Scaling in Gravitational Clustering, 2D and 3D Dynamics

D. Munshi\textsuperscript{1}, F. Bernardeau\textsuperscript{2}, A. L. Melott\textsuperscript{3}, R. Schaeffer\textsuperscript{2}

\textsuperscript{1} Queen Mary and Westfield College, London E1 4NS, United Kingdom
\textsuperscript{2} CEA, Service de Physique Théorique, F-91191 Gif-sur-Yvette cédex, France
\textsuperscript{3} Department of Physics and Astronomy, University of Kansas, Lawrence, Kansas 66045, U.S.A.

1 September 2021

ABSTRACT

Perturbation Theory (PT) applied to a cosmological density field with Gaussian initial fluctuations suggests a specific hierarchy for the correlation functions when the variance is small. In particular quantitative predictions have been made for the moments and the shape of the one-point probability distribution function (PDF) of the top-hat smoothed density. In this paper we perform a series of systematic checks of these predictions against N-body computations both in 2D and 3D with a wide range of featureless power spectra. In agreement with previous studies, we found that the reconstructed PDF-s work remarkably well down to very low probabilities, even when the variance approaches unity. Our results for 2D reproduce the features for the 3D dynamics. In particular we found that the PT predictions are more accurate for spectra with less power on small scales.

The nonlinear regime has been explored with various tools, PDF-s, moments and Void Probability Function (VPF). These studies have been done with unprecedented dynamical range, especially for the 2D case, allowing in particular more robust determinations of the asymptotic behavior of the VPF. We have also introduced a new method to determine the moments based on the factorial moments. Results using this method and taking into account the finite volume effects are presented.

Key words: Cosmology: theory – large-scale structure of the Universe – Methods: statistical

1 INTRODUCTION

The general frame of the current theory for the large–scale structure formation is generally believed to be based on the gravitational amplification of small density perturbations. The detailed study of such a mechanism started in the seventies with a complete treatment of the linear growth of the initial fluctuations (e.g. Peebles 1980), but progress towards an understanding of the nonlinear aspects of the dynamics has been very slow. Zel’dovich (1970) proposed an interesting approximation for a qualitative description of some nonlinear aspects but this approximation cannot provide accurate quantitative predictions, and is certainly not valid in the late stage of the nonlinear dynamics. The discovery of the self-similar solutions with the assumption of stable clustering (Davis & Peebles 1977) provided a valuable insight into the fully nonlinear regime. Subsequent numerical analysis, however, partially confirmed the existence of such a regime but only for very large mean density fluctuations, for rms density fluctuations above ten. More recently a lot of theoretical and numerical efforts have been devoted to the study of the quasi-linear regime corresponding to density fluctuations below unity. These results suggest we should distinguish three different regimes:

(i) the linear or quasi-linear regime at large scale when the variance is below unity;
(ii) the intermediate regime when the variance is between unity and 10;
(iii) the nonlinear regime for the smallest scales for which the rms density fluctuation exceeds 10.

The nonlinear regime is expected to be reasonably well described by the self-similar solutions, which is very useful because it provides a well-defined frame for theoretical predictions. We recall here the general results. For an initial power-law spectrum, \( P(k) \propto k^n \), the small scale two-point correlation function, \( \xi_2 \), is expected to follow a power-law behavior,

\[ \xi_2(r) \sim r^{-\gamma} \]  

(1)

with an index \( \gamma \) related to \( n \) (Davis & Peebles 1977). Moreover the higher order correlation functions are also expected to follow a specific behavior for their global scale dependence,

\[ \xi_p(\lambda r_1, \ldots, \lambda r_p) = \lambda^{-\gamma(p-1)} \xi_p(r_1, \ldots, r_p). \]  

(2)
From this general property it is possible to show that the average $p$-point correlation functions,

$$\xi_p = \frac{1}{V^p} \int_V d^3 r_1 \ldots \int_V d^3 r_p \, \xi_p(r_1, \ldots, r_p),$$

in a cell of volume $V$ are related to the average two-point correlation function with

$$\xi_p = S_p \xi_2^{-1},$$

where the $S_p$ coefficients are scale independent. Note that $\xi_p$ is identical to the $p$-order cumulant of the one-point density probability distribution function. There are however no definitive theories for the $S_p$ parameters, although different models have been proposed (Hamilton 1988, Fry 1984, Schaeffer 1992, Balian & Schaeffer 1988, 1989a, Bernardeau & Schaeffer 1992, Munshi & Padmanabhan 1996).

The intermediate regime is certainly the most poorly understood from a theoretical point of view. It has been the subject of only few semi-analytic investigations (Jain, Mo & White 1995, Mo & White 1996, Padmanabhan 1996, Munshi & Padmanabhan 1996).

The linear and quasi-linear regime have been investigated in details over the last few years using perturbation theory techniques. In particular it is possible to show that for Gaussian initial conditions the average $p$-point correlation functions follow the hierarchy (Fry 1984, Goroff et al. 1986, Bernardeau 1992),

$$\xi_p = S_p^{PT} \xi_2^{-1},$$

where the $S_p^{PT}$ coefficients (not necessarily identical to the $S_p$ in eq. [4]) depend on the local shape of the power spectrum (Goroff et al. 1984, Bouchet et al. 1992). A lot of effort has been devoted recently to the analytic calculation of these coefficients in different cases (Juszkiewicz, Bouchet & Colombi 1993, Juszkiewicz et al. 1995, Bernardeau 1994a). Thus $S_3$ and $S_4$ are known analytically or semi-analytically for a Gaussian filter and a power-law spectrum (Lokas et al. 1995). More interesting is the case of a top-hat filter for which the whole series of the coefficients $S_p$ is known for any cosmological model and any power spectrum in 3D (Bernardeau 1994b) and in 2D (Bernardeau 1995). The fact that the complete series is known allows us to build the shape of the one-point density Probability Distribution Function (PDF). These predictions have been checked in peculiar cases at various levels, for the moments (Bouchet et al. 1992, Lokas et al. 1995, Baugh, Gaztañaga & Efstathiou 1995, Colombi et al. 1995) or for the shape of the PDF (Bernardeau 1994b). All these checks have been made for the 3D dynamics.

