

Halo Definition and Environmental Effects

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ABSTRACT

Recent work has shown the importance of environment to the properties of dark matter halos. This brings conflict to standard implementations of the halo model and excursion set theory which assume that the properties of a population within the halo is determined by the mass of the halo alone. We seek to find a definition of the size of a halo that allows us to minimize the impact of assembly bias on halo model calculations. We analyze the dependence on environment of our properties using the method of marked correlation functions for several different halo definitions, utilizing the Diemer & Kravtsov (2015) simulations. We find that environmental dependencies are dramatically different as we vary the definition of the halo radius in terms of the overdensity Δ . At large length scales, we find that the majority of assembly bias is removed through suitable redefinition of Δ . We are able to determine that the majority of the reduction in assembly bias is related to the elimination of host halos that would cease to be hosts in catalogs at lower values of Δ . Further, we analyze how different mass cuts affect this methodology. We note that unresolved halos leads to assembly bias being missed and that the most massive halos seem to exhibit minimal assembly bias. We further note that the choice of halo definition can induce assembly bias and consider how this may be interpreted in the context of previous results in the literature.

Key words: cosmology: dark matter – cosmology: large-scale structure of Universe – galaxies: formation – galaxies: halos – methods: numerical

[ARZ: We should add Andrew Hearin and Frank van den Bosch. Look into formatting the title correctly. Once I'm happy with the quality of the writing, we'll also need to send this to Benedikt Diemer and Andrey Kravtsov and offer them authorship.]

1 INTRODUCTION

In the concordance cosmology, galaxies and clusters form within merging dark matter halos (e.g., White & Rees 1978; Blumenthal et al. 1984). Numerical simulations have provided a solid understanding of the abundances, properties, and clustering of dark matter halos in the standard cosmological model. Accordingly, it is possible to compute the clustering statistics of galaxies given a model for the (probabilistic) relationship between galaxies and dark matter halos. Such halo occupation models have been used to interpret large-scale structure measurements and constrain models of galaxy evolution (e.g., ?????????????????????). To date, the vast majority of halo occupation models rely on a key assumption, namely that the probability of a halo hosting a certain number of galaxies of a particular type depends only upon

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halo mass. It is now well known that the clustering strength of halos depends upon properties such as halo formation time (Gao et al. 2005; Wechsler et al. 2006; ?, Croton et al. 2007; Zentner 2007; ?), concentration (Wechsler et al. 2006; Mao et al. 2015), and other halo properties **[ARZ: NEED SOME ADDITIONAL CITATIONS HERE.]** If galaxy properties depend upon these halo properties, a phenomenon known as galaxy *assembly bias*, then standard halo occupation modeling will fail (e.g., ?) and more complex models (e.g., ?) are necessary.

In this work, we explore the possibility of simplifying halo occupation modeling, at least for some applications, by altering the definition of a halo. Traditional definitions of halo boundaries are not particularly well motivated, so it seems natural to explore alternative definitions of halo size. Halo definitions have become a matter of convention that vary considerably in the literature. Many authors define halos using a friends-of-friend (FoF) algorithm applied to the particle distribution that designates halo boundaries as regions of roughly constant density. More often, authors define halos by spherical overdensity (SO) regions within which the mean density exceeds a particular threshold. The threshold used varies significantly. One difference is that some studies define the density threshold with respect to the mean background density, while others define the density threshold with respect to the critical density of the universe. Even when the same reference density is used, the threshold level varies. With respect to the mean density of the universe, commonly used thresholds are 178, 180, 200, and ~ 340 times the mean background density, the last of these being the “virial” overdensity in a concordance Λ CDM universe. Significantly higher values of the overdensity parameter are often used in studies of X-ray emission from cluster-sized halos. These definitions exclude particles that are gravitationally bound to the halo (e.g., Kazantzidis et al. 2006).

In this paper we explore the possibility of using halo definition to work for the convenience of the practitioner. We study the strength of various halo assembly bias signals as a function of halo definition. This is motivated, in large part, by recent literature that suggests that a large portion of the assembly bias signals are driven by halos in the relatively density environments surrounding larger halos. **[ARZ: Need citations here to work by Wang (2007), the recent work on splash-back (backsplash) and flyby halos, and the recent work on the splashback radius.]** These halos have anomalous properties (e.g., formation times, concentrations, etc.) compared to field halos in part because of their interactions with their larger, neighbor halos. It seems interesting to ask whether or not defining halos to contain many of these anomalous neighbor halos, thus using the halo concept to draw a more meaningful boundary around regions within which highly nonlinear effects are important, can mitigate halo assembly bias. This is the aim of our paper.

In the present paper, we restrict ourselves to halos that are defined by spherical regions and we study the strength of halo assembly bias as a function of the density threshold used to delineate these spherical halos. We find

that **[ARZ: two or three sentence, relatively non-technical summary of our results should go here.]**

In § 2 of this paper, we discuss the cosmological simulations that we analyze and how halos are identified. In § 3, we consider the properties of interest within our halo simulation and their standard definitions. In § 4, we discuss the statistics that we use to measure assembly bias and the removal of known mass scaling from our halo properties. In § 5, we present our results and consider how the change of halo definition impacts measures of assembly bias. In § 7, we discuss the significance of reducing environmental effects through a redefinition of halo environments and discuss possible applications of our methodology. We also consider the nature of assembly bias as a function of halo definition.

2 SIMULATIONS AND HALO IDENTIFICATION

In order to study the effects of halo redefinition, we use three cosmological N -body simulations of structure formation. The Diemer & Kravtsov (2015) simulations each utilize a Planck best-fit cosmology with $\Omega_M = 0.32$, $\Omega_\Lambda = 0.68$, and $h_0 = 0.67$. We use three simulation boxes with comoving sizes of 125, 250, and 500 $h^{-1}\text{Mpc}$ respectively. The particle masses are 1.6×10^8 , 1.3×10^9 , and $1.0 \times 10^{10} h^{-1}M_\odot$ respectively, implying a total of 1024^3 particles in each simulation. Furthermore, the three simulations have different force softening scales of 2.4, 5.8, and 14 $h^{-1}\text{kpc}$. We refer to each simulation as L0125, L0250, or L0500 for the remainder of the paper. This set of simulations allows us to probe the resolution effects inherent in halo finding (due to the varying resolutions of the simulations) and to probe the mass dependence of halo clustering over a wider range of halo masses than would be possible with only one simulation from the set. In particular, L0125, with its higher resolution, contains the least massive resolved halos, while L0500 has the most robust statistics for the most massive halos as a result of the larger simulation volume.

To identify halos, we use the ROCKSTAR halo finder, which works on the phase space algorithm described in Behroozi, Wechsler & Wu (2013). In short, ROCKSTAR determines initial groupings of particles using a Friends-of-Friends algorithm in phase space before applying the spherical overdensity halo definition in order to determine halo properties of interest. Unbound particles are removed prior to the calculation of halo mass and other halo properties. Our method of halo redefinition is to change how halo size is calculated as part of the ROCKSTAR pipeline. A halo is given a radius, r_Δ , determined by

$$\bar{\rho}(r_\Delta) = \Delta \rho_m, \quad (1)$$

where the mean density within a spherical volume of radius r is $\bar{\rho}(r)$, Δ is the overdensity parameter, and ρ_m is the mean background mass density of the simulation. The resulting halo size calculation can have a large variation depending on the choice of Δ . The number chosen for Δ varies throughout the literature from $\Delta \approx 178$ to $\Delta \approx 340$ or higher for X-ray studies of clusters. We choose

to vary the size of a halo by treating the overdensity parameter as tunable, expanding the range from $\Delta = 340$ to $\Delta = 10^1$. While this range is greater than the differences between most the common overdensity choices, this broad range enables a fairly extensive exploration of environmental effects on halo properties as a function of halo definition.

3 HALO PROPERTIES

We investigate the clustering of halos as a function of a number of halo properties that have been shown to influence halo clustering at fixed halo mass. We explore two definitions of halo concentration. The first stems from a fit of the spherically-averaged halo density profile, $\rho(r)$, to a Navarro et al. (1997, ; hereafter, NFW) density profile,

$$\rho(r) = \frac{\rho_0}{\frac{r}{r_s}(1 + \frac{r}{r_s})^2}, \quad (2)$$

where the density scale, ρ_0 , and the scale radius, r_s , are parameters that are fit to the density profile of each halo. The standard definition of halo concentration is then

$$c_{\text{NFW}} = \frac{r_\Delta}{r_s}, \quad (3)$$

where r_Δ is the radius of the halo given an overdensity parameter, Δ , defining the halo and r_s is the inferred halo scale radius. We use the NFW concentrations computed by the ROCKSTAR halo finder.

The NFW concentration defined in the previous paragraph has the shortcoming that it depends upon a parametric description of dark matter halos, so we study the clustering dependence of halos as a function of a non-parametric description of halo concentration. In particular, we define the velocity concentration as

$$c_V = \frac{V_{\text{max}}}{V_\Delta}, \quad (4)$$

where V_{max} is the maximum circular velocity achieved within the halo and V_Δ is the circular velocity at the halo radius, r_Δ . All halos of the same mass have the same value of V_Δ ; however, they exhibit a variety of V_{max} values depending upon the degree to which their mass is concentrated toward the halo center. The quantity c_V is a non-parametric measure of halo concentration and can be measured from simulation snapshots without any need for fitting halo density profiles. Consequently, c_V is robust to halo density profile parameterizations and halo profile fitting procedures.

