former, calibrations with mock catalogues give \( \alpha_{V} = 1.8 \), and the latter \( \alpha_{V} = 1.0 \). We set a lower limit to the wavenumber range of \( k = 0.1 \, h \, \text{Mpc}^{-1} \), to avoid the effects of convolution with the window function (Percival et al. 2001). We set an upper limit of \( k = 0.5 \, h \, \text{Mpc}^{-1} \) where, for the 2dFGRS power spectrum, shot noise begins to dominate the signal so that little further is gained by increasing the limit. In addition, we impose a constraint \( k < 0.35 \, h \, \text{Mpc}^{-1} \) for the second configuration choice, where perturbation theory for this configuration begins to break down. This leaves us with 80 million triangles. To cope with so many triangles, we define a new estimator for \( b_1 \) and \( b_2 \) in the Appendix.
The estimator is not minimum-variance, it is unbiased and has the big advantage that it allows us to analyze many triangles, thus increasing the signal to noise. The estimator does not give error bars; these are obtained by Monte Carlo simulation from 16 mock 2dFGRS catalogues (see Section 4). In Fig. 1 we show the measured (redshift-space) 2dFGRS bispectrum from the SGP and NGP for the two chosen triangle configurations. The dotted line shows the perturbation theory prediction for $b_1 = 1$, $b_2 = 0$ while the dashed line shows the shot noise contribution. For the bispectrum, the shot noise contribution becomes dominant around $k = 0.5$ h Mpc$^{-1}$.

4 MOCK CATALOGUES AND TESTS

We have used 16 mock catalogues from a Hubble Volume N-body simulation with $\Omega_m = 0.3$, $\Omega_\Lambda = 0.7$ (‘$\Lambda$CDM’ model), with the same selection function as the 2dF galaxy redshift survey (Cole et al. 1998)\(^\dagger\). This includes both the radial selection function and an angular mask that reflects the varying completeness of the survey in February 2001. The catalogues contain mock galaxies whose positions are determined according to the prescription described in Cole et al. (1998). This is a 2-parameter exponential model based on the final density field, i.e. it does not conform to our assumption of equation (3). We will show that nevertheless the bispectrum method recovers the bias parameter $b_1$, defined by the square root of the ratio of the galaxy and matter power spectra\(^\ddagger\). This is the crucial test, since it is this ratio that we wish to determine, as we can then translate the galaxy power spectrum to the underlying mass spectrum. For the simulated galaxies in the Hubble volume as a whole the square root of the power spectrum ratio varies between about 0.9 at large scales and 0.75 at $k = 0.5$ h Mpc$^{-1}$\(^\S\). Over the scales probed by our bispectrum triangles, the bias is roughly 0.8, but with an uncertainty which ultimately limits our error determination. The power spectrum of the simulated 2dFGRS catalogues varies a little, so we estimate $b_1 \equiv (P_g/P)^{1/2}$ individually for each catalogue (and individually for SGP and NGP), from the wavenumber range $0.1 - 0.5$ h Mpc$^{-1}$. This sets $b$ for each sample, and we fit the pairwise velocity dispersion $\sigma_p$, individually by requiring a good fit to the redshift-space power spectrum, using the real-to-redshift mapping of equation (9).

We then analyse the set of triangles for each simulation, as detailed above, and in Fig. 2 we show the error in the bias from 16 simulations in the SGP and NGP. The average bias in the estimator is consistent with zero for the SGP: $-0.01 \pm 0.03$, but shows a small bias of $0.10 \pm 0.04$ for the NGP\(^\S\). Fig. 2 shows that the approximations made in the analysis (e.g. the functional form of the bias, the window function, the distant observer approximation etc.) do not significantly bias the result. We will use the sample rms of 0.11 for the SGP and 0.16 for the NGP as the errors in the analysis of the 2dFGRS. We also carried out the same analysis on 16 mock catalogues of the SGP obtained from the Hubble volume simulation for a $\tau$CDM model. The true underlying bias parameter is correctly recovered with a 20% error ($b = 1.7 \pm 0.3$). The error in the natural parameter $1/b_1$ is comparable for the $\Lambda$CDM and $\tau$CDM models, but the value of $b_1$ itself is larger in the latter case. In Fig. 3 we show $b_1$ recovered with the bispectrum method versus the underlying (true) $b_1 \equiv \sqrt{P_g/P}$ for 16 mock SGP simulations for the $\tau$CDM and $\Lambda$CDM models.