Applications to observational data have, so far, given contradictory results. For instance Gaztañaga & Frieman (1994), Gaztañaga (1995) concluded that the observed galaxy distribution in the APM survey reproduces the theoretical predictions, but Bernardeau (1995) claimed a significant discrepancy. In any case the fact that quantitative predictions, that are direct consequences of the gravitational instability scenario, could be checked in observational data boosted the theoretical investigations in this domain. In particular the calculation of the next-to-leading order term in the perturbative expansion of $\xi_2$ have been contemplated by Scoccimarro & Frieman (1995a, b), Lokas et al. (1995). In this paper we propose a systematic checks of the PT results for power law spectra, completing previous results in 3D and investigating the 2D case. Our objective is to circumvent the validity range of all the PT results at the level of the PDF's, and also for the void probability function.

The paper is divided as follows. In section 2 we recall the mathematical tools that originally have been developed for the fully nonlinear regime by Balian & Schaeffer (1989a, b), and turn out to be useful for the quasi-linear regime as well. In the subsequent section we present the specific results obtained in the quasi-linear regime, and compare them to the numerical results in quasi-linear and nonlinear regime. Most of the mathematical details about calculating $S_p$ parameters using method based on factorial moments are presented in the appendix.

## 2 SCALING AND COUNTS IN CELLS STATISTICS

To explore the statistical properties of the particle distribution in an N-body simulation we consider the Count Probability Distribution Function (CPDF), $P_r(N)$, the probability of having $N$ particles in a spherical cell of radius $l$. In order to relate the CPDF to the $p$-point correlation functions we can consider its generating function,

$$P(\lambda) = \sum_{N=0}^{\infty} \lambda^N \frac{P_r(N)}{N!},$$

that can be shown to be given by (White 1979, Schaeffer 1984, Balian & Schaeffer 1989a, Szapudi & Szalay 1993),

$$P(\lambda) = \exp \left( \sum_{p=1}^{\infty} \frac{(nV)^p (\lambda - 1)^p}{p!} \, \xi_p \right),$$

where $n$ is the mean number density of particles in the considered sample. Assuming scale invariant $p$-point correlation functions we can further write,

$$P(\lambda) = \exp \left( -\phi((1-l) n V \xi_2) \right),$$

where the function $\phi$ is defined by,

$$\phi(y) = -\sum_{p=1}^{\infty} \frac{S_p}{p!} (-y)^p,$$

(we have set $S_1 = S_2 = 1$). We can notice that the Void Probability Function (VPF) is equal to the generating function for $\lambda = 0$. As a result the probability $P_1(N)$ can be derived from the VPF through the relation,

$$P_1(N) = \frac{\partial^N P_r(N)}{\partial \lambda^N} \bigg|_{\lambda=0} = \frac{(-n)^N}{N!} \frac{P_r(0)}{\partial n^N}$$

where the partial derivatives are taken at fixed volume. From eq. [10] one can show that for a continuous field

$$p(\delta) \, d\delta = d\delta \int_{-1\infty}^{+1\infty} \frac{dy}{2\pi \xi_2} \, \exp[-\phi(y)\xi_2 + (1 + \delta) \, y/\xi_2].$$

For small $\xi$ and very small $\delta$, $p(\delta)$ reduces to the well-known Gaussian form, $p(\delta) \propto \exp(-\delta^2/2\xi_2)$, but even for moderate small values of $\delta$, strong deviations from the Gaussian behavior are expected (Balian & Schaeffer 1989a, Bernardeau 1992).
In the nonlinear regime, under the assumption of scale-invariance of the correlation functions (in the sense that the coefficients $S_n$ are scale-independent), $P(N)$ and $p(\delta)$ exhibit characteristic scaling laws. This implies that the VPF has a specific scale dependence, namely that,

$$
\sigma(N_c) = -\ln[P(0)]/(n V),
$$

is a function of the combination

$$
N_c = n V \xi/g,
$$

(13)

only. The only reasonable physical models are those for which $\sigma(N_c)$ decreases to 0 when $N_c$ is large (Balian & Schaeffer 1989a). It is thus quite reasonable to assume that, at large $N_c$, $\sigma(N_c)$ follows a power-law behavior,

$$
\sigma(N_c) \sim N_c^{-\omega},
$$

(14)

where $\omega$ is a model dependent parameter that should be comprised between 0 and 1.

Let us describe in more details the consequences of these assumptions in the fully nonlinear regime as they were found by Balian & Schaeffer (1989a). Thus, for the shape of $P_1(N)$ three domains are expected. They are delimited by $N = N_c$ and $N = N_v$, with

$$
N_v = N_c (\xi_2/\rho)^{-1/(1-\omega)}.
$$

(15)

This number is smaller than $N_c$ when the variance is large (we will see that $\alpha$ is of the order of 1). Then, when $N < N_v$, $P_1(N)$ is expected to follow a specific scaling,

$$
P_1(N) = \frac{1}{N_v} g\left(\frac{N}{N_v}\right)
$$

(16)

where the function $g$ is given by

$$
g(z) = \frac{-z^{-1/\omega}}{\frac{1}{2\pi i}} \int_{-i\infty}^{i\infty} \frac{dt}{t} \exp\left[z^{1/(1-\omega)}(t - t^{-1})\right].
$$

(17)

It depends on $\omega$ only, and when $z$ is small it reads,

$$
g(z) \approx \frac{1}{z} \left[1 - \frac{1 - \omega}{2\pi \omega} \frac{1}{z} \left(1 - \frac{z}{1 - \omega}\right)^{1/(1-\omega)} \times \exp\left(-\omega \left(1 - \frac{z}{1 - \omega}\right)^{1/(1-\omega)}\right)\right]^{\frac{1}{1 - \omega}}.
$$

(18)

When $N_c < N < N_v$, $P_1(N)$ is expected to follow a power-law behavior,

$$
P_1(N) = \frac{a}{N_c \xi_2} \frac{1 - \omega}{1/(\omega)} \left(\frac{N}{N_c}\right)^{-\omega^2}.
$$

(19)

