HOW MANY GALAXIES FIT IN A HALO? CONSTRAINTS ON GALAXY FORMATION EFFICIENCY FROM SPATIAL CLUSTERING

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ABSTRACT

We study galaxy clustering in the framework of halo models, where gravitational clustering is described in terms of dark matter halos. At small scales, dark matter clustering statistics are dominated by halo density profiles, whereas at large scales, correlations are the result of combining nonlinear perturbation theory with halo biasing. Galaxies are assumed to follow the dark matter profiles of the halo they inhabit, and galaxy formation efficiency is characterized by the number of galaxies that populate a halo of given mass. This approach leads to generic predictions: the galaxy power spectrum shows a power-law behavior even though the dark matter does not, and the galaxy higher order correlations show smaller amplitudes at small scales than their dark matter counterparts. Both are in qualitatively agreement with measurements in galaxy catalogs. We find that requiring the model to fit both the second- and third-order moments of the Automatic Plate Measuring Facility (APM) galaxies provides a strong constraint on galaxy formation models. The data at large scales require that galaxy formation be relatively efficient at small masses, \( m \approx 10^{10} M_\odot h^{-1} \), whereas data at smaller scales require that the number of galaxies in a halo scale approximately as the mass to the 0.8th power in the high-mass limit. These constraints are independent of those derived from the luminosity function or Tully-Fisher relation. We also predict the power spectrum, bispectrum, and higher order moments of the mass density field in this framework. Although halo models agree well with measurements of the mass power spectrum and the higher order \( S_p \) parameters in \( N \)-body simulations, the model assumption that halos are spherical leads to disagreement in the configuration dependence of the bispectrum at small scales. We stress the importance of finite-volume effects in higher order statistics and show how they can be estimated in this approach.

Subject headings: cosmology: theory — galaxies: clusters: general — large-scale structure of universe

1. INTRODUCTION

Understanding galaxy clustering is one of the main goals of cosmology. The wealth of information provided by galaxy surveys can only be used to extract useful cosmological information if we understand the relation between galaxy and dark matter clustering—biasing. On large enough scales, galaxy biasing can be described as a local process, and so galaxy clustering can be used as a direct probe of the primordial spectrum and Gaussianity of initial conditions (Fry & Gaztañaga 1993; Frieman & Gaztañaga 1994; Fry 1994; Fry & Scherrer 1994; Gaztañaga & Māhōnen 1996; Matarrese, Verde, & Heavens 1997; Gaztañaga & Fosalba 1998; Frieman & Gaztañaga 1999; Scoccimarro et al. 2000; Durrer et al. 2000). On smaller scales, however, nonnegligible contributions from complex astrophysical processes relevant to galaxy formation may complicate the description of galaxy biasing.

Galaxy formation is not yet understood from first physical principles. However, following White & Rees (1978) and White & Frenk (1991), a number of prescriptions based on reasonable recipes for approximating the complicated physics have been proposed for incorporating galaxy formation into numerical simulations of dark matter gravitational clustering (see, e.g., Kauffmann et al. 1999, Somerville & Primack 1999, or Benson et al. 2000 for some of the most recent work). These "semianalytic galaxy formation" schemes can provide detailed predictions for galaxy properties in hierarchical structure formation models, which can then be compared with observations.

The basic assumption in the semianalytic approach is that galaxy biasing can be described as a two-step process. First, the formation and clustering of dark matter halos can be modeled by neglecting nongravitational effects.4 This can be done reasonably accurately following the analytic results of Mo & White (1996), Mo, Jing, & White (1997), and Sheth & Lemson (1999). Second, the distribution of galaxies within halos, which in principle depends on complicated physics, can be described by a number of simplifying assumptions regarding gas cooling and feedback effects from supernovae. For the purposes of this paper, the main outcome of this second step is the number of galaxies that populate a halo of a given mass, \( N_{\text{gal}}(m) \).

In this paper we consider the problem of galaxy clustering from a complementary point of view to semianalytic models. We construct clustering statistics from properties of dark matter halos and the \( N_{\text{gal}}(m) \) relation, and show how these simple ingredients can be put together to make reasonably accurate analytic predictions about the galaxy power spectrum, bispectrum, and higher order moments of the galaxy field. We also consider the inverse problem: we

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4 Halos here are defined in the sense of Press & Schechter (1974). There is not necessarily a one-to-one correspondence between halos and galaxies.
show how measurements of galaxy clustering can constrain the \( N_{gal}(m) \) relation. We show in particular that the variance and skewness of the galaxy distribution in the Automatic Plate Measuring Facility (APM) survey provide significant constraints on the \( N_{gal}(m) \) relation.

Our approach to gravitational clustering has a long history, dating back to Neyman & Scott (1952), and explored further by Peebles (1974) and McClelland & Silk (1977a, 1977b, 1978). These works considered perturbations described by halos of a given size and profile, but distributed at random. A complete treatment that includes the effects of halo-halo correlations was first given by Scherrer & Bertschinger (1991). Recent work has focused on applications of this formalism to the clustering of dark matter, e.g., the small-scale behavior of the two-point correlation function (Sheth & Jain 1997), the power spectrum (Seljak 2000; Peacock & Smith 2000; Ma & Fry 2000a; Cooray & Hu 2000), and the bispectrum for equilateral configurations (Ma & Fry 2000a; Cooray & Hu 2000). Interest in this approach has undoubtedly been sparked by recent results from numerical simulations on the properties of dark matter halos (Navarro, Frenk, & White 1996, 1997; Bullock et al. 1999; Moore et al. 1999).

This paper is organized as follows. In § 2 we review the halo model formalism for the power spectrum, bispectrum, and higher order moments of the smoothed density field, and we compare the predictions with numerical simulations in § 3. We discuss in detail the role of finite-volume effects, which, if neglected, can lead to incorrect conclusions regarding higher order statistics. In § 4, we use the \( N_{gal}(m) \) relation to make predictions for galaxy, rather than dark counts-in-cells of the APM survey. Section 6 summarizes our conclusions.

2. DARK MATTER CLUSTERING

2.1. Formalism

In this section we follow the formalism developed by Scherrer & Bertschinger (1991). The dark matter density field is written as

\[
\rho(x) = \sum_i f(x - x_i, m_i) \equiv \sum_i m_i u(x - x_i, m_i)
\]

where \( f \) denotes the density profile of a halo of mass \( m_i \) located at position \( x_i \). The mean density is

\[
\bar{\rho} = \langle \rho(x) \rangle = \left\langle \sum_i m_i u(x - x_i, m_i) \right\rangle = \int n(m) dm \int d^3x' u_m(x - x'),
\]

where we have replaced the ensemble average by the average over the mass function \( n(m) \) (which gives the density of halos per unit mass) and an average over space, i.e., \( \langle \sum_i \delta(m - m_i) \delta^3(x' - x_i) \rangle = n(m) \). Our normalization convention is such that \( \int d^3x' u_m(x - x') = 1 \) and \( \int n(m) dm m = \bar{\rho} \). The two-point correlation function can be written as

\[
\bar{\rho}^2 \xi(x - x') = \int n(m) dm \int d^3y u_m(y) u_m(y + x - x') + \int n(m_1) dm_1 \int n(m_2) dm_2 \int d^3x_1 u_{m_1}(x - x_1) \times \int d^3x_2 u_{m_2}(x' - x_2) \xi(x_1 - x_2; m_1, m_2),
\]

where the first term describes the case in which the two particles occupy the same halo, and the second term represents the contribution of particles in different halos, with \( \xi(x - x'; m_1, m_2) \) being the two-point correlation function of halos of mass \( m_1 \) and \( m_2 \). Since we are dealing with convolutions of halo profiles, it is much easier to work in Fourier space, where expressions become multiplications over the Fourier transform of halo profiles. We use the following Fourier space conventions:

\[
A(k) = \int \frac{d^3x}{(2\pi)^3} \exp(-ik \cdot x) A(x),
\]

and

\[
\langle \delta(k_1) \delta(k_2) \rangle = \delta_{\Omega(k_1)} P(k_1),
\]

\[
\langle \delta(k_1) \bar{\delta}(k_2) \bar{\delta}(k_3) \rangle = \delta_{\Omega(k_1, k_2, k_3)} B_{123},
\]

where \( P(k) \) and \( B_{123} \equiv B(k_1, k_2, k_3) \) denote the power spectrum and bispectrum, respectively. Thus, the power spectrum reads

\[
\bar{\rho}^3 P(k) = (2\pi)^3 \int n(m) dm \int d^3x' |u_m(x - x')|^2
\]

\[
+ (2\pi)^6 \int u_{m_1}(k_1) n(m_1) dm_1 \times \int u_{m_2}(k_2) n(m_2) dm_2 P(k; m_1, m_2),
\]

where \( P(k; m_1, m_2) \) represents the power spectrum of halos of mass \( m_1 \) and \( m_2 \). Similarly, the bispectrum is given by

\[
\bar{\rho}^2 B_{123} = (2\pi)^3 \int n(m) dm \int \Pi_{i=1}^3 u_m(k_i)
\]

\[
+ (2\pi)^6 \int u_{m_1}(k_1) n(m_1) dm_1 \times \int u_{m_2}(k_2) u_{m_3}(k_3) n(m_2) n(m_3) dm_2 dm_3 B(k_1; m_1, m_2) + \text{cyc}.
\]

\[
+ (2\pi)^6 \left[ \sum_{i=1}^3 \int u_{m}(k_i) n(m_i) dm_i \right] B_{123}(m_1, m_2, m_3),
\]

where \( B_{123}(m_1, m_2, m_3) \) denotes the bispectrum of halos of mass \( m_1, m_2, m_3 \). So far, the treatment has been completely general. To make the model quantitative, we must specify
the halo profile, \( u_{\delta}(x) \); the halo mass function, \( n(m) \); and the halo-halo correlations encoded in \( P(k; m_1, m_2) \), \( B_{i\alpha\beta}(m_1, m_2, m_3) \), etc.