5 RESULTS

We analyse the catalogue as compiled in February 2001 (i.e. the same catalogue as in Percival et al. 2001), and we analyse the NGP and SGP regions separately; we excise the random fields entirely. This leaves us with 75792 and 51862 galaxies in the SGP and NGP respectively. The two survey regions give us a useful consistency check.

We initially present results fixing $\beta = 0.43$ and $\sigma_p = 385/H_0$ km s$^{-1}$ (3.85 $h^{-1}$ Mpc). These are the best-fitting values from the analysis of the 2dFGRS redshift-space correlation function (Percival et al. 2001; Norberg et al. 2001a; see also Tegmark, Hamilton

\(^\dagger\) See http://star-www.dur.ac.uk/~cole/mocks/hubble.html

\(^\ddagger\) For a more general discussion of definitions of bias, see Lahav et al. (2001)

\(^\S\) This is understandable since the NGP is smaller than the SGP, thus the effects of the convolution with the window are more important.
The dashed and dot-dashed lines are the perturbation theory predictions, relative to the bispectrum for a fiducial unbiased model ($B_{\text{fid}}$). Dashed line $b_1 = 1.3$, $b_2 = 0$, dot-dashed line $b_1 = 1.0$, $b_2 = 0.5$. To produce this figure only the SGP data were used and only configurations with two wavevectors of common length. This means that only 12 million triangles were used from the total 80 millions. Inclusion of the remainder excludes both models at high confidence. The error bars are obtained via Monte Carlo from the 16 N-body simulations, and are placed centrally on the mean of the estimates from the mock catalogues. This illustrates the level of bias in the estimator. The figure also shows that there is no evidence of scale dependent bias.

$$b_2 = -0.02 \pm 0.07,$$

with SGP and NGP individually giving $b_2 = 0.01 \pm 0.00$ and $-0.07 \pm 0.11$ respectively.

Thus we see that the 2dFGRS galaxies are perfectly consistent with tracing the underlying mass distribution on these scales (wavelengths $\lambda = 2\pi/k = 13$ to $63 \, h^{-1} \, \text{Mpc}$) i.e. $b_1$ is consistent with unity, and $b_2$ with zero.

The results depend mildly on the values of $\beta$ and $\sigma_p$ used. Marginalising over the distribution of these quantities estimated from the redshift-space correlation function (Peacock et al. 2001), the final results are virtually unchanged: $b_1 = 1.04 \pm 0.06$, $b_2 = -0.054 \pm 0.04$; adding these errors in quadrature with those of (15) gives our final estimates:

