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The completed SDSS-IV extended Baryon Oscillation Spectroscopic Survey: N -body mock challenge for galaxy clustering measurements

Graziano Rossi,^{1★} Peter D. Choi,¹ Jeongin Moon,¹ Julian E. Bautista^{1,2}, Hector Gil-Marín^{1,3,4}, Romain Paviot,^{5,6} Mariana Vargas-Magaña,⁷ Sylvain de la Torre,⁵ Sebastien Fromenteau^{1,8}, Ashley J. Ross^{1,9}, Santiago Ávila,^{10,11} Etienne Burtin,¹² Kyle S. Dawson,¹³ Stéphanie Escoffier,⁶ Salman Habib,^{14,15} Katrin Heitmann,¹⁴ Jiamin Hou,¹⁶ Eva-Maria Mueller,¹⁷ Will J. Percival,^{18,19,20} Alex Smith,¹² Cheng Zhao²¹ and Gong-Bo Zhao^{2,22,23}

Affiliations are listed at the end of the paper

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ABSTRACT

We develop a series of N -body data challenges, functional to the final analysis of the extended Baryon Oscillation Spectroscopic Survey (eBOSS) Data Release 16 (DR16) galaxy sample. The challenges are primarily based on high-fidelity catalogues constructed from the Outer Rim simulation – a large box size realization ($3h^{-1}\text{Gpc}$) characterized by an unprecedented combination of volume and mass resolution, down to $1.85 \times 10^9 h^{-1} M_{\odot}$. We generate synthetic galaxy mocks by populating Outer Rim haloes with a variety of halo occupation distribution (HOD) schemes of increasing complexity, spanning different redshift intervals. We then assess the performance of three complementary redshift space distortion (RSD) models in configuration and Fourier space, adopted for the analysis of the complete DR16 eBOSS sample of Luminous Red Galaxies (LRGs). We find all the methods mutually consistent, with comparable systematic errors on the Alcock–Paczynski parameters and the growth of structure, and robust to different HOD prescriptions – thus validating the robustness of the models and the pipelines used for the baryon acoustic oscillation (BAO) and full shape clustering analysis. In particular, all the techniques are able to recover α_{\parallel} and α_{\perp} to within 0.9 per cent, and $f\sigma_8$ to within 1.5 per cent. As a by-product of our work, we are also able to gain interesting insights on the galaxy–halo connection. Our study is relevant for the final eBOSS DR16 ‘consensus cosmology’, as the systematic error budget is informed by testing the results of analyses against these high-resolution mocks. In addition, it is also useful for future large-volume surveys, since similar mock-making techniques and systematic corrections can be readily extended to model for instance the Dark Energy Spectroscopic Instrument (DESI) galaxy sample.

Key words: methods: analytical – methods: numerical – methods: statistical – galaxies: formation – cosmology: theory – large-scale structure of Universe.

1 INTRODUCTION

The Sloan Digital Sky Survey (SDSS; York et al. 2000), currently in its fourth generation (SDSS-IV; see Blanton et al. 2017 for a review), has established a remarkable legacy in astronomy and set new standards for precision cosmology. A key component of the SDSS-IV, the Extended Baryon Oscillation Spectroscopic Survey (eBOSS; Dawson et al. 2016) is now releasing the final cosmological catalogues (Lyke et al. 2020; Raichoor et al. 2020; Ross et al. 2020) with the Data Release 16 (DR16), summarizing the efforts of more than 10 yr of operations. eBOSS spectroscopically targets four distinct astrophysical populations: luminous red galaxies (LRGs), the primary focus of this work), emission-line galaxies (ELGs), clustering quasars (QSOs), and the Lyman- α (Ly α) forest of quasars at high redshift. In a novel and yet uncharted redshift interval, eBOSS has built the most complete, unprecedented large volume map of the universe usable for large-scale structure (LSS) to date.

Exquisite high-quality data from the SDSS have been pivotal in firmly establishing the standard minimal six-parameter concordance cosmological scenario dominated by cold dark matter (CDM) and a dark energy (DE) component in the form of a cosmological constant Λ , known as the Λ CDM model. Traditionally, this has been achieved by using the baryon acoustic oscillation (BAO) feature as measured in galaxy and quasar clustering, to estimate the angular diameter distance D_M and the Hubble parameter H , as well as their product from the Alcock–Paczynski effect (AP; Alcock & Paczynski 1979), and the growth of structure quantified by $f\sigma_8(z)$ from redshift-space distortions (RSD) – with $f(z)$ the logarithmic growth rate of the linear fluctuation amplitude with respect to the expansion factor, and $\sigma_8(z)$ the normalization of the linear theory matter power spectrum at redshift z via the rms fluctuation in $8h^{-1}$ Mpc spheres. Since the very first BAO detections (Colless et al. 2003; Cole et al. 2005; Eisenstein et al. 2005), measurements of the BAO peak have been sharpening and expanding in redshift range, allowing for multiple accurate cosmological constraints and solid confirmations of the Λ CDM framework. Noticeably, the eBOSS team has recently presented the first measurement of the BAO signal

★ E-mail: graziano@sejong.ac.kr

in a novel uncharted redshift range ($0.8 < z < 2.2$) using the clustering properties of 147 000 new quasars (Ata et al. 2018), and reported a BAO detection with a significance $>2.8\sigma$ along with detailed high- z distance measurements (within 3.8 per cent), a remarkable result that confirms and extends the validity of the standard Λ CDM cosmological model to an unprecedented large volume.

To this end, multiple techniques involving RSD methods and clustering estimators along with BAO reconstructions in configuration or Fourier space are generally adopted for the analysis of the various LSS tracers, to extract cosmological information. The most up-to-date SDSS results involving LRGs can be found in Beutler et al. (2017a, b), Gil-Marín et al. (2017), Bautista et al. (2018), Mueller et al. (2018), Vargas-Magaña et al. (2018), Wang et al. (2018b), Zheng et al. (2019), and Icaza-Lizaola et al. (2020). Regarding ELGs, one of the novelties in eBOSS, recent studies have been carried out by Comparat et al. (2016), Raichoor et al. (2017), Guo et al. (2019). For the QSO population, see e.g. Gil-Marín et al. (2018), Hou et al. (2018), Wang et al. (2018a), Zarrouk et al. (2018), Zhu et al. (2018), Ruggeri et al. (2019), Zhao et al. (2019); and for Ly α QSOs see Blomqvist et al. (2019) and de Sainte Agathe et al. (2019).