Halo concentrations are interesting to investigate for these purposes for a number of reasons. First of all, the environment dependence of halo concentrations is of interest in the modeling of galaxy clustering and gravitational lensing statistics. In the case of galaxy clustering, the relevance is indirect, because galaxies will not necessarily trace the mass density of their host halos. In the case of gravitational lensing, the mass distribution is the

primary quantity of interest and halo concentrations are directly related to predictions for lensing statistics. Secondly, concentrations can be measured from individual simulation snapshots easily and halo concentrations are known to be strongly correlated with the formation histories of dark matter halos with earlier forming halos having higher concentrations at fixed halo mass (Wechsler et al. 2002, 2006) [ARZ: Cite papers here like Wechsler et al. 2002. I think there are at least two Zhao et al. papers from the last decade relevant here. I don't remember the years, but one of them is at least 8 years old. I think another might be from around 2011.]. As such, exploring the concentration dependence of halo clustering may yield insight into the age dependence of halo clustering without the need for constructing merger trees. This is particularly important in the present study in which the halo finding is performed repeatedly for many different values of Δ . Constructing a self-consistent merger tree from which halo formation history can be studied requires halo finding at all simulation snapshots for each new value of Δ , which is a computationally prohibitive task. In the present paper, we limit our study to halo properties that can be measured from a single simulation snapshot. However, halo formation histories correlate with halo concentrations with significant scatter and this correlation may depend upon environment, so the reader should be wary of drawing conclusions about the environmental dependence of halo formation by extrapolating our results on halo concentration. We will explore measures of halo age directly in a forthcoming follow-up study dedicated to halo formation histories.

Figure 1 shows the mean $c_{\text{NFW}}-M_\Delta$ relation for halos defined with $\Delta = 200$ in L0125, L0250, and L0500. For each simulation, we consider halos only above a minimum mass threshold to ensure that concentration measurements are not compromised by resolution effects. The minimum mass thresholds are shown as the vertical lines in Fig. 1 and listed in Table 1. Similar to Fig. 1, Figure 2 shows the relation between the velocity-ratio concentration, c_V , and halo mass. In the interest of completeness, Figure 3 shows the relationship between c_{NFW} and c_V on a halo-by-halo basis. As is evident, the two concentration proxies are strongly correlated and exhibit a $\sim 10\%$ scatter indicating that c_{NFW} and c_V indeed encode similar information about each halo [ARZ: I eyeballed this from your figure, check the scatter in this relationship please to confirm that it is roughly 10%.].

In addition to halo concentrations, we examine halo clustering as a function of a variety of other halo properties. We study halo clustering as a function of halo shape, s , quantified by the ratio of the halo minor and major axis lengths,

$$s = \frac{c}{a}, \quad (5)$$

where a is the major axis length and c is the minor axis length. The mean relations for halo shapes as a function of halo mass for $\Delta = 200$ are shown in Figure 4 along with the mass thresholds used to ensure that our results are not compromised by resolution.

We study halo clustering as a function of halo an-

¹ It is worth noting that the ROCKSTAR linking length parameter must be adjusted as Δ is adjusted in order to ensure that SO halos contain all relevant particles.

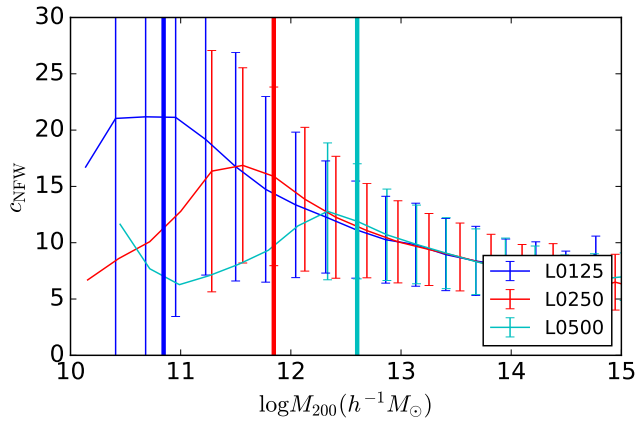


Figure 1. The relationship between the NFW concentration and halo mass for each of our simulations with $\Delta = 200$. In order of increasing simulation volume, the blue line corresponds to the concentration-mass relation from simulation L0125, the red line corresponds to L0250, and the cyan line corresponds to L0500. The error bars designate the dispersion on NFW concentration at fixed halo mass. Each simulation is subject to resolution limitations at different halo masses. We show with vertical lines the minimum M_{200} mass thresholds that we adopt in our analyses using the same color code as the concentration-mass relations. Note the deviation from a monotonic trend as a result of resolution effects. [ARZ: This figure is still way too busy to view. First, the legend covers some of the data. Second there are too many error bars. Try using error bars only for the L0250 simulation and just state in the caption that the dispersion in each simulation is similar in the regime in which the results are not compromised by resolution. Use a different line type for each line so that this can make sense in black and white if necessary. Lastly, is there some way to make the thresholds stand out even more? Perhaps they will stand out more when there is only one set of error bars? You will have to play with this. As far as the caption, this caption is way too brief. It should describe each line and it should state what the error bars are showing. Please make sure this and subsequent captions are sufficiently informative such that the figures are self contained. I will take a stab at this caption so that you can see what I mean.]

gular momentum quantified by the spin parameter, λ , as introduced by (Peebles 1969),

$$\lambda = \frac{J\sqrt{|E|}}{GM_{\Delta}^{2.5}} \quad (6)$$

where J is the halo angular momentum, E is the total energy of the halo, and M_{Δ} is the mass at the halo radius, r_{Δ} . The mean relations for halo spin as a function of halo mass for $\Delta = 200$ are shown in Figure 5 along with the mass thresholds enforced to ensure that our results are not compromised by resolution. These thresholds are summarized in Table 1, where we also show the threshold masses at various values of the overdensity parameter, Δ .

[ARZ: I really don't understand what this last paragraph is saying.] In practice, the mean relations between the various halo properties and the mass thresholds for our analyses must be determined separately for each combination of simulation, halo property (e.g., c_{NFW}

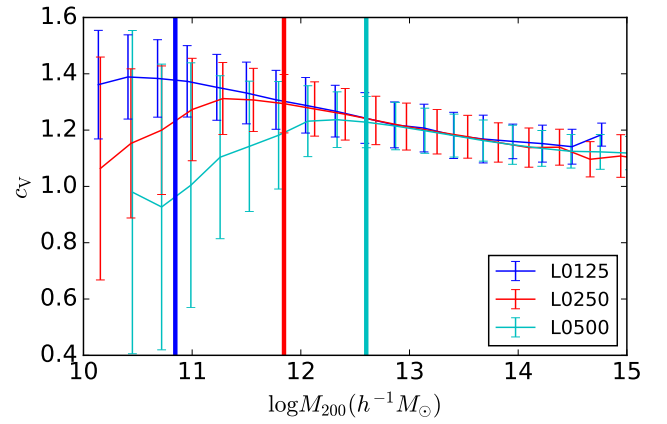


Figure 2. An example of the relationship between the velocity ratio concentration and halo mass for each of our simulations for $\Delta = 200$. The chosen lower limit on halo mass for our sample is marked as a blue, red, or cyan line for L0125, L0250, and L0500 respectively. Note the deviation from a monotonic trend as a result of resolution effects. [ARZ: Same comments as for Fig. 1.]

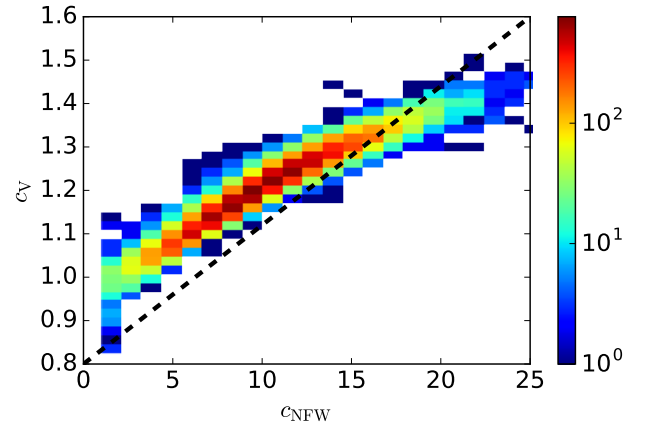


Figure 3. The relationship between the two different marks of concentration, using halos in L0250. The color scale, shown at the right, encodes the number of halos within a single two-dimensional bin in the $c_{\text{NFW}} - c_v$ space. Redder pixels are populated with a greater number of halos and show where the bulk of the halos lie. The black dashed line shows something... [ARZ: I don't know what the black dashed line shows, so you have to specify it.]. [ARZ: Previous caption was STILL not sufficiently specific. Consider carefully what I have written here as an example of a sufficiently specific caption.]

or s), and halo definition (the value of Δ in the case of the present study). [ARZ: The first sentence makes sense and I understand it. It is everything that follows that I don't understand. Are you saying that you have a unique threshold for each halo property? That is what I get from this statement, but I think that is not what you are doing. Can you please clarify this discussion?] We use thresholds determined by the particular case under consideration so as to ensure that resolution limitations do not drive any of our primary results. We summarize the mass thresh-

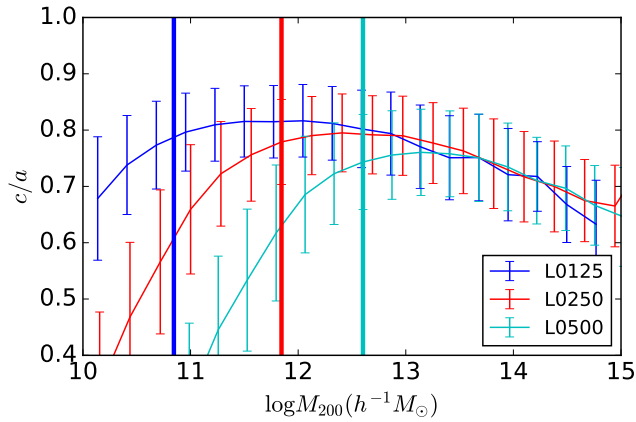


Figure 4. [ARZ: All the same comments as for Fig. 1. In addition, I was under the impression that you were going to select thresholds that guaranteed that the $s(M)$ relation is monotonic? Am I mistaken?] The relationship between halo shape and halo mass for each of our simulations with $\Delta = 200$. The chosen lower limit on halo mass for our sample is marked as a blue, red, or cyan line for L0125, L0250, and L0500 respectively. Note the deviation from a monotonic relation as a result of resolution effects.