And when $N > N_v$ another scaling behavior is expected,

$$
P_1(N) = \frac{1}{N_v \xi} h\left(\frac{N}{N_v}\right),
$$

(20)

with

$$
h(x) = -\int_{-\infty}^{+\infty} \frac{dy}{2\pi i} \phi(y) \exp(xy).
$$

(21)

Furthermore, it is quite natural to expect that $\phi(y)$ has a singular behavior for a negative and small value of $y$,

$$
\phi(y) \sim \varphi_3 - a_3 \Gamma(\omega_3) (y - y_3)^{-\omega_3}.
$$

(22)

This singularity induces an exponential cut-off for $h(x)$,

$$
h(x) \sim a_3 x^{\omega_3 - \omega} \exp(-|y_3| x),
$$

(23)

when $x$ is large. the scaling functions $\sigma(N_c)$, $g(z)$ and $h(x)$ have been studied in great detail computationally including spurious effects that may alter their measurements in finite $N$-body catalogues (Bouchet & Hernquist 1992, Colombi et al. 1992, 1994, 1995).

3 SCALING PARAMETERS FROM THE QUASI-LINEAR REGIME

In the quasi-linear regime the scaling for the dominant part of the correlation functions is a consequence of the dynamical evolution. The picture obtained in the previous section for the shape of $P_1(N)$ is however no more valid when $\xi_2$ is small. But actually the functions $h(x)$ and $g(z)$ can still be used to describe the shape of the density PDF's. The main difference with the highly non-linear regime is that the low and large density cut-offs merge together, thus suppressing the domain of the power law behavior. It is then still relevant to describe the results in terms of $\alpha$, $\omega$ for the low density domain and in terms of $\alpha_s$, $y_s$ for the large density tail.

We take advantage here of the fact that, for top-hat filtering, the whole series of the $S_p^{PT}$ parameters is known (Bernardeau 1994b, Colombi et al. 1996), and we will make this assumption in the following, thus considering the global properties of $\phi_T(y)$ as a valid model for the calculation of the scaling functions, $\sigma(N_c)$, $h(x)$, $g(z)$. So the function, $\phi_T(y) = -\sum_{p=1}^{GP} S_p^{PT} (-y)^p/p!,$ is given by the system,

$$
\phi_T(y) = y + y \mathcal{G}_T[y] + \frac{1}{2} y^2 \tau(y)
$$

(25)

and

$$
\tau(y) = -y \frac{d \mathcal{G}_T[y]}{dy},
$$

(26)

where the function $\mathcal{G}_T(\tau)$ can be deduced from the spherical model dynamics, since we have

$$
\mathcal{G}_T(\tau) = \mathcal{G}_s \left[ \frac{\sigma_M \left(M_0 [1 + \mathcal{G}_T(\tau)]\right)}{\sigma_M(M_0)} \right]^{\tau - \tilde{\tau}}
$$

(27)

and

$$
\mathcal{G}_s(\tau) = \mathcal{G}_s \left( \frac{\tau}{\sigma_M(M_0)} \right). \left[1 + \mathcal{G}_T(\tau)\right]^{-2/\eta_f},
$$

(28)

To get quantitative predictions from PT one needs to know the expression of $\mathcal{G}_s(\tau)$ in both cases. There are no simple

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analytical expressions for $G_3(\tau)$. It is however possible to develop $G_3(\tau)$ with respect to $\tau$ and one obtains,

$$G_3(\tau) = -\tau + 17 \frac{\tau^2}{21} - 341 \frac{\tau^3}{567} + \ldots$$

(29)

for the 3D case (Bernardeau 1994b) and

$$G_3(\tau) = -\tau + 18 \frac{\tau^2}{21} - 29 \frac{\tau^3}{42} + \ldots$$

(30)

for the 2D case (Bernardeau 1995). One can then compute the values of the first $S_3$ coefficients for the 3D case,

$$S_{3D} = \frac{34}{7} - (n + 3),$$

(31)

$$S_{4D} = \frac{6712}{1323} - 62 \frac{(n + 3)}{3} \frac{(n + 3)^2}{7},$$

(32)

and for the 2D case,

$$S_{3D} = \frac{36}{7} - 5 \frac{(n + 2)}{2},$$

(33)

$$S_{4D} = \frac{2540}{49} - 33(n + 2) + 24 \frac{(n + 2)^2}{4}.$$  

(34)

To have more global properties of $\phi(y)$ it is necessary to know the global shape of $G_3(\tau)$ and one can actually show that,

$$G_3(\tau) = \left(1 - \frac{\tau}{\nu}\right)^{-\nu} - 1,$$

(35)

with

$$\nu = \frac{3}{2}$$

for 3D,

(36)

and

$$\nu = \frac{\sqrt{13} - 1}{2} \approx 1.3$$

for 2D,

(37)

is actually a good approximate function. The parameters $\nu$ are chosen to reproduce the exact asymptotic behavior of $G_3(\tau) + 1$ for large $\tau$. Note that the Zel’dovich approximation would have given $\nu = d$ in (23), (Bernardeau & Kovacs, 2005). It is interesting to remark also that the asymptotic behavior of $\phi(y)$ for large values of $y$, and thus the value of $\omega$, is given by the asymptotic behavior of $G_3(\tau) + 1$ for large $\tau$ (Bernardeau & Schaeffer 1992). From (29) it is easy to show that

$$G_3(\tau) + 1 \sim \tau^{-\nu/(1-\nu(n+d)/(2d))}$$

(38)

so that

$$\omega^{PT} = \frac{d}{d(2 + \nu) / \nu - (n + d)}.$$  

(39)

As a result for the 3D case we have,

$$\omega^{3D} = \frac{3}{7 - (n + 3)},$$

(40)

and for the 2D case,

$$\omega^{2D} = \frac{2}{2(3 + \sqrt{13})/(\sqrt{13} - 1) - (n + 2)} \approx \frac{2}{5.1 - (n + 2)}.$$  

(41)

Note that these results are exact, in the sense that they do not depend on the approximation (23) made for $G_3(\tau)$. These parameters entirely determine the shape of the function $g(z)$. It can be however more interesting to consider a

