On the Onset of Stochasticity in ΛCDM Cosmological Simulations

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March 24, 2008

ABSTRACT
The onset of stochasticity is measured in ΛCDM cosmological simulations using a set of classical observables. It is quantified as the local derivative of the logarithm of the dispersion of a given observable (within a set of different simulations differing weakly through their initial realization), with respect to the cosmic growth factor. In an Eulerian framework, it is shown here that chaos appears at small scales, where dynamics is non-linear, while it vanishes at larger scales, allowing the computation of a critical transition scale corresponding to $\sim 3.5 \text{Mpc}/h$. This picture is confirmed by Lagrangian measurements which show that the distribution of substructures within clusters is partially sensitive to initial conditions, with a critical mass upper bound scaling roughly like the perturbation’s amplitude to the power 0.15. The corresponding characteristic mass, $M_{\text{crit}} = 2 \times 10^{13} M_\odot$, is roughly of the order of the critical mass of non linearity at $z = 1$ and accounts for the decoupling induced by the dark energy triggered acceleration.

The sensitivity to detailed initial conditions spills to some of the overall physical properties of the host halo (spin and velocity dispersion tensor orientation) while other “global” properties are quite robust and show no chaos (mass, spin parameter, connectivity and center of mass position). This apparent discrepancy may reflect the fact that quantities which are integrals over particles rapidly average out details of difference in orbits, while the other observables are more sensitive to the detailed environment of forming halos and reflect the non-linear scale coupling characterizing the environments of halos.

Key words: Cosmology, Chaos, N-body, etc.

1 INTRODUCTION

Concerns regarding the predictability of cosmological measurements in simulations have been with us for some time. In the neighbouring field of secular galactic evolution, it has been known for a while [Sellwood and Wilkinson (1993)] that the significant undersampling of resonances could mislead the dynamical evolution of N-body systems when the evolution time becomes large compared to the local dynamical time. Over the course of the last decade, various “universal” relationships [Navarro et al. (1997), Zhang and Fall (1999), Richer et al. (1991)] have been extracted numerically from cosmological N-body simulations. In this context, significant efforts [Power et al. (2003)] have been invested in comparing different numerical schemes and codes, but, with the development of very high resolution “zoom” re-simulations [Weil et al. (1998), Diemand et al. (2004), Hansen and Moore (2006a), Sales et al. (2007), Strigari et al. (2007)] one question remains: how sensitive is a given run with respect to its initial conditions? In particular, what set of observables is likely to be robust with respect to a specific choice in the “phases” of the draw (the whitened initial realization)? In the context of cosmology, the general assumption has been that, even though the detailed orbits of dark matter particles are likely to be poorly resolved by the numerical schemes implemented, the properties of structures would nevertheless be well represented statistically. In other words, so long as the simulated region was large enough to represent a fair sample of dynamically independent regions, the stochastic exponential departure from the unperturbed trajectories was expected to average out when considering such a statistical sample. The question remains for features specific to a given realization, such as the relative position of objects.

The sensitivity of the gravitational N-body problem to small changes in initial conditions has been investigated in details by Kandrup and collaborators in a series of papers [Kandrup and Smith (1991, 1992), Kandrup et al. (1993, 1994)] in the context of a Newtonian (time-independent) Hamiltonian. They have shown in particular that the growth of small perturbations in initial conditions is exponential, with a mean e-folding time that is asymptotically independent of the number of particles at large N, and a distribution of e-folding times that is reproducible from simulation to simulation for sufficiently large N. In the cosmological context, the N-body description is an approximation of the collisionless Boltzmann equation for the evolution of the dark matter, so that another related question is in which sense a limit to the
continuum can be established as the number of particles increases. Indeed, it has been argued in the literature (Kandrup (1990); Goodman et al. (1993)) that the discretization of smooth, and possibly integrable potentials invariably leads to strongly chaotic orbits in the N-body framework, independently of the number of particles; this has been confirmed numerically (Kandrup and Sideris (2001); Sideris and Kandrup (2002)), both for integrable and non-integrable underlying distributions by evolving orbits in “frozen-N body” samplings of the smooth mass distributions. However, Kandrup and Sideris also showed that when one follows the deviation of orbits evolved in the frozen-N and smooth potentials with identical initial conditions, and when the deviation amplitude is allowed to reach large fractions of the system size (macroscopic view), a continuum limit can be well defined in the sense that these macroscopic departures from the orbits of the smooth potentials (which can be themselves either regular or chaotic) grow as a power law with time, and that the characteristic divergence time is growing with the number of particles. These results, together with the claim of universality (halo profiles (Navarro et al. (1997)), their shape (Hansen and Moore (2006)), the mass functions (Zhang and Fall (1999), Richer et al. (1991)) that is widely used in the cosmology community, lead us to revisit the problem of the sensitivity of N-body simulations to slight changes in the initial conditions at fixed power spectrum in the cosmological context, and to numerically investigate the presence (or absence) of “chaotic” behaviour of different statistical quantities derived from N body simulations. Our focus will be on the transition between large scale linear dynamics and small scale stochastic properties (Strigari et al. (2007)). A possible concern in this context is the development of stochasticity induced by the ill-conditioning/non-linearity of the estimator of the chosen set of observables. Another concern lies in the specificities of the numerical code used. Finally, numerical noise induced by round-off errors should also be kept at bay, since they by themselves will lead to some level of stochasticity. Since the topic of this paper is not optimal estimation, no attempt will be made to argue that the set of estimators used here is superior or offers a better trade-off in bias versus variance. Similarly, a standard integrator (Springel et al. (2001)) is used to carry the simulations with a set of conservative parameters. Round-off errors are assumed to be irrelevant. Specifically, this paper will investigate what scale and mass is expected to play a role and will identify which quantities are found to be robust with respect to such exponential divergence; it will also find out if stochasticity breaks in as soon as non linearity occurs or if it is possible to identify two distinct time scales in the dynamics of large scale structures.