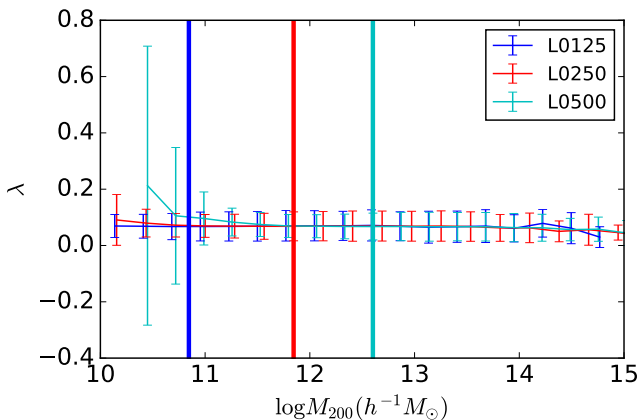


Figure 5. [ARZ: All of the same comments as for Fig. 1. Also, there is no reason here that the y-axis has to have such a large range. This crams all of the data into a small portion of the plot and makes most of the plot useless, wasted, white space. A more appropriate y-axis range looks to be from -0.1 to about 0.25 or so.] An example of the relationship between halo spin λ and halo mass for each of our simulations for $\Delta = 200$. The chosen lower limit on halo mass for our sample is marked as a blue, red, or cyan line for L0125, L0250, and L0500 respectively. Note the deviation from a monotonic relation as a result of resolution effects.

olds we have used for a subset of Δ values in Table 1. At most values of Δ , the minimum mass thresholds are driven by the requirement that the halo properties do not suffer significantly from finite resolution effects. [ARZ: What else could the thresholds be driven by?]

[ARZ: In the next section, you discuss subhalos. However, in order for the paper to have the correct flow, you should introduce the concept of

subhalos here and state that we will study halo clustering as a function of satellite number.]

[ARZ: In the table, don't "low mass," "mid mass," and "high mass" correspond to the different simulations? For example, low mass corresponds to L0125 and so on? If not, then I don't understand. If so, then you should label each cut by the relevant simulation name as well.] [ASV: We use the various mass cuts across simulations, so I was going for a more general name - e.g., we use high mass on both L0250 and L0500, which allows us to probe if the relation is driven by the simulation size or the actual mass threshold. I can swap out to more specific names though, but I thought that might cause confusion. It seems like we have swapped away from that methodology in the final version, so that might be best.] [ARZ: I still prefer using the simulation name to specify the threshold, otherwise the name seems arbitrary.]

[ARZ: I would like to see another figure here. I would like to see M_{200} on the x-axis and M_{Δ} on the y-axis. What should be plotted is the mean M_{Δ} as a function of M_{200} for all halos common to both catalogs. This will be a nice informative plot for the reader and will help the reader to digest the table.] [ASV: Will do using the halo matched catalogs - in progress.]

4 HALO CLUSTERING AS A FUNCTION OF AUXILIARY HALO PROPERTIES

4.1 Auxiliary Halo Properties

We are interested in studying the clustering behavior of halos as a function of properties other than mass. As mass is the dominant halo property determining halo clustering strength and environment, we refer to the other properties that we study as "auxiliary halo properties" (those properties other than mass, such as concentration c_{FW} or shape s). As has been demonstrated extensively in the literature, the auxiliary properties that we consider are a function of mass (?Allgood et al. 2006; Duffy et al. 2008; ?). [ARZ: Cite the Bullock et al. (2001) paper on halo concentrations here. Also, search to see if there are other papers with mass dependences of the other parameters. There probably will be and we probably need more citations here.] This mass dependence induces auxiliary-property-dependent halo clustering. Most contemporary cosmological N -body simulations, and specifically the suite of simulations that we study in this work, do not have a sufficiently large number of halos to make isolating halos of fixed mass, and then further splitting these halos by an auxiliary property, a powerful method with which to study the dependence of clustering on auxiliary properties. Therefore, it is necessary to remove or mitigate the mass dependence of the auxiliary properties that we study.

We mitigate the mass dependence of the auxiliary properties as follows. First, we take all host halos more massive than our minimum mass thresholds and sort

Table 1. Minimum mass thresholds for each of our analyses depending upon the value of the overdensity, Δ , used to define the halos. In the columns below the values of Δ we show the minimum host halo masses considered in units of $h^{-1}M_{\odot}$.

Cutoff Name	$\Delta = 340$	$\Delta = 200$	$\Delta = 100$	$\Delta = 75$	$\Delta = 50$	$\Delta = 10$
low mass	N/A	7×10^{10}	8×10^{10}	9×10^{10}	1×10^{11}	2×10^{12}
mid mass	N/A	7×10^{11}	8×10^{11}	9×10^{11}	1.5×10^{12}	N/A
high mass	3×10^{12}	4×10^{12}	5×10^{12}	6×10^{12}	7×10^{12}	N/A

them by their halo masses, M_{Δ} . We place these halos into ten logarithmically-spaced bins of halo mass, ensuring that no bin has fewer than XXX halos. [ARZ: I was under the impression that you used the median value? Can you check on this.] The mean value of each halo property is calculated within each bin, and subtracted from the recorded value for each halo in the bin. This removes the strong mass trend in each of these properties. We use these new measures of halo auxiliary properties, those with the mean mass trend removed, to study halo clustering strength in the remainder of this work. [ARZ: You should make sure that your results do not change qualitatively under a few modifications. First, make sure that if you use finer binning that nothing qualitative changes (small quantitative changes can be acceptable). We have to be sure that our results are not sensitive to the number of bins that you choose here. Second, you should make sure that your results are not qualitatively different if you use medians rather than means. Third, I believe that you also explore removing the mass trend by removing a best fit power law $c(M)$ (or $s(M)$ or $\lambda(M)$ or whatever) relationship? If so, that is great. You should conclude this sentence with a statement along the lines of “We have experimented with a variety of methods to remove the gross mass dependence of these auxiliary properties and have verified that our qualitative results are not sensitive to the number of bins we use or the choice of median or mean property. Further, we have experimented with removing the gross mass dependence of auxiliary halo properties by subtracting away a best-fit power law relation for the property of interest and again, we find that our qualitative results are not altered using this method.” We have to be able to give the reader some sort of confidence that our results are not driven by these seemingly arbitrary choices.] [ASV: I’m not necessarily comfortable with the current ten bin situation and could conceivably switch back to equally populated bins, which allows for more total bins with halos to look at, but unevenly spaced in log mass.] [ARZ: You should do whatever is most robust. Using either method, you should choose sufficiently large number of bins that our primary results are not altered by changing the number of bins.]

[ARZ: Flesh this out a bit. See the discussion in Wechsler et al. 06 as a model. The specific

values of the thresholds need to be given] [ASV: Slight change - following Wechsler et al 06 does not ever put a specific limit on subhalo velocity and instead places a limit on the ratio between the subhalo velocity and the host velocity as the prime limit. Just making sure that lines up well.]

Host halo size and the number of satellite halos (above some threshold in a proxy of satellite halo size) are also strongly correlated. To mitigate the correlation between halo mass and abundance of subhalos, we follow the prescription of Wechsler et al. (2006). First, we select host halos with $v_{\max, \text{host}} \geq 240 \text{ km s}^{-1}$. This ensures that the satellite counts will be minimally affected by finite resolution. We then select subhalos from the sample based on the criteria that the ratio $V_{\max, \text{sub}}/V_{\max, \text{host}} \geq 0.3$, where $V_{\max, \text{sub}}$ and $V_{\max, \text{host}}$ are the maximum circular velocities of the subhalos and host halos respectively. This scaling exploits the fact that the subhalo velocity function is a very nearly self-similar function (e.g., Zentner et al. 2005, and references therein), so that scaling subhalo $V_{\max, \text{sub}}$ by host $V_{\max, \text{host}}$ eliminates the gross mass dependence of satellite number. The threshold value for this velocity ratio was chosen such that all host halos that make the above cut contain, on average, more than one satellite halo. Subhalo maximum circular velocities were used as a proxy for subhalo size rather than subhalo masses because subhalo maximum circular velocities can be more robustly measured from simulation data, making comparison to other work easier.

4.2 Clustering Statistics

We assess the influence of assembly bias on two-point statistics of host halos. In order to do so, we study both the standard two-point correlation functions of halos selected by properties other than mass (e.g., the auxiliary properties concentration, shape, and spin) as well as halo mark correlation functions (MCFs). MCFs quantify the manner in which a halo property (the “mark”) correlates among halo pairs as a function of the distance between the pairs. MCFs have the advantage that they effectively stack signal from all values of the halo auxiliary property, or mark, in contrast to selecting subsets of halos based on the auxiliary property and MCFs stack signal from all environments and do not require any specific definition of halo environment in order to detect environmental trends. Absent halo assembly bias, the halo marks are uncorrelated among pairs. MCFs have been used in many previous papers to quantify environmental dependence of halo properties other than mass (Sheth & Tormen 2004;

Harker et al. 2006; Wechsler et al. 2006; Mao et al. 2015) [ARZ: I think there are also a couple of Sheth papers that should be cited here, maybe more. Check on that.].