Table 1. Parameters of the singularity, eq. (43), for the 2D PT case

| $n$ | $y_s$ | $\phi_s$ | $r_s$ | $a_s$ |
|-----|-------|----------|-------|-------|
| -2  | -0.171979 | -0.19731 | 1.60005 | -1.71747 |
| -1.5 | -0.211979 | -0.251915 | 1.80602 | -2.24518 |
| -1  | -0.276897 | -0.345685 | 2.22626 | -4.0954 |
| -0.5 | -0.402865 | -0.580901 | 6.63172 | -7.73243 |

Table 2. Parameters of the singularity, eq. (43), for the 3D PT case

| $n$ | $y_s$ | $\phi_s$ | $r_s$ | $a_s$ |
|-----|-------|----------|-------|-------|
| -3  | -0.1848 | -0.214286 | 1.6565 | -1.83824 |
| -2.5 | -0.213447 | -0.253791 | 1.80389 | -2.20845 |
| -2  | -0.252692 | -0.3113 | 2.0344 | -2.80188 |
| -1.5 | -0.309854 | -0.402947 | 2.44269 | -3.93026 |
| -1  | -0.401244 | -0.572949 | 3.3437 | -6.69931 |
| -0.5 | -0.57352 | -1.00758 | 6.63172 | -18.6473 |

more general expression given by the saddle point approximation in eq. (11),

$$p(\delta)d\delta = -\frac{d\delta}{G_3^{PT}(\delta) G_3^{PT}(\delta)} \left[1 - \tau \frac{G_3^{PT}(\delta)/G_3^{PT}(\delta)}{2\pi\sigma^2}\right]^{1/2} \times \exp\left(-\frac{\tau^2}{2\sigma^2}\right),$$

(42)

with $G_3^{PT}(\delta) = \delta$.

This expression reduces to (10) when the functions $G_3^{PT}(\tau)$ and $G_3^{PT}(\tau)$ are replaced by their asymptotic power-law behavior.

To obtain the position of the cut-off in $h(x)$ one needs to know the dominant singular value of $\phi^{PT}(y)$ and the behavior of $\phi^{PT}(y)$ near this value. It is actually not given by the singularity appearing in the expression of $G_3(\tau)$ but generally by the 2nd order equation of the system (23). It is quite easy to see that there is a value of $y$ for which $(dy/d\tau) = 0$. At this point we have

$$\phi^{PT}(y) = \phi_s + r_s (y - y_s) + a_s (y - y_s)^{3/2} + \ldots$$

(43)

As a result we get an asymptotic shape for $h(x)$ given by $^*$

$$h(x) \approx \frac{3\sigma a_s}{4\sqrt{\pi}} (\delta + 1 - r_s)^{-5/2} \times \exp[-|y_s|(1/\delta)^{2} + |\phi_s|/\sigma^2].$$

(44)

The parameters $\phi_s$, $a_s$, $y_s$ and $r_s$ are given in table 1 for the 3D case as a function of the power law index $n$ and in table 2 for the 2D case.

One can see that the singularity is sharper for low values of $n$, and for the 2D case. Note that for $n \geq 0$ there is no singularity anymore, and the form (42) only can be used to describe $P(N)$. In such a case the asymptotic behavior of $G_3^{PT}(\tau)$ is

$$1 + G_3^{PT}(\tau) \approx \left(\frac{\tau}{\tau_c}\right)^{2d/(n+d)},$$

(45)

so that $|\tau_c| \approx 1.47$ for the 2D case and $|\tau_c| \approx 1.69$ for the 3D case. As a result the large density tail takes the form,

$$p(\delta)d\delta = -\frac{\tau_c(n+d)}{2d} (1 + \delta)^{(d-n)/(2d)} \times \exp[-|y_s|(1/\delta)^{2} + |\phi_s|/\sigma^2].$$

(46)

$^*$ In Bernardau (1992), the regular term $r_s (y - y_s)$, was neglected so there is a slight change in eq. (44)

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\[ x \sqrt{2n/(n+d)} \exp \left( -\frac{x^2}{2\sigma^2} (1 + \delta)^{(n+d)/d} \right) \, d\delta \]

which gives a sharper cut-off than in the expression [1].

4 COUNT PROBABILITY DISTRIBUTION FUNCTION IN NUMERICAL SIMULATIONS

4.1 The simulations

The simulations used here are numerical models for the gravitational dynamics of collisionless particles in an expanding background. We are studying evolution of initial Gaussian perturbations in \( \Omega = 1 \) universe. All the simulations are done with a particle-mesh (PM) code with 512\(^3\) particles in 2D in an equal number of grid points and 128\(^3\) particles with 128\(^3\) grid points in 3D (Melott 1986; Melott, Weinberg & Gott 1988, hereafter MWG). More details about the peculiarities of the simulations used here can be seen in Melott & Shandarin (1993). The code has at least twice the dynamical resolution of any other PM code with which it has been compared.

We use a subset of initial conditions used by Beacom et al. (1991) and Kofman et al. (1992) in their studies. The time evolution of the N-body models can be seen in the video accompanying the paper of Kauffmann & Melott (1992). The models for which we carry out our analysis are featureless power-law spectra of the general form,

\[ P(k) \propto k^n \text{ for } k \leq k_c, \quad (47) \]

\[ = 0 \text{ for } k > k_c. \quad (48) \]

We have analyzed power-law models with \( n = 2, 0, -2 \) in 2D and \( n = 1, 0, -1, -2 \) in 3D with a cutoff in each case at the Nyquist wave number: \( k_c = 256k_f \) for 2D and \( k_c = 64k_f \) for 3D where \( k_f = 2\pi/L_{\text{box}} \) is the fundamental mode associated with the box size.