$$b_1 = 1.04 \pm 0.11,$$

$$b_2 = -0.054 \pm 0.08.$$  

### 5.2 Luminosity-dependent and scale-dependent bias

The 2dFGRS exhibits luminosity-dependent clustering (Norberg et al. 2001a; Norberg et al. 2001b), so it is reasonable to ask in what sense the galaxies in the 2dFGRS are unbiased. We have consistently used all the galaxies in the 2dFGRS NGP and SGP regions, to determine $\beta$ and the bias parameters. With the weighting scheme employed to measure $\beta$ and $b_1$, the weighted mean luminosity is $1.9L_\star$. From studies of the correlation function, Norberg et al. (2001a) found a luminosity dependence in the relative bias of 2dFGRS galaxies: $b/b_\star = 0.85 \pm 0.15L/L_\star$. If we use this relation to adjust our recovered bias to apply to $L_\star$ galaxies, we find they could be slightly anti-biased, but they are still consistent with $b_1 = 1$ ($b_\star = 0.92 \pm 0.11$ assuming $b_1 = b$). The 2dFGRS galaxies, as a population, trace the mass extremely well.
On scale-dependence of the bias, it is worth noting that theoretical arguments (e.g. Mann, Peacock & Heavens 1998; Peacock & Smith 2000; Seljak 2000; Benson et al. 2000; Blanton et al. 2000; Berlind & Weinberg 2001) suggest that bias is expected to be constant on large scales (larger than a few \( h^{-1} \) Mpc) and a scale-dependent bias is plausible on intermediate scales, but the scale dependence is expected to be weak. We find no evidence of scale-dependent bias (see Fig. 4). Both scale-dependence and luminosity-dependence may be explored in detail when the catalogue is complete.

5.3 The matter density of the Universe

Analysis of the redshift-space correlation function allows one to estimate \( \beta = \Omega_m^0 b_1 \) (Peacock et al. 2001), and this can be combined with our determination of \( b_1 \) to estimate the matter density parameter \( \Omega_m \). A strength of the result is that it is obtained from the 2dFGRS alone, independently of all other datasets. Following the same procedure as in the previous subsection, we marginalise over the uncertainty in \( \beta \) and \( \sigma_p \), to get:

\[
\Omega_m = 0.27 \pm 0.06. \tag{18}
\]

The small size of the error bar is worth remarking on. For each pair of values of \( \beta \) and \( \sigma_p \), we obtain an estimate of \( b_1 \). The estimates of \( \beta \) and \( b_1 \) are slightly anti-correlated, which leads to a slightly smaller scatter in \( \Omega_m \) than one would expect from the errors on \( \beta \) and \( b_1 \).

This result is consistent with other recent determinations by other methods such as combining CMB with large-scale structure or supernova measurements (e.g. Percival et al. 2001; Jaffe et al. 2001; Efstathiou et al. 2001) and from early weak lensing measurements (Hoekstra, Yee & Gladders 2001).

5.4 Comparison with previous work

Previous work has concentrated on using cumulants, although there have been two analyses (Feldman et al. 2001; Scoccimarro et al. 2001) of the bispectrum of infrared-selected galaxies observed with the Infrared Astronomy Satellite (IRAS; Fisher et al. 1995; Saunders et al. 2000). The IRAS samples are much smaller than the current 2dFGRS, and are also rather shallow, so shot noise and the radial nature of the redshift distortion is more severe than for the 2dFGRS. Nevertheless, bias estimates have been made, and reported with remarkably small error bars. The most accurate quoted values are from Feldman et al. (2001) \( 1/b_1 = 1.2^{\pm 0.18} \) for the PSCz survey (Saunders et al. 2000). This value of the linear bias parameter is consistent with our determinations, since IRAS galaxies have a power spectrum which is lower than optically selected galaxies, by a factor \( \sim 1.3^2 \) on the relevant scales (Peacock & Dodds 1994; Seaborne et al. 1999).

Studies of the skewness, kurtosis and higher-order moments of the optical galaxy distribution (e.g. Gaztanaga 1994; Szapudi & Szalay 1997; Hui & Gaztañaga 1999; Hoyle, Szapudi & Baugh 2000) have also shown consistency with a linear bias of unity, but here we are able to derive both the linear bias and the quadratic bias simultaneously. In addition, our present results are more accurate and convert these general indications of a low degree of bias into a strong constraint on theoretical models.