For a specific halo property, or mark m , we use the MCF normalization of Wechsler et al. (2006), namely

$$\mathcal{M}_m(r) \equiv (\langle m_1 m_2 \rangle_p(r) - \langle m \rangle^2) / \mathcal{V}(m), \quad (7)$$

where m_i is the value of the mark for halo i , $\langle m \rangle$ is the mean of the mark, and $\mathcal{V}(m)$ the variance of the mark. The notation is intended to indicate that the average is taken over all pairs of halos separated by a distance r . In the absence of any correlation between a halo property among neighbors a separation r away, $\mathcal{M}_m(r) = 0$. Deviations of the MCF from zero indicate such correlations exist and the size of the $\mathcal{M}_m(r)$ gives the excess of the mark among pairs compared to the one-point mean of the mark $\langle m \rangle$ in units of the one-point variance.

It is necessary to assess statistical fluctuations in the statistics that we measure in these simulations in order to determine the significance of the signals. For two-point correlation functions, we determine the covariance of the measurement through jackknife resampling of the eight octants of the simulation cube. We assess the statistical significance of the MCFs by randomly re-assigning each of our marks to the halos in the sample. We then compute the MCF of these randomized marks. As the mark re-assignment is random, the MCF computed on these re-assigned marks *cannot* exhibit any environment dependence other than that induced by statistical fluctuations. We perform this reassignment 400 times and approximate a 2σ error region by the span of the MCF between our 10th lowest and 10th highest (that is the 390th if the MCFs were sorted in ascending order) values of the MCF. In the event that there are no environmental correlations with halo auxiliary properties, the MCFs measured in the simulations would fall within this error band 95% of the time.

5 RESULTS

5.1 Correlation Functions

[ARZ: Fix up and fill in numbers as needed, but follow this general format.] We begin by studying the correlation functions of halos in our mass threshold samples, sub-selected by auxiliary properties. Figure 6 exhibits the difference between the clustering strength of halos in the 20th percentile highest concentrations and the halos with the halos that have the 20th percentile lowest concentrations as a function of the overdensity parameter Δ , used to define the halos. [ARZ: which concentration?] First, it is clear that clustering strength generally depends upon the auxiliary properties of halos at all halo mass thresholds. [ARZ: How can you make a statement about auxiliary properties in general, when you only plot concentration?] Second, it is clear that the strength and sign of assembly bias is strongly mass dependent for any fixed halo definition, a result that agrees with the significant previous literature on halo assembly bias using conventional halo definitions (Wechsler et al. 2002; Gao et al. 2005; Zentner 2007;

Wechsler et al. 2006; Harker et al. 2006; Croton et al. 2007; Dalal et al. 2008; Mao et al. 2015; Sunayama et al. 2016)[ARZ: Cite all the usual papers here, Gao, Wechsler, Harker, Zentner07, Dalal07, probably several more.][ASV: I think that is most of the usual suspects, but I'll do a double check for more important ones!] At relatively low mass (the L0125 panel, $M_{200} > 7 \times 10^{10} h^{-1} M_\odot$), high-concentration halos are considerably more strongly clustered than low-concentration halos using the more conventional $\Delta = 200$ definition for halos. At somewhat higher halo masses (the L0250 panel, $M_{200} > 7 \times 10^{11} h^{-1} M_\odot$), this difference is markedly reduced. Finally, for the highest mass halos that we have the capability of studying (the L0500 panel, $M_{200} > 4 \times 10^{12} h^{-1} M_\odot$), the effect is of opposite sign; low-concentration halos are more strongly correlated than high-concentration halos.

Focus first on the middle panel of Fig. 6. In this panel, corresponding to the L0250 mass threshold, the difference in large-scale clustering between high- and low-concentration halos is dramatically reduced for a halo definition with $\Delta = 75$. Further decreasing Δ leads to concentration-dependent clustering of opposite sign. Comparing the differing clustering strengths across the three panels of Fig. 6, it is clear that this is not a universal conclusion. For low-mass halos, very low values of Δ (and correspondingly large definitions of halo radii, as $R_\Delta \propto \Delta^{-1/3}$) are necessary in order to mitigate the concentration dependence of halo clustering. Conversely, for higher mass halos (the L0500 panel), values of $\Delta \approx 200$ yield little concentration-dependent clustering. In this case, decreasing Δ (increasing R_Δ) results in significantly *increased* concentration dependent halo clustering. The reasons for these changes are of interest and we return to the interpretation of these results below.

Notice that in all panels of Fig. 6, the effect of concentration-dependent clustering is scale-dependent. Moreover, the effect is scale-dependent for all values of Δ . In these cases, simply defining halos with a different value of Δ does not suffice to eliminate concentration-dependent clustering on all scales. In this discussion and throughout, we focus primarily on the large scale clustering, which we take to mean clustering on scales significantly larger than the radii, R_Δ of the halos in our samples.

[ARZ: I assume that if you made the same plot for c_V we would reach the same basic conclusions as for c_{FW} ? If this is so, then you need to add a specific statement in this regard to the end of the discussion above. In this case, no new figure would be needed. If not, then you need to show the figure.]

[ARZ: You need to make some statement that the same conclusions would be reached with c_V .]

[ARZ: In this subsection, I advocate at least one and perhaps two new figures. First, can you make an analogous plot to Fig. 6 that shows some other property, such as shape or spin? This would be a nice addition for completeness. Second, what would happen if you used 50% or 10% cuts on concentration? Would we reach nearly the same conclusions? If so, then you do not need to show

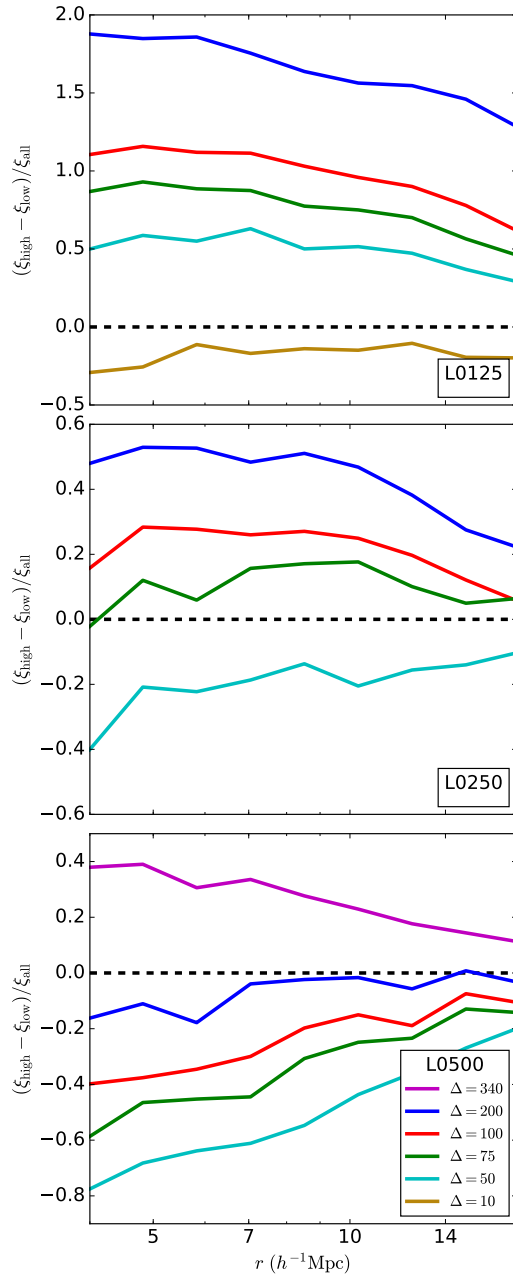


Figure 6. The difference of the correlation function for only the top 20% most concentrated halos and the bottom 20% in concentration, normalized by the overall correlation function of the entire sample. From the top panel to the bottom panel, we show the results for the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. [ARZ: You need to describe each line type in the panels.] [ASV: May still need error bars, depending on how clean we want the overall image, but should be quick enough to whip together.] [ARZ: For errors, you don’t want the jackknife error bars. What you want is to show the range you get when you randomly sub-sample 20% of all halos repeatedly.]

the plots explicitly. However, in this case, you DO need to state explicitly that this is the case. In other words, you need to state the the broad conclusions we reach are roughly independent of the specific percentiles that we use to select halos so that showing results for a variety of different percentiles doesn’t add very much to the discussion. It is also possible that selecting on a different percentile yields an entirely different conclusion. I don’t suspect that this will be the case, but it is possible. If that happens, then you need to show an example of such a figure for, say, concentration.] [ASV: These have all been tested at this point. They do not offer any unique conclusions, so I am uncertain if they want to be included beyond the concentration one, which shows the equivalence of the two methods.]

[ARZ: If you have tested these things, then you really need to have a paragraph in here at this point stating that you’ve checked these things and that they do not lead to any significantly different conclusions.]