### 2.2. Halo Profiles

For the halo profile we use (Navarro, Frenk & White 1997; hereafter NFW)

\[
U_g(r) = \frac{f_1^3}{4\pi^2 v^3} \frac{1}{cr/R_v} (1 + cr/R_v)^{-3} \tag{9}
\]

where \( f = 1/\ln(1 + c) - c/(1 + c) \), \( R_v \) is the virial radius of the halo, related to its mass by \( m = (4\pi R_v^3/3)\Delta \rho \), where \( \Delta = 200 \) or 340 for an \( \Omega = 1 \) or 0.3 universe, respectively. We also use the Lagrangian radius \( R \) (the initial radius from which the mass \( m \) came) to specify halo sizes; \( R = R_{vir} \Delta^{-1/3} \). It is convenient to work in units of the characteristic nonlinear mass, \( m_* \), or the equivalent scale \( R_* \). [defined such that \( \sigma(R_*) = \delta_c \); note that \( m_* = (4\pi R_*^3/3)\bar{\rho} \). Since \( m = 1.16 \times 10^{12} \Omega(R_{vir} \text{Mpc }^{-1})^3 \text{M}_\odot \text{h}^{-1} \), for \( \Lambda CDM \) (\( \Omega = 0.3, \Omega_\Lambda = 0.7 \)) with \( \sigma_8 = 0.90, m_* = 3.135 \text{Mpc }^{-1} \text{h}^{-1} \), so \( m_* = 1.07 \times 10^{13} \text{M}_\odot \text{h}^{-1} \).

The Fourier transform of the halo profile reads

\[
u(k) = \int \frac{d^3x}{(2\pi)^3} \exp(-i k \cdot x) U_g(r) = \frac{1}{(2\pi)^3} u(k, y), \tag{10}\]

where \( y = R/R_* \), \( k = kR_* \Delta^{-1/3} \), and

\[
u(k, y) = \int \frac{d^3x}{(2\pi)^3} \exp(-i k \cdot x) \left\{ \sin[k(\kappa(1 + c) - \kappa(1 + c))] + \cos[k(\kappa(1 + c) - \kappa(1 + c))] \right\}, \tag{11}\]

where \( \kappa \equiv \kappa y/c, \kappa(x) = \int_0^x dt \sin(t)/t \) is the sine integral, and \( \kappa(x) = -\int_0^x dt \cos(t)/t \) is the cosine integral function. The concentration parameter quantifies the transition from the inner to the outer slope in the NFW profile. For the concentration parameter of halos, we use the result (Bullock et al. 1999)

\[
c(m) \approx 9 \left( \frac{m}{m_*} \right)^{-0.13}. \tag{12}\]

Note that halos defined as above are somewhat different from the prescription in NFW, who define \( R_{vir} \) as the scale within which the mean enclosed density is 200 times the critical density, independent of cosmology. Both approaches agree for \( \Omega = 1 \), whereas for the model we use \( \Omega = 0.3 \), and thus our virial radius is defined at the scale at which the mean enclosed density is \( 340 \times 0.3 \approx 100 \) times the critical density. Thus, our characteristic density is smaller than NFW's by a factor of 2, and our virial radius is correspondingly larger. These two effects partially cancel, and lead to a very similar prediction for halo profiles (also taking into account that the concentration parameters are slightly different as well).

### 2.3. Mass Function

The mass function in normalized units reads

\[
n(m)m dm = \rho \frac{dy}{y} n(y) = \rho \frac{dy}{y} \gamma A \sqrt{\frac{\gamma}{2\pi}} \left[ 1 + (\gamma y)^{-p} \right] \exp \left( -\frac{\gamma y^2}{2} \right), \tag{13}\]

where \( \gamma = 1/\sigma \) with \( \sigma = 1.68 \), the collapse threshold given by the spherical collapse model, \( A = 0.5, 0.322, p = 0, 0.3 \), and \( \gamma = 1, 0.707 \) for the PS (Press & Schechter 1974) and ST (Sheth & Tormen 1999) mass functions, respectively (see Jenkins et al. 2000 for a recent comparison of these mass functions against \( N \)-body simulations). Note that in this formula, the linear variance is \( \sigma^2 = \sigma^2(R_*) \), and \( \gamma(R) = R(4\gamma^2/R_d) \) is the logarithmic slope of the linear variance as a function of scale. The ST mass function was derived by fitting to the GIM/Virgo \( N \)-body simulations (Kauffmann et al. 1999). Sheth, Mo, & Tormen (2000) show that it may be related to a model in which objects form from an ellipsoidal, rather than a spherical, collapse.

### 2.4. Halo-Halo Correlations

Following Mo & White (1996; see also Mo, Jing, & White 1997; Sheth & Lemson 1999; Sheth & Tormen 1999 for extensions), halo-halo correlations are described by nonlinear perturbation theory plus a halo biasing prescription obtained from the spherical collapse model.\(^5\) For the PS and ST mass functions, we will need

\[
b_1(m) = 1 + e_1 + E_1, \tag{14}\]

\[
b_2(m) = 2(1 + a_2 \epsilon_1 + E_1) + e_2 + E_2, \tag{15}\]

\[
b_3(m) = 6(a_2 + a_3 \epsilon_1 + E_1) + 3(1 + 2a_2 \epsilon_2 + E_2) + e_3 + E_3, \tag{16}\]

\[
b_4(m) = 24(a_2 + a_3 \epsilon_1 + E_1) + 12[a_2^2 + 2(a_2 + a_3)] \times (e_2 + E_2) + 4(1 + 3a_2 \epsilon_3 + E_3) + e_4 + E_4, \tag{17}\]

where

\[
e_1 = \frac{\gamma y^2 - 1}{\delta_c}, \quad e_2 = \frac{\gamma y^2}{\delta_c} (\gamma y^2 - 3), \quad e_3 = \frac{\gamma y^2}{\delta_c} (\gamma y^2 - 4), \quad e_4 = \frac{\gamma y^2}{\delta_c^2} (\gamma y^2 - 15 \gamma y^2 + 31), \quad E_1 = \frac{2p/\delta_c}{1 + (\gamma y^2)^p}, \quad E_3 = \left[ \frac{4\gamma^2}{\delta_c} (\gamma y^2 - 3) \right]^3, \quad E_4 = \frac{2\gamma y^2}{\delta_c} \left[ \gamma y^2 - \gamma y^2 - 15 \gamma y^2 + 31 \right] + 2 \frac{1 + p}{\delta_c^2}, \tag{18}\]

\[
a_2 = -17/21, \quad a_3 = 341/567, \quad a_4 = -5805/130977. \quad E_3 = \left[ \frac{4\gamma^2}{\delta_c} (\gamma y^2 - 3) \right]^3, \quad E_4 = \frac{2\gamma y^2}{\delta_c} \left[ \gamma y^2 - \gamma y^2 - 15 \gamma y^2 + 31 \right] + 2 \frac{1 + p}{\delta_c^2}, \tag{18}\]