This paper is organized as follows: in Section 2 the method to characterize the statistical onset of stochasticity in numerical N-body simulations is presented. In Section 3 the corresponding Lyapunov exponents are computed for Eulerian quantities and (in Section 4) for Lagrangian ones. Section 5 discusses other issues and wraps up.

2 METHOD & SETTINGS

In this paper, we address the problem of the sensitivity of N-body simulations to initial conditions. To do so, we choose to study how slight changes in these initial conditions affect the evolution of the dispersion of a number of statistical quantities with time. This is achieved by generating several realizations of identical simulations where a small amount of random noise has been added to the initial realization. The generic procedure goes as follows:

(i) Generate a cosmological simulation using grafic (Bertschinger (1995)) and gadget (Springel et al. (2001)).
(ii) Start with the same noise file (i.e. the “phases”), but add a Gaussian white noise with RMS 1/30 of the previous white noise (except in Sections 3.2 and 4.4 where this amplitude is varied). This only affects the relative positions of clumps, not their spectral distribution (the expectation of the power spectrum remains unchanged).
(iii) Rerun the simulation with the new white noise;
(iv) Iterate the above two steps ~50 times;
(v) Compute a set of observables in each simulation;
(vi) Compute the RMS (or the relative RMS) of the distribution of observables for various expansion factors.
(vii) Fit the corresponding evolution of the log RMS vs the expansion factor.
(viii) Possibly find the scaling of its corresponding Lyapunov exponent (see below), with the smoothing scale associated with the observable (see Sec. 3.2) or the corresponding mass (see Sec. 4.3).

Let us define the “Lyapunov exponent”, \( \lambda_X \), as the rate of change of the logarithm of the fluctuation of the relevant quantity, \( X \), as a function of the scale factor, \( a \):

\[
\lambda_X \equiv \frac{d \ln \sigma_X}{da}.
\]

This stochasticity parameter is not strictly speaking a Lyapunov exponent since it corresponds neither to an asymptotic limit at large time, nor to an asymptotic limit at small fluctuation. It is closer in spirit to the short time Lyapunov exponent defined by Kandrup et al. (1997).

In practice two distinct sets of simulations are considered in this paper, one composed of 65 realisations of \( 128^3 \) particles each (\( S_1 \) hereafter) and the other of 27 realisations with \( 256^3 \) particles each (\( S_2 \) hereafter). The box size is \( 100 h^{-1}\text{Mpc} \), the cosmological standard ΛCDM model (\( \Omega_m = 0.3, \Omega_\Lambda = 0.7, H_0 = 70 \)), the softening parameter is \( 39.5 h^{-1}\text{kpc} \) and the expansion factor ranges from 0.05 up to 1 for \( S_1 \) and from 0.05 up to 0.4 for \( S_2 \). These two sets allow us to check the robustness of our finding with respect to resolution. Lyapunov exponents will also be expressed as characteristic timescales, \( \tau \), using the relationship between time and expansion factor in a CDM model (or equivalently in a ΛCDM model below \( a \leq 0.5 \)), \( \alpha \propto \tau^{2/3} \). Note that the resolution in mass of FOF halos containing more than 100 particles corresponds here to \( 4 \times 10^{12} M_\odot \) for the set \( S_1 \) and \( 5 \times 10^{11} M_\odot \) for the set \( S_2 \).

3 EULERIAN EXPONENTS

In this Section, we investigate the “global” chaos in the evolution of the Eulerian properties of the density field with respect to the expansion factor \( a \), as opposed to chaos in the Lagrangian properties of objects which are specific to the matter distribution in the universe (such as halos and filaments). This will be addressed in Section 4.

