A NEW STATISTICAL INDICATOR TO STUDY NONLINEAR GRAVITATIONAL CLUSTERING AND STRUCTURE FORMATION

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ABSTRACT

In an $\Omega = 1$ universe dominated by nonrelativistic matter, the velocity field and the gravitational force field are proportional to each other in the linear regime. Neither of these quantities evolves in time, and they can be scaled suitably so that the constant of proportionality is unity and the velocity and force fields are equal. The Zeldovich approximation extends this feature beyond the linear regime, until the formation of pancakes. Nonlinear clustering that takes place after shell crossing breaks this relation, and the mismatch between these two vectors increases as the evolution proceeds. We suggest that the difference of these two vectors could form the basis for a new indicator of nonlinear clustering. We define an indicator called velocity contrast, study its behavior using $N$-body simulations, and show that it can be used effectively to delineate the regions where nonlinear clustering has taken place. We discuss several features of this statistical indicator and provide simple analytic models to understand its behavior. Particles with velocity contrast higher than a threshold have a correlation function that is biased with respect to the original sample. This bias factor is scale dependent and tends to a constant value at large scales.

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

1. INTRODUCTION

Large-scale structures such as galaxies are believed to have formed out of small density perturbations via gravitational instability. This process, in most popular models, is driven by dark matter, which is the dominant constituent of the universe. We can compute the rate of growth of clustering using linear theory when the perturbations are small. Linear theory, however, has a very limited domain of validity, and we have to resort to numerical simulations for studying evolution of inhomogeneities at late epochs.

In the linear regime the density field is related to the velocity field in a unique manner (in the growing mode), and the density field alone specifies the system completely. Evolution of perturbations is described by a second-order differential equation, and specification of the initial density field and velocity field completely determines the state of the system at any later time. However, for a given nonlinear density field there is no practical method for computing the corresponding velocity field. Our understanding of the nonlinear regime will improve if we have a simple physical indicator of the velocity field. We introduce velocity contrast, a new statistical indicator that may be used to quantify some features of the velocity field.

Velocity contrast can be used for isolating nonlinear regions in the output of numerical simulations. It provides a simple and stable algorithm, in contrast to some other methods that are used for this purpose. These methods are required, as studies of nonlinear gravitational clustering have focused mainly on aspects relating to dark matter. Comparison of these studies with observations is made difficult by the fact that we observe only sources of light. Many techniques have been devised for isolating regions that can host galaxies in numerical simulations. Some of these are quite elaborate, like DENMAX (Bertschinger & Gelb 1991), and hence computationally very intensive. Other schemes, such as the use of a density threshold or the friend-of-friend algorithm, are very simple to implement but have problems with interlopers, since these algorithms do not use any dynamical information. We show that velocity contrast can be used to isolate regions of interest in a relatively simple and robust manner.

In the next section we briefly review dynamical evolution of trajectories in a system undergoing gravitational collapse. This is used to motivate the form of a new indicator, which is introduced in § 3. In § 4 we use $N$-body simulations to study velocity contrast for cold dark matter (CDM) and for a hot dark matter–like spectrum. Section 5 contains discussion of the new indicator using a spherical model and nonlinear approximations. In § 6 we compare density and velocity contrast and study the average relation between them as well as dispersion around it. We also discuss clustering properties of nonlinear mass and with respect to total mass.

2. EVOLUTION OF TRAJECTORIES

The problem of gravitational dynamics in an expanding universe can be simplified considerably for nonrelativistic matter, at scales that are much smaller than a Hubble radius. In this domain, we can take the Newtonian limit of relativistic equations. If the mass of the universe is dominated by collisionless matter (e.g., dark matter), then the system is entirely characterized by the following equations:

$$\frac{dv}{dt} + \frac{\dot{a}}{a} v = - \frac{1}{a^2} \nabla \varphi, \quad \nabla^2 \varphi = 4\pi G \rho_p a^2 \delta = \frac{3}{2} H_0^2 \Omega_0 \frac{\delta}{a},$$