We choose \( \sigma(k_{NL}) \), the epoch when the scale \( 2\pi/k_{NL} \) is going nonlinear as a measure of time,

\[ \sigma(k_{NL}) = \left( \frac{\int_{k_f}^{k_{NL}} P(k) k \, dk}{\int_{k_f}^{\infty} P(k) k \, dk} \right)^2 \quad (49) \]

The first scale to go nonlinear is the one corresponding to the Nyquist wave number. This happens, by definition, when the variance \( \sigma \) is unity. Of course as \( \sigma \) increases successive larger scales enter in the nonlinear regime. The simulations were stopped at \( \lambda_{NL} = 2L_{\text{grid}}, 4L_{\text{grid}}, 8L_{\text{grid}}, \ldots, L_{\text{box}}/2 \). The growth rate of various modes in linear theory were studied in MWG for this PM code. The results given by our code at \( \lambda = 3L_{\text{grid}} \) are equivalent to the ones obtain by a usual PM code at \( \lambda = 8L_{\text{grid}} \). This is due to the staggered mesh scheme. So we expect that our code performs well at the wavelength associated with four cells and since the collapse of \( 4L_{\text{grid}} \) size perturbations will give rise to condensations of diameter \( 2L_{\text{grid}} \) or less, the smallest cell size that we take into account should be bigger than \( 2L_{\text{grid}} \). On the other hand Kauffman and Melott (1992) found that for voids of size greater than size \( L_{\text{box}}/4 \) self-similarity was broken in a model equivalent to our index \( n = -1 \) in 3D, see also Gramann (1992) and Melott and Shandarin (1993). We therefore restrict our cell sizes to be less than \( L/10 \). As a result our cell sizes vary between \( 2L_{\text{grid}} \) and \( L_{\text{box}}/10 \). We also do not use cells with \( \sigma < 0.1 \), since in this case shot noise becomes comparable to the fluctuation power impressed on the simulation.

Our simulations were started by using the Zel’dovich approximation (Klypin & Shandarin 1983) but we wait long enough before doing a comparison with the PT results so that the effects of the Zel’dovich approximation have died away (see for instance Baugh Gaztañaga & Efstathiou, 1995).

4.2 The Count Probability Distribution Functions

To evaluate the CPDF we calculate the occupancy of spherical cells of size \( l \) disposed on a regular mesh. The number of cells to be used is to be as large as possible so that all structures are fully taken into account. This can be achieved by considering cells that are about \( l_c \) apart where \( l_c \) is the typical separation between particles in clusters \( (N_c(l_c) = 1) \). Actually the major constraint comes from the resolution of the \( N \)-body code. If \( l_{\text{box}} \) is the resolution scale of the \( N \)-body, it is clear that probabilities below \( (l_{\text{box}}/L)^d \) are meaningless. As a result we probe probabilities as low as \( 10^{-6} \) both in 2D and 3D.

Finite size of the sample affects mainly the large \( N \) tail of the CPDF, and this effect is all the more important that the cell size \( l \) is large. The main effect is that the large density tail of the PDF is dominated by a single cluster. It creates a bump in the PDF which is followed by an abrupt cut-off of the distribution. These features have been recognized as faked by Colombi et al (1995) and methods to correct for it have been proposed. This is important in particular for the derivation of the high order moments that are extremely sensitive to the defects in the large density tails. In the method proposed by Colombi et al., this effect is corrected using a theoretical prejudice, that is that the density PDF is assumed to follow an exponential cut-off. This hypothesis is supported by the PT theoretical results and by our understanding of the nonlinear regime. We will use this method as well to compute the high order moments.

We have followed the evolution of CPDF for all the spectra for different scales with time. The different scales that we have studied are in the range \(-2.2 \leq \log(l/L_{\text{box}}) \leq -1.2 \) and they are separated by equal logarithmic intervals \( \Delta \log(l/L_{\text{box}}) = 0.2 \).

4.2.1 Results in the Quasi-linear Regime

In figs. 1 and 2 we present the results of the PDF-s for the 2D and 3D dynamics for different values of the variance. The latter is at most of the order of 1.5 (see table) and the agreement is found to be extremely good for the smaller variances. For variances approaching unity, the departure from the PT results depends on the value of the initial index.

When there is a lot of power at large scale \( (n \) small), the agreement is better then for the other case. This is not too surprising since when \( n \) is large there is a lot of power at small scales that can affect the behavior of the largest scales.

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4.2.2 Results in the nonlinear Regime

The nonlinear regime has been explored mainly in terms of the function $h(x)$ (see eq. [20]). Indeed when the variance is large it is natural to expect that $N_c \Sigma_i \xi_i P_i(N)$ is a function of $N/N_c$ only (for a given power law index). As a result one expects that, when plotted with the appropriate variables, the PDF measured at different smoothing scales coincide. This test is presented on figs. 3 and 4. Here we see that the locations of the PDF are indeed the same when the adequate change of variable is made. It should be noted however, that not all the PDF have been plotted. As the function $h$ is pertinent is the large density tail only, the PDF-s have been truncated in the low density domain. We have removed part of $P(N)$ where it is dominated by shot noise. Typically $N < 10$ while increasing with large scale power. Very large $N$ part of $P(N)$ is dominated by large statistical fluctuations, we applied smoothing to reduce such fluctuations.

In order to have quantitative results we used the parameterized fit proposed by Bouchet, Schaeffer & Davis (1991) to describe the function $h(x)$,

$$h(x) = a(1 - \omega)^{x-2} \exp(-x|y_s|) \frac{\Gamma(\omega)}{(1 + bx)^{\omega}}$$  \hspace{1cm} (50)

The values of $\omega$ and $a$ are estimated from CPDF (see in the next subsection), $|y_s|$ is found from fitting the large $N$ tail of CPDF which shows an exponential cut-off. The other

| Table 3. Parameter values for the $P(N)$ in 2D dynamics |
|-------------------------------------------------------|
| $n = 2$  $n = 0$  $n = -2$ |
| $\xi$ | 0.61 | 0.47 | 0.88 |
| $N_v$ | 4.11 | 10.33 | 69.55 |
| $N_v$ | 1.63 | 1.63 | 4.11 |
| $N_v$ | 0.65 | 0.65 | 0.65 |

| Table 4. Parameter values for the $P(N)$ in 3D dynamics |
|-------------------------------------------------------|
| $n = 1$  $n = 0$  $n = -1$  $n = -2$ |
| $\xi$ | 0.40 | 0.48 | 0.57 | 0.71 |
| $N_v$ | 33.23 | 33.23 | 33.23 | 33.23 |
| $N_v$ | 8.33 | 8.33 | 8.33 | 8.33 |
| $N_v$ | 2.09 | 2.09 | 2.09 | 2.09 |

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Figure 3. The measured $h(x)$ in 2D N-body simulations for $n = 2$, $n = 0$ and $n = -2$ in highly nonlinear regime. Long dashed lines correspond to fitting function of the form (50) with the parameters of table 5.