6 SUMMARY AND CONCLUSIONS

We have demonstrated through analysis of the bispectrum of the 2dFGRS that the optically-selected galaxies of the sample trace the matter density extremely well on large scales (Fourier modes with \( 0.1 < k < 0.5 h^{-1} \) Mpc\(^{-1} \) that approximately correspond to \( 30 < r < 5 h^{-1} \) Mpc). Specifically, the linear bias parameter is very close to unity, and the quadratic (nonlinear) bias is very close to zero. Ironically, this is exactly the assumption which used to be made decades ago in large-scale structure analysis, but which was questioned when the concept of biased galaxy formation was introduced (e.g. Kaiser 1984; Peacock & Heavens 1985; Bardeen et al. 1986; Dekel & Rees 1994). Theoretical arguments suggest that an initial bias at formation approaches unity with time, provided that galaxies are neither created nor destroyed (Fry 1996), and that the Universe does not become curvature- or vacuum-dominated in the meantime (Catelan, Matarrese & Porciani 1998). In any case, these assumptions will clearly fail at some level. Note that the effective depth of the survey is \( z = 0.17 \) and our measurement should be interpreted as the bias and \( \Omega_m \) at this epoch (for additional discussion see Lahav et al. 2001). Because of a tendency for bias to approach unity with time, we do not by this measurement rule out a significant bias of galaxies at formation time, but the unbiased nature of the galaxies today puts a significant constraint on theoretical models of galaxy formation.

Currently, we find no evidence of scale-dependent bias (see Fig. 4). The size of the survey at the present time does not allow us to place strong constraints on scale dependence or luminosity dependence of the bias parameter, or on the nature of biasing (e.g. deterministic, stochastic, Eulerian, Lagrangian etc.), but these issues will be explored with the completion of the survey. At some level nonlinear bias must appear on small scales, from the morphology-density relation (Dressler 1980; Hashimoto & Oemler 1999), but this is probably on too small scales for the perturbative method of this paper to be valid.

Our measurement of the matter density of the Universe, \( \Omega_m = 0.27 \pm 0.06 \), is in agreement with other independent determinations such as cosmic microwave background together with Hubble constant constraints (e.g. Jaffe et al. 2001; Efstathiou et al. 2001; Freedman et al. 2001) which also give comparable error bars.

It is worth re-emphasising that the analysis presented here relies only on the 2dFGRS dataset. It is also important that we not only conclude that the linear bias is essentially unity, but also the quadratic (nonlinear) bias term is constrained to be very close to zero. Taken together, these measurements argue powerfully that the 2dFGRS galaxies do indeed trace the mass on large scales. In addition to our findings, joint analysis of the 2dFGRS and CMB data (Lahav et al. 2001) also supports our conclusion that the 2dFGRS galaxies are unbiased. Different methods of measuring linear bias require different assumptions, and it is remarkable that such different methods agree on the basic cosmological model.

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References

Appendix

Here we outline the new estimator for the bias parameters. Although not optimal, it allows many millions of triangles to be analysed.

The likelihood of \( c_1 \equiv 1/b_1 \) and \( c_2 \equiv b_2/b_1^2 \) from each triangle (labelled by \( \gamma \)) is degenerate, since it constrains only a linear combination of \( c_1 \) and \( c_2 \), namely

\[
B_\gamma = R_\gamma c_1 + S_\gamma c_2 + T_\gamma
\]

where the expressions for \( R_\gamma \), \( S_\gamma \), and \( T_\gamma \) are at the end of this Appendix. The bispectrum estimate \( B_\gamma \) has an intrinsic statistical error \( \sigma_\gamma \), as discussed e.g. in Matarrese, Verde & Heavens (1997). In computing the bias parameter, we wish to weight the determination obtained from each triangle by the inverse of the variance of the corresponding bispectrum. Since this is a cumbersome expression (see Matarrese, Verde & Heavens 1997; Verde et al. 1998) we approximate this by its dominant Gaussian term,

\[
\langle \hat{F}_1 \hat{F}_2 \hat{F}_3 \rangle = (P_g + P_{SN})(P_g + P_{SN})(P_g + P_{SN})
\]

where \( P_g \equiv P_g(k_1) \) etc. and \( P_{SN} \) is the shot noise. Inclusion of higher-order terms leaves the results practically unchanged for this dataset, and adds considerably to the computing time. If we now assume that this error is Gaussian distributed we can immediately see that the likelihood contours will be lines in the \( c_1 - c_2 \) plane:

\[
\ln \mathcal{L}(c_1, c_2) \propto -\ln \sigma_\gamma \frac{(B_\gamma - R_\gamma c_1 + S_\gamma c_2 + T_\gamma)^2}{2\sigma_\gamma^2},
\]