(1)

where $v = \dot{a}$ is the peculiar velocity, $\varphi$ is the perturbed gravitational potential, $\rho_p$ is the background density and $a(t)$ is the scale factor. $H_0$ is the present value of the Hubble constant, and $\Omega_0$ is the density parameter. These equations can be recast in an elegant form by using the growing mode.
of density perturbations as the time parameter:

\[
\frac{du}{db} = -\frac{3Q}{2b} (u - g),
\]

\[
\nabla^2 \psi = \left(\frac{\delta}{b}\right),
\]

\[
g \equiv -\nabla \psi \equiv -\frac{2}{3H_0^2 \Omega_0} \left(\frac{a}{b}\right) \nabla \phi,
\]

\[
Q = \left(\frac{\rho_b}{\rho_c}\right) \left(\frac{\dot{a}b}{\dot{a}}\right)^2,
\]

where \(u = dx/db\) is the "velocity" at time \(t\), \(g\) is the rescaled gravitational force, \(\rho_c\) is the critical density, and \(b(t)\) is the growing solution to the equation

\[
\dot{b} + \frac{2\dot{a}}{a} b = 4\pi G \rho_c b.
\]

In a matter-dominated universe with \(\Omega = 1\), we have \(b = a\) and \(Q = 1\). (We shall consider only this case here, although generalization of our analysis to other models is straightforward.)

In this form, the relations between various quantities in the linear regime are simplified to a great degree. This simplicity also makes the transition to the quasi-linear regime more apparent through a transparent expression for the Zeldovich approximation.

Equations (2) describe a complicated many-body system even in the limit of a smooth gravitational potential. From the structure of equations (2), we can distinguish four different epochs in the evolution of clustering: linear, Zeldovich, quasi-linear, and nonlinear.

At sufficiently early time, \(\delta \ll 1\) and we can use linear perturbation theory to study growth of perturbations. In this limit, we can easily solve equations (2) and show that

\[
\delta(a, x) = a f(x),
\]

\[
u(a, x) = u(x) = g(x),
\]

\[
g(x) = -\nabla \psi(x),
\]

\[
\nabla^2 \psi(x) = f(x),
\]

Clearly, \(u(a, x)\) and \(g(a, x)\) are independent of \(a\) and \(\delta \propto a\).

Also note that in the linear regime, \(u(a, x) = g(a, x)\). [Velocity here is, of course, defined in a dynamically relevant manner as \(u = (dx/da)\); the conventional definition of peculiar velocity is \(v = dx - \dot{a}da\), and it scales as \(v \propto a^{1/2}\)]. We shall work with \(u\), since we can always obtain \(v\) by a simple rescaling.]

Linear theory becomes invalid when density contrast becomes comparable to unity; however, we can understand some aspects of dynamics by using the Zeldovich approximation (Zeldovich 1970). This approximation extrapolates equality of velocity and force beyond the linear regime:

\[
u(a, q) = g(a_{in}, q) = -\nabla \psi(a_{in}, q),
\]

\[
x = q + b(a)u(a, q).
\]

Here \(q\) is the initial position of the particle, also called its Lagrange position. Particles move with a constant velocity \(u\) that is related to the force at its initial position at the initial epoch. This approximation compares well with true motion before shell crossing. The validity of this approximation can be used to infer that \(u(a, x) \approx g(a, x)\) in the Zeldovich regime.

The Zeldovich regime ends with the formation of pancakes and shell crossing. After shell crossing, particles oscillate about pancakes and form small clumps. These clumps move toward each other or have some bulk motion toward deep potential wells. Therefore, in the quasi-linear regime, velocities of particles are not aligned with the gravitational force, except in the direction of bulk motion of the clump to which the particles belong.

The mismatch between velocity and force steadily increases until velocities are randomized inside clumps and dominate over any residual bulk motion. Such a situation is expected in highly nonlinear clusters that have either virialized or are close to it.

3. NEW STATISTICAL INDICATOR: VELOCITY CONTRAST

The above discussion shows that mismatch between the peculiar velocity and the gravitational force is a good indicator of nonlinearity in dynamics for pancake-like models. However, it is as good an indicator for hierarchical models because nonlinearity in dynamics at small (mass) scales does not influence the evolution of larger scales. Shell crossing is necessary for formation of nonlinear objects at a given scale, independent of the dynamical state of smaller scales. Virialized structures cannot form without shell crossing and mixing in the phase space. Merger of smaller structures leading to a larger virialized object is always accompanied by shell crossing at the new scale. Therefore, while studying a given mass scale we can neglect nonlinearity in dynamics at much smaller scales. In fact, numerical simulations of hierarchical models are based on the assumption that nonlinearity at small scales does not influence larger scales.