Traditionally, all the main results from different SDSS tracers are eventually combined in a ‘consensus’ publication (e.g. Aubourg et al. 2015; Alam et al. 2017; Ata et al. 2018), and confronted with measurements from other state-of-the-art surveys – such as Planck (2018). This consensus is then of utmost importance, as it represents a legacy for the entire science community. We are now releasing the final eBOSS DR16 consensus analysis that summarizes the full impact of the SDSS spectroscopic surveys on the cosmological model (eBOSS Collaboration 2020), which encapsulates all the supporting clustering measurements presented in Bautista et al. (2020) and Gil-Marín et al. (2020) for LRGs, Hou et al. (2021) and Neveux et al. (2020) for QSOs, de Mattia et al. (2021) and Tamone et al. (2020) for ELGs, as well as du Mas des Bourboux et al. (2020) for the Ly α forest.¹

In this respect, quantifying the systematic error budget in RSD methods and BAO clustering estimators for all of the eBOSS tracers as well as characterizing the robustness of the analysis pipelines are essential tasks, in order to obtain unbiased cosmological parameters, accurate $f\sigma_8$ constraints, and reliable consensus likelihoods. This is indeed the central aim of our work: here we focus on galaxies, and assess the performance and robustness of the BAO fitting methods and of three complementary RSD full shape (FS) models in configuration and redshift space, adopted in Bautista et al. (2020) and Gil-Marín et al. (2020) for the analysis of the complete DR16 eBOSS LRG sample – briefly described in Section 5. See also Smith et al. (2020) for an analogous effort on the QSO sample, and Alam et al. (2020), Avila et al. (2020), and Lin et al. (2020) for ELGs.

With this primary goal in mind, we have devised a targeted galaxy mock challenge. In embryonic form, a similar mini-challenge was already present in the consensus eBOSS Data Release 12 (DR12) LRG analysis (Alam et al. 2017, see their Section 7). Here we expand on that, and carry out a more systematic investigation. Specifically, in our challenge (detailed in Section 6) we test the performance of BAO/RSD LRG fitting techniques against different galaxy population schemes and bias models having analogous clustering properties, with the main objective of validation and calibration of such methods and the quantification of theoretical systematics.

Assessing the robustness and accuracy of RSD models is only possible via high-fidelity (N -body-based) synthetic realizations. In this work, we construct new heterogeneous sets of galaxy mocks from the Outer Rim (Heitmann et al. 2019, see Section 4) – a large box size run ($3 h^{-1}$ Gpc) characterized by a high mass resolution, down to $1.85 \times 10^9 h^{-1} M_\odot$. We base our methodology on Halo Occupation Distribution (HOD) techniques, in an increasing level of complexity (as thoroughly explained in Section 3): in particular, moving from the most conventional HOD framework, we explore more sophisticated scenarios able to distinguish between quiescent and star-forming galaxies and with the inclusion of assembly bias, that generalize further the standard HOD framework. Since our primary goal is to test and validate LRG analysis pipelines under a common set of high-resolution mocks sharing similar clustering properties, rather than improve the HOD modelling, for this study we select a few representative galaxy–halo connection schemes from the plethora of HODs available in the literature: the inclusion of further models that go beyond the more conventional HOD formulation is left to future studies. We also exploit a small homogeneous set of cut-sky mocks (the NSERIES) – which has been previously used in the SDSS DR12 galaxy clustering analysis (Alam et al. 2017) – to address the impact of cosmic variance and related theoretical systematics, and make use of a new series of DR16 EZMOCKS (Zhao et al. 2020) for determining the rescaled covariance matrices functional to all the analyses. The mock-making procedure is explained in detail in Section 4.

By confronting the different BAO and RSD LRG fitting techniques on a common ground against a subset of those high-fidelity mocks having different HOD prescriptions, we are thus able to assess their performance, quantify the systematic errors on the AP parameters and the growth of structure, and eventually confirm the effectiveness of the LRG analysis pipelines. In particular, we anticipate that we find all the methods mutually consistent, and robust to different HOD prescriptions, validating the models used for the LRG clustering analysis.

Furthermore, the mock challenge developed here is suitable to a number of applications. Beside being directly useful for the final eBOSS DR16 ‘consensus cosmology’ (eBOSS Collaboration 2020), as the systematic error budget for the ultimate $f\sigma_8$ constraint are informed by testing the results of analyses against these high-resolution mocks, our work may be relevant for future large-volume surveys. For example, similar mock-making techniques and systematic corrections can be readily extended to model the Dark Energy Spectroscopic Instrument (DESI; DESI Collaboration 2016) and the Large Synoptic Survey Telescope (LSST; Ivezić et al. 2019) galaxy samples.

The layout of the paper is organized as follows. Section 2 briefly describes the eBOSS DR16 data release, and the final LRG sample. Section 3 provides the theoretical foundation for modelling the galaxy–halo connection, and explains the different HOD schemes adopted in the mock-making procedure – along with the rationale behind our choices. Section 4 describes the tools and methodology used to construct high-fidelity mocks; the expert reader may wish to jump directly to the next sections, while a reader less familiar with HOD modelling could benefit from these parts, without the need of consulting extensive literature works. Section 5 briefly presents the different RSD models, that are described in depth in companion papers. Section 6 shows selected results from the mock challenge, and compares the various LRG BAO and RSD models in configuration and Fourier space. Section 7 presents the global error budget for the completed eBOSS DR16 LRG sample, with a primary focus on theoretical systematics. Finally, we conclude in Section 8, where we summarize the main findings and indicate future avenues. We

¹A description of eBOSS and a link to its associated publications can be found at this URL: <https://www.sdss.org/surveys/eboss/>.

leave in Appendix A an extensive description of all of the mock products developed and publicly released with this study, and report in Appendix B some useful tables.

Throughout the paper and if not specified otherwise, all numerical values of length and mass are understood to be in $h = 1$ units.

2 SDSS-IV EBOSS AND DR16 LRG SAMPLE

2.1 SDSS legacy and eBOSS

SDSS observations, carried out on the 2.5-m Sloan Foundation telescope at Apache Point Observatory (Gunn et al. 2006), first begun in 2014 July (SDSS-I and SDSS-II; York et al. 2000). Since then, thanks to the remarkable efforts of more than 10 yr of operations, the survey has evolved till its current fourth generation (SDSS-IV), collecting an increasing number of high-quality data for high-precision cosmology – outperforming on the targets that drove the initial survey design. eBOSS, a key component of the SDSS-IV and ranked in the highest tier in the 2018 DOE-HEP Portfolio Review, is a continuation of the Baryon Oscillation Spectroscopic Survey (BOSS) – part of the SDSS-III (Eisenstein et al. 2011) – and a pre-cursor for DESI (DESI Collaboration 2016). eBOSS lies at the leading edge of cosmological experimentation: by spectroscopically targeting four distinct astrophysical populations in a unique redshift interval, eBOSS has built the largest volume and most complete map of the Universe to date of any redshift survey. The primary innovation in eBOSS is extending BAO measurements with ELGs and a much larger number of quasars, enabling a per-cent-level measurement in the critical epoch of transition from deceleration to acceleration (i.e. $0.8 < z < 2.2$). This is why the eBOSS data set allows exploration of DE in epochs where no precision cosmological measurements currently exist (improving the DE Figure of Merit by a factor of 3), by addressing three Particle Physics Project Prioritization Panel (P5) science drivers and pursuing four key goals: BAO measurements of the Hubble parameter and distance as a function of z , RSD measurements of the gravitational growth of structure, constraints on the neutrino mass sum, and constraints on inflation through measurements of primordial non-Gaussianity. In particular, the exquisite BAO and RSD measurements that eBOSS provide (see e.g. eBOSS Collaboration 2020) are key for DE and gravity studies. Moreover, eBOSS has the spectroscopic capabilities to complement and enhance other current and future cosmological probes, representing a strategic asset and a pathfinder for upcoming experiments.