Figure 4. The measured $h(x)$ in 3D N-body simulations for $n = 1$, $n = 0$, $n = -1$ and $n = -2$ in highly nonlinear regime. Long dashed lines correspond to fitting function of the form (50) with the parameters of table 6.

Table 5. Parameters of the fitting function $h(x)$ (eq.50) for the 2D case

| $n$ | $\omega$ | $a$ | $|y_s|$ | $b$ | $c$ |
|-----|----------|-----|--------|-----|-----|
| 2   | .85      | 3.35| 1.31   | 3.2 | -.83|
| 0   | .72      | 2.51| .58    | .06 | 1.2 |
| -2  | .30      | 1.2 | .09    | 1.1 | .80 |

Table 6. Parameters of the fitting function $h(x)$ (eq.50) for the 3D case

| $n$ | $\omega$ | $a$ | $|y_s|$ | $b$ | $c$ |
|-----|----------|-----|--------|-----|-----|
| 1   | .70      | 2.38| .64    | .00 | .00 |
| 0   | .55      | 1.59| .41    | .38 | .60 |
| -1  | .40      | 1.31| .23    | .65 | .70 |
| -2  | .33      | 1.25| .11    | .80 | .95 |

two parameters are adjusted to reproduce the constraints $S_1 = 1$ and $S_2 = 1$ to at least 5 percent accuracy.

We give the parameter values that have been used in tables 5 and 6.

The variation with the power law index is extremely large, in particular for the value of $\omega$. It is interesting to note that the results are better when $n$ is large, for which there is a convincing overlapping of curves. The situation is more questioning when $n$ is small, but it is still not clear whether it is due to some numerical difficulties (finite volume effects are large when the power law index is small) or to a genuine physical effect. In particular it has been found that the $S_p$ parameters reach their asymptotic value for quite large values of the variance. The curve presented here may therefore still be in the intermediate regime for which the function $h(x)$ is rapidly changing.

We can also note that the CDM case as analyzed by Bouchet, Schaeffer & Davis (1991) matches roughly with our calculations of $n = -1$ or $n = -2$ for 3D dynamics with their computed value of $\omega = .4 \pm .05$ and $|y_s| = .08$.

4.3 The Moments

Once CPDF has been calculated one can use this information to compute moments of this distribution and hence $S_p$ parameters (eq.[4]). As recalled in the introduction these parameters have been studied extensively in recent past, both analytically and computationally. In the limit $\sigma^2 \rightarrow 0$ and for Gaussian conditions they are constant (Bernardeau 1992) and can be calculated using perturbation theory. The resulting values of $S_3$ and $S_4$ are given in the first section.

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Figure 5. The measured $S_p$ parameters in 3D N-body simulations for $n = 1$, $n = 0$, $n = -1$ and $n = -2$ are plotted against $\bar{\xi}_2$. The filled dots are measurements of $S_p$ parameters for $p = 3$ to $p = 5$ after taking all the corrections. Crosses represent results from Lucchin et al. (1994). Open circles correspond to measurements of $S_p$ parameters by Colombi et al. (1996). Solid lines represent predictions from perturbation theory.

Figure 6. The measured $S_6$ parameter after (filled circles) and before (open circles) finite volume corrections for different initial power spectra $n$. 

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Numerical results show good agreement with theoretical predictions.

Of course these formula are not expected to be valid in the intermediate and in the nonlinear regime. Note however, that the spherical model might be of help to get insights into those two regimes as pointed out by Mo & White (1996), Munshi & Padmanabhan (1996).

In this paper we have developed a new method based on factorial moments (see Appendix) to compute $S_p$ parameters, with corrections due to finite volume effects taken into account. Using this method we have computed $S_p$ parameters up to $p = 6$ for both 2D and 3D and compared them with earlier results for power law spectra. We found that finite volume corrections are extremely important as scale of nonlinearity increases. They are most important for spectra with lot of large scale power e.g. $n = -2$. Comparison of our results with earlier results of Colombi et al. (1996) where different method of volume corrections were applied shows reasonable agreement within the dynamical range studied by us. We however find that the $S_p$ coefficients in the non-linear regime are reasonably constant: the drift $S_p \propto \xi_n^{-0.25 (n-2)}$ seen by Colombi et al. (1996) is no longer present. The difference in the finite volume corrections explain why the agreement is better for spectra with less large scale power. To do these corrections we are using a systematic method - see Appendix - which makes sure that all normalization constraints are well respected. Note however that the corrections depend mainly on proper parameterization of $m(x)$. With increasing large scale power accurate determination of $m(x)$ and $\sigma(N_c)$ becomes difficult. This may also partially explain the lack of agreement between our result and Colombi et al. (1996) for $n = -2$ spectra.

The $S_p$ parameters were also studied by Lucchin et al. (1992) using central moments. Our results shows good agreement for all spectra with their results. Although finite volume corrections were not taken into account they used several realizations of the same spectra which made their results agree with ours where volume corrections were taken care of and only one realization has been used. Since variation of $S_p$ was not studied for large values of $\xi_2$ in Lucchin et al. (1994) it is difficult to compare their results with our results.

4.4 The Void Probability Function

In the previous section we have investigated mainly the large density tails. As mentioned earlier the void probability function is directly related to the clustering properties but contains complementary information in numerical measurements. The scaling in the CPDF statistics is related to the scaling in $\sigma \equiv -\ln[P(0)]/(nV)$. Note that in the absence of clustering, for a Poisson distribution, $\sigma$ is exactly unity and when some clustering is present the VPF is expected to grow (it is more probable to find a void) and hence $\sigma$ decreases. Moreover from the previous sections we know that the function $\sigma$ when expressed as a function of $N_c$ is expected to have a power-law behavior for large $N_c$. For small $N_c$ it tends to the Poisson limit, $\sigma(N_c) = 1$.