Mismatch between velocity and gravitational force is a vector quantity and is therefore difficult to handle and interpret. In addition, it is a dimensional quantity, and the numerical value of the mismatch must be compared with something else to make a model-independent estimate of the level of nonlinearity in dynamics. These considerations led us to suggest the following form for velocity contrast:

\[
D_{uu} \equiv \frac{(u - g)^2}{u^2}.
\]

In this equation we can think of \(u\) and \(g\) as the velocity of a given particle and the force acting on it, respectively. Alternatively, if smooth velocity and force fields are given, equation (6) defines a scalar field \(D_{uu}(a, x)\). These two descriptions are equivalent for our purpose. If linear theory (or the Zeldovich approximation) is valid for most particles (regions), then this quantity will be nearly zero for a large fraction of particles (mass). With further evolution of clustering, more and more particles (mass) acquire significant values for this parameter. Regions containing particles (mass) with \(D_{uu}\) larger than a threshold value will exhibit significant mixing, and a study of these regions will offer insight into the study of nonlinear clustering. Luminous objects that we see have nonlinear density contrasts; therefore it is more meaningful to compare clustering properties of nonlinear objects in simulations with observations. \(D_{uu}\) provides a simple method for selecting nonlinear structures, of mass much larger than the mass of individual particles, in

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numerical simulations. Velocity contrast shares this limitation with all other methods that use numerical simulations.

This algorithm has two specific advantages over the method of density threshold or the friends-of-friends algorithm, where one finds nonlinear structures by requiring local density to be higher than some cutoff value, or particles to be within some linking length of nearest neighbors. The density threshold (linking length) for an object that has decoupled from expansion depends on the local symmetry of collapse, whereas deviations of velocity from acceleration depend only on the stage of dynamical evolution. When particles in a particular region are counted, in order to ascribe a density to that region, one does not take into account velocities of particles. A fast-moving particle that is merely passing through a region will be counted in this process even if it is following a Zeldovich trajectory with \( u = g \). In using \( D_{\text{gr}} \) as an indicator, we exclude such particles until the alignment between \( u \) and \( g \) is disrupted. Our algorithm will select “nonlinear particles” in underdense regions as well. These are found in regions where pancakes are forming in otherwise underdense regions or where density is low but shear is important.

Velocity contrast suggests another method for studying approximation schemes. Note that \( D_{\text{gr}} \) directly characterizes what could arguably be considered the most significant factor in the formation of bound structures: the ability of a local mass inhomogeneity to pull back particles toward it and thereby increase the local potential depth. It is precisely the failure to do this that makes the Zeldovich approximation break down at the onset of significant nonlinearity. For comparison, consider the “frozen potential approximation” (FPA; see Bagla & Padmanabhan 1994; also suggested independently by Brainerd, Scherrer, & Villumsen 1993), which may be thought of as a logical continuation of the Zeldovich approximation (ZA); in the ZA, velocities are frozen to initial values and force is ignored; in the FPA, gravitational force is frozen to the initial value and the particles are moved in this given background potential. Studies have shown that the FPA correctly reproduces the behavior of particles near the mass concentrations, since \( (u - g) \) is not reassigned to vanish in this approach. Therefore, the structure of pancakes in the FPA is similar to that in N-body simulations; pancakes do not thicken, although they are not as thin as those seen in N-body simulations. Comparison of \( D_{\text{gr}} \) in the FPA and the N-body simulations will be useful in understanding both.

Note that this statistical indicator takes into account both the velocity and the force. It is possible to devise approximation schemes, like the frozen flow (Matarrese et al. 1992) and adhesion models (Gurbatov, Saichev, & Shandarin 1989), which will move particles to the right regions but will give a physically unacceptable picture for the velocity field. In our opinion, approximation schemes should also provide the correct velocity field if they are to offer insight into dynamics. \( D_{\text{gr}} \) may be used to discriminate between the dynamical content of different approximation schemes.

Velocity contrast can be used to compare analytical models with the N-body model. For simple analytic models of structure formation, such as the spherical top-hat model, it is possible to compute \( D_{\text{gr}} \) as a function of the density contrast \( \delta \). As we shall see, this helps one to form an intuitive picture of the nonlinear evolution.

4. N-BODY SIMULATIONS

To understand velocity contrast, we study it using N-body simulations. Use of analytical methods for this study is made difficult by the fact that velocity contrast vanishes in the linear regime. In this section we present results for two representative models of structure formation.