2.2 The eBOSS DR16 LRG sample

The LRG spectroscopic sample allowed the first SDSS detection of the BAO peak in the galaxy large-scale correlation function (Eisenstein et al. 2005). Since its original version, comprised by 46 748 LRGs over 3816 deg^2 at $0.16 < z < 0.47$, the SDSS LRG catalogue has considerably grown both in size and redshift depth, thanks to over about a decade of observations. In particular, BOSS was designed to measure BAOs with LRGs over the redshift range $0.2 < z < 0.75$, while eBOSS increases the redshift coverage up to $z = 1$. With the final eBOSS DR16, completed on 2019 March 1, the LRG eBOSS-only released sample contains 174 816 galaxies with good redshifts in the interval range $0.6 < z < 1.0$, with an effective redshift $\bar{z} = 0.698$, spanning a total area of 4104 deg^2 and an effective volume of 1.241 Gpc^3 . LRG targets were selected via optical and infrared imaging over 7500 deg^2 angular footprint, using photometry with updated calibration (Dawson et al. 2016): full details of LRG selections are provided in Prakash et al. (2016). To this

end, Bautista et al. (2018) recently demonstrated that the sample is well-suited for LSS studies. The final DR16 eBOSS-only LRG sample is combined with the high redshift tail of the BOSS galaxy sample (denoted as CMASS), in order to provide one catalogue of luminous galaxies with $z > 0.6$. Overall, BOSS CMASS galaxies make up slightly more than half of the total sample, and the area they occupy is more than twice that of eBOSS LRGs – over an effective volume of 1.445 Gpc^3 , hence the total effective volume of the combined DR16 LRG sample is 2.654 Gpc^3 . Most of the BOSS CMASS footprint was re-observed by the eBOSS LRG program, which covered 37 per cent and 65 per cent of the original Northern Galactic Cup (NGC) and Southern Galactic Cup (SGC) CMASS areas, respectively. The projected number density of galaxies with $0.6 < z < 1.0$ is more than twice as high for the eBOSS LRGs (44 deg^{-2} compared to 21 deg^{-2}).

Regarding redshift assignment, a different philosophy both for redshift estimates and spectral classification has been designed specifically for the eBOSS clustering catalogues, motivated by new challenges due to low signal-to-noise eBOSS galaxy spectra. In fact, previous routines used for BOSS were not optimized for the fainter and higher redshift LRG galaxies that comprise the eBOSS LRG sample, and therefore a new approach and software development to provide accurate redshift estimation (indicated as REDROCK²) was necessary. As a result, the redshift completeness approaches 98 per cent for the eBOSS LRG sample with a rate of ‘catastrophic failures’ estimated to be less than 1 per cent – hence such redshift failures are not a concern for the LRG sample.

About sector completeness (a sector being an area covered by a unique set of plates), for the eBOSS LRG sample the 100 per cent completeness requirement was relaxed, to increase the fibre efficiency and total survey area. To this end, the completeness of the eBOSS LRG sample exceeds 95 per cent in every relevant chunk (i.e. an area tiled in a single software run) of the survey, where completeness statistics are determined on a per-sector basis.

A technical description of the LRG observational strategy, and on how spectra are turned into redshift estimates, can be found in Ross et al. (2020). Extensive details on the LRG catalogue creation, observing strategy, matching targets and spectroscopic observations, veto masks etc., as well as observational effects such as varying completeness, collision priority, close pairs, redshift failures, systematics related to imaging and their correction are also presented in Ross et al. (2020).

3 MODELLING THE GALAXY–HALO CONNECTION: THEORETICAL BACKGROUND

In this section, we provide a concise overview of the theoretical formalism underlying our mock-making procedure, in an increasing level of complexity. Starting from the most conventional HOD approach, we then consider more sophisticated scenarios able to distinguish between quiescent or star-forming galaxies, and with the inclusion of assembly bias – that generalize further the standard HOD framework.

3.1 HOD modelling: basics

The Halo Occupation Distribution (HOD; see e.g. Peacock & Smith 2000; Seljak 2000; Scoccimarro et al. 2001; Berlind & Weinberg 2002; Kravtsov et al. 2004; Zheng et al. 2005 for pioneering works,

²See github.com/desihub/redrock

and e.g. Guo et al. 2016; Yuan, Eisenstein & Garrison 2018; Tinker et al. 2019; Alpaslan & Tinker 2020 for more recent implementations and extensions) is a popular framework able to establish a statistical connection between galaxies and dark matter haloes bypassing the complex galaxy formation physics, useful to inform models of galaxy formation, interpret LSS measurements, and eventually constrain cosmological parameters. The core assumption underlying any HOD modelling is that all galaxies reside in dark matter haloes, and that haloes are biased tracers of the dark matter density field. In this regards, knowledge of how galaxies populate, and are distributed within, dark matter haloes enables a complete description of all the statistics of the observed galaxy distribution.

In the most conventional HOD formulation, the central quantity is the probability distribution function (PDF) of galaxies within haloes $P(N_g|M_h)$, namely the probability that a halo of mass M_h hosts – on average – N_g galaxies in a pre-defined sample. Galaxies are further split into centrals and satellites, and conventionally the occupation statistics of central galaxies (N_{cen}) are modelled separately from satellites (N_{sat}), so that:

$$\langle N_g|M_h \rangle = \langle N_{\text{cen}}|M_h \rangle + \langle N_{\text{sat}}|M_h \rangle. \quad (1)$$

In the standard mass-only ‘ansatz’ (i.e. halo bias b_h is only a function of halo mass), central galaxies are commonly assumed to reside at the centre of their host haloes, inheriting the corresponding halo velocity and concentration values, while satellite galaxies are typically designed to follow a radial number density distribution that traces the NFW density distribution (Navarro, Frenk & White 1997) of the underlying dark matter halo. Hence, the starting point for any HOD-style model is choosing an analytic form for $\langle N_{\text{cen}} \rangle$ and $\langle N_{\text{sat}} \rangle$. In the vast majority of HOD studies it is assumed that centrals and satellite HODs are completely uncorrelated, so that $\langle N_{\text{cen}}N_{\text{sat}}|M_h \rangle = \langle N_{\text{cen}}|M_h \rangle \langle N_{\text{sat}}|M_h \rangle$. This means that satellites have no knowledge of the central galaxy occupation of their host halo. Moreover, motivated by the occupation statistics of subhaloes in high-resolution N -body simulations, the PDF of satellite occupation is commonly assumed to be Poissonian, so that $\langle N_{\text{sat}}(N_{\text{sat}} - 1)|M_h \rangle = \langle N_{\text{sat}}|M_h \rangle^2$.