We have considered seven different epochs starting from the epoch when the grid scale goes nonlinear to the epoch when the box scale goes nonlinear. We have also considered different level of dilution to the full set of $512^2$ and $128^3$ data.

Different spurious effects affect the VPF differently, unlike CPDF it is less affected by finite volume of the sample but it is very much affected by grid effects. Over-dense regions lose memory of the initial grid just after the first shell crossing but in under-dense regions the grid structure is present till very late stages and these under-dense regions contribute more to VPF. It was shown that this effect is negligible if we restrict ourselves to those cells where the conditions $P(0) > 1/e$ is satisfied (Colombi et al. 1995).

Grid effects are more significant for spectra with more power on larger scales where as for spectra with lots of power in smaller scales collapse of smaller objects erases the memory of initial grid very fast.

Scatter in plots increases from $n = 2$ to $n = 0$ and for $n = -2$ the $\sigma(N_c)$ diagrams reveal significant evolution with time. For $n = -2$ spectra larger and larger modes have more power and results are affected badly by finite volume corrections. Since we superpose all scales from quasi-linear to highly nonlinear regime in the same plots, any scatter gives a measure of variation of $S_p$ parameters in these two regimes. In earlier studies, it was noticed that variation of $S_p$ parameter from quasi-linear to highly nonlinear regime increased with presence of large scale power. This may explain the small scatter in our plot for $n = 2$ spectra. We do not see such a trend in our 3D analysis which may be due to...
Figure 8. The measured $\sigma(N_c)$ in 2D N-body simulations for $n = 2$, $n = 0$ and $n = -2$. The open symbols correspond to cases where the variance is below unity, filled symbols for a variance above unity. The slope of the solid lines are given in table 7.

Figure 9. The measured $\sigma(N_c)$ in 3D N-body simulations for $n = 1$, $n = 0$, $n = -1$ and $n = -2$. The open symbols correspond to cases where the variance is below unity, filled symbols for a variance above unity. The slope of the solid lines are given in table 8.

Table 7. Measured values of $\omega$ for the 2D dynamics, Fig. 6, compared to the PT prediction, eq. (39).

| $n$  | $\omega_{\text{meas}}$ | $\omega_{\text{PT}}$ |
|------|------------------------|----------------------|
| 2    | 0.85                   | -0.65                |
| 0    | 0.72                   | 0.39                 |
| -2   |                        |                      |

Table 8. Measured values of $\omega$ for the 3D dynamics, Fig. 7, compared to the PT prediction, eq. (38).

| $n$  | $\omega_{\text{meas}}$ | $\omega_{\text{PT}}$ |
|------|------------------------|----------------------|
| 1    | 0.7                    | 1                    |
| -1   | 0.55                   | 0.75                 |
| 0    | 0.4                    | 0.6                  |
| 1    | 0.33                   | 0.5                  |

The smaller available dynamic range. Computation of $\sigma(N_c)$ for very small values of $\bar{\xi}_2$ is constrained by the restriction $P(0) > 1/e$ which is to be satisfied for avoiding grid effect and discreteness effect.

The values of $\omega$ we get are given in table 7 for the 2D case and in table 8 for the 3D case where the errors on the measured values are about 0.05. It is interesting to see that PT predicts the right trend for the $n$ dependence, although there are significant discrepancies between the measured $\omega$ and the ones predicted by PT.

5 CONCLUSION

We have shown that PDF constructed from PT formalism works extremely well for $\sigma^2 < 1$. Since all the loop corrections to $S_p$ parameters are neglected in this kind of approach it seems that loop corrections may not be very important in the perturbative regime for construction of PDF.

The constructed PDF is more accurate for spectra with less power on smaller scales. The existence of lots of small scale power produces a flow of power from smaller scale to larger scales which contradicts the basic assumptions of perturbation theory where density evolution at sufficiently large scales can always be described by linear theory.

For spectra with more power in larger scales the evolution is quite rapid and $P_N$ take the characteristic nonlinear power-law form quite early even when $\sigma < 1$.

We developed a method based on factorial moments to calculate higher order correlation functions and used them to study evolution of $S_p$ parameters for power law spectra in 2D and 3D. Comparison with earlier studies shows reasonable agreement.
ACKNOWLEDGMENT

It is pleasure for D.M. to acknowledge Varun Sahni, his thesis supervisor for constant encouragement and active support during the course of work. D.M. was financially supported by the Council of Scientific and Industrial Research, India, under its SRF scheme. D.M. also acknowledges financial support from CEA (Saclay) during his stay there where most of the work was completed. A.L.M. wishes to acknowledge the National Center for Super-computing Applications for support to perform the ensemble of simulations, and financial support under NASA grant NAGW-2832.

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APPENDIX

5.1 $P(N)$ and its continuous analogue $\Pi(\nu)$
Using the relation between count in cell and void probability function one can write,

$$ P(N) = \frac{(-1)^N}{N!} \frac{d}{d\mu^N} \exp \left( -\frac{\phi(\mu N)}{\xi_2} \right)_{\mu=1}, $$

(51)

which can also be written in the following form

$$ P(N) = \frac{1}{2\pi i} \int d\lambda N^N \exp \left( -\frac{\phi((1-\lambda)N)}{\xi_2} \right). $$

(52)

The above integral has to be evaluated along a contour around $\lambda = 0$.

We then define the function $\Psi(t) = -\phi(-t)$ with which the function $\Pi(\nu)$ is defined,

$$ \exp \left( \frac{\Psi(t)}{\xi_2} \right) = \int_0^\infty dv \exp \left( \frac{\nu t}{N_c} \right) \Pi(\nu). $$

(53)

Using the definition of $P(N)$ now one can easily show that

$$ P(N) = \int_0^\infty d\nu \exp \left( -\frac{\nu N}{N!} \right) \Pi(\nu). $$

(54)

Therefore $\Pi(\nu)$ can be viewed as the continuous limit of $P(N)$ in the limit of large number densities. This can be seen by change of variable $\lambda = 1 + t/N_c$ in equation (2) and then taking the limit $N_c \to \infty$ and $N \to \infty$ with the ratio $N/N_c$ remaining finite which gives $P(N) = \Pi(N)$. More precisely $P(N)$ is the convolution of the function $\Pi(\nu)$ with a Poisson distribution describing the shot noise effects.