All the \( e_i \)'s are zero, and \( a = 1 \) if \( m(m) \) is given by the PS formula. In this case, our formulae reduce to those in Mo, Jing, & White (1997), although our expression for \( b_4(m) \) corrects a typographical error in their equation (15c).\(^5\) A treatment of halo biasing beyond the spherical collapse approximation using perturbation theory is given in Catelan et al. (1998).
Halo-halo correlations read
\[
P(k; m_1, m_2) = b_1(m_1)b_1(m_2)P_L(k),
\]
\[
B_{123}(m_1, m_2, m_3) = b_1(m_1)b_2(m_2)b_2(m_3)B_{123}^{PT}
+ b_1(m_1)b_2(m_2)b_3(m_3)
\times P_L(k_1)P_L(k_2) + \text{cyc.},
\]
and similarly for higher order moments (see eqs. [44] and [45]). The terms \(P_L(k)\) and \(B_{123}^{PT}\) denote respectively the linear power spectrum and the second-order perturbative bispectrum (Fry 1984),
\[
B_{123}^{PT} = 2F_2(k_1, k_2)P_L(k_1)P_L(k_2) + \text{cyc.},
\]
where \(F_2(k_1, k_2) = \hat{k}_1 + \frac{i}{2} \cos \theta_1 \theta_2 + \frac{1}{2} \cos^2 \theta_1 \theta_2\), with \(k_1 \cdot k_2 = k_1 k_2 \cos \theta_{12}\). By construction, the bias parameters obey \((n = 2, 3, \ldots)\)
\[
\int dy \tilde{n}(y)b_1(y) = 1, \quad \int dy \tilde{n}(y)b_0(y) = 0.
\]

2.5. Results

Using the ingredients above, we can rewrite the power spectrum and bispectrum as
\[
P(k) = [M_{11}(\hat{k})]^2P_L(k) + M_{03}(\hat{k}, \hat{k}),
\]
\[
B_{123} = \sum_{i=1}^{3} M_{1i}(\hat{k}_i)B_{123}^{PT}
+ [M_{11}(\hat{k}_1)M_{11}(\hat{k}_2)M_{11}(\hat{k}_3)P_L(k_1)P_L(k_2) + \text{cyc.}]
+ [M_{1i}(\hat{k}_1)M_{1j}(\hat{k}_2, \hat{k}_3)P_L(k_1) + \text{cyc.}]
+ M_{03}(\hat{k}_1, \hat{k}_2, \hat{k}_3),
\]
where \((b_0 = 1)\),
\[
M_i(\hat{k}_1, \ldots, \hat{k}_i) = \int dy \tilde{n}(y)b_i(y)
\times [u(\hat{k}_1, y) \ldots u(\hat{k}_i, y)](R_s^2/k)^{i-1}.
\]

It is convenient to define the reduced bispectrum \(Q_{123}\),
\[
Q_{123} = \frac{B_{123}}{P_1P_2 + P_2P_3 + P_3P_1},
\]
which shows a much weaker scale dependence than the bispectrum itself, since at large scales, perturbation theory (PT) predicts \(B \propto P^2\), and at small scales the hierarchical Ansatz also predicts such a behavior.

We can also obtain the one-point moments smoothed at scale \(R\) from Fourier space correlations by integrating with a top-hat window function in Fourier space, \(W(kR)\). For example, from equation (23), the variance is
\[
\sigma^2(R) = \int d^3k P(k)W(kR)^2 \approx \sigma^2[R] + \int dy \tilde{n}(y)y^3\bar{u}^2(R, y),
\]
where
\[
\bar{u}^2(R, y) = \frac{2\Lambda}{3\pi} \int \hat{k}^2 d^3k [u(\hat{k}, y)]^3W^2(kR),
\]
and we have assumed that \(\int M_{11}^2(\hat{k})P_L(k)d^3k W(kR)^2 \approx \sigma^2[R]\), since \(M_{11}(\hat{k}) \to 1\) as \(k \to 0\), and at smaller scales the power spectrum is dominated by the second term in equation (23). Similarly, the third moment reads \([W_i \equiv W(k_i), R_i]\),
\[
\langle \delta^3(R) \rangle = \int d^3k_1 d^3k_2 d^3k_3 \delta(\hat{k}_{123})B_{123}W_1 W_2 W_3
\approx S_{12}^P \sigma^2[R] + 3\sigma^2[R]
\times \int dy \tilde{n}(y)b_1(y)y^3\bar{u}(R, y)
\times \int dy \tilde{n}(y)y^3\bar{u}(R, y).
\]

There are several approximations involved in this result. First, we take the large-scale limit, \(M_{11} \approx 1, M_{12} \approx 0, \) valid to a good approximation because of the consistency conditions, equation (22). In addition, we assume that the configuration dependence of the one-halo and two-halo terms in equation (24) can be neglected (this holds very well for the two-halo term and approximately for the one-halo term, as we discuss below; e.g., Fig. 4, bottom right). We can thus evaluate these terms for equilateral configurations, and simplify the angular integration by further assuming \(W_1 W_2 W_3 \approx W_1^2 W_2^2\), the leading-order term in the multipolar expansion. With similar approximations, we can derive higher order connected moments. Define
\[
A_i(R) = \int dy \tilde{n}(y)b_i(y)y^{3(j+1)}[\bar{u}(R, y)]^j\bar{u}(R, y),
\]
so that \(\sigma^2 = \sigma_{L}^2 + A_{00}\) and \(\langle \delta^3 \rangle = S_{12}^P \sigma_{L}^4 + 3\sigma_{L}^2 A_{10} + A_{01}\). Then it follows that
\[
\langle \delta^4 \rangle_c = S_{12}^P \sigma_{L}^6 + 6 \frac{S_{12}^P}{3} \sigma_{L}^4 A_{10} + 4 \frac{S_{12}^P}{7} A_{11} + A_{02},
\]
\[
\langle \delta^5 \rangle_c = S_{12}^P \sigma_{L}^8 + 10 \frac{S_{12}^P}{16} \sigma_{L}^2 A_{10} + 25 \frac{3S_{12}^P}{5} \sigma_{L}^4 A_{11}
+ 15 \sigma_{L}^2 A_{12} + A_{03},
\]
where the terms in \(\langle \delta^m \rangle_c\) are ordered from \(n\)-halo to one-halo contributions. The coefficient of an \(m\)-halo contribution to \(\langle \delta^m \rangle_c\) is given by \(s(n, m)\) (e.g., 6 and 7 in the second and third terms of eq. [31]), the Stirling number of second kind, which is the number of ways of putting \(n\) distinguishable objects \((\delta)\) into \(m\) cells (halos), with no cells empty (Scherrer & Bertschinger 1991). Thus, in general we can write
\[
\langle \delta^m \rangle_c = S_n^P \sigma_{L}^{2(n-1)} + \sum_{m-n=2}^{n-1} s(n, m) \sigma_{mn}^P \sigma_{L}^{2(m-1)} A_{mn},
\]
where the first term in equation (33) represents the \(n\)-halo term, the second the contributions from \(m\)-halo terms, and the last is the one-halo term. The coefficients \(\sigma_{mn}\) measure how many of the terms contribute as \(A_{mn}\), the other contributions being subdominant. For example, in equation (31) the two-halo term has a total contribution of seven terms; four of them contain three particles in one halo and one in the other, and three of them contain two particles in each. The factor \(\frac{3}{2}\) is included to take into account that the
3-1 amplitude dominates over the 2-2 amplitude. Note that in these results we neglected all the contributions from the nonlinear biasing parameters, in view of the consistency conditions, equation (22). When neglecting halo-halo correlations, equation (33) reduces to those in Sheth (1996) in the limit that halos are point-size objects \( \mu(k, y) = 1 \).

For the perturbative values, we use (Bernardeau 1994)

\[
S_3^\text{PT} = \frac{34}{7} - \gamma, \quad S_4^\text{PT} = \frac{60712}{1323} - \frac{62}{3} \gamma + \gamma^2, \quad (34)
\]

\[
S_5^\text{PT} = \frac{200575880}{305613} - \frac{1847200}{3969} \gamma + \frac{6940}{63} \gamma^2 - \frac{235}{27} \gamma^3, \quad (35)
\]

where for simplicity we neglect derivatives of \( \gamma \) with respect to scale, which is a good approximation for \( R \lesssim 20 \) Mpc \( h^{-1} \).

Before we compare these predictions for dark matter clustering with numerical simulations, it is important to note that, within this framework, there are many ingredients that can be adjusted to improve agreement with simulations. Rather than exploring all possible variations, we have chosen to always use the NFW halo profile and the dependence of the concentration parameter on mass given earlier, and only change the mass function between PS and ST; it turns out that these two models usually bracket the results of numerical simulations. Other choices are considered in Seljak (2000), Ma & Fry (2000a), and Cooray & Hu (2000). The sensitivity of the results to these choices reflects the underlying uncertainty in this type of calculation. As numerical simulation results converge on the different ingredients, however, the predictive power of this method will increase.

3. COMPARISON WITH NUMERICAL SIMULATIONS

We have run two sets of N-body simulations using the adaptive particle-particle/particle-mesh (P^3M) code HYDRA (Couchman, Thomas, & Pearce 1995). Both have 128^3 particles and correspond to a \( \Lambda \)CDM model (\( \Omega_m = 0.3 \), \( \Omega_\Lambda = 0.7 \)) with \( \sigma_8 = 0.9 \). The first set contains 14 realizations of a box size 100 Mpc \( h^{-1} \) and softening length \( l_{\text{soft}} = 100 \) kpc \( h^{-1} \). The second set has four realizations with a box size of 300 Mpc \( h^{-1} \), and \( l_{\text{soft}} = 250 \) kpc \( h^{-1} \), which allows us to check for finite-volume effects. We have also studied the effects of changing the softening, number of time steps, and \( \Omega \), as described below.