3.1 Chaos in density fluctuations

In order to study the density fluctuations, the density fields of \( S_1 \) and \( S_2 \) are sampled on a \( 64^3 \) grid using a simple NRG (nearest grid point) method allowing the computation of statistical quantities on the resulting grid, such as the average density or the density fluc-
3.2 Chaos transition scale

Transition to chaotic behaviour of the density field that started with linear evolution is fundamentally linked to the development of the nonlinearity. Since different scales enter nonlinear regime at different epochs, one expects that at a given time there exist a transition scale, $L_c$, below which variation of the density in pixels of the sampled field is clearly chaotic. Figure 2 represents the behaviour of the average value of $\lambda_P$ for different perturbation amplitudes $A$, as a function of the scale $L$. These measurements are derived by computing the average Lyapunov exponents in pixels on the sampled maps shown in figure 2 smoothed using a Gaussian kernel of width 128 pixels.

In fact, in denser regions, the evolution becomes non-linear earlier, which favors the development of chaos. But at later times, non-linearities have had time to develop at all considered overdensity levels, which explains the asymptotic merging of the curves. The exponential growth of the dispersions demonstrates that the evolution is chaotic as defined in the introduction and allows for the computation of Lyapunov exponents, $\lambda_P$, as the rate of change of the logarithm of the average relative density fluctuation as a function of the scale factor.

The fact that the non-linearity in the evolution increases chaos is illustrated by Figure 2 where maps of the average density (top panel) and the corresponding Lyapunov exponent $\lambda_P$ (bottom panel) are plotted. Each map represents the projection of a $10 h^{-1} \text{Mpc}$ slice from a sample of $S_2$ at $a = 0.35$. The correlation between the two maps confirms the dependence of chaos on overdensity level (see also the projections of different realisations of the same halos on Figure 3 where substructures are clearly different even though the shape of main halos remains mostly the same). These results must nonetheless be interpreted with care as the use of a finite sampling grid may bias the measurements. Indeed, considering higher density regions amounts to considering smaller scale regions, of order the size of the grid pixels ($\approx 1.5 h^{-1} \text{Mpc}^3$), which may affect the measured value of $\lambda_P$.

1 we also considered a $128^3$ grid and found no difference in the measured exponents.

2 note that the number of pixels above a given threshold is going to depend on redshift, but for the contrasts considered here, the error on the dispersion due to shot noise is always negligible, as we have at least 8000 particles above the highest threshold, at the highest redshift.

Figure 1. top: Evolution of the pixel-averaged PDF of the density values while sampling $S_1$ on a $64^3$ grid for different values of $a \in [0.1(\text{light}), \cdots, 0.4(\text{dark})]$ (only the pixels where $\rho/\bar{\rho} > 2$ were considered, where $\bar{\rho}$ is the cosmic mean density). As expected the full width half max (FWHM) of the distribution increases exponentially with the expansion factor reflecting the chaotic behaviour of the PDF of the density field; middle: temporal evolution of the dispersion in the sampled density field per unit of the mean density, for sub regions of $S_1$ corresponding to thresholds in overdensity $\rho/\bar{\rho}$ of 0.5, 1, 1.5 and 2 respectively as labelled. The asymptotic merging for different thresholds reflects the fact that at later times, regions of different overdensity levels are all in the nonlinear regime; bottom: same as middle frame but for $S_2$. 

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FWHM $L_h$ and considering only the overdense regions ($\rho/\bar{\rho} > 1$). Density is computed by making a histogram of particles in the grid using the NGP method, and by smoothing it with a Gaussian kernel afterwards. The measurements are performed at the present time, $a = 1$ in $S_1$ simulation and at $a = 0.4$ for $S_2$ set.

The plot demonstrates a rather sharp transition to chaotic behaviour at scales below the critical smoothing length $L_0 \approx 3.5h^{-1}$ Mpc with Lyapunov exponent increasing for ever smaller scales, whereas on larger scales the Lyapunov exponent is small and constant. This behaviour is indicative of the $\Lambda$CDM background cosmology of the standard model. Indeed, in the pure CDM cosmology with the critical density of the matter, the gravitational clustering would have continued to escalate to present time and one expect to see Lyapunov exponent falling smoothly to $L \approx 8h^{-1}$Mpc, the present-day nonlinear scale. In contrast, in $\Lambda$CDM cosmology the hierarchical clustering saturates when the dark energy begins to accelerate the expansion of the Universe. Numerical simulations show that in the standard $\Lambda$CDM model the clustering largely ceases by $z \approx 1$ (Hatton et al. 2003). The non-linear scale at this redshift is $L \approx 3.7h^{-1}$Mpc, which corresponds to the mass scale $M \approx 2 \times 10^{13} M_{\odot}$. The halos of smaller mass collapse en masse at earlier times passing by $z = 1$ through a period of hierarchical mergers with similar-mass halos as well as accretion that contributes to the formation of the chaotic features. Whereas the larger overdense patches, even the rare ones that turned around by $z \sim 1$ and will collapse by the present time, evolve in a quiescent environment of frozen hierarchy (van den Bosch 2002; Aubert and Pichon 2005). This argues for $L \approx 3.7h^{-1}$Mpc providing the fixed critical length between chaotic and regular regimes for all $z < 1$, which is in general agreement with our measurements.