5.2 Mark Correlation Functions

We now move toward a discussion of halo assembly bias as diagnosed by MCFs. MCFs have the advantage that it is not necessary to specify particular auxiliary property subsamples, such as the percentiles above, in order to assess assembly bias.² The NFW concentration, c_{NFW} , MCF is shown in Figure 7. The shaded bands in the figure delineate the statistical fluctuations in MCFs induced by finite sampling as discussed in the previous section. Qualitatively, Fig. 7 exhibits the same features that are evident in Fig. 6: more concentrated halos are significantly more clustered in the low-mass L0125 halo sample; concentration-dependent halo clustering weakens and reverses sense as halo mass increases, consistent with work on assembly bias by Sunayama et al. (2016); for the L0250 sample with $\Delta = 75$, the large-scale concentration dependence of halo clustering has been reduced so as to be consistent with zero within the statistical limitations of the simulation. [ARZ: It seems like we need $\Delta \sim 65 - 70$ for L0250?]

Figure 8 is a similar plot for the velocity-defined concentration, c_v MCF. This figure exhibits qualitatively and quantitatively similar features to Fig. 7, a fact that is not surprising given that we already know that c_{NFW} and c_v quantify largely redundant information about their halos.

Moving on from concentrations, Figure 9 illustrates MCFs in which the mark is the shape parameter, s , of

² Of course, this comes at the cost of averaging over all values of the auxiliary properties. Using MCFs it is not evident if the assembly bias effect is a smooth function of the auxiliary property, dominated by the tails of the auxiliary property, or has some more complex dependence upon the auxiliary property. Our results in § 5.1 suggest that assembly bias is a fairly smooth function of the auxiliary properties.

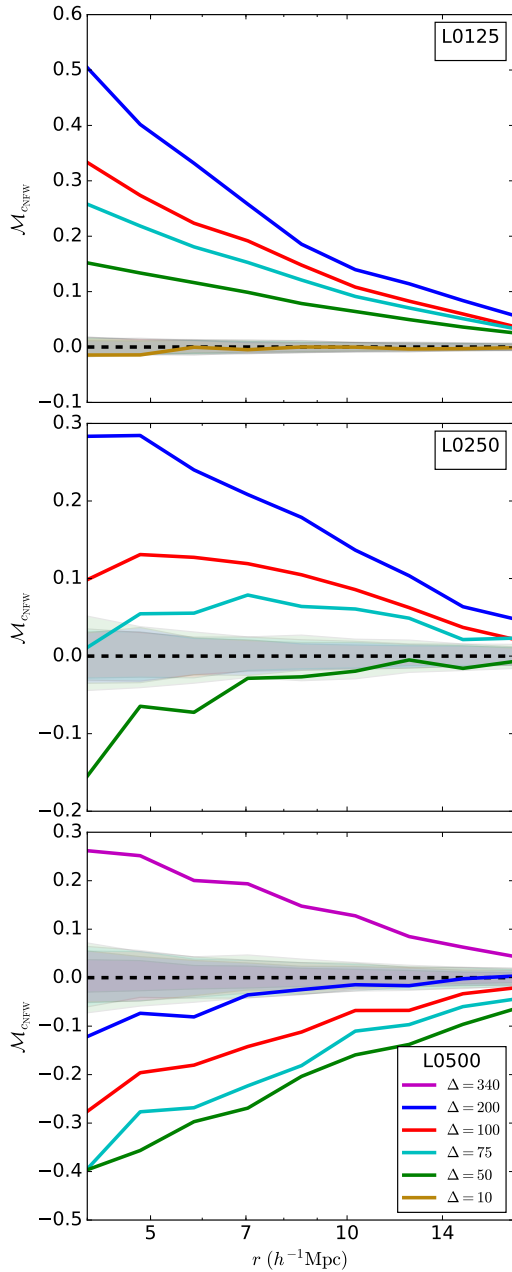


Figure 7. The marked correlation function for the concentration defined according to the NFW profile. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. The shaded bands represent 2-sigma confidence regions generated by randomization of the marks.

the halo. The environmental dependence of shape parameter is distinct from that of concentration in a number of ways. Notice that at all masses and at all halo definitions, less spherical halos (halos with smaller s) are more strongly clustered. Furthermore, decreasing Δ (increasing halo radius) only serves to increase this environmental dependence in all cases. This is indicative of halo shapes being driven by structure on scales larger than halo radii, in particular R_{200} , and is not entirely surprising given our significant amount of knowledge of the filamentary nature of large-scale structure. [ARZ: If possible, a line here

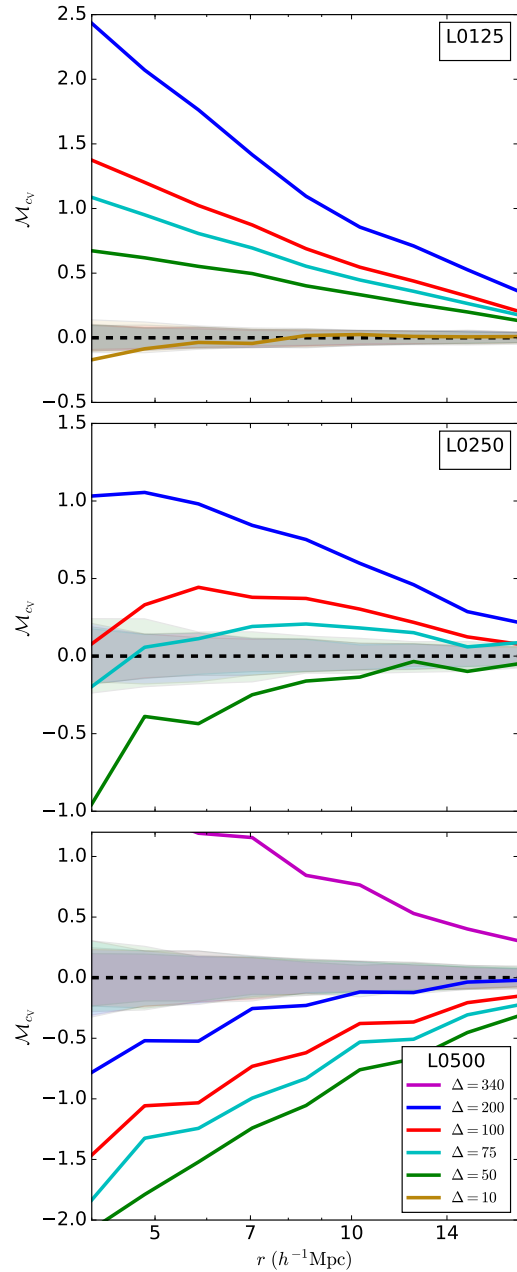


Figure 8. The marked correlation function for the concentration defined according to the velocity ratio. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. The shaded bands represent 2-sigma confidence regions generated by randomization of the marks.

with $\Delta \gg 200$. Seems like it could be instructive, let me know if this is going to take a long time or be very onerous.][ASV: Added and continues the trend in the direction as expected. Note reversal in monotonic relationship for lowest mass objects though. Looking over the cuts, this may imply that the mass cuts are not aggressive enough for the shape mark, though. Vega et al 2016 paper suggests this is the case.]

The spin MCFs are shown in Figure 10. Qualitatively, spin-dependent halo clustering is quite similar

to shape-dependent halo clustering. Halos of high-spin cluster more strongly than halos of low-spin. [ARZ: Search the literature, particularly Brandon Allgood’s papers and Andreas Faltenbacher whose names come to mind, for spin-dependent assembly bias. I’m trying to figure out if yours is the first paper to point this out.][ASV: Need to track down all the specific citations, but after changing our normalization method, it seems to match up better with existing texts.] Likewise, increasing halo radii by decreasing Δ in halo definitions only drives spin-dependent halo clustering to be stronger.

The clustering of halos as a function of the number of satellite galaxies at fixed mass is of particular practical interest. Efforts to model survey data on the large-scale galaxy distribution, such as the halo occupation distribution (HOD) or conditional luminosity function (CLF) formalisms typically make the assumption that the multiplicity of satellite galaxies within a host dark matter halo depends solely upon halo mass. If this assumption is violated, then the phenomenological modeling of galaxy clustering can be more complicated than in these simplest scenarios.

Figure 11 shows clustering marked by satellite number as described in § 4. For all values of Δ , halo clustering is strongly dependent upon satellite occupation at all masses. Interestingly, altering halo definitions as we have makes little difference to this dependence. In fact, defining halos to have larger radii (smaller Δ) generally makes the environmental dependence of halo clustering more significant. Of course, our results pertain to satellite halos, or subhalos, rather than satellite galaxies, so the connection to observations and how one might model observed galaxy clustering is indirect, yet suggestive.

6 DISCUSSION

[ARZ: Repeating this comment here because of its importance. Look in the literature to see if spin-dependent halo clustering has been measured in the literature before. Look at <http://arxiv.org/abs/1207.4476>.]

We have confirmed that for conventional halo definitions halo clustering strength is a strong function of “auxiliary properties” halo concentration (either measured through a fit to the NFW profile or assigned non-parametrically as the ratio of the maximum circular velocity to the virial velocity), halo shape, halo spin, and number of subhalos for host halos over a wide range of masses. These findings are consistent with the now significant literature on the subject of halo assembly bias. (Peacock & Smith 2000; Wechsler et al. 2002; Sheth & Tormen 2004; Gao et al. 2005; Zentner et al. 2005; Allgood et al. 2006; Harker et al. 2006; Wechsler et al. 2006; Croton et al. 2007; Dalal et al. 2008; Mao et al. 2015; Sunayama et al. 2016) [ARZ: Reiterate citations here.][ASV: I THINK this covers our bases, but my lit notes might be bad.]