In this paper we concentrate in the “large-scale” predictions relevant to current large-scale structure surveys (\( r \gtrsim 0.5 \) Mpc \( h^{-1} \) or \( k \lesssim 3 \) h Mpc^{-1}). This requires a large volume, and thus our simulations do not have the resolution to resolve the inner halo profiles. However, at the range of scales we have access to, the outer part of profiles are important, since we probe the regime dominated by one-halo terms (where all particles are in the same halo).

3.1. Bispectrum Measurements: Finite-Volume Effects

Figure 1 shows the results on the reduced bispectrum for equilateral triangles as a function of scale. The squares show the measurements from the “large” box (\( L_{\text{box}} = 100 \) Mpc \( h^{-1} \)), while the triangles show the measurements from the “small” box (\( L_{\text{box}} = 300 \) Mpc \( h^{-1} \)). Error bars are obtained from the scatter among 4 and 14 realizations, respectively. The disagreement between the results of the large and small boxes is a result of finite-volume effects; the bispectrum is much more sensitive to the presence or absence of massive halos than the power spectrum (we quantify this below), so the smallness of the small box is important. Note that the total volume in the 14 small-box realizations only adds up to about half of the volume of a single large-box realization. This translates into a large scatter among realizations of the small box; three such realizations are shown as solid lines in Figure 1. Note that not only the amplitude of \( Q_{eq} \) but also its dependence on scale fluctuates significantly from realization to realization, so one must interpret measurements made using only a small number of small-volume simulations very cautiously. A similar situation holds for higher order moments (e.g., Colombi, Bouchet, & Hernquist 1996), as we see below.

As is well known, the distribution of higher order statistics is non-Gaussian, with positive skewness (e.g., Szapudi & Colombi 1996; Szapudi et al. 2000); the mean value of a higher order statistic is a consequence of having a few large excursions above the mean, with most values underestimating the mean. This is exactly what we see in the small-box realizations: most of them are closer to the bottom solid line than the top one in Figure 1. Even 14 realizations of the small box are not enough to recover the correct answer given by the large-box mean (Fig. 1, squares). In other words, the skewness of the \( Q \) distribution makes convergence toward the true value much slower for the small-box simulations (Szapudi & Colombi 1996; see also Scoccimarro 2000 for the bispectrum case). Note that this is not bias in a statistical sense: given a sufficient number of realizations, the mean will always converge to the true value; it is just that the convergence is slow. It is also important to note that when measuring \( Q \) from multiple realizations, one should always obtain the average of \( B \) and the
average of $P$ separately from the realizations, and only at the end divide to obtain $Q$ (which is what we have done in making Fig. 1). Otherwise, a “ratio” bias would result (Hui & Gaztanaga 1999), and the skewness measurements from the small-box simulations would be lower than are shown in Figure 1. Such an estimator bias will certainly affect measurements of $Q$ from, say, a single realization of a size similar to our small box.

Figure 1 also shows the predictions of (tree-level) PT, which agree very well with the large-box results at large scales, as well as the predictions of hyperextended PT (HEPT; Scoccimarro & Frieman 1999), which has been proposed as a description of clustering in the nonlinear regime ($k \gtrsim 1 \, h \, \text{Mpc}^{-1}$). The agreement with HEPT is good up to the resolution scale of our simulations, which we estimate as $k \approx 4 \, h \, \text{Mpc}^{-1}$. It is not straightforward to assign a resolution scale in Fourier space (i.e., it is not just $2\pi/l_{\text{soft}}$, since a given Fourier mode has contribution from a range of scales). Two-body relaxation causes the two-point correlation function to be underestimated at scales comparable to $l_{\text{soft}}$, which in turn implies an overestimate of the reduced bispectrum $Q$. To illustrate this, we have run the same realization after halving $l_{\text{soft}}$ to 50 Kpc $h^{-1}$ (Fig. 1, dashed line); this shows that beyond $k \approx 4 \, h \, \text{Mpc}^{-1}$ the bispectrum results are sensitive to the resolution, so we only plot the small-box results up to this scale. Similarly, we only show results for the large box up to $k \approx 2 \, h \, \text{Mpc}^{-1}$. For our particle number, $l_{\text{soft}}$ cannot be pushed too much smaller than the values we used, or else there would be not enough particles in a cell of radius $l_{\text{soft}}$ to satisfy the fluid limit. We have also checked the sensitivity of our results to changes in the number of time steps used in the $N$-body integrator and found no difference. Changing the density parameter $\Omega$ leads to extremely small differences in the reduced bispectrum $Q$; we thus present results only for $\Lambda$CDM. This is true not only in the weakly nonlinear regime, but also in the nonlinear regime, as expected from the general nature of the $\Omega$ dependence in the equations of motion (Scoccimarro et al. 1998).

Recently, Ma & Fry (2000a) claimed that the hierarchical Ansatz is not obeyed in $N$-body simulations. They based their claim on analysis of one single realization—the equivalent to just one of our small boxes. Their measurement approximately follows the lower solid line in Figure 1, which, as we have shown, is seriously affected by finite-volume effects (we will quantify these effects shortly). To reliably test the hierarchical Ansatz at smaller scales than probed here, one must resort to higher resolution simulations, preferably keeping the box size as large as possible to avoid finite-volume effects. For example, a 300 Mpc $h^{-1}$ box $512^3$ particle simulation would be able to probe up to $k \approx 10 \, h \, \text{Mpc}^{-1}$ reliably.

### 3.2. Comparison with Predictions

The top panel in Figure 2 shows the ratio of the power spectrum in our two models (PS and ST mass functions) to the power spectrum fitting formula (Hamilton et al. 1991; Jain, Mo, & White 1995; Peacock & Dodds 1996) as a function of scale $k$. The dashed and dotted lines show the contributions to the one-halo term in the PS and ST cases, respectively, from halos having masses in the range $10 < m/m_\ast < 100$, $1 < m/m_\ast < 10$, and $0.1 < m/m_\ast < 1$, from left to right ($m_\ast = 1.07 \times 10^{13} \, M_\odot \, h^{-1}$). Since the PS mass function has more halos than the ST when $m \lesssim 40 m_\ast$, the one-halo term is enhanced. Note the dip in both predictions at $k \approx 0.5 \, h \, \text{Mpc}^{-1}$, where the amplitudes of the one-halo and two-halo terms are comparable. This is due to our treatment of the two-halo term; we approximate it by simply using linear PT. In practice, nonlinear corrections enhance this term at scales smaller than the nonlinear scale, $k \approx 0.3 \, h \, \text{Mpc}^{-1}$. However, when including this term one must also take into account exclusion effects (halos cannot be closer than their typical size); otherwise, the power spectrum at intermediate scales would be overestimated. Since exclusion effects are nontrivial to compute (although Sheth & Lemson 1999 suggest how one might do so), the simplest solution is to ignore these effects, because they approximately cancel each other. This is a reasonable approximation, because at the scales where exclusion effects become important, one-halo contributions dominate.

The bottom panel in Figure 2 shows the prediction of halo models for the reduced bispectrum for equilateral configurations as a function of scale (solid lines). For the ST case we also show the partial contributions from one-halo, two-halo, and three-halo terms by dashed lines, which dominate at small, intermediate, and large scales, respectively. For the PS case we show as dotted lines the contributions to the one-halo term in $Q$ when the bispectrum is restricted to the mass range $10 < m/m_\ast < 100$ (which dominates at all scales shown in the plot) and $1 < m/m_\ast < 10$. In this case, when taking the ratio in equation (26), we have used the full power spectrum. As expected, comparing the two panels, we see that at a given scale the bispectrum is dominated by larger mass halos than the power spectrum.
For the bispectrum at $k \approx 1 \, h \, \text{Mpc}^{-1}$, this implies that halos with $m > 40 \, m_\text{h}$ contribute more significantly, thus leading to a higher $Q$. At smaller scales, say $k \approx 10 \, h \, \text{Mpc}^{-1}$, the PS mass function has more halos of the relevant masses ($m < 40 \, m_\text{h}$), so the bispectrum is larger for PS than ST (by a factor slightly smaller than the ratio of power spectra); thus, the reduced bispectrum $Q$ is higher again for the ST case. We have also varied the concentration parameter to test the sensitivity of our results. Doubling the concentration parameter (with the same mass dependence) leads to a significant increase of the power spectrum at small scales (this is consistent with Seljak 2000) and increases $Q$ by about 10% at small scales. On the other hand, changing the scaling of the concentration parameter by a factor of 2, to $c(m) = 9(m/m_\text{h})^{-0.26}$, decreases (or increases) the power spectrum at scales where $m/m_\text{h} > 1$ (or $m/m_\text{h} < 1$, respectively) contributes, as expected. For the bispectrum, larger concentration leads to higher $Q$, although at a given scale larger masses contribute than for the power spectrum, so the effects are shifted in scale with respect to the power spectrum case.