4 LAGRANGIAN EXPONENTS

In the previous section, we studied the development of chaos in the density field of cosmological simulations. We measured the evolution of the variance of this density field on a grid (i.e. at peculiar Eulerian locations) and showed that chaos tends to be more pronounced in higher density regions as well as on smaller scales. Let us now focus instead on Lagrangian properties of peculiar objects with a physical significance such as dark matter halos or filaments.

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3 The nonlinear scale is usually defined with top-hat smoothing as $\sigma^2(R_{TH}) = 1$. The FWHM of the Gaussian smoothing filter $L_h$ that we use gives similar variance to the top-hat filter at $R_{TH} \approx 0.9L$. Our simulations are normalized to $\sigma(8h^{-1}\text{Mpc}) = 0.92$ which at $a = 1$ corresponds to nonlinear scale $R_{TH} \approx 7.2h^{-1}\text{Mpc}$, i.e. $L \approx 8h^{-1}\text{Mpc}$. 

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4.1 Inter skeleton distance

Filaments correspond to a central feature of the large scale distribution of matter: large void regions are surrounded by a filamentary web linking haloes together. Studying the properties of the filaments isn’t an easy thing and one first needs to find a way of extracting their location from a simulation. The skeleton gives a mathematical definition of the filaments as the locus where, starting from the filament type saddle points (i.e. those where only one eigenvalue of the Hessian is positive), one reaches a local maximum of the field by following the gradient. This is equivalent to solving the equation:

\[
\frac{dx}{dt} \equiv v = \nabla \rho,
\]

for \( x \), where \( \rho(x) \) is the density field, \( \nabla \rho \) its gradient, and \( x \) the position. Although apparently simple, solving this equation is quite difficult which is why a local approximation was introduced in (Sousbie et al. 2006, 2007): the local skeleton. One can show that, up to a second order approximation, solving Equation 2 is equivalent to finding the points in the field where the gradient is an eigenvector of the Hessian matrix together with a constraint on the sign of its eigenvalues. This approach leads to a system of two differential equations, solved by finding the intersection of two isodensity surfaces of some function of the density field and its first and second derivatives. This procedure is very robust and allows for a fair detection of the dark matter filaments. Figure 3 displays the skeleton of different realisations of \( S_2 \) at \( a(t) = 0.1 \) (top) and \( a(t) = 0.4 \) (bottom). Note that the dispersion of the skeleton location has increased with the scale factor. Using the method described in (Sousbie et al. 2006, 2007), the local skeleton provides a list of small segments. In order to measure the distance between two skeletons, for each segment, the distance to the closest segment in the other skeleton is computed leading to the PDF of this distribution. The mean distance between the two skeletons is set to the position of the first mode of their inter-distance PDF (See also Caucci et al. 2008). We then define a mean inter-skeleton distance among all the realisations within a set as the arithmetic average of their pairwise distances. This means that the normalized inter-skeleton distance, \( \langle D \rangle / L_0 \), is our measure of the dispersion in the skeleton location. It is a Lagrangian property since it follows the flow. Its evolution as a function of the scale factor is plotted on Figure 3 for different smoothing lengths \( L_0 \), for \( S_2 \) (top) and \( S_1 \) (bottom). The smoothing operation is achieved, as previously, by convolving the density field with a Gaussian kernel of FWHM \( L_0 \), ranging from \( L_0 = 1.2 h^{-1} \text{Mpc} \) (3 pixels) up to \( L_0 = 3.5 h^{-1} \text{Mpc} \) (9 pixels). It is clear that whatever the smoothing scale or the resolution used, the evolution of the dispersion is linear with the scale factor. A shift in the skeleton of the initial conditions will evolve linearly with time and not exponentially: the skeleton at present time won’t be affected very much. There is no chaotic drift of the position of the skeleton and thus no chaos in the evolution of the cosmic web. Note nonetheless that the smaller the smoothing length, the stronger the increase of \( \langle D \rangle / L_0 \). This implies that smaller scales are more sensitive to initial conditions, which is confirmed by the fact that \( \langle D \rangle / L_0 \) is larger for lower values of \( L_0 \), whatever the value of \( a \).