The widespread success of the standard HOD framework in interpreting galaxy clustering statistics, the galaxy–halo connection, and testing cosmology at small scales is not free from several drawbacks. There are in fact a number of simplifying assumptions that enter in the conventional HOD modelling, and may represent a limitation of its predicative power – see for example the recent interesting study by Hadzhiyska et al. (2020).

To start with, while certainly halo mass is the dominant parameter governing the environmental demographics of galaxies, in reality semi-analytical models and hydrodynamical simulations of galaxy formation and evolution predict significant correlations between galaxy properties and halo properties other than mass (i.e. halo formation time, concentration, halo spin, merger history, star formation rate, etc.). Such dependence of the spatial distribution of dark matter haloes upon properties besides mass is generically referred to as halo assembly bias. To date, a clear detection of assembly bias remains still controversial, as some earlier claims of detection in a sample of SDSS galaxy clusters (Miyatake et al. 2016) have been called into significant question – see e.g. Zu et al. (2017). However, some level of assembly bias may be present in the eBOSS LRG sample, as recent studies seem to indicate (Zentner et al. 2019; Obuljen, Percival & Dalal 2020; Yuan et al. 2020). Moreover, it has been shown that ignoring assembly bias in HOD modelling yields constraints on the galaxy–DM connection that may be plagued by significant systematic errors (Yang et al. 2006; Blanton & Berlind 2007; Zentner, Hearin & van den Bosch 2014). In addition, while in standard HOD

studies centrals and satellite HODs are completely uncorrelated, likely, a degree of central–satellite correlation is always present: such correlations are induced by interesting astrophysics, rather than being simply a nuisance systematics. In fact, the correlation encodes the extent to which the properties of satellite galaxies (stellar mass, colour, etc.) may be correlated with the properties of its central galaxy at fixed halo mass (i.e. galactic cannibalism or conformity). Moreover, central galaxies may not be located at the central (minimum) of the halo potential well, the occupation statistics of subhaloes in host haloes of fixed mass has been shown to deviate from a Poisson distribution especially in the limit where the first occupation moment become large, the satellite distribution may not track the NFW spatial profile of the dark matter halo, and much more – see for example Yuan et al. (2018) and the most recent study by Duan & Eisenstein (2019) for an extensive discussion on such challenges.

Within this complex framework, the main goal of our study is to produce a series of synthetic galaxy catalogues spanning a variety of HODs exhibiting similar clustering properties, in order to assess the robustness of different fitting methodologies relevant for LRG clustering. In this respect, the primary focus is not to improve the HOD modelling and the galaxy–halo connection. However, we have devised the mock challenge in an increasing order of HOD complexity by exploring various methodologies, so that our study may be helpful for ameliorating the galaxy–halo connection in future works. Motivated by these reasons, we start from the simplest and most conventional HOD framework, and gradually increase the complexity till considering models with assembly bias, particularly useful in exploring intermediate correlations between central–satellites, as well as more generalized HOD approaches. As a byproduct of our work, we are thus able to draw some interesting conclusions regarding the galaxy–halo connection, based on our high-fidelity mocks.

In this work, unless specified otherwise, we always consider two galaxy populations (referred as centrals and satellites); moreover, as commonly adopted in the most conventional HOD implementations, we generally assume that the central phase space model requires central galaxies to be located at the exact centre of the host halo with the same halo velocity, and that the satellite phase space model follows an unbiased NFW profile with a phase space distribution of mass and/or galaxies in isotropic Jeans equilibrium, where the concentration of galaxies is identical to one of the parent halo.

3.2 Traditional HOD: Zheng model

The most traditional composite HOD model is the one first proposed by Zheng et al. (2007): it represents the backbone for any other HOD framework, and the starting point of this work. The central occupation statistic (N_{cen}) is described by a nearest integer distribution with first moment given by an error function introduced by Zheng et al. (2005), namely³:

$$\langle N_{\text{cen}}(M_h) \rangle = \frac{1}{2} \left[1 + \text{erf} \left\{ \frac{\log(M_h) - \log(M_{\text{min}})}{\sigma_{\log M}} \right\} \right], \quad (2)$$

³Throughout the paper, we implicitly assume base 10 for all the logarithmic notations indicated with \log , namely $\log \equiv \log_{10}$, and drop the understood subscript for clarity of notation.

Table 1. HOD parameters adopted for the Zheng et al. (2007) model, corresponding to different ‘luminosity threshold’ values.

ZHENG MODEL					
Threshold	$\log M_{\min}$	$\sigma_{\log M}$	α	M_0	M_1
Th1 ($M_r = -19$)	11.60	0.26	1.02	11.49	12.83
Std ($M_r = -20$)	12.02	0.26	1.06	11.38	13.31
Th2 ($M_r = -21$)	12.79	0.39	1.15	11.92	13.94

Table 2. HOD parameters for central and satellite galaxies assumed for the calibration of the Leauthaud et al. (2011) model.

Leauthaud model	
Centrals	
$\log [M_{1,0}]$	12.35
$\log [M_{1a}]$	0.30
$\log [M_{0,0}]$	10.72
$\log [M_{0a}]$	0.59
β_0	0.43
β_a	0.18
δ_0	0.56
δ_a	0.18
γ_0	1.54
γ_a	2.52
$\sigma_{\log M_*}$	0.20
Satellites	
α_{sat}	1.0
β_{sat}	0.859
B_{sat}	10.62
β_{cut}	-0.13
B_{cut}	1.47

where M_h is the halo mass, M_{\min} is the characteristic minimal mass for a halo to host a central galaxy above a luminosity threshold,⁴ and $\sigma_{\log M}$ is the rate of transition from $\langle N_{\text{cen}} \rangle = 0$ to $\langle N_{\text{cen}} \rangle = 1$, representing the width of the cutoff profile. Hence, central galaxies are characterized just by two HOD parameters. The satellite occupation statistic $\langle N_{\text{sat}} \rangle$ is represented by a Poisson distribution with first moment given by a power law that has been truncated at the low-mass end (Kraivsov, Nagai & Vikhlinin 2005), and described by three parameters:

$$\langle N_{\text{sat}}(M_h) \rangle = \left(\frac{M_h - M_0}{M_1} \right)^\alpha, \quad (3)$$

where α is the power-law slope of the relation between halo mass and $\langle N_{\text{sat}} \rangle$, M_0 a low-mass cutoff in $\langle N_{\text{sat}} \rangle$, and M_1 is the mass where approximately there is an average of one satellite galaxy per halo, namely $\langle N_{\text{sat}}(M_h = M_1) \rangle \sim 1$ – or more specifically $\langle N_{\text{sat}}(M_h = M'_1) \rangle = 1$ with $M'_1 = M_1 + M_0$. Note that the previous distribution can be optionally modulated by the central distribution $\langle N_{\text{cen}} \rangle$. Redshift has no impact on this model.