The first model we consider is the standard unbiased CDM model normalized to COBE. Simulation of this model used a particle-mesh code on a 128³ box with 128³ particles. The size of the box in physical units is 90 Mpc. Figure 1 shows projected density and velocity contrast fields for slices 14 Mpc thick. These slices correspond to redshifts \( z = 3 \), \( z = 1 \), and \( z = 0 \).

In a hierarchical model like CDM, collapse occurs at very small mass scales at very early times. However, since small scales do not influence the dynamics of much larger scales, we can safely ignore the fact that shell crossing has already occurred at scales smaller than those considered in a simulation. To some extent the results of simulations, for velocity contrast or any other indicator, depend on the smallest mass scale probed by the given simulation and can only be trusted for mass scales much larger than the limit of resolution. With this note of caution we proceed with the discussion of velocity contrast.

Comparison of panels at a given redshift shows that density and velocity contrast select almost the same set of nonlinear regions. These tend to disagree for some regions that appear dense to the eye but have not reached a sufficiently high level of nonlinearity in dynamics. On the other hand, some regions in which appears to be a void have a high value of velocity contrast. Highly nonlinear regions appear to have a larger radius of influence as diagnosed by velocity contrast in comparison with density. This may indicate that velocity contrast does not increase substantially after virialization, so that almost all members of a virialized cluster have similar values of velocity contrast, even though density varies rapidly with distance from the center. Evolution with redshift follows the expected pattern, with more regions becoming nonlinear at later times.

Regions selected for contours of high velocity contrast tend to be spherical. This indicates that planar and filamentary structures cannot attain a very high level of nonlinearity in dynamics; for nonspherical structures there is always some direction in which bulk velocity dominates over random motions.

An interesting feature that emerges from Figure 1 is that velocity contrast filters out voids (apart from regions within them where large shear leads to formation of structures) more effectively as compared to density contrast.

The CDM model is a prime example of a hierarchical clustering scenario with power at all scales and effective index varying from \(-3\) to \(1\). To see how this algorithm works in other cases, we studied the other extreme example. We considered a hypothetical universe which has power peaked at one dominant scale. The power spectrum for this model is taken to be

\[
P(k) = \frac{A}{\Delta k(2\pi)^{1/2}} \exp \left[-\frac{(k - k_0)^2}{2(\Delta k)^2}\right],
\]

with \( k_0 = \pi/16 \) and \( \Delta k = \pi/64 \). Clearly, this has power only in a band of width \( \Delta k \) peaked at \( k_0 \). We chose \( A = 68,000 \), so that the linearly extrapolated amplitude of the peak is unity at \( z \sim 2 \), giving us sufficient coverage of both the
Fig. 1.—These figures show gray-scale maps and contours of projected density and velocity contrast for slices taken from N-body simulations of CDM. The left frames show the density field, and the right frames show the velocity contrast field for the corresponding slices. The top frames are for $z = 3$, the middle frames are for $z = 1$, and the bottom frames show the same slice at $z = 0$. 
linear and the nonlinear regime. Our choice of $k_0$ and $\Delta k$ ensures that the power is peaked at a large scale, and it is concentrated in a very narrow range of scales. Note that this spectrum is somewhat similar to HDM models as far as the peak and smaller scales are concerned; standard hot dark matter has $P(k) \propto k$ for small $k$, but this model has an exponentially low power at small $k$. This is not very important, since a $k^4$ spectrum will be generated due to dynamics.
Fig. 3.—These figures show close-ups of two clusters for the CDM simulation. The left panels show the projected density field, and the right panels show the projected velocity contrast field. A cube of size $14\, h^{-1} \, \text{Mpc}$ centered at the clusters was used for these plots.

Fig. 4.—Same as Fig. 3, but for a model with a Gaussian power spectrum. We have shown a close-up of only one cluster for this model.
Fig. 5.—Fraction of particles above a threshold velocity contrast, shown as a function of threshold for three redshifts: $z = 3$, $z = 1$, and $z = 0$. Thick lines are for CDM, and dashed lines are for a model with a Gaussian power spectrum. The top axis has been labeled by the equivalent density contrast for spherical top-hat collapse.

(Zeldovich 1965, §§ 4, 10, 24, and 28). The evolution of this model provides an interesting paradigm for structure formation and will be discussed in detail elsewhere (Bagla & Padmanabhan 1996). Here we merely use this model to study velocity contrast in order to test it in the widest possible range. This simulation used a $128^3$ box with $64^3$ particles.