4.2 Positions of halos

Turning to stochasticity on smaller scales in a Lagrangian framework (i.e. ignoring the absolute shift in position relative to a fixed unit of length), we use a mean of the pairwise distances. This means that the normal-

\[
\text{distance among all the realisations within a set as the arithmetic average of their pairwise distances.}
\]

\[
\langle D \rangle / L_0 \text{ is our measure of the dispersion in the skeleton location. It is a Lagrangian property since it follows the flow. Its evolution as a function of the scale factor is plotted on Figure 3 for different smoothing lengths } L_0, \text{ for } S_2 \text{ (top) and } S_1 \text{ (bottom). The smoothing operation is achieved, as previously, by convolving the density field with a Gaussian kernel of FWHM } L_0, \text{ ranging from } L_0 = 1.2 h^{-1} \text{Mpc} \text{ (3 pixels) up to } L_0 = 3.5 h^{-1} \text{Mpc} \text{ (9 pixels). It is clear that whatever the smoothing scale or the resolution used, the evolution of the dispersion is linear with the scale factor. A shift in the skeleton of the initial conditions will evolve linearly with time and not exponentially: the skeleton at present time won’t be affected very much. There is no chaotic drift of the position of the skeleton and thus no chaos in the evolution of the cosmic web. Note nonetheless that the smaller the smoothing length, the stronger the increase of } \langle D \rangle / L_0. \text{ This implies that smaller scales are more sensitive to initial conditions, which is confirmed by the fact that } \langle D \rangle / L_0 \text{ is larger for lower values of } L_0, \text{ whatever the value of } a.\]

\[
\text{Figure 4. The local skeletons of the different realisations of } S_2, \text{ computed at } a(t) = 0.1 \text{ (top) and } a(t) = 0.4 \text{ (bottom) and for a smoothing length } L_0 = 1.2 h^{-1} \text{Mpc. Each figure corresponds to the projection of a } 10 h^{-1} \text{Mpc thick slice. Each color represents a different realisation of the simulation, the color coding is not consistant between the top and the bottom panels. The dispersion in the position of the skeletons appears to have grown from } a(t) = 0.1 \text{ to } a(t) = 0.4.\]

\[
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\]
of $S$ not change from one simulation to another, but the substructures are quite different; this is confirmed via automated substructure identification using ADAPTAHOP. For different runs, haloes are first identified using the FOF algorithm (Davis et al. 1985; Suginoara and Suto 1992) with a percolation length of $0.25h^{-1}\text{Mpc}$ for $S_1$ and $0.5h^{-1}\text{Mpc}$ for $S_2$ corresponding to $0.2\times$ mean interparticle distance. In order to tag different FOF haloes in different realisations as counterparts, all particles of a given halo are matched in another realisation using their initial index (Figure 6). The halo of the other simulation containing most of these particles is tagged as its counterpart. The procedure is carried out over all pairs of simulations, allowing the measurement of the variation in the halo properties like their spins, their positions, their velocity dispersion tensors or their masses.

As shown on Figure 6 the haloes locations seem relatively insensitive to small changes in the initial conditions. The evolution of the mean distance between a halo in a given simulation and the same halo in another realisation is linear, as for the skeleton, which confirms the first impressions: no chaos is observed at linear scales and so $\lambda_Q$, the Lyapunov exponent of the inter halo distance, is null. But the most interesting results involve the substructures. The halo pictured in Figure 6 is a good example of the generic behaviour. The number of substructures changes from one realisation to another (here, 1 or 2 substructure(s)) and their positions also differ. These results are confirmed by an automated detection of the substructures using ADAPTAHOP (Aubert et al. 2004). Both the locations and the number of substructures are possibly subject to chaos, but the lack of a cross identification procedure makes it difficult to quantify it and is somewhat beyond the resolution of these sets of simulations (Section 4.3 addresses this problem for the FOF halos). This trend confirms quantitatively the findings of Section 4.4.4 from the point of view of Eulerian estimators which are sensitive to the detailed extension of the distribution of matter within halos: denser regions were found to be chaotic, and will be addressed in more details in Section 4.4 in terms of halo density and velocity moments.

4.3 Connexity and mass of clusters

The connexity of haloes can be defined as follows: considering the $p^{th}$ halo, $H_p^r$, in the $r^{th}$ realisation, its particles are spanned amongst $n$ haloes in the $r^{th}$ realisation, and a fraction $f_{pk}^{r'}$ ($k \in 1, \ldots, n$) of them belong to a halo $k$ among $n$ in the $r^{th}$ realisation. Hence, its relative connexity $C_{pr}^{r'}$ can be defined as:

$$C_{pr}^{r'} = \sum_{i=1}^{n} \left( \prod_{j=1}^{r-1} \sum_{k=j}^{n} f_{pk}^{r'} \right),$$

where by construction $C_{pr}^{r'}$ is equal to one if both haloes are identical in realisations $r$ and $r'$; $C_{pr}^{r'}$ is equal to $n$ if the halo $p$ splits into $n$ haloes with equal fractions $f_{pk}^{r'} = 1/n$ in realisation $r'$, while preserving continuity when the values of $f_{pk}^{r'}$ differ, see Appendix A. The mean connexity, $C$, is obtained by averaging over all haloes containing more than 100 particles in every possible combinations of realisations and is a measure of the dispersion of the particles.