Figure 3 compares these results with the measurements in the numerical simulations presented in Figure 1. We see that generally, there is good agreement between predictions and the simulations; the simulations seem to be roughly in that generally, there is good agreement between predictions and the numerical simulations presented in Figure 1. We see although at a given scale larger masses contribute than for the power spectrum, so the effects are shifted in scale with respect to the power spectrum case.

$m_{\text{max}} = 10^{14} \, M_\odot \, h^{-1}$ (dashed lines). These values are those for which the mass functions would predict just one halo with mass larger than $m_{\text{max}}$ in a $(100 \, \text{Mpc} \, h^{-1})^3$ volume; however, since these are cumulative numbers and both mass functions actually overestimate the number of halos when compared to simulations (more so PS), a smaller number, such as $m_{\text{max}} = 10^{14} \, M_\odot \, h^{-1}$, is perhaps a more reasonable cutoff. In any case, we see that the predictions change significantly; in particular, introducing such a cutoff makes the scale dependence of $Q$ much more like that seen in most of the realizations of the small box (Fig. 1, bottom solid line).

One key element in the halo model is that we are using the spherical average (rather than the actual shapes) of halo profiles. On the other hand, it is well known that halos found in N-body simulations are not spherical, but rather triaxial (Barnes & Efstanthiou 1987; Frenk et al. 1988; Zurek, Quinn, & Salmon 1988). The bispectrum is the lowest order statistic that is sensitive to the shapes of structures, so one expects to find differences for the bispectrum as a function of triangle shape at small scales where halo profiles (one-halo terms) determine correlation functions. Figure 4 shows such a comparison at different scales, for triangles where $k_2 = 2k_1$, as a function of the angle $\theta$ between $k_1$ and $k_2$. The top left panel shows that, at large scales, the bispectrum agrees reasonably well with simulations; this is of course by construction, since nonlinear PT holds. At smaller scales (Fig. 4, top right), however, the predictions become independent of triangle shape at scales at which there is still noticeable configuration dependence. In fact, this is understood from the bottom right panel, which shows the partial contributions for the ST case. We see that the one-halo term (which is determined by the halo profiles) has the opposite configuration dependence to the three-halo term, which comes from nonlinear PT.

The fact that $Q_{1h}$ is convex can be understood from the spherical approximation. If halos were exactly spherical and $Q$ were scale-independent, then one would expect the maximum of $Q$ to occur for equilateral configurations. When $k_2 = 2k_1$, the closest configurations to equilateral are isosceles triangles with $\theta \approx 0.6 \pi$. Aside from an overall slight scale dependence ($\theta = 0$ configurations are somewhat more nonlinear than $\theta = \pi$), we see that this is indeed the case. Note also that if the contribution $Q_{1h}$ were flat, the residual configuration coming from $Q_{3h}$ would be enough to produce agreement with the N-body simulations. At even smaller scales (Fig. 4, bottom left), the numerical simulation results become approximately flat, but the halo models predict a convex configuration dependence due to the fact that $Q_{1h}$ dominates. From these results we conclude that although halo models predict bispectrum amplitudes that are in reasonable agreement with simulations, the configuration dependence is in qualitative disagreement with simulations. Of course, if we knew the actual halo shapes, then they could be incorporated into the models (at the expense of complicating the calculations!).

Finally, in Figure 5, we compare counts-in-cells measurements of the higher order moments of the smoothed density field in our simulations with the predictions of halo models. Symbols and error bars are as in Figure 1: the top solid line in each case corresponds to the ST prediction, the bottom solid line to the PS. The dashed lines show the predictions of HEPT (Scoccimarro & Frieman 1999), and the vertical lines show the softening scale of the large and small boxes. The disagreement of the average of 14 small-box measurements with the large-box average is, again, a manifestation

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

**Figure 3.**—Comparison of the N-body results of Fig. 1 (**filled symbols with error bars**) with the predictions of Fig. 2 (**solid lines**). The dashed lines show the predictions of halo models when there are no halos of mass larger than $m = 10^{14} \, M_\odot \, h^{-1}$ in the simulation volume, to illustrate finite-volume effects, for both PS (lower curve) and ST mass functions. The dot-dashed lines represent the predictions if halos of mass larger than $m = 5.9 \times 10^{14} \, M_\odot \, h^{-1}$ (or $m = 6.8 \times 10^{14} \, M_\odot \, h^{-1}$) are excluded for PS (or ST) mass functions.
of finite-volume effects. As expected, the difference becomes increasingly important for the higher order moments. Despite the many approximations made in the calculations of $S_n$ parameters in halo models, the agreement with simulations is quite good. We also see a very good agreement with the HEPT predictions, and that the scales at which halo models predict a strong scale dependence are beyond the limits of our resolution. This also confirms that our prescription for the resolution limit in Fourier space was reasonably accurate. Thus, contrary to Ma & Fry (2000a), we conclude that higher resolution simulations in larger boxes are essential if one is to test models of the higher order correlations reliably.

4. GALAXY CLUSTERING

We now turn to a discussion of how to use the halo models described above to predict the clustering of galaxies. Our treatment follows ideas present in the semianalytic galaxy formation models (Benson et al. 2000; Kauffmann et al. 1999) and has been also explored by Seljak (2000) and Peacock & Smith (2000) for the case of the power spectrum.

4.1. Galaxy Correlation Functions

To describe galaxy clustering, we need to know the distribution, the mean, and the higher order moments of the number of galaxies that can inhabit a halo of mass $m$. In addition, we need to know the spatial distribution of galaxies within their parent halo. To illustrate our model predictions, here we will assume that the galaxies follow the dark matter profile (we discuss what happens if we change this requirement shortly). This implies that equation (7) for galaxies reads

$$\bar{n}_g^2 P_g(k) = (2\pi)^3 \int n(m) \langle N_{\text{gal}}^2(m) \rangle dm |u_g(k)|^2$$

$$+ (2\pi)^6 \int u_{m_1}(k) n(m_1) \langle N_{\text{gal}}(m_1) \rangle dm_1$$

$$\times \int u_{m_2}(k) n(m_2) \langle N_{\text{gal}}(m_2) \rangle dm_2$$

$$\times P(k; m_1, m_2),$$

(36)
Similarly, the galaxy one-point connected moments satisfy

$$\langle \delta_g^3 \rangle_c = (S_3^{\text{PT}})_{\text{gal}} (\sigma_g^2)_{\text{gal}} + 3 (\sigma_g^2)_{\text{gal}} B_{10} + B_{01} ,$$

and

$$\langle \delta_g^5 \rangle_c = (S_5^{\text{PT}})_{\text{gal}} (\sigma_g^2)_{\text{gal}} + 5 (\sigma_g^2)_{\text{gal}} B_{10} + 5 (\sigma_g^2)_{\text{gal}} B_{12} + B_{03} ,$$

where

$$B_{ij}(R) = \frac{1}{b_i R_g^{j+2}} \int \frac{dy}{y} \tilde{n}(y) b_i(y) y^{3(j+1)} \left[ \frac{\tilde{\sigma}(R, y)}{w^2(R, y)} \frac{N_{g}^{j+2}}{m^{j+2}} \right] ,$$

and the perturbative moments are given by their local bias counterparts (Fry & Gaztaña 1993),

$$(\sigma_g^2)_{\text{gal}} = b_1^2 \sigma^2 , \quad (S_3^{\text{PT}})_{\text{gal}} = \frac{1}{b_1} (S_3^{\text{PT}} + 3c_2) ,$$

$$(S_4^{\text{PT}})_{\text{gal}} = \frac{1}{b_4^2} (S_4^{\text{PT}} + 12c_2 S_3^{\text{PT}} + 4c_3 + 12c_3^2) ,$$

and

$$(S_5^{\text{PT}})_{\text{gal}} = \frac{1}{b_1} [ S_5^{\text{PT}} + 20 c_2 S_4^{\text{PT}} + 15 c_2 (S_3^{\text{PT}})^2$$

$$+ (30 c_2 + 120 c_2^2) S_5^{\text{PT}} + 5 c_4 + 60 c_2 c_3 + 60 c_3^2 ] ,$$

where $c_i \equiv b_i/b_1$ and $b_i$ are the effective bias parameters in equation (39). The halo-halo skewness and kurtosis are given by these expressions upon replacing $c_i$ by $c_i B_{0i}/B_{10}$.