This should give an unbiased a posteriori probability for \( c_1 \) and \( c_2 \), if we assume uniform priors. We define a non-optimal estimator by simply multiplying these likelihoods together, ignoring the correlations between different triangles, forming what we term the pseudolikelihood. The maximum of the pseudolikelihood is still calculable analytically, yielding the following estimates for \( c_1 \) and \( c_2 \):

\[
\hat{c}_1 = \frac{2A_1A_{22} - A_2A_{12}}{A_{12}^2 - 4A_{11}A_{22}}
\]

\[
\hat{c}_2 = \frac{2A_2A_{11} - A_1A_{12}}{A_{12}^2 - 4A_{11}A_{22}}
\]

where

\[
A_{11} = \sum_\gamma \frac{R_\gamma^2}{2\sigma_\gamma^2}
\]

\[
A_{22} = \sum_\gamma \frac{S_\gamma^2}{2\sigma_\gamma^2}
\]

\[
A_1 = \sum_\gamma \frac{(T_\gamma - B_\gamma)R_\gamma}{\sigma_\gamma^2}
\]

\[
A_2 = \sum_\gamma \frac{(T_\gamma - B_\gamma)S_\gamma}{\sigma_\gamma^2}
\]

We can also, if desired, compute \( c_1 \) on the assumption that \( c_2 = 0 \):

\[
\hat{c}_1 = \frac{A_1}{2A_{11}} \quad (c_2 \equiv 0).
\]

The expressions for \( R_\gamma \), \( S_\gamma \), and \( T_\gamma \) can easily be obtained from the expression for the bispectrum of Verde et al. (1998); Sections 2.3 and 2.4. We report them here for completeness. To obtain their corresponding real space quantities just set \( \sigma_\gamma = 0 \) and \( \beta = 0 \).

\[
R_\gamma = \left\{ [J(k_1, k_2) + \mu^2\beta K(k_1, k_2)]2P_g(k_1)P_g(k_2) \right. 
\] 

\[
\left. \times (1 + \beta\mu_1^2)(1 + \beta\mu_2^2) + \text{cyc.} \right\} D_3
\]

where \( \mu = -\mu_3 \) for the term explicitly written and the expression for the kernel \( K \) can be found e.g. in Catelan & Moscardini (1994),

\[
S_\gamma = [(1 + \beta\mu_1^2)(1 + \beta\mu_2^2)P_g(k_1)P_g(k_2) + \text{cyc.}] D_3
\]

\[
T_\gamma = \left\{ \beta^2\mu_1\mu_2 \left[ \beta \left( \mu_1^2 + \mu_2^2 \right) + \beta\mu_1\mu_2 \left( \frac{k_1}{k_2} + \frac{k_2}{k_1} \right) \right] \right. 
\]

\[
\left. \times (1 + \beta\mu_1^2)(1 + \beta\mu_2^2) + \text{cyc.} \right\} D_3
\]

\[
+ \left[ P_g(k_1) \left( 1 + \frac{\beta\mu_1^2}{1 + \alpha^2/2k_1^2\mu_1^2} \right) + \text{cyc.} \right] \frac{I_{32}}{I_{33}} + \left( 1 - \alpha \frac{I_{11}}{I_{33}} \right)
\]

where \( D_3 \) denotes the damping term due to the incoherent small scale velocity dispersion (see eq. 6):

\[
D_3 = \left[ 1 + \frac{\alpha^2 k_2^2 \sigma_p^2 \mu_1^2}{2} \right] \left[ 1 + \frac{\alpha^2 k_2^2 \sigma_p^2 \mu_2^2}{2} \right]^{-1/2}
\]