As shown on Figure 7, $C$ increases with the scale factor, ranging from 1.17 (i.e., statistically, 90% of the particles belong to a unique halo in other realisations) to 1.37 (85%) from $\alpha = 0.2$ up to $\alpha = 1.0$. The connexity clearly does not vary exponentially with the scale factor: there are statistically no halo fission during evolution (thanks to the efficiency of dynamical friction). Moreover, two haloes marginally linked by FOF would almost always end up
The influence of chaos on the spin of haloes is estimated by computing the cosine of the angle, \( \cos(\theta_{pq}) \), between the spins \( J_p \) and \( J_q \) of corresponding haloes in two different realisations \( p \) and \( q \):

\[
\cos(\theta_{pq}) = \frac{J_p \cdot J_q}{\|J_p\| \|J_q\|}.
\]

For every bin of mass, a measure of the dispersion, \( \sigma \), of the orientation is given by the average angle:

\[
\sigma = \arccos \left( \frac{1}{N_c} \sum_{i=1}^{N_c} \cos \theta_{pq} \right),
\]

where the sum is over all the \( N_c \) possible pairwise combinations of realisations. Note that only bins of masses containing more than 30 haloes have been retained.

Figure 7 displays the exponential growth of this dispersion with time, and shows that the precise value of its associated Lyapunov exponent \( \lambda_\sigma \) depends on the selected bin of mass. It also shows that the exponent does not seem to be sensitive to shot noise, as its value is left unchanged when resolution is increased between \( S_1 \) and \( S_2 \).

Also, as it is seen in Fig. 6, the detailed distribution of satellites within a given cluster varies from one realisation to another; the angular momentum orientation (in contrast to say, its modulus or the halo mass) is quite sensitive to the outer region of the distribution. Recall that the spin parameter (i.e. the \( \Lambda = J/(\sqrt{2} MV_{200} R_{200}) \) (Bullock et al. (2001), Aubert et al. (2004)) of a halo displays no chaotic behaviour. It stays quite constant from a simulation to another and the evolution of its dispersion is not exponential.

The measured Lyapunov exponent ranges from 0 to 0.3. The value of the mean dispersion of the orientation of the spin for heavier haloes is about 35 degrees (= \( \exp(3.55) \)). Globally this suggests that the orientation of the spin varies with the tidal field, which in turn depends on the relative position of structures within the environment of the halo.

For lighter haloes, the measured value of \( \lambda_\sigma \) is higher than for heavier ones (Figure 8) which may be partly explained by the fact that a slight change in a few clumps within the haloes has a larger influence on its spin when they represent a significant fraction of it. Faltenbacher & al. (Faltenbacher et al. (2003)) showed that if the lightest halo has a mass less than 10% of the mass of the larger halo, the orientation of the resulting post merging halo will remain statistically the same. In contrast, if its mass is greater than 20% of the more massive halo, the final orientation of the merged halo depends on the speed vector of the two progenitors. Our results corroborate well their finding since the lightest haloes that are formed by merging of two substructures of comparable masses have chaotic spins (substructures being chaotic, see Section 3.2) and Figure 6, while heavier ones have spin that are relatively stable with time (they only merge with much smaller haloes). It emerges from these measurements that there is a critical mass, \( M_c \), above which chaotic behaviour disappears. Haloes heavier than this mass are too heavy to feel the influence of incoming clumps and their spins are clearly defined. They are therefore not subject to chaos and their Lyapunov exponents are null at the one sigma level, in contrast to lighter ones whose spin are sensitive to the initial conditions and whose Lyapunov exponents are positive.

As for the critical smoothing length (see Section 3), we can study the evolution of this critical mass as a function of the amplitude, \( A \), of the perturbations. Figure 8 shows this evolution. A good fit of this transition mass is given by \( M_c = 2 \times 10^{13} M_\odot A^{1.15} \). The higher the amplitude of the perturbations, the higher the required time for haloes to have a spin clearly defined. Consequently, the critical mass increases with the perturbation amplitude, and haloes that can be considered stable are heavier. Note that it means that
4.5 Orientation of the velocity dispersion tensor