Following HALOTOOLS conventions (Hearin et al. 2017; see Section 4.4), the setting of these parameters in our mock-making procedure is controlled by a luminosity threshold, intended as the r -band absolute magnitude of the luminosity of the galaxy sample. The HOD parameters used in our modelling are those of table 2 in Zheng et al. (2007), and conveniently reported in Table 1 as a function of threshold: specifically, we consider three thresholds in this work, referred globally as ‘threshold 1’ (Th1; $M_r = -19$), ‘standard’ (Std;

⁴As pointed out by Zheng et al. (2007), M_{\min} can also be interpreted as the mass of such haloes for which half of them host galaxies above the given luminosity threshold, i.e. $\langle N_{\text{cen}}(M_{\min}) \rangle = 0.5$.

$M_r = -20$), and ‘threshold 2’ (Th2; $M_r = -21$); the latter one is closer to the characteristics of the eBOSS DR16 LRG sample. Since the meaning of the parameter ‘threshold’ in the Zheng model, according to HALOTOOLS conventions, is effectively different from that of the other HOD models considered (based on stellar mass rather than luminosity, thus not direct correspondent because we choose to maintain HOD literature parameters – see again Section 4.4 and Table A2), we opted to keep the Zheng framework separate from the other models in the following presentation, in order to avoid confusion; we also notice that in general stellar mass is a more faithful tracer of the halo mass than galaxy luminosity – see e.g. Leauthaud et al. (2011, 2012), Tinker et al. (2013). However, we reiterate that the five-parameter Zheng HOD framework is the starting point of our work, and results involving the Zheng model can be found in Appendix B (Table B1), as well as in our companion paper Gil-Marín et al. (2020) – see in particular their fig. 11. Simply, in the main analysis presented in Section 6 we have chosen to display only results involving the HOD models discussed next, as their selected HOD parameters for Th2 provide mocks closer in terms of number density to the eBOSS LRG sample (i.e. Table A2), and thus more suitable for our main science targets.

The shapes of the Zheng HODs used in this work are shown in Fig. 1, for the three conventional choices of HOD parameters corresponding to the previously mentioned threshold values (see Table 1). Note also that in our HOD modelling we do not modulate the satellite distribution by the central one, unless specified otherwise.

3.3 Adding the SHMR complexity: Leauthaud model

The second model we consider is the Leauthaud prescription (Leauthaud et al. 2011, 2012), a composite HOD framework that extends the standard Zheng formalism by including a parametrization of an underlying stellar-to-halo mass relation (SHMR), which specifies the mean mass of a galaxy as a function of halo mass. The main assumption in this picture is that the SHMR is valid only for central galaxies, as satellites and centrals experience distinct stellar growth rates and so it is necessary to model them separately. To this end, the conditional stellar mass function (CSMF) – denoted as $\Phi(M_*|M_h)$, with M_* the mass in stars and M_h the mass of the parent halo – is divided into centrals and satellites, namely $\Phi(M_*|M_h) = \Phi_{\text{cen}}(M_*|M_h) + \Phi_{\text{sat}}(M_*|M_h)$. Moreover, $\Phi_{\text{cen}}(M_*|M_h)$ is modelled stochastically as a lognormal PDF with a lognormal scatter $\sigma_{\log M_*}$, and normalized to unity. The exact form is (Leauthaud et al. 2012):

$$\Phi_{\text{cen}}(M_*|M_h) = \frac{1}{\ln(10)\sigma_{\log M_*}\sqrt{2\pi}} \cdot \exp \left[-\frac{\{\log M_* - \log [f_{\text{SHMR}}(M_h)]\}^2}{2\sigma_{\log M_*}^2} \right], \quad (4)$$

where f_{SHMR} is the logarithmic mean of the stellar mass given the halo mass for the Φ_{cen} distribution function. Equation (4) incorporates the scatter associated with the determination of stellar masses, as well as the intrinsic scatter in stellar mass at fixed halo mass due to astrophysical processes. The functional form for f_{SHMR} is described by five shape parameters ($M_0, M_1, \beta, \delta, \gamma$), and defined via its inverse function as (Behroozi et al. 2010):

$$\begin{aligned} \log [f_{\text{SHMR}}^{-1}(M_*)] &= \log(M_1) + f_{\text{low}}(M_*/M_0) + f_{\text{high}}(M_*/M_0) \\ &= \log(M_1) + \beta \log \left(\frac{M_*}{M_0} \right) + \\ &\quad \frac{(M_*/M_0)^\delta}{1 + (M_*/M_0)^{-\gamma}} - \frac{1}{2}, \end{aligned} \quad (5)$$

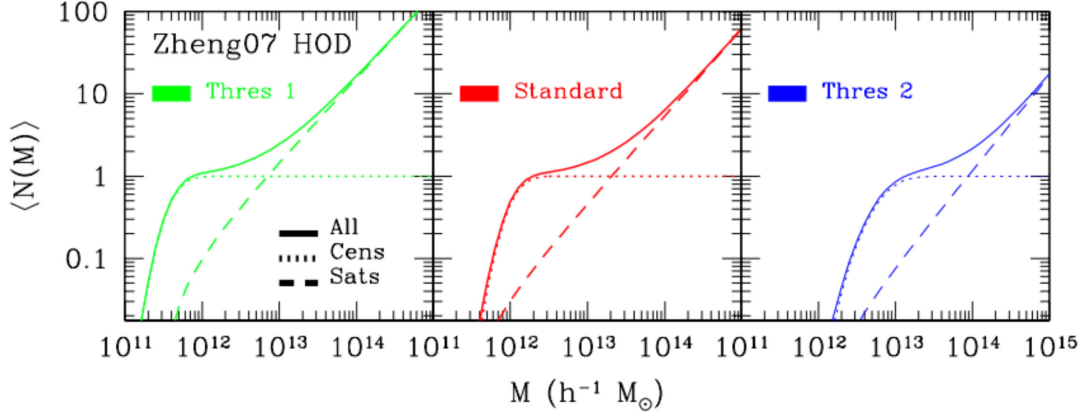


Figure 1. HOD shapes in the Zheng, Coil & Zehavi (2007) model, used for the production of galaxy mocks. In the various panels, dotted lines describe the central occupation statistics (equation 2), dashed lines are used for the satellite occupation statistics (equation 3), and solid lines represent the composite HODs. Three luminosity thresholds are considered, corresponding to different HOD parameter choices, as reported in Table 1. See the main text for more details.