Figure 2 shows projected density and velocity contrast fields for this model. The thickness of the slice used here is $20L_c$, where $L_c$ is one grid length in the simulation box. Panels of this figure follow the pattern of panels in Figure 1, showing that the general behavior of velocity contrast is independent of the type of model considered. Particular values of velocity contrast for particles will of course depend on the mass resolution used in simulations, but the smoothed field has the same properties for a large range in mass scales: $M_{\text{particle}} < M < M_{\text{box}}$.

One key difference between density and velocity contrast is that density is a function of space and is the same for particles located at the same position, whereas velocity contrast can be different for such particles. This feature of velocity contrast suggests that it can give us insight about some facets of clustering in highly nonlinear regions that are inaccessible through density contrast. We bring out these differences by studying some clusters in greater detail. Figure 3 shows a close-up of two rich clusters from the CDM simulation. For each cluster we have plotted projected density and velocity contrast fields for a cube of size 14 $h^{-1}$ Mpc centered at the cluster. Comparison of the panels for density and velocity contrast shows that for high-density regions the information content of these indicators differs significantly. Contours of equal density indicate positions of density peaks and outlines of pancakes, whereas contours of velocity contrast show us regions that have significantly nonlinear dynamics.

Some regions that appear as pancakes or small density peaks are not seen in contours of velocity contrast; particles in these regions have not undergone any shell crossing but are only falling into the cluster. In all such regions, bulk flow toward the nearby cluster dominates over any random motions within the infalling mass. Many density peaks show complex substructure within. This can result from two scenarios: either there has been some recent merger of two small clusters and these have not mixed sufficiently, or a group of particles with a much lower level of nonlinearity in dynamics has fallen into the cluster and it has not become an integral part of the cluster in a dynamical sense.

Figure 4 shows a similar close-up of a cluster for the model with a Gaussian power spectrum. Here the substructure is not as complex as in Figure 3, because there was no small-scale power in the initial spectrum. However, broad features are similar for both these models.

We now give the above results in a less picturesque manner. Figure 5 shows the fraction of particles with velocity contrast above a threshold as a function of threshold velocity contrast at different epochs for the CDM model. The same figure shows corresponding curves for the second model as dashed lines. For reference, we have labeled the equivalent density contrast for top-hat spherical collapse on the top axis. These curves show that as clustering proceeds, the fraction of particles with $D_{rv}$ larger than a given threshold increases. Also, the fraction of particles selected with a high cutoff are significantly less than those selected by a smaller value of $D_{rv}$. These features are expected in any indicator of nonlinearity, and it is heartening to see that velocity contrast satisfies these criteria. These features can also be seen, at a qualitative level, in Figures 1 and 2.

5. ANALYTICAL MODELS AND NONLINEAR APPROXIMATIONS

In this section we study velocity contrast with help of nonlinear models and analytical approximations. Spherical top-hat (STH) collapse is one model of nonlinear collapse in which it is possible to calculate almost all quantities of interest. Here we compute velocity contrast as a function of density contrast for this model. Initial velocities are related to initial density contrast through the potential as

$$ u(a, x) = g(a, x) = -\nabla \psi(a, x), \quad \nabla^2 \psi(a, x) = \frac{\delta(a)}{a}. \quad (8) $$

Here $x$ is the comoving coordinate and relates to the proper coordinate $r$ in the usual way. These initial conditions put the system into growing mode, and this leads to the following solution:

$$ r = \frac{r_{\text{max}}}{2} (1 - \cos \theta), $$

$$ a = \frac{3a}{5\delta} \left[ \frac{3}{4} (1 + \delta)(\theta - \sin \theta) \right]^{2/3}, \quad (9) $$

$$ \delta = \frac{9 (\theta - \sin \theta)^2}{2 (1 - \cos \theta)^2} - 1. $$

We can also write down expressions for $g$ and $u$, leading to an expression for velocity contrast:

$$ g = -\frac{\delta}{3a^2} r, $$

$$ u = -\frac{1}{3a} \frac{\partial \ln (1 + \delta)}{\partial a} r. \quad (10) $$

From these, given a value of velocity contrast we can determine the corresponding phase angle and solve for density
A comparison with equation (2) shows that the viscosity term tries to mimic gravitational force, and

$$\mathbf{g}_{\text{eff}} = \mathbf{u} + \frac{2b}{3Q} \nabla^2 \mathbf{u}. \tag{12}$$

In this equation, gravitational force is some kind of an effective acceleration modeled by the viscosity term.