The orientation of the velocity dispersion tensor is also a quantity of interest from the point of view of stochasticity since it is related to the shape of the halo via the Virial theorem. The corresponding estimator involves computing the orientation of the eigenvector $V$ associated to the largest eigenvalue of the velocity dispersion tensor. As for the orientation of the spin, (Sec. 3.1) the angle $\theta_{pq}$ between the eigenvector $V_p$ of the halo in simulation $p$ and its corresponding eigenvector $V_q$ in a simulation $q$ is computed as:

$$\cos(\theta_{pq}) = \frac{V_p \cdot V_q}{\|V_p\| \|V_q\|}. \tag{6}$$

For every bin of mass, a measure of the dispersion, $\sigma$, is also given by Equation 5. Once again, only bins of masses containing more than 30 haloes were considered. As shown in Figure 10, this estimate is consistent with the exponents of the orientation of the spin: only the lightest masses are sensitive to the initial conditions, while the dispersion of the orientation for the heavier masses is constant (about 40 degrees). The measured Lyapunov exponent, $\lambda_T$, ranges from 0 up to 0.65. These results corroborate well those for the spin axis, given that its orientation follows the third eigenvector of the dispersion matrix (i.e. the axis along which dispersion is the smallest). Faltenbacher et al. (2003) showed that the orientation of the principal axis of the halo is correlated with the vector linking the two mergers (i.e. their relative positions) particularly in the case where one of the mergers has a mass smaller than 10% of the second one. The chaos found at small scales (substructures scale) is once again responsible for the chaos in these geometrical properties of haloes since changing the initial conditions amounts to changing the relative positions of the substructures (see sec. 3.1) and thus to changing the orientation of the resulting halo’s dispersion tensor. As for the spin, there is evidence of a critical mass above which chaotic evolution disappears: more massive haloes only merge with lighter ones that do not affect their global properties.

5 CONCLUSION AND DISCUSSION

Let us first emphasize here again that the term chaos is used in this paper in the loose sense, as the age of the universe does not allow for many e-foldings on larger scales. Table 1 summarizes the different Lyapunov exponents computed in this paper. As shown in section 3.1 (Figure 3), chaos appears below a critical scale which corresponds roughly to cluster scales. The higher the density, the more chaotic is the corresponding region. We also found that both Lagrangian and Eulerian measurements are consistent: super clusters and filaments, whose dynamics is globally linear (large scale structures), are not stochastic: a shift in the initial conditions will increase linearly with the scale factor. By contrast, the distributions of substructures within clusters, whose characteristic size is smaller than $\sim 3.5h^{-1}$ Mpc, are governed by non-linear dynamics and may undergo a stochastic evolution for some observables.

Nevertheless, this chaos at substructures scale does not occur for all physical characteristics of the cluster’s halo. A main fraction of particles remains in the same halo from one realisation to another, while a difference arises (in part) from the delay in merging times of the substructures. These timing effects are however averaged out yielding, to first order, a constant halo mass. It follows that the mass function derived from a simulation is quite consistent from one realisation to another. Similarly, the dispersion of the amplitude of the total spin of haloes does not increase exponentially with time.

The mass of a given halo is an integrated quantity which does
not trace which specific particle entered the FOF halo; similarly, the spin parameter is also an adiabatic invariant, and the trace of the dispersion tensor will relax rapidly to its virial expectation (which is mass dependent) in a few short dynamical times; in contrast, the spin orientation or the orientation of the dispersion tensor will depend precisely on the orientation of velocities of the entering particles and has no direct relation to the mass of the halo; it also reflects the initial environment of the proto halo. For instance it has been shown in Sousbie et al. (2007), Aubert et al. (2004) that the halos preferentially anti-align their spin with the axis of the filament in which they are embedded, while we have shown in Section 4.1 that the filament’s locus was not stochastic.

It is possible to recast these interpretations in the context of the peak-patch (Bond and Myers (1996)) description of haloes. In this framework, massive haloes correspond to large quasi spherical patches around density peaks, which non-linear evolution will decouple from their neighbouring large patches thanks to the cosmic acceleration below \( z \sim 1 \). Conversely, small haloes correspond to small typically aspherical peak-patches, and will acquire tidal torques early on which depend specifically on the detailed white noise realisation (which fixes the shape of the peak-patches). In this framework, massive haloes correspond to large quasi spherical peak-patches, and will acquire tidal torques early on which depend specifically on the detailed white noise realisation (which fixes the shape of the peak-patches). In the tidal torque theory, the mass and the spin parameter are essentially integral functions over the volume of these patches, hence will not depend on the initial perturbations, whereas the spin orientation itself is sensitive to these perturbations, at least at the lower end of the mass spectrum. This is consistent with the low scatter relationship between the spin parameter and the mass (Aubert et al. (2004)).

Thanks to angular momentum leverage, the orientation of haloes is itself affected by stochasticity mostly at small scales, a result which seems insensitive to shot noise as the lyapunov exponents are consistent between sets \( S_1 \) and \( S_2 \). In fact, as long as the haloes merging together have similar sizes (masses), the orientation of both spin and velocity dispersion tensors is determined by the relative positions and velocities of the two mergers, whose dynamics is non-linear and whose characteristic size is below the critical scale (Figure 3). These results seem robust with respect to resolution.