### 4.2 Galaxies

As a first example, we use the results of the semi-analytic models of Kauffmann et al. (1999); these $N$-body simulations and halo and galaxy catalogs are publicly available. Sheth & Diaferio (2000) show that in these catalogs, the mean number of galaxies, $N_{\text{gal}}$, per halo of mass $m$ is well fitted by

$$\langle N_{\text{gal}} \rangle = \langle N_B + N_R \rangle , \quad \langle N_B \rangle = 0.7 (m/m_B)^{\eta_B} ,$$

$$\langle N_R \rangle = \langle m/m_R \rangle^{\eta_R} ,$$

where $N_B$ and $N_R$ represent the number of blue and red galaxies, respectively, per halo of mass $m$, and $m_B = 4 \times 10^{12} M_{\odot} h^{-1} \leq m \leq m_B$, $\eta_B = 0.8$ for $m > m_B$, $m_B = 4 \times 10^{12} M_{\odot} h$, $\eta_R = 0.9$, and $m_R = 2.5 \times 10^{12} M_{\odot} h$ (no lower mass cutoff for $R$). The physical basis for this relation is as follows. At large masses, the gas cooling time becomes larger than the Hubble time, so galaxy formation is suppressed in large-mass halos; therefore, $\langle N_{\text{gal}} \rangle$ increases less rapidly than the mass. In small-mass halos, however, effects such as supernova winds can blow away the gas from halos, also suppressing galaxy formation; thus the cutoff at small masses.

To calculate the power spectrum and higher order statistics, we also need the second and higher order moments of $N_{\text{gal}}$. The second moment is also obtained from the semi-analytic models, and obeys

$$\langle N_{\text{gal}} (N_{\text{gal}} - 1) \rangle = \langle \tilde{x} \rangle (m/m_{11})^{1/2} (m_{11} \equiv 10^{11} M_{\odot} h^{-1})$$

where the function $\tilde{x}(m)$ quantifies deviations from Poisson statistics $\tilde{x}(m) \approx \log (m/m_{11})^{1/2}$ for
no. 1, 2001 how many galaxies fit in halo? 29

\[ m < 10^{11} M_\odot h^{-1}, \text{ and } \alpha(m) = 1 \text{ otherwise; that is, for large masses, the scatter about the mean number of galaxies is Poisson, whereas for small masses it is sub-Poisson.} \]

To model the higher order moments, we assume that the number of galaxies in a halo of mass \( m \) follows a binomial distribution:

\[ p(N_{\text{gal}} = n | m) = \binom{N_{m}}{n} p_m^n (1 - p_m)^{N_{m} - n}. \tag{48} \]

The binomial distribution is characterized by two parameters, \( N_{m} \) and \( p_m \), which we set by requiring that the first two moments of the distribution equal those from the semi-analytic. Specifically, the first and second factorial moments are \( N_{m} p_m \) and \( N_{m} p_m (N_{m} p_m - p_m) \), and we require that they equal \( \langle N_{\text{gal}} \rangle \) and \( \langle N_{\text{gal}} (N_{\text{gal}} - 1) \rangle \), respectively. One can think of \( N_{m} \) as the maximum number of galaxies that can be formed with the available mass \( m \), and of \( p_m \) as the probability of actually forming a galaxy. For a constant \( N_{m} p_m \), the small \( p_m \) limit gives a Poisson distribution. Here \( N_{m} \) is an increasing function of mass, whereas \( p_m \) peaks at \( m \approx 10^{12} M_\odot h^{-1} \) with \( p_m \approx 0.8 \). The higher order factorial moments are completely determined once the first two moments have been specified; they obey

\[ \langle N_{\text{gal}} (N_{\text{gal}} - 1) \ldots (N_{\text{gal}} - j) \rangle = x^2 (2x^2 - 1) \ldots (jx^2 - j + 1) \langle N_{\text{gal}} \rangle^{j+1}. \tag{49} \]

In this model, all the moments become Poisson-like at the same mass scale, i.e., when \( x(m) = 1 \), all the factorial moments become Poisson, \( \langle N_{\text{gal}} (N_{\text{gal}} - 1) \ldots (N_{\text{gal}} - j) \rangle = \langle N_{\text{gal}} \rangle^{j+1} \). However, at small scales, where small halos contribute, the galaxy counts per halo are significantly sub-Poisson. Our binomial model provides a simple way of accounting for this. To correct the power spectrum and bispectrum for shot noise, we use the same form as in the Poisson case (Peebles 1980):

\[ P_q(k) = P(k) - \varepsilon, \]
\[ B_{123} = B_{123} - \varepsilon(P_1^2 + P_2^2 + P_3^2) - \varepsilon^2, \tag{50} \]

but where the parameter \( \varepsilon \) is set to \( \varepsilon \equiv P(k \to \infty) \), to avoid making the corrected power spectrum negative at small scales. In the Poisson case, \( \varepsilon^{-1} = (2\pi)^3 \bar{n}_g \), we find that in our prescription \( \varepsilon \) can be smaller than the Poisson value by almost a factor of 2. Although this model is somewhat arbitrary, our results are insensitive to shot-noise subtraction for \( k \lesssim 5 \ Mpc^{-1} \) and smoothing scales \( \bar{R} \gtrsim 1 \ Mpc h^{-1} \).

Figure 6 shows the resulting galaxy correlation functions. The top panel shows the predictions of the ST (dot-dashed line) and PS (solid line) mass functions for galaxies using the relations given above. For comparison, we also show the mass power spectrum predicted by the fitting formula of Peacock & Dodds (1996). Note that the galaxy power spectrum is well approximated by a power law, even though the dark matter spectrum is not. This is due to the fact that galaxy formation is inefficient in massive halos; this suppresses the one-halo term compared to the dark matter case. In addition, at small scales the sub-Poisson statistics of galaxy counts per halo suppresses the galaxy spectrum relative to the dark matter. The dotted line in the top panel shows how the predictions change for the ST model when the low-mass cutoff in the \( N_{\text{gal}}(m) \) relation is raised from \( m_{\text{cut}} = 10^{11} M_\odot h^{-1} \) to \( m_{\text{cut}} = 10^{11.5} M_\odot h^{-1} \). At large scales, suppressing low-mass halos leads to an increase of the bias factor from \( b_1 = 0.86 \) to \( b_1 = 1.12 \). At small scales, since these low masses do not contribute significantly, the overall amplification of the one-halo term is due to the lower galaxy number density, which decreases by almost a factor of 2. Therefore, our galaxy clustering predictions are quite sensitive to the low-mass cutoff for galaxy formation.

The bottom panel in Figure 6 shows the galaxy reduced bispectrum, \( Q_{\text{gal}} \), for equilateral configurations as a function of scale (solid lines), compared to the dark matter value (dashed lines) for the ST mass function. At large scales, the halo models predict a negative quadratic bias, which suppresses the bispectrum. At smaller scales, both the power spectrum and bispectrum are suppressed, but at a given scale the bispectrum is sensitive to larger mass halos than the power spectrum, so \( Q_{\text{gal}} \) rises more gradually than \( Q_{\text{dm}} \). At small scales, \( k \gtrsim 1 \ Mpc^{-1} \), the different mass weighting of the galaxy bispectrum leads to a suppression of the scale dependence of \( Q \). The dot-dashed curve shows again the sensitivity to the low-mass cutoff \( (m_{\text{cut}} = 10^{11.5} M_\odot h^{-1}) \), as expected, the distribution at small scales is more biased. To be more quantitative, let us note that at small scales, where one-halo terms dominate, the scaling of the p-point spectrum is

\[ T_p(k) \sim \int \left[ \frac{n(m) m^p u_{\text{halo}}(m)}{m^p} \right]^p \frac{\langle N_{\text{gal}} \rangle}{m^p} dm, \tag{51} \]

and at small masses, \( m < m_s, n(m) \sim m^{-2} v^2(m), c(m) \sim m^{-b}, \) and \( u_{\text{halo}}(m) \sim [\sin\kappa\text{Si}(\kappa) - \cos\kappa\text{Ci}(\kappa)]/\ln c, \) where \( \text{Si}(\kappa) \equiv \pi/2 - \text{Si}(\kappa). \) This expression for the profile follows when \( c \gg 1, \) at large \( \kappa, u_{\text{halo}}(m) \sim (\kappa\ln c)^{-1}. \) Furthermore, if \( v(m) \sim m^{\alpha} \), as in the scale-free case, and \( \langle N_{\text{gal}} \rangle \sim m^{\alpha+1} \), changing the variable from \( m \) to \( \kappa, \) we have

\[ T_p(k) \sim k^{6(1 - p + p_s) - (n + 3)a}/[2(3b + 1)], \tag{52} \]
where we have neglected the weak logarithmic dependence of the profile on the concentration (which leads to additional suppression at small scales) and used the fact that the integral over \( \kappa \) converges. This means that the reduced spectra \( Q_p(k) \sim T_p(k)/[P(k)]^{p-1} \) scale as \( Q_p(k) \sim k^{\gamma_p} \), with

\[
\gamma_p = (p - 2) \frac{a(n + 3) - 6\epsilon}{2(3b + 1)}.
\]