with f_{low} and f_{high} the low and high mass parts of the SHMF, respectively. In the previous expression, M_0 and M_1 are characteristic stellar and halo masses – respectively – in the $\langle M_* \rangle(M_h)$ map, β is a low-mass slope of the $\langle M_* \rangle(M_h)$ map, δ is the high-mass slope of the same map, and γ represents the transition between the low- and high-mass behaviour of the $\langle M_* \rangle(M_h)$ mapping. The redshift evolution of the SHMR is modelled by allowing the parameters that define f_{SHMR} in equation (5) to vary linearly with the scale factor a as:

$$\log[M_1(a)] = \log(M_{1,0}) + \log(M_{1,a})(a - 1) \quad (6)$$

$$\log[M_0(a)] = \log(M_{0,0}) + \log(M_{0,a})(a - 1) \quad (7)$$

$$\beta(a) = \beta_0 + \beta_a(a - 1) \quad (8)$$

$$\delta(a) = \delta_0 + \delta_a(a - 1) \quad (9)$$

$$\gamma(a) = \gamma_0 + \gamma_a(a - 1). \quad (10)$$

Therefore, the model for the SHMR effectively requires 10 parameters. For our modelling, we adopt the same literature values as in the second column of Table 2 in Behroozi et al. (2010) at $z = 0$, and also reported in the upper part of Table 2 for convenience. Furthermore, equation (4) requires a functional form for the lognormal scatter $\sigma_{\log M_*}$ in the SHMR: motivated by Leauthaud et al. (2012), who found that a halo mass-varying scatter produced no better fit than a model with constant scatter, we assume a constant scatter in this work – which can be thought as the sum in quadrature of an intrinsic component plus a measurement error component. Specifically, we set $\sigma_{\log M_*} = 0.20$ in our modelling for all the redshifts considered.

The top panel of Fig. 2 provides an example of the effects of varying in turn the five main parameters that control the SHMF (equation 5) at $z = 0$ by 10 percent and up to 50 percent – as specified in the plot with different line styles and colours, when the lognormal scatter is kept constant. The baseline parameters are fixed as in Behroozi et al. (2010) at $z = 0$ (solid black line, and upper part in Table 2). The bottom panel of the same figure displays the SHMF underlying the Leauthaud model at $z = 0$, along with its redshift evolution for the two main redshift snapshots considered in our mock-making procedure ($z = 0.695$ and $z = 0.865$, respectively – see Section 4): we use these SHMRs in our modelling.

The key difference with respect to the Zheng model relies in the assumption that the stellar mass, rather than the galaxy luminosity, is

used to implement the HOD as a more reliable tracer of the halo mass – see e.g. Leauthaud et al. (2011, 2012), Tinker et al. (2013). To this end, for a volume-limited sample of galaxies such that $M_* > M_*^{\text{thr}}$, with M_*^{thr} a galaxy threshold mass, the central occupation function $\langle N_{\text{cen}}(M_h | M_*^{\text{thr}}) \rangle$ is fully specified given $\Phi_{\text{cen}}(M_* | M_h)$ according to (Leauthaud et al. 2011):

$$\langle N_{\text{cen}}(M_h | M_*^{\text{thr}}) \rangle = \int_{M_*^{\text{thr}}}^{\infty} \Phi_{\text{cen}}(M_* | M_h) dM_*. \quad (11)$$

Assuming that $\sigma_{\log M_*}$ is constant, the previous expression can be readily integrated and becomes:

$$\langle N_{\text{cen}}(M_h | M_*^{\text{thr}}) \rangle = \frac{1}{2} \left[1 - \text{erf} \left\{ \frac{\log(M_*^{\text{thr}}) - \log[f_{\text{SHMR}}(M_h)]}{\sqrt{2}\sigma_{\log M_*}} \right\} \right]. \quad (12)$$

Equation (12) represents a generalization of the Zheng HOD formula (equation 2), and it is controlled by five parameters that enter in the SHMR – plus 1, if we allow for a varying scatter in the SHMR, and plus additional five parameters if we also consider redshift evolution in the SHMR. Interestingly, equation (2) – i.e. the Zheng model – can be readily recovered from (11) as a limiting case by assuming a constant scatter in the SHMR and by setting f_{SHMR} to be a power law; however, this latter assumption is not realistic.

Regarding the satellite occupation function, in the Leauthaud model it is parametrized as a power law of host mass with an exponential cutoff, and can be optionally scaled by $\langle N_{\text{cen}} \rangle$ (this is not done in our case). Specifically:

$$\langle N_{\text{sat}}(M_h | M_*^{\text{thr}}) \rangle = \left(\frac{M_h}{M_{\text{sat}}} \right)^{\alpha_{\text{sat}}} \exp \left(- \frac{M_{\text{cut}}}{M_h} \right), \quad (13)$$

where M_{sat} defines the amplitude of the power law and M_{cut} sets the scale of the exponential cutoff; haloes with masses $M_h < M_{\text{cut}}$ are extremely unlikely to host a satellite galaxy. Instead of simply modelling M_{sat} and M_{cut} as constant factors of $f_{\text{SHMR}}^{-1}(M_*^{\text{thr}})$, flexibility is added by enabling M_{sat} and M_{cut} to vary as power-law functions of $f_{\text{SHMR}}^{-1}(M_*^{\text{thr}})$:

$$\frac{M_{\text{sat}}}{M_{12}} = B_{\text{sat}} \left[\frac{f_{\text{SHMR}}^{-1}(M_*^{\text{thr}})}{M_{12}} \right]^{\beta_{\text{sat}}} \quad (14)$$

$$\frac{M_{\text{cut}}}{M_{12}} = B_{\text{cut}} \left[\frac{f_{\text{SHMR}}^{-1}(M_*^{\text{thr}})}{M_{12}} \right]^{\beta_{\text{cut}}}, \quad (15)$$