5.2 Factorial moments and $S_p$ parameters

The factorial moments of $P(N)$ can be related with moments of $\Pi(\nu)$ by the following expression.

$$ \sum_{N=0}^\infty N(N-1)\ldots(N-p+1)P(N) = \int_0^\infty x^p\Pi(x)dx. $$

(55)

Now let us expand $\Psi(t)$ in a power series of $t$, $\Psi(t) = \sum_{p} \Psi(t)p^p$. One can convince himself that the coefficients $\Psi_p$ have the following relation with the $S_p$ parameters; $S_p = p^p/\Psi_p$. Since in realistic scenarios, $S_p$ behaves as $p^p$ the $\Psi_p$ coefficient are expected to be of order unity.

Similarly one can define the normalized factorial moments of $P(N)$ by the following expression

$$ \Sigma_p = \frac{\xi_2}{p!N_c} \sum_{N=0}^\infty N(N-1)\ldots(N-p+1)P(N). $$

(56)

These numbers are also of order unity. One can define a generating function $\Sigma(t)$ for $S_p$ parameters $\Sigma(t) = \sum_1^\infty \Sigma_p t^p$. We have the following expression connecting these two generating functions $\Psi(t)$ and $\Sigma(t)$. 

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These formulae are valid for arbitrary values of $\bar{\xi}$, of course in the limit $\bar{\xi}$ one has to work with sufficiently large cells, so that the field

The particle positions in simulations sample an underlying

Expanding the above relation in powers of $1/\bar{\xi}$ we can write,

One can then write down the expressions of $\Psi_p$’s as a function of $\Sigma_p$’s. We present some lower order relations here,

These formulae are valid for arbitrary values of $\bar{\xi}$. Note that of course in the limit $\bar{\xi} \rightarrow \infty$ we have $\Sigma_p = S_p$.

5.3 Minimum and maximum length scales to be probed

The particle positions in simulations sample an underlying continuous field. To extract this field from actual output, one has to work with sufficiently large cells, so that the field points appear continuous. The condition is given by $N_c \gg 1$.

In practice if $l_c$ is the scale where $N_c = 1$, scales of few times $l_c$ start to be usable and obviously the code resolution provides the minimum length scale that can be probed. Choice of maximum length scale is slightly arbitrary as larger and larger cells starts finite volume effects. These effects can be visualize however with irregularities in the shape of the measured PDF-s.

5.4 Number of trials

The typical distances between particles in a cluster is $l_c$. If the sample is divided in cells of size $l$ some information at scales smaller than $l$ is erased. To recover all informations available in the sample one has to use many grids displaced by a distance $l_c$ from each other. As mentioned earlier in 3D there are $(l/l_c)^3$ grids and $(L/l_c)^3$ cells, but only $(L/l)^3$ cells are completely independent.

Events corresponding to $P(N)$’s smaller than $(l/L)^3$ will be either over represented in case there is one such event in the sample, or under represented in case there is none in the sample. So it is clear that one has limited access to $P(N)$’s smaller than the above value. From the exponential decay of $\Pi(\nu) \propto \exp(-\nu/\nu_*)$ with $\nu_* = x_s N$ one gets a limit for $N$ above which the information about $P(N)$ is not contained in the sample.

For $N > N_{max}$, $P(N)$ abruptly drops to zero due to the finiteness of the sample. The moments $\Sigma_p$ are dominated by values of $\nu$ that can be inferred to be $\nu_{comp} \approx (p - 5/2)\nu_*; \quad p \ll 5/2$

These coefficients are not known for $\nu_0 > N_{max}$ that is for the simple for (19) $p$ must satisfy

$P(N)$ whose value is between $(l/L)^3$ and $(l_c/L)^3$ contains systematic wiggles due to the fact that only $(L/l)^3$ cells out of $(L/l_c)^3$ cells are fully independent but averages of $P(N)$’s such as the moments (5) are less sensitive to this. It is nevertheless better to use this information than to drop it. With $(L/l)^3$ trials $P(N)$ will become inaccurate much earlier. $\Sigma_p$ are systematically under estimated for large $p$ due to abrupt drop in $P(N)$ for $N > N_{max}$.

5.5 Correction for finite volume effect

A simple way to correct $\Sigma_p$ is to use the form (17) to supplement the missing information at large $N$, and to use $P(N)$ for $N < N_{max}$.

$P^{corr}(N) = \{1 - \frac{a}{\xi^2} - b \frac{(N - N_c)}{N_c}\} P^{com}(N); \quad N < N_{max}$

$P^{corr}(N) \approx \Pi(\nu); \quad N > N_{max}$

It is clear that the corrected $P(N)$ so constructed have to be normalized properly before using it to calculate $S_p$ parameters.

Using the constrain $\sum P(N) = 1$ we get $a = H_0$ and $\sum N P(N) = 1$ gives us $b = H_1$. Where we have used the following notation

$H_p = \int_{N_{max}/N_c}^{\infty} x^p h(x)dx$

Where we have neglected the corrections of order $1/\xi^2$.

The corrected $N_c$ now can be written as

$N_{c}^{corr} = (1 - 6H_1 \Sigma_3 + H_2) N_c^{comp}$

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and finally the corrected $\Sigma_p$'s are of the form

$$\Sigma^\text{corr}_p = \frac{\Sigma^\text{comp}_p - (p + 1)H_1\Sigma^\text{comp}_{p+1} + \frac{1}{p}H_p}{(1 - 6H_1\Sigma_3 + H_2)^{p-1}}$$ (74)

Corrected $S_p$ parameters can now be recovered by using relation between $\Psi_p$ and $\Sigma_p$ as described already. This way to correct has the advantage of being rather simple and preserve all normalizing conditions.

This method of calculating $S_p$ parameter is an alternative to the method generally used based on central moments.