This generalizes the derivation in Ma & Fry (2000b) for galaxies. The validity of the hierarchical Ansatz, \( \gamma_p \approx 0 \), thus depends on the low-mass behavior of the concentration parameter, mass function, and \( N_{\text{gal}} \). For scale-free initial conditions, the validity of the hierarchical Ansatz for the mass correlation functions is linked to that of stable clustering (Peebles 1980), although for higher order correlation functions stable clustering takes a different form than the usual two-point requirement that pairwise peculiar velocities cancel the Hubble flow (Jain 1997). We see that for \( p = 3 \) (the bispectrum), \( a = 0.4 \) (ST mass function), \( \epsilon = 0.1 \), \( n_{\text{eff}} \approx -2.5 \) \((k = 3 \ h \ Mpc^{-1})\), and \( b = 0.13 \), \( \gamma_3 \approx -0.3 \), which agrees approximately with the behavior of \( Q_{\text{gal}} \) in Figure 6. This is only qualitative, due to the many approximations involved. On the other hand, it captures the general behavior of mass weighing; for \( b > 0 \), suppressing contributions from massive halos \((e > 0)\) preferentially places galaxies in more concentrated halos, and thus at a given \( k \) one is probing outer regions of the NFW profile compared to the mass case, since in the outer regions \( u(r) \sim r^{-3} \), \( Q_{\text{gal}} \) is less scale dependent than \( D \).

Figure 7 shows the \( S_p \) parameters as a function of smoothing scale \( R \) for the mass (dashed line; same as ST in Fig. 5), the galaxies with \( m_{\text{cut}} = 10^{10.5} M_{\odot} h^{-1} \) (dot-dashed line), and the galaxies with \( m_{\text{cut}} = 10^{11.5} M_{\odot} h^{-1} \) (solid line). Although we see the same general behavior as in the bispectrum case, it is interesting that the galaxy \( S_p \) parameters are smaller than the dark matter ones at small scales, which is similar to the trend seen in comparisons of dark matter predictions with real galaxy catalogs. Both galaxy plots assume that the galaxy per halo moments obey equation (49). We have repeated the calculation assuming Poisson statistics \([\bar{a}(m) = 1]\) and found that the results are the same for scales \( R \geq 5 \ Mpc^{-1}\), and that at \( R = 1 \ Mpc^{-1}\) the Poisson values are about 15% below those shown in Figure 7. The sensitivity of the clustering to the details of the relation of the number of galaxies as a function of halo mass can be used to probe aspects of galaxy formation, as we now discuss.

5. COMPARISON WITH APM SURVEY: CONSTRAINTS ON GALAXY FORMATION

Measurements of two-point and higher order moments of the galaxy field in the APM survey (Maddox et al. 1990) provide important constraints on models of galaxy clustering. Here we use the measurements of counts-in-cells, deprojected into three dimensions, by Gaztañaga (1994, 1995) to constrain the \( N_{\text{gal}}(m) \) relation.

As discussed above, galaxy clustering at large scales is given by the standard local-bias model, with bias parameters obtained from equation (39). To constrain the \( N_{\text{gal}}(m) \) relation, we use the parametrized form

\[
\langle N_{\text{gal}} \rangle = \begin{cases} 
(m/m_0)^{g_1} & m_{\text{cut}} \leq m \leq m_0, \\
(m/m_0)^{g_2} & m \geq m_0,
\end{cases}
\]

with \( \langle N_{\text{gal}} \rangle = 0 \) for \( m < m_{\text{cut}} \), and the second moment obeys equation (47) with

\[
\bar{a}(m) = \frac{\log (m/m_{\text{cut}})}{\log (m_0/m_{\text{cut}})}.
\]

for \( m \leq m_0 \), and \( \bar{a}(m) = 1 \) for \( m > m_0 \). For the higher order moments, we adopt the binomial model in equation (49). Requiring \( p_m \) in the binomial model to be positive implies that \( a_1 \geq 0 \). Note that in equation (54) we have set \( \langle N_{\text{gal}} \rangle = 1 \) at \( m = m_0 \), since clustering statistics do not depend on the overall number density of galaxies; however, constraints such as the luminosity function and the Tully-Fisher relation are sensitive to the overall amplitude of the \( N_{\text{gal}}(m) \) relation.

For a given luminosity cut, say as the whole APM sample, it is clear that we can use the freedom in the amplitude of the \( N_{\text{gal}}(m) \) relation (which does not affect clustering) to satisfy the luminosity function (LF). However, this does not guarantee that the shape of the LF is correct. In order to test this, one needs to know exactly how \( N_{\text{gal}}(m) \) changes as a function of luminosity cut; this can then be used in our expressions to predict clustering statistics for these different cuts and the shape of the LF. In a similar way, one could incorporate constraints from the Tully-Fisher relation. Another interesting point is that detailed knowledge of how \( N_{\text{gal}}(m) \) changes as a function of color can be used to make predictions and compare surveys selected in different bands.

For the measured values in APM at large scales, we use \( \sigma_N^2 = 0.167 \pm 0.021 \) at \( R = 20 \ Mpc h^{-1} \), and for the higher order moments we use \( (S_3)_{\text{gal}} = 3.3 \pm 0.5 \) and \( (S_4)_{\text{gal}} = 15 \pm 4 \) at \( R = 20 \ Mpc h^{-1} \). The latter are fiducial values obtained by averaging over the large-scale \((R > 10)\) skewness and kurtosis, rather than the value at a particular scale. The skewness at large scales is rather flat, so averaging is reasonable (see Fig. 10, below); for the kurtosis, the large-

![Figure 7.](image-url)
scale limit is not so well defined, but in practice this constraint is not very important because it comes with large error bars.

The constraint on the variance at large scales sets the linear bias parameter, $1 \leq b_1 \leq 1.15$, whereas the skewness also depends on the nonlinear bias, $b_2$. In halo models, $b_1$ and $b_2$ are not independent—for a given mass function, they have a specific relation as a function of halo mass, equations (14) and (15), shown in Figure 8 in terms of the threshold parameter $v = \delta_c / \sigma(m)$. As a result of this relation, it turns out to be nontrivial to satisfy both $\sigma^2$ and $(S_3)_{gal}$ constraints simultaneously. Essentially, since galaxies in cluster-normalized ΛCDM are constrained by the APM variance to be almost unbiased at large scales ($b_1 \approx 1$), a high skewness (as high as the mass skewness, shown as dashed lines in Fig. 10) requires that which is not easy to obtain if galaxy formation is inefficient at small and large halo masses (the latter is required to suppress the one-halo term contribution to the variance and match the observations at small scales). To quantify constraints on the $N_{gal}(m)$ relation, we run a Monte Carlo with varying parameters for the $N_{gal}(m)$ relation,

$$10^9 \leq m_{cut} \leq 10^{13} \, M_\odot \, h^{-1}, \quad m_{cut} \leq m_0 \leq m_{cut} \times 10^4, \quad -1 \leq a_1 \leq 4, \quad 0 \leq a_2 \leq 1.5, \quad (56)$$

and use the bias parameters from equation (39) in equation (44) to decide whether a given model is accepted. We considered both PS and ST mass functions. For the mass, we take that at $R = 20$ Mpc $h^{-1}$, $\sigma^2 = 0.145$, $S_3 = 2.9$, and $S_4 = 14.6$. Note that the inferred deprojected variance of the APM corresponds to a median redshift of $\bar{z} = 0.15$, so we extrapolate the predictions from ΛCDM $\sigma_8 = 0.90$ to this redshift.

If we impose no further constraint on the high-mass slope $a_2$, we find that the maximum skewness (at $R = 20$ Mpc $h^{-1}$) is $S_3 = 3$ for the ST mass function (with $a_2 = 1$) and $S_3 = 3.1$ for the PS mass function (with $a_2 = 1.1$). However, such a high value for $a_2$ means that one-halo terms are not suppressed with respect to the mass, and thus the variance and skewness at scales $R \approx 1-5$ Mpc $h^{-1}$ are much larger than observed. If we restrict $a_2 \leq 0.9$, we find that the maximum skewness for the PS mass function becomes $S_3 = 2.4$, uncomfortably small for APM galaxies (but see below for discussion of APM deprojection issues). For the ST mass function, we find that values as high as $S_3 = 2.64$ are possible (with $S_4 = 12.6$); this requires in addition that $a_1 \approx a_2 = 0.9$ and $m_{cut} \approx 10^{10} \, M_\odot \, h^{-1}$, so galaxy formation is relatively efficient in small-mass halos. This values is insensitive to $m_0$, since $a_1 \approx a_2$, and the large-scale clustering is insensitive to the distribution of galaxies within halos. Essentially, in this model galaxies trace as much as possible the dark matter. However, as shown in Figures 9 and 10 by solid lines ($m_{cut} = 8 \times 10^9 \, M_\odot \, h^{-1}, m_0 = 6 \times 10^{10} \, M_\odot \, h^{-1}$, $a_1 = 1.2$, $a_2 = 0.9$), the small-scale variance and skewness are overestimated in this model. Therefore, further suppression of galaxy formation in high-mass halos is required, $a_2 < 0.9$. Similarly, the galaxy model of equation (46) (Figs. 9 and 10, dot-dashed lines) suffers from the same problem for the skewness.