The general solution to equation (11) is well known, and it allows us to compute $\mathbf{u}$ for any particle, given the initial velocity potential. We can use this solution to compute velocity contrast for the effective force in an adhesion model. It is apparent that $D_{\text{in}}$ is significantly large only at the caustics in the limit $\nu = 0$. At present we are studying $D_{\text{in}}$ as a function of $\nu$, which could lead to an understanding of the origin of effective viscosity in the adhesion model.

6. EVOLUTION AND CLUSTERING OF NONLINEAR STRUCTURES

Velocity contrast and its relation to density contrast is shown in Figure 7, where we have plotted contours of equal population on the $(1 + \delta)D_{\text{in}}$ plane. This plot shows contours for two epochs each for the two models we are using in this paper: CDM and the Gaussian power spectrum. For comparison we have also plotted the curve obtained from STH collapse on each of the panels. These plots show that there is an average relation between velocity contrast and density that is independent of epoch and model, and holds for a large range of scales. For CDM simulations, it is apparent that the relation evolves at high velocity contrast and the density for a given velocity contrast increases with time. This could imply a shift from mostly planar collapse occurring at early epochs toward more instances of spherical collapse at later stages.

There is a large dispersion about the average relation between density and velocity contrast. Part of this dispersion arises from the fact that collapse of various kinds, from planar to spherical, is occurring in the simulation volume. As indicated by the relation of velocity contrast to density for symmetric collapse in FPA, the variation in density contrast for a given velocity contrast may be as large as an order of magnitude. Contours for the second model show a very large dispersy at large velocity contrast. This indicates ongoing collapse of objects with different local symmetries.

Velocity contrast is a good indicator of nonlinearity. In a simulation output we can identify nonlinear regions by requiring velocity contrast to be higher than some threshold. Regions selected in this manner will in general have clustering properties that differ from those of the total underlying mass. We study clustering properties of these models and define a scale-dependent bias parameter. Only for those regions where particles have undergone shell crossing (at the scale of a galaxy) can we form virialized structures like galaxies that we see. Therefore, the clustering properties of regions with velocity contrast above a threshold are essentially clustering properties of regions that can host visible structures. Figure 8 shows an averaged correlation function for the underlying mass and also for regions with velocity contrast above a threshold in a CDM simulation at $z = 0$. [The bias parameter for particles with velocity contrast between two thresholds $(D_{\text{in}1} < D_{\text{in}} < D_{\text{in}2})$ may show antibias for small values of the upper threshold $D_{\text{in}2}$. See Bagla 1996a for more details.] The threshold values
used here were 0.1, 1, and 17. The averaged correlation function $\xi$ and bias $b$ are defined here as

$$\bar{\xi}(a, x) = \frac{3}{x^3} \int_0^x \xi(a, y) y^2 \, dy,$$

$$b^2(x) = \frac{\bar{\xi}(x; D_{\text{mu}} > D_{\text{max}})}{\bar{\xi}(x; \text{all particles})},$$

where $\xi$ is the two-point correlation function. The relative bias decreases with the scale, approaching a constant value at very large scales, and increases with the cutoff used in selecting particles. This scale-dependent bias may be used in understanding relative bias between different kinds of objects, like different populations of galaxies and clusters of galaxies. Unlike the method of peaks in the initial distribution, the algorithm described in this paper clearly picks out regions that are dynamically “nonlinear” and therefore are potential sites for galaxy formation. It may be possible to model the environmental effects in galaxy formation, as $D_{\text{mu}}$ describes the level of “churning” for nonlinear regions, which is likely to be an important factor.

7. CONCLUSIONS

The purpose of this paper was to introduce the new statistical parameter described above, to study its behavior, and to establish a prima facie case that it is worth being considered further. It will be interesting to see whether one can develop an analytic model for the evolution of this statistical indicator, or a closely related one. This and related questions are under investigation.
Fig. 8.—Parameter $\xi$ vs. $x$ for CDM at $z = 0$. We have also plotted $\xi$ for particles with $D_{\text{vir}} > 0.1, 1, \text{and } 17$. The left panel shows the bias parameter for these threshold values as a function of scale.

We are studying the bias parameter introduced here in greater detail. As a part of this study, we are also investigating effects of mass resolution in numerical simulations on this bias. (Results of $N$-body simulations depend, to some extent, on the smallest scale in the simulations set by mass of particles.) This work is in progress and will be reported elsewhere (Bagla 1996b).

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REFERENCES

———, 1996, in preparation