We have explored the efficacy of alternative halo definitions to mitigate the dependence of halo clustering on these “auxiliary properties.” Rather generally, we find

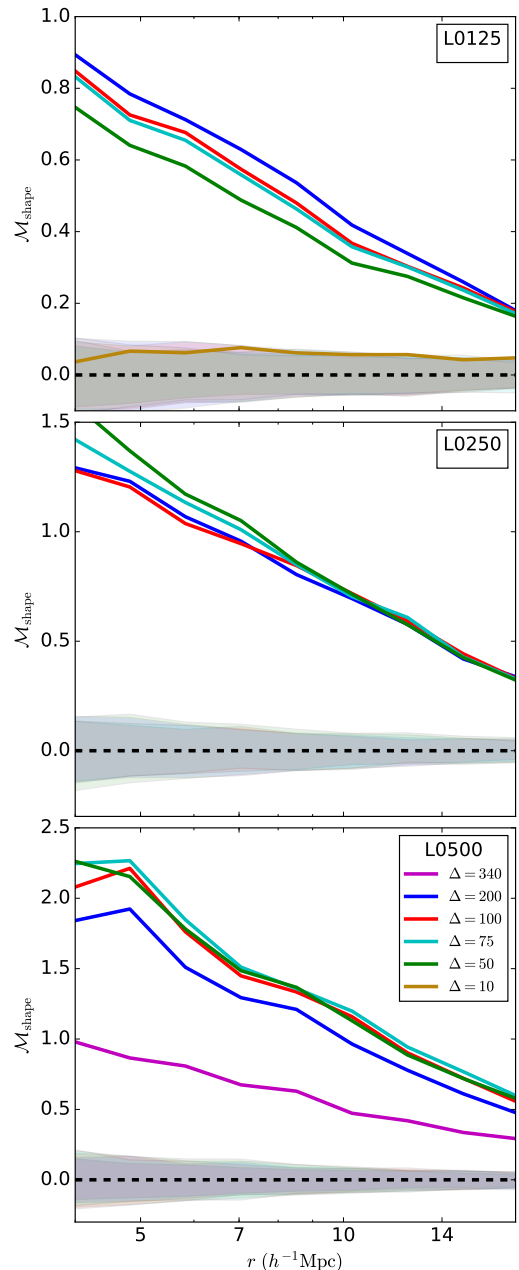


Figure 9. The marked correlation function for the shape of the halo. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. The shaded bands represent 2-sigma confidence regions generated by randomization of the marks. [ASV: it is possible that this is driven by insufficiently aggressive resolution cuts. This may drive us to do separate mass cuts for each mark - vexing, but should be feasible in the new framework.]

that these alternative definitions do *not* mitigate the effects of assembly bias. In most cases, defining halos to have significantly larger radii (lower Δ) than in conventional halo definitions had only a modest influence on these assembly bias effects. Moreover, to the degree that these modified halo definitions had any effect at all, it was often of the sense of making the assembly bias effect stronger, rather than weaker.

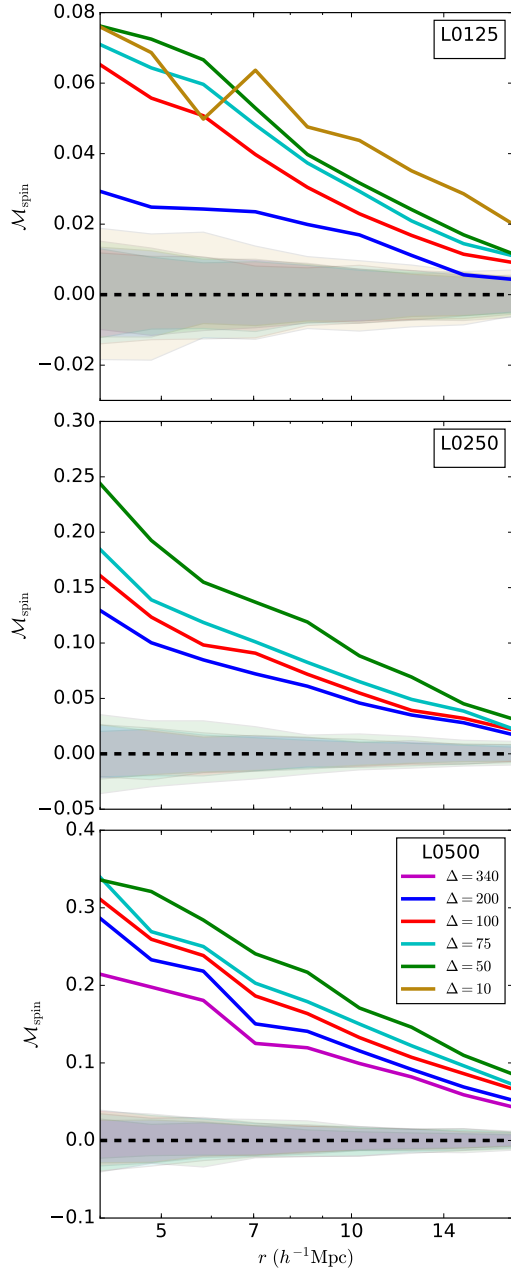


Figure 10. The marked correlation function for the spin parameter of the halo. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. The shaded bands represent 2-sigma confidence regions generated by randomization of the marks.

One exception to this general conclusion is the case of halo concentration. Our results suggest that halo redefinition may be able to mitigate concentration dependent halo clustering. This is evident in Fig. 7 and Fig. 8. Halo concentration is strongly correlated with halo formation time, so this suggests that such a redefinition may also aid in reducing assembly bias associated with halo formation time; however, this is a non-trivial extrapolation of our results and a follow-up study to assess halo formation times in alternative halo definitions is both interesting and warranted.

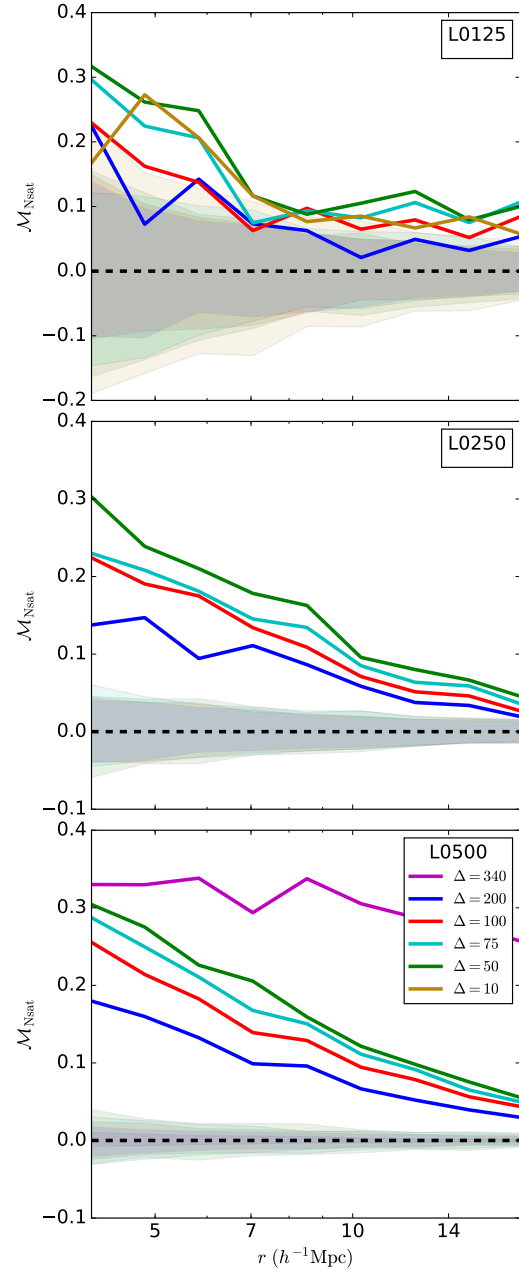


Figure 11. The marked correlation function for the number of satellites for a host halo. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. The shaded bands represent 2-sigma confidence regions generated by randomization of the marks.

Clearly, the halo definition that best mitigates concentration-dependent assembly bias must be mass dependent. Low values of Δ ($\Delta \sim 25$ with $R_{25} \sim 2R_{200}$) seem appropriate near for our lowest mass-threshold sample (with $M_{200} \geq 7 \times 10^{10} h^{-1} M_{\odot}$) whereas $\Delta \sim 200$ or slightly higher is adequate for our highest mass threshold sample (with $M_{200} \geq 4 \times 10^{12} h^{-1} M_{\odot}$). This result is reminiscent of much recent work on the so-called halo “splashback radius.” (More et al. 2015) **[ARZ: Here, compare our results to the splashback radius work.]**

[ARZ: Here is where we could use ANOTHER FIGURE! What I would like to see is a visual comparison of our “best” Δ as a function of halo mass threshold compared to the equivalent for the splashback radius. To be clear, it is trivial to go back and forth between Δ and radius (since $\text{radius} \propto \Delta^{-1/3}$) so this can be represented with either Δ or radius. If you choose to use radius, then it should probably be normalized, such as R_{Δ}/R_{200} and $R_{\text{splashback}}/R_{200}$ and so on. This could be an important figure and point to future work on this subject.] [ASV: Working on creating this plot from the splashback radius paper.]