Before we turn to constraints derived from small-scale clustering, we should note that these depend on at least two additional assumptions [rather than just the $N_{gal}(m)$ relation]. First, we are assuming that galaxies trace the dark matter profile. Figure 2 in Diaferio et al. (1999) shows that this cannot be true for both red and blue galaxies. The blue semianalytic galaxies, which should be more like the ones in the APM survey, are preferentially located in the outer regions of their parent halos. In the semianalytic models, this happens because, in the time it takes for a galaxy’s orbit to decay (by dynamical friction) from the edge of its parent

![Fig. 8.—Halo linear ($b_1$, solid lines) and nonlinear ($b_2$, dashed lines) bias parameters as a function of threshold $v = \delta_c / \sigma(m)$ for the PS and ST mass functions. The dot-dashed lines show $b_1 = 1$ and $b_2 = 0$ for comparison.](image)

![Fig. 9.—Variance of APM galaxies as a function of smoothing scale $R$ (filled circles) compared to the predictions for galaxies as in eq. (46) with $m_{cut} = 10^{12} \, M_\odot \, h^{-1}$ (dot-dashed line) and galaxies from eq. (54) (solid line) with $a_1 = 1.2$, $a_2 = 0.9$, $m_{cut} = 0.8 \times 10^{10} \, M_\odot \, h^{-1}$, and $m_0 = 6 \times 10^{10} \, M_\odot \, h^{-1}$. Dotted lines show the predictions for using eq. (54) with the parameter values in eq. (57).](image)
we find that at 3 Mpc one-halo term from these halos is suppressed. For example, since many small-mass halos host at most one galaxy, so the For galaxies this may not be true anymore, particularly terms dominate over halo-halo correlation contributions.

Hence, this is well justified, because at scales at which exclusion effects become important (invalidating the extrapolation from PT), one-halo terms dominate over halo-halo correlation contributions. For galaxies this may not be true anymore, particularly since many small-mass halos host at most one galaxy, so the one-halo term from these halos is suppressed. For example, we find that at 3 Mpc, the contributions of one- and two-halo terms to the variance and the third moment of the galaxy counts are comparable, whereas this scale is about 5 Mpc for the dark matter. If one assumes that exclusion effects suppress the contribution of two-halo terms (see, e.g., Fig. 5 of Mo & White 1996, or the analytic treatment of exclusion effects in Sheth & Lemson 1999), one might conclude that the skewness is higher than what we get by including two-halo contributions using PT, since these affect more the square of the variance than the third moment. On the other hand, Mo, Jing, & White (1997) show that, on scales smaller than the nonlinear scale, our formulae for the three- and four-halo terms significantly overestimate the skewness and kurtosis of the halos in their simulations (see the bottom right panels of their Figs. 3 and 4). Therefore, although these exclusion effects may conspire to approximately cancel out in the end, we should bear in mind that nontrivial behavior from exclusion effects can invalidate our conclusions from clustering statistics at small scales.

Modulo these caveats, if we impose the additional constraint that at small scales, e.g., \( R = 1 \) Mpc \( h^{-1} \), \( (S_3)_{\text{gal}} \approx 6 \), we find models with parameters

\[
m_{\text{cut}} = 4 \times 10^9 \, M_{\odot} \, h^{-1}, \quad a_1 = 1, \quad a_2 = 0.8,
\]

\[
m_{\text{cut}} = 2.5 \times 10^{10} \, M_{\odot} \, h^{-1}, \quad m_0 = 10^{12} \, M_{\odot} \, h^{-1}, \quad a_1 = 0.8, \quad a_2 = 0.8,
\]

\[
m_{\text{cut}} = 6 \times 10^{10} \, M_{\odot} \, h^{-1}, \quad m_0 = 1.2 \times 10^{12} \, M_{\odot} \, h^{-1}, \quad a_1 = 0.6, \quad a_2 = 0.8,
\]

the first of which is shown in Figures 9 and 10 as dotted lines (the others have very similar behavior). However, all these models predict a small-scale variance that is too large; in fact, in Figure 9 we have used a concentration parameter that is a factor of 2 smaller than equation (12) to decrease the small-scale variance. We were unable to find a model that matched both the variance and skewness of the APM survey at all scales. Therefore, we conclude that the skewness provides a stringent constraint on models of galaxy formation. In particular, the relatively large skewness at large scales and relatively small skewness at small scales provides opposite requirements for the number of galaxies per halo for massive halos. The “large” value of the skewness at large scales requires that galaxies trace mass; however, the small value of the skewness at small scales requires that galaxy formation be suppressed in massive halos.

In deprojecting from angular to three-dimensional space, Gaztaña (1994, 1995) assumed the validity of the hierarchical Ansatz for the third and higher order correlation functions. At large scales, however, PT predicts that the hierarchy of correlation functions is not a simple hierarchical model with constant amplitudes, but rather that the amplitudes depend strongly on the shape of the configuration (i.e., the bispectrum depends on the triangle configuration). This can affect the deprojection (Bernardeau 1995; Gaztaña & Bernardeau 1998); in particular, it can lower the three-dimensional skewness and higher order moments deduced from the angular data, perhaps by as much as 20\% at \( R \gtrsim 10 \) Mpc \( h^{-1} \) (E. Gaztaña, private communication). At large scales, at least, this would improve the agreement with halo-model predictions.

6. CONCLUSIONS

We used the formalism of Scherrer & Bertschinger (1991) to construct an analytic model of the dark matter and of galaxies. For the dark matter, we find that the predictions are in good agreement with numerical simulations, except for the configuration dependence of the bispectrum at small scales. This (numerically small) disagreement can be traced to the assumption that halo profiles are spherically symmetric; in reality, the halos generically found in CDM simulations are triaxial. In general, halo models provide accuracy no better than 20\% when compared to simulations. However, as N-body results converge on the values of the ingredients of halo models (profiles, mass functions, etc.), their predictions will improve.
We showed how the finite volume of the simulation box can significantly affect the results of higher order statistics, due to the fact that small boxes have a deficit of massive halos. In particular, we showed that, by making suitable cuts in halo mass function, we were able to reproduce the observed behavior of the bispectrum in small-box (100 Mpc $h^{-1}$) simulations. At small scales, halo models predict a significant departure from the hierarchical scaling, unless the low-mass dependence of the concentration parameter, the mass function, and the small-scale slope of the halo profile are different from the currently accepted values. Unfortunately, the limited resolution of our simulations cannot test these predictions. We caution that Ma & Fry’s (2000a) conclusion about the breakdown of the hierarchical Ansatz in numerical simulations is premature, since their results are likely to suffer from finite-volume effects and inadequate resolution.

If galaxies in individual dark matter halos trace the dark matter profiles, galaxy clustering is completely determined within halo models by specifying the moments of galaxy counts as a function of halo mass. In general, suppression of galaxy formation in large-mass halos leads to a power-law-like behavior for the galaxy power spectrum and higher order moments that are smaller than for the dark matter. This is similar to what is observed in galaxy surveys. However, a quantitative comparison with counts-in-cells statistics in the APM survey puts stringent constraints on the galaxy counts as a function of halo mass. At large scales, these require that galaxies trace the mass as closely as possible, implying that galaxy formation is relatively efficient even in small-mass halos. In addition, the small-scale behavior of the skewness requires a high-mass slope for the $N_{gal}(m)$ relation of about $a_s = 0.8$, although we found no model that simultaneously fits the small-scale value of the second moment. The parameters of our ‘‘best-fit models’’ are given in equations (57)–(59). These constraints are independent of those derived from the luminosity function and Tully-Fisher relation, which are sensitive to the overall amplitude of the galaxy counts as a function of halo mass.

Clearly, halo models provide a useful framework within which to address many interesting questions about galaxy clustering, and also related topics such as galaxy-galaxy and quasar-galaxy lensing. Our treatment should be of particular interest for the interpretation of clustering in future galaxy surveys, such as the Sloan Digital Sky Survey (SDSS) and Two-Degree Field (2dF). In these cases, however, redshift distortions of clustering must also be taken into account. We plan to address this issue and others relevant to upcoming galaxy surveys in the near future.

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