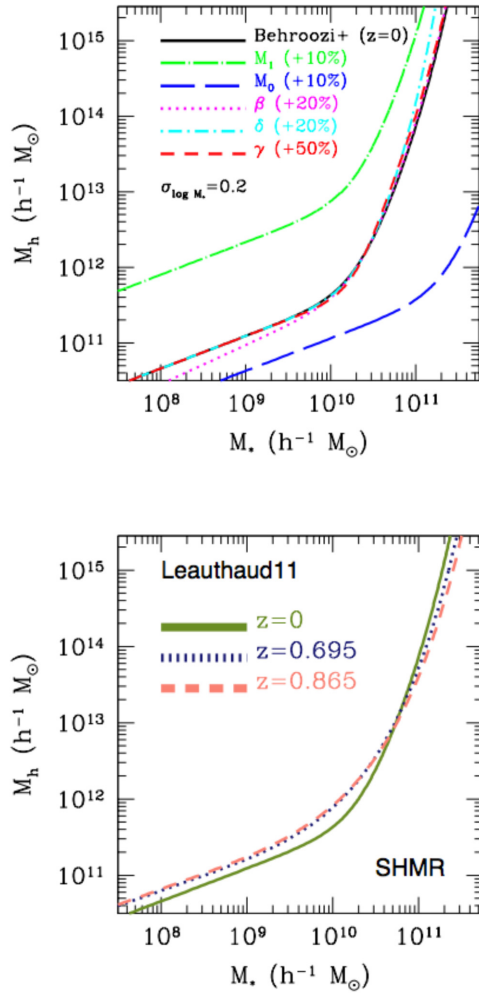


Figure 2. [Top] Effects of varying the main parameters controlling the SHMF for central galaxies (equation 5), modelled as a mean–log relation in the Leauthaud model and heavily based on Behroozi, Conroy & Wechsler (2010). The key shape parameters are altered in turn by 10 per cent (M_0 , M_1), 20 per cent (β , δ), and 50 per cent (γ), respectively, from the baseline Behroozi et al. (2010) model at $z = 0$ (solid black line and upper part in Table 2), when $\sigma_{\log M_*} = 0.20$. Different line styles and colors refer to such variations, as clearly indicated in the panel. [Bottom] SHMF underlying the Leauthaud model at $z = 0$, and its redshift evolution at $z = 0.695$ and $z = 0.865$: we use these SHMRs in our mock-making procedure.

where $\bar{M}_{12} = 10^{12} M_\odot$. Hence, satellite occupation statistics, independent of binning schemes and parametrized with threshold samples, are modelled by five parameters: α_{sat} , β_{sat} , B_{sat} , β_{cut} , B_{cut} . In detail, α_{sat} is the power-law slope of the relation between halo mass and the satellite mean occupation function $\langle N_{\text{sat}} \rangle$, β_{sat} and B_{sat} control the amplitude of the power-law slope of $\langle N_{\text{sat}} \rangle$, and β_{cut} and B_{cut} control the low-mass cut off in $\langle N_{\text{sat}} \rangle$. These satellites parameters are fixed as in table 5 of Leauthaud et al. (2011) for the first redshift bin, and also reported in the bottom part of Table 2 for convenience.

In summary, the Leauthaud framework is completely determined by 11 HOD parameters: six controlling the central galaxies and five for satellite galaxies, plus additional five if we also take into account the redshift evolution of the SHMR. In this framework, ‘threshold’ should be intended as the minimal stellar mass of the galaxy sample, rather than galaxy luminosity.

The shapes of the HODs in the Leauthaud model adopted in this work are shown in the top panels of Fig. 4 for three thresholds in mass, indicated as ‘Threshold 1’ (Th1; $M_*^{\text{thr}} = 10^{10} h^{-1} M_\odot$), ‘Standard’ (Std; $M_*^{\text{thr}} = 10^{10.5} h^{-1} M_\odot$), and ‘Threshold 2’ (Th2; $M_*^{\text{thr}} = 10^{11} h^{-1} M_\odot$), when $z = 0.695$. In particular, Th2 is one closer to the eBOSS LRG sample, and we mainly focus on this mass interval in our analysis. Note finally that in the Leauthaud framework $\langle N_{\text{sat}}(M_h | M_*^{\text{thr}}) \rangle$ depends on $\langle N_{\text{cen}}(M_h | M_*^{\text{thr}}) \rangle$, indicating that in this particular model the occupation statistics of centrals and satellites are correlated: this constitutes an important difference with respect to the traditional five-parameter HOD. In fact, as pointed out by Contreras et al. (2017) and Duan & Eisenstein (2019), if the goal is not to constrain HOD parameters and this fitting difficulty is not of concern (as in our specific case), there is no advantage or motivation for insisting on no correlations between centrals and satellites.

3.4 HOD with colour/SFR: Tinker model

The next HOD framework we consider has been introduced by Tinker et al. (2013): it is an extension of the Leauthaud formalism previously described, to samples defined by both stellar mass and star formation (SF) activity. In this context, galaxies can be roughly categorized into the star-forming sequence of blue, discy, gas-rich galaxies, and the quiescent, ellipsoidal galaxies with old stellar populations and red colours: the bimodality is firmly in place at $z = 1$ (Tinker et al. 2013, 2019). Therefore, in this model galaxies are divided into quiescent, to indicate galaxies that have little to no star formation and are intrinsically located on the red sequence, and the set of star-forming galaxies. Hence, the Tinker model represents a minimal modification of the Leauthaud prescription to adapt it to passive and SF subsamples of galaxies. The HOD behaviour is in fact governed by an assumed underlying SHMR as first introduced in Behroozi et al. (2010), but that is now distinct for star-forming and quiescent populations: each subsample will then have a separate f_{SHMR} , with two different sets of HOD parameters related to their specific SHMRs. A constant scatter $\sigma_{\log M_*}$ in the SHMR is adopted here, but the scatter is different and independent for passive and SF central galaxies. The main difference with respect to the Leauthaud formalism is the following requirement:

$$\int \left\{ f_q(M_h) \times \Phi_{\text{cen}}^q(M_* | M_h) + [1 - f_q(M_h)] \times \Phi_{\text{cen}}^{\text{SF}}(M_* | M_h) \right\} dM_* = 1, \quad (16)$$

where $f_q(M_h)$ is a function specifying the fraction of times that a halo of mass M_h contains a quenched central galaxy (independent of galaxy mass), and Φ_{cen}^q is the conditional stellar mass function for central quiescent or star-forming galaxies, each normalized to unity. The function $f_q(M_h)$ does not have a parametric form, but five halo mass points are chosen at which to specify $f_q(M_h)$ and smoothly interpolate between them, where the five masses are evenly spaced in $\log M_h$. Moreover, to avoid explicit dependencies of HOD parameters on bin sizes, all HODs are defined as threshold quantities, which provides maximal flexibility.

Fig. 3 shows the SHMFs adopted in the Tinker model, as well as in our mock-making procedure, along with their redshift evolution: solid lines refer to $z = 0$, dotted and dashed lines are at $z = 0.695$ and $z = 0.865$, respectively. Active galaxies are represented in blue, while quiescent galaxies are displayed in brown.