When the halo is formed and well-isolated by cosmic acceleration, it merges only with satellites/substructures whose masses represent a small fraction of the host’s mass. Consequently, their orientations are globally preserved after merging, and thus, the chaotic behaviour stops and the dispersion in the orientation remains at the same level (i.e. the resulting average angle is unchanged). Hence, a critical mass can be defined as the mass above which this chaotic behaviour of the orientation stops. The measured value of this critical mass, \( M_c = 2 \times 10^{13} M_\odot A^{0.15} \), is just below the scale of non-linearity at \( z \sim 1 \) and shows weak dependence on the amplitude of the added perturbative noise: \( M_c = 2 \times 10^{13} M_\odot A^{0.15} \). Although some slow increase of \( M_c \) with \( A \) is expected since adding power to inhomogeneities shifts the nonlinear scale to higher masses, the details of the dependence require further investigation.

This paper has concentrated on a realistic \( \Lambda \)CDM cosmology: it would also be interesting to rerun this investigation on scale-free power spectra to confirm that the dark energy is indeed responsible for the saturation of \( M_c \). A natural extension of this work, clearly beyond its current scope, would also involve computing lyapunov exponents for the properties of substructures within halos (see for instance Valluri et al. (2007)), and parameters corresponding to the inner structure of halos, such as NFW concentration parameter, the phase space density \( Q = \rho_0 / \sigma^3 \) (Petranic et al. (2006)), the Gini index or the asymmetry (Conselice et al. (2007)) within the FOF.

In closing, the answer to our riddle is that chaos and non-linearities are very strongly linked, and both occur at small scales (substructures scales) though some non linear halo parameters (spin, mass etc...) do not seem to be subject to chaos. While the large scale structures in a simulation (filaments and haloes) are quite robust both in their locus and properties, the distribution of substructures is more sensitive to initial conditions since their numbers and positions vary when initial conditions vary. This in turn may prove to be a concern when generating zoomed resimulations.
Acknowledgments

We thank the anonymous referee for helpful suggestions, D. Aubert, S. Colombi, and R. Teyssier for fruitful comments during the course of this work, and D. Munro for freely distributing his Yorick programming language and opengl interface (available at http://yorick.sourceforge.net/). This work was carried within the framework of the Horizon project, www.projet-horizon.fr.

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APPENDIX A: CONNEXITY

Let us consider a halo, $H$, split into $n$ parts, $P^n_i$, $i \leq n$, with a fraction, $f_i$, of its particles in each of them, the indices $i$ being sorted following the decreasing values of $f_i$ ($f_i > f_j$ if $i < j$). A measure $C_n$ of the connexity of $H$ should indicate the number of clumps into which it was split, this number does not necessarily have to be an integer, depending on the fraction of the mass of $H$ that went into each $P^n_i$. For instance, if $n = 2$, we want to obtain $C_2 = 2$ when $f_1 = f_2 = 1/2$ and $C_2 \rightarrow 1$ when $f_1 \rightarrow 1$ and $f_2 = 1 - f_1 \rightarrow 0$; as the indices are sorted, $0 < f_2 < 1/2$. So, in this case, we could write the connexity of $H$ as:

$$C_2 = 1 + 2f_2.$$  \hspace{1cm} (A1)

Now, considering that $H$ was split into 3 parts $P^n_i$, then $0 < (f_2 + f_3) < 2/3$ and $0 < f_3 < 1/3$. So $C_3' = 2 + 3f_3 \rightarrow 2$ if $f_3 \rightarrow 0$ and $C_3' \rightarrow 3$ when $f_3 \rightarrow 1/3$. It follows that $C_3'' = (f_2 + f_3)C_3' \rightarrow 2$ when $f_2 \rightarrow 1/3$ (which implies that $f_3 \rightarrow 1/3$) and that $C_3'' \rightarrow 0$ when $f_2 \rightarrow 0$ (which implies that $f_3 \rightarrow 0$ also). So

$$C_3 = 1 + (f_2 + f_3)(2 + 3f_3)$$  \hspace{1cm} (A2)

has the right properties to represent the connexity of a halo split into 3 parts. Hence, by generalizing recursively this formula, we obtain:

$$C_n = 1 + (f_2 + \cdots + f_n)[2 + (f_3 + \cdots + f_n)[\cdots [n - 2 + (f_{n-1} + f_n)[(n - 1) + nf_n]]]],$$  \hspace{1cm} (A3)

which can be developed as:

$$C_n = 1 + 2(f_2 + \cdots + f_n) + 3(f_2 + \cdots + f_n)(f_3 + \cdots + f_n) +$$

$$\cdots + n(f_2 + \cdots + f_n)(f_3 + \cdots + f_n)\cdots (f_n),$$  \hspace{1cm} (A5)

$$= \sum_{i=1}^{n} i \left( \prod_{j=1}^{n} \sum_{k=j}^{n} f_k \right),$$  \hspace{1cm} (A6)

which corresponds to Eq. (3). Note that by construction, the braket in Eq. (A7) is smaller than $1/i$, so that $C_n$ is always smaller or equal to $n$. 