It is useful to investigate the reasons why halo redefinitions may be helpful in the case of concentrations. On the positive side, it is possible that these redefinitions do define halos in a more practically useful way, better isolating objects that have been strongly altered by nonlinear evolution from the larger-scale environment. In this case, halo redefinition would be a step forward. However, it is also possible that the details of measuring halo properties using these new halo definitions introduce new sources of noise into the measurements. If this is the case, then the reduction in environmental effects stems from the fact that their measurement of the halo property introduces noise and is *less* informative about the halo itself. For the case of halo concentration, which is the most interesting to follow up, introducing noise can happen in numerous ways. For example, the NFW concentration c_{NFW} is determined by a fit to the NFW profile. Inferred values of c_{NFW} will depend upon the degree to which the density profiles of the halos follow the NFW functional form within some radius R_{Δ} that is different from traditional halo radii, such as $\sim R_{200}$. At large halocentric distances ($r \gtrsim R_{200}$) halo profiles are known to deviate from the NFW form. It is worth noting that the velocity-defined concentration c_V is a non-parametric measure of concentration and should be less subject to such effects.

[ARZ: New figure needed here. This should be, for the L0250 simulation (my guess this one is most instructive, but use your judgment here), a comparison of the concentration-mass relation for halos defined with $\Delta = 200$ to the concentration-mass relation for halos defined with $\Delta = 70$ (of whatever you deem best). The figure must represent both the mean (or median) relation AS WELL AS the dispersion in this relation at fixed mass. The plot should exhibit this for both c_{NFW} and c_V . Two panels may be necessary to make these points.] [ASV: contemplating how best to show this - currently making a running median plot, as in the masscut plots above.]

As such, it is useful to examine the concentration-mass relations for halos in the simulations for various halo definitions. This is shown in [ARZ: NEW FIGURE.] Notice that [ARZ: Now you mention the interesting part about the new figure. This should be a discussion of how much larger/smaller the dispersion in concentrations gets for, say $\Delta = 70$ compared to $\Delta = 200$.]

We can explore in more detail the degree to which the mitigation of environmental effects by halo redefi-

nition are due to introducing noise that is uncorrelated with environment into the measurement of halo properties when halos are defined in a manner distinct from the more traditional definitions. We proceed as follows. All host halos that are found in the halo catalogs constructed from lower values of Δ (for example, $\Delta = 70$ which is an interesting value for exploring concentration in the L0250 simulation) are present as host halos in the halo catalogs constructed with higher values of threshold density (e.g., $\Delta = 200$). We match each halo in the lower threshold (lower Δ) simulation to its corresponding halo in the higher threshold catalog. We then consider the clustering of only those halos that we have matched across catalogs. We refer to these as the “matched” halo samples between two values of threshold overdensity Δ .

[ARZ: I removed the correlation function comparison here. It is too busy to read and not necessary. It also opens up the whole Pandora’s box of “why choose 20%?” I’d rather avoid that in this discussion. Let’s just jump straight to the MCF. This will reduce the proliferation of figures as well.]

[ARZ: Please check to ensure that I know what you mean by your matched samples. Your language was not specific enough for me to be 100% sure. Modify as necessary.] Figure 12 shows the same statistic as Figure 7, except for matched subsamples. The matched subsamples are matched to a catalog generated for the value of Δ most likely to remove assembly bias at large scales for the concentration marks; $\Delta = 10$ for the ‘low mass’ cut, $\Delta = 70$ for the ‘mid mass’ cut, and $\Delta = 200$ for the ‘high mass’ cut. These matched halo catalogs only differ from the standard halo samples in that they contain only those host halos common to both the catalog in question and the best fit Δ catalog. The most interesting matched sample to examine in this case is the $\Delta = 200$ sample. In this sample, all halos are defined as they would be defined in the $\Delta = 200$ catalog, including all inferred halo properties; however, the matched catalog contains only those host halos that also appear in the $\Delta = 70$ halo catalog. Therefore, many host halos have been removed from the sample because they have become subhalos in the $\Delta = 70$ catalogs.

From Fig. 12, it is apparent that some degree of assembly bias persists in the matched samples. Yet, what is interesting is that a very significant fraction of the assembly bias effect has been removed compared to the standard $\Delta = 200$ result. The halos in the matched catalogs have the same properties (including c_{NFW}) as those in the standard catalogs, so that removal of assembly bias is *not* due to introducing noise or other systematics into the measurement of concentration. That mitigation of assembly bias is due to the halo redefinition and, in particular, subsuming those halos most subject to assembly bias effects as subhalos of the $\Delta = 70$ halos. This is an interesting result suggesting that seeking optimal halo definitions may be one avenue to more completely separating the strongly nonlinear evolution occurring within halos from large-scale evolution and mitigating assembly bias.

[ARZ: Put a discussion of Fig. 13 here. It does not need to reiterate the discussion of Fig. 12.]

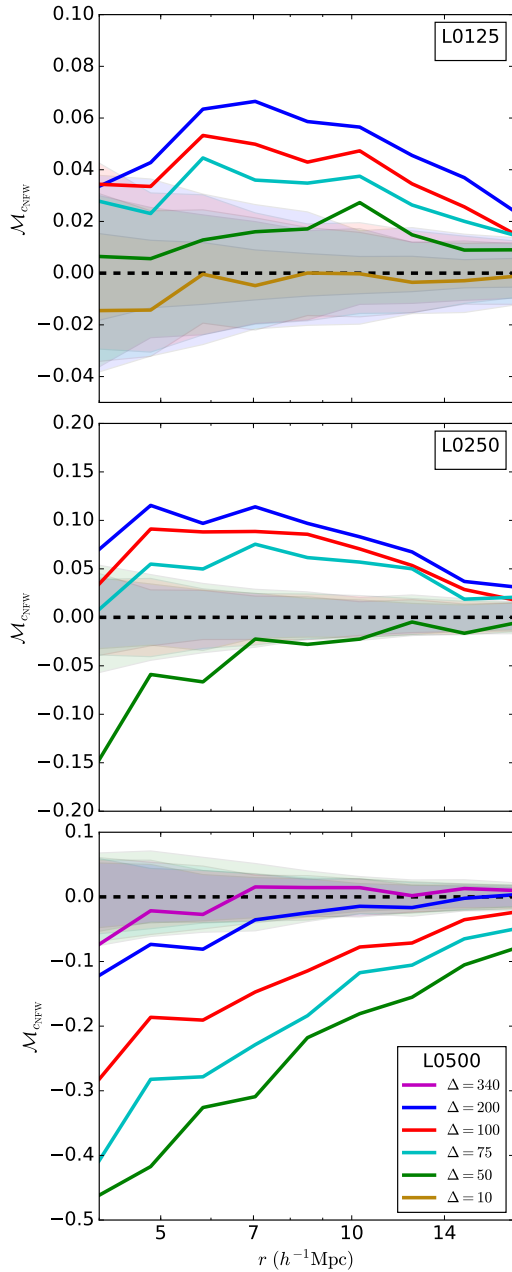


Figure 12. The marked correlation function for the concentration defined according to the NFW profile. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. Only host halos consistent with the best-fit catalog from above are included in the analysis.

However, it should state that we reach the same broad conclusion and that this is good in part because c_v is a nonparametric measure of halo concentration.] Figure 13 follows the same exercise as above, with a comparison drawn to the Fig. 8. Notice that the same trends as Fig. 12 can be seen: a significant fraction of assembly bias is removed as compared to the unmatched catalog, though statistically significant assembly bias remains. Only in combination with halo redefinition do we remove assembly more completely on large scales.

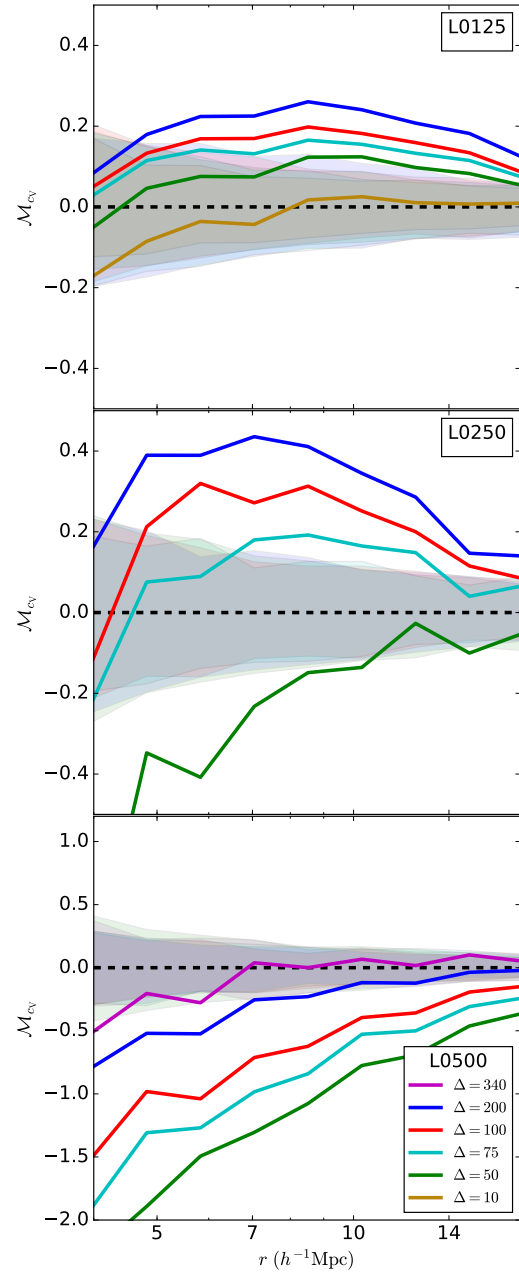


Figure 13. [ARZ: All the same comments as for the c_{NFW} figure.] The marked correlation function for the concentration defined according to the velocity ratio. From top to bottom we show the ‘low mass’, ‘mid mass’, and ‘high mass’ cutoffs. Only host halos consistent with the best-fit catalog from above are included in the analysis.

[ARZ: Would it be easy for you to plot the concentration-mass relation for only those host halos in the “matched” $\Delta = 200$ sample? If so, that would be potentially interesting for our interpretation.][ASV: In progress for above.]

[ARZ: This paragraph is good, but needs to be written a bit more professionally. Start with a sentence like, “It is interesting to explore the reasons that halo shape, spin, and satellite number are not amenable to having their assembly

bias mitigated through simple halo redefinitions.” Then move on to some specifics.] [ASV: attempted to rewrite more professionally!]

While halo shape, spin, and satellite number are not amenable to having their assembly bias mitigated through the simple halo redefinition technique we have suggested, the underlying reasons for this behavior remains of interest for exploration. In the case of halo shape, we suggest that the assembly bias may be driven through interactions with large scale structure. Studies have shown a statistically significant alignment between filaments and satellite galaxy position (Tempel et al. 2015; Velliscig et al. 2015)[**ASV: need to grab paper from arxiv 2016.05.09 for this**]. Figure 14 demonstrates the distribution of halos for the $\Delta = 200$ and $\Delta = 10$ catalogs of L0250. [**ASV: should probably re-make this with halotools results just to be robust**] Notice the preferential distribution of satellites being subsumed into host halos. Notice the preferential direction in which host halos are subsumed in the latter definition. This material is likely to have significant impact on the measurement of halo shape, potentially driving our results. This would also impact the satellite number count, as the most clustered regions will see the largest increase as a result of this halo radius change.

Note that the mass dependence of assembly bias is implicitly explored with our suite of simulations. The nature of the MCFs emphasizes halos just above the threshold mass; e.g. L0125 probes the smallest mass halos, while L0250 probes the largest mass halos. One single halo definition is insufficient to account for the entirety of assembly bias across the entire mass range. However, it is clear that assembly bias can be a strong function of halo definition at fixed halo mass. Meta-analysis of the literature remains difficult as a natural consequence, as different choices in the process of defining a halo can lead to significantly different results. [**ARZ: Instead, add one summarizing paragraph here discussing the mass dependence of assembly bias. Emphasize that our findings suggest that the strength of assembly bias can be a strong function of halo definition and that this may already be making it difficult to compare the results of various different research groups using different halo definitions.**]

7 CONCLUSIONS

[**ARZ: After dealing with the comments above, let’s come back to rewriting the conclusions section.**] We have looked at how to use CFs and MCFs in order to analyze the environmental effects upon the properties of the halo. We have suggested a method of removing the mass dependence that is not subject to the small number statistics at large halo masses. Taking our various tests, we then apply a change to the threshold density Δ in an attempt to remove the effect that environment has upon these properties. We come to the following conclusions from our simulation data.

- Our halo redefinition method does not cause any substantial breakdown in the ROCKSTAR halo finding

algorithm, though this may not be the case for every halo finding methodology. This is something that should be considered prior to utilization of this method, unless working directly from particle data. As our initial halo sizes and locations are determined through spherical overdensities, it cannot be assumed that starting from a FoF grouping and then determining values through particle data directly will produce identical results. Similarly, different cosmologies may remove environmental effects at different scales.

- When looking at the two-point correlation function, there appears to be a “sweet spot” that appears to remove environmental effects the most efficiently. Going beyond that seems to reintroduce environmental effects, possibly as an extreme side effect of halo exclusion.

- For our marked correlation functions we see that both proxies of concentration that we use as marks show significant removal of environmental effects at large scales for similar values of the overdensity parameter Δ . In cases where one is only interested in the concentration of dark matter halos and large scales (or correspondingly small values of k), this method will allow you to compensate for bias that environment could introduce to calculations dependent upon the halo model. This may prove valuable for calculations such as that of the shear power spectrum calculated through weak lensing.

- The environmental effects on the shape of the host halo and the satellite number of the host halo cannot be removed regardless of the chosen redefinition of Δ . We propose that this may be intrinsically tied to the nature of the filaments, whose effects cannot be removed by a simple redefinition of the halo radius.

- This method is definitively related to the mass of the halos that are being observed. Furthermore, it appears that the majority of the reduction in assembly bias is tied to the exclusion of halos from the catalog as a result of being subsumed into larger halos. This information does not seem to be contradictory; it can be intuitively understood that the region about the most massive halos will be different than the region around the least massive halos, leading to a different frequency at which halos are being excluded. It does however warrant that careful consideration be given to the sample of halos that are of interest.

- The selection of halo size is intrinsically related to the assembly bias and varies across scales. This might help to resolve contradictory results in the search for halo assembly bias in the literature.

This methodology, while certainly not perfect in accounting for assembly bias, may be of significance when applied to galaxy formation models and give insight into seemingly conflicting results. Provided that the properties of interest in a given model behave well under our redefinition, it will allow us to create better mock galaxy catalogs without resorting to more complicated models requiring halo formation histories - giving us another powerful tool to test observation against.

There remain possible uncertainties to study in the future. One possible area of follow-up is the matter of simulation cosmology, which is not explored in this text. It is possible that the choice of cosmology may change

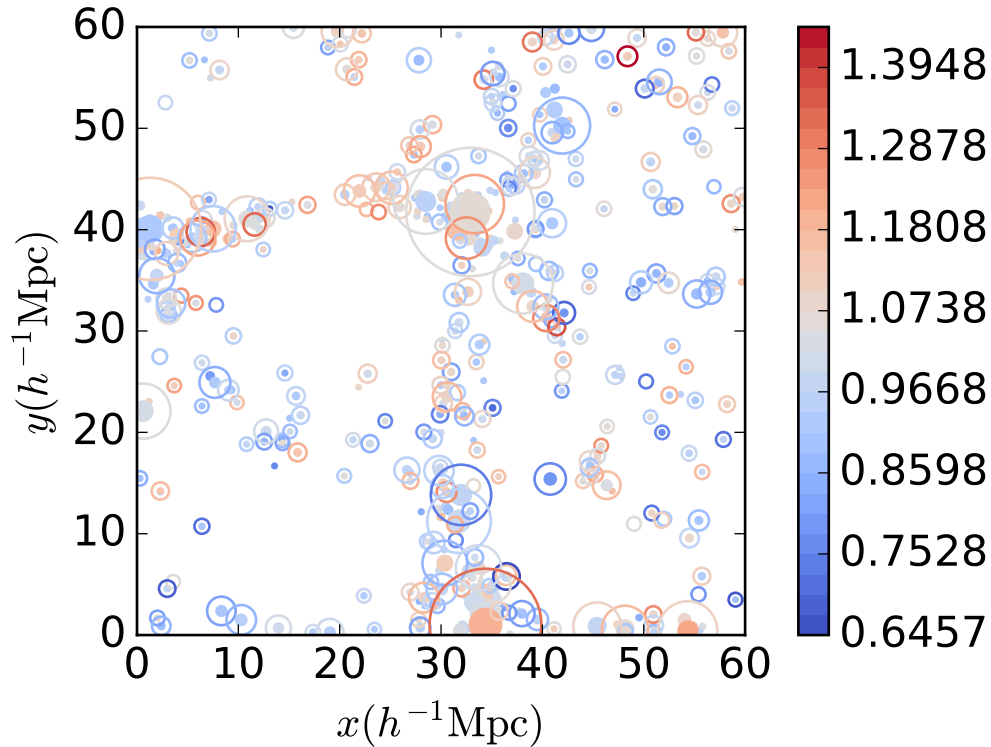


Figure 14. A $20 h^{-1}\text{Mpc}$ deep cut of L0250 along the z -axis. This zoom-in demonstrates the process that decreases the shape parameter as a function of clustering. The size of each circle represents the projection of a spherical dark matter halo with a given halo radius onto the x - y plane. Filled circles use the $\Delta = 200$ catalog and unfilled circles use the $\Delta = 10$ catalog in order to make the effect more visible. Color scale refers to the shape mark, normalized by halos of the same mass.

observed assembly bias as a function of the halo masses, something that our methodology should be capable of observing. Furthermore, we can determine if the choice of halo size that best reduces assembly bias is a function of the chosen cosmology. This may be of interest in attempting to determine signatures of assembly bias in observational samples in the future.

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APPENDIX

One natural question that might arise in the analysis of this work is the nature of the resulting assembly bias trends. Our focus in the main sections of this paper is on the nature of the assembly bias changing as a function of the mass cut chosen. Our conclusions include the fact that there is a strong mass dependence on halo assembly bias that must be accounted for separately depending on the halos of interest in a study. However, while the existence of this trend is clear within our analysis, the determination that this is solely due to the masses of the halos included in our calculation is less clear upon closer inspection. One possibility that might be particularly concerning is the potential that the different simulations have created halos that have fundamentally different clustering and this is leading to the result that we are interpreting as a mass dependence on assembly bias. Thankfully, though our statistics become less meaningful to carry out this calculation, we can carry out a comparison using the same mass cut across two of our simulations, knowing that these will only contain well resolved halos.

While not addressed directly, Figure ?? through Figure ?? contain a demonstration of the result that we are seeking in the left column of panels. The lower left panels show various marks of interest for L0250 using the “mid mass” cut on the data set. In comparison, the upper left panel contains the same marks of interest for the L0125 using the same mass cut. In the latter, there are fewer halos in this mass cut range, as a result of the smaller simulation box size. However, we note that despite the additional noise in the data set, the behavior of the assembly bias measurement is nearly identical within tolerances accounting for differences between simulations and noise. This motivates our conclusion that the driver behind the behavior is the mass cut of the data sets rather than the resolution of the simulation.