As the Tinker model inherits almost all the features and methods of the Leauthaud framework, the HOD for centrals is the same as in equation (12), but the parameters of the f_{SHMR} are independent for each subsample: this is indeed an important aspect that clearly differentiates the two models. Moreover, for red central galaxies,

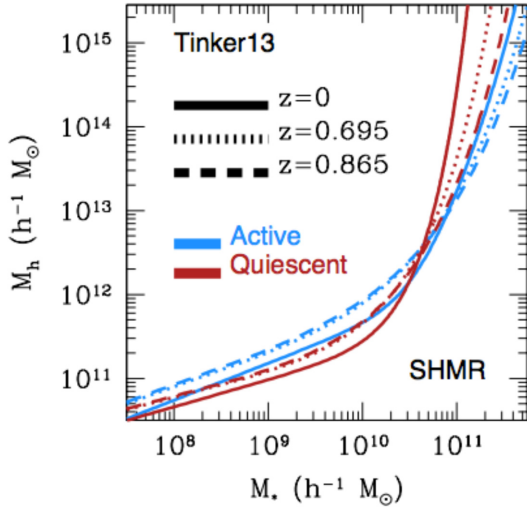


Figure 3. SHMFs for central galaxies adopted in the Tinker model and in our mock-making procedure at $z = 0$ (solid lines), $z = 0.695$ (dotted lines), and $z = 0.865$ (dashed lines). Active galaxies are displayed in blue, while quiescent galaxies are represented in brown.

the HOD is multiplied by $f_q(M_h)$, and by $1 - f_q(M_h)$ for SF central galaxies.

The occupation statistics of satellite galaxies as a function of halo mass are similar to those of Leauthaud et al. (2011), although the satellite occupation of passive and star-forming galaxies subsamples are treated independently. Hence, a modification is introduced in order to produce a proper cutoff scale by including f_{SHMR}^{-1} to the numerator in the exponential cutoff, so that:

$$\langle N_{\text{sat}}(M_h | M_*^{\text{thr}}) \rangle = \left(\frac{M_h}{M_{\text{sat}}} \right)^{\alpha_{\text{sat}}} \exp \left(- \frac{[M_{\text{cut}} + f_{\text{SHMR}}^{-1}(M_*)]}{M_h} \right). \quad (17)$$

This guarantees that satellite occupation fully cuts off at the same halo mass scale as central galaxies of the same mass. In addition, while in Leauthaud et al. (2011) $\alpha_{\text{sat}} = 1$, here the fraction of satellites that are star forming depends on halo mass, so $\alpha_{\text{sat}} = 1$ is allowed to be free for both passive and star-forming subsamples.

In summary, the Tinker model is characterized by 27 free parameters: 11 are needed for the composite HOD of a given subsample (5 for the central SHMR, one additional for the SHMR scatter, plus 5 for the satellite occupation statistics), and 5 pivot points are necessary to specify $f_q(M_h)$. Each set of 27 parameters describes the galaxy–halo relation at a given redshift, and clearly additional quantities are required to characterize the redshift evolution. In this work, we adopt literature values from the lowest redshift bin in table 2 of Tinker et al. (2013), as reported in Table 3; we note that adopting these parameters makes the model differing from that of Leauthaud, even when the full HOD shape is considered.

The central panels of Fig. 4 show the shapes of the HODs in the Tinker model adopted in this work at $z = 0.695$, for the same three thresholds in mass described before in relation to the Leauthaud framework, and also distinguishing between centrals and satellites. Active and quiescent galaxy HODs are represented by different colours, as indicated in the panels, and the global HODs are also displayed. As noted by Tinker et al. (2013) and also evident from our figures, the number of quiescent satellites exhibits minimal redshift evolution; all evolution in the red sequence is due to low-mass central galaxies being quenched of their star formation. Moreover, the efficiency of quenching star formation for centrals increases with

Table 3. HOD parameters for central and satellite galaxies assumed for the calibration of the Tinker et al. (2013) model.

TINKER MODEL	
Centrals	
$\log[M_{1,0,\text{active}}]$	12.56
$\log[M_{1,0,\text{quiescent}}]$	12.08
$\log[M_{0,0,\text{active}}]$	10.96
$\log[M_{0,0,\text{quiescent}}]$	10.70
$\beta_{0,\text{active}}$	0.44
$\beta_{0,\text{quiescent}}$	0.32
$\delta_{0,\text{active}}$	0.52
$\delta_{0,\text{quiescent}}$	0.93
$\gamma_{0,\text{active}}$	1.48
$\gamma_{0,\text{quiescent}}$	0.81
$\sigma_{\log M_*,\text{active}}$	0.21
$\sigma_{\log M_*,\text{quiescent}}$	0.28
Satellites	
$\alpha_{\text{sat, active}}$	0.99
$\alpha_{\text{sat, quiescent}}$	1.08
$\beta_{\text{sat, active}}$	1.05
$\beta_{\text{sat, quiescent}}$	0.62
$B_{\text{sat, active}}$	33.96
$B_{\text{sat, quiescent}}$	17.90
$\beta_{\text{cut, active}}$	0.77
$\beta_{\text{cut, quiescent}}$	− 0.12
$B_{\text{cut, active}}$	0.28
$B_{\text{cut, quiescent}}$	21.42

cosmic time, while the mechanisms that quench the star formation of satellite galaxies in groups and clusters is losing efficiency.

3.5 Decorated HOD: Hearin model

The fourth galaxy–halo prescription we consider is the Hearin model (Hearin et al. 2016), a decorated HOD framework designed to account for galaxy assembly bias, that naturally extends the standard HOD approach, minimally expands the parameter space, and maximizes the independence between traditional and novel HOD parameters. Galaxy assembly bias, namely the correlation between galaxy properties and halo properties at fixed halo mass, is a challenging and yet important ingredient in the galaxy–halo connection framework. The model builds on early work by Wechsler et al. (2006). The halo occupation statistics are described in terms of two halo properties rather than just one, and the extra degree of freedom has relevant impact on galaxy clustering. The formalism of the model is general and flexible, with parametric freedom, and it can be applied to any halo property in addition to halo mass; it is also readily extendable to describe HODs that depend upon numerous additional halo properties. Interestingly, the decorated HOD formalism allows one to characterize and quantify the degree of central–satellite correlation at fixed halo mass, which is an indication of compelling astrophysics – such as galactic cannibalism or conformity. We refer the reader to Hearin et al. (2016) for extensive modelling details, and report here only a few key aspects relevant to our work – see also, e.g. Xu, Zehavi & Contreras (2020) and Xu & Zheng (2020) for recent studies.

In particular, the core idea is based on the principle of ‘HOD conservation’, which preserves the moments of the standard HOD formalism: namely, it is required that the marginalized moments of a new decorated framework are equal to those of the standard HOD model, in order to minimize the modifications needed for assembly bias. In this regard, any model $P_{\text{dec}}(N_{\text{gl}} | M_h, x)$ with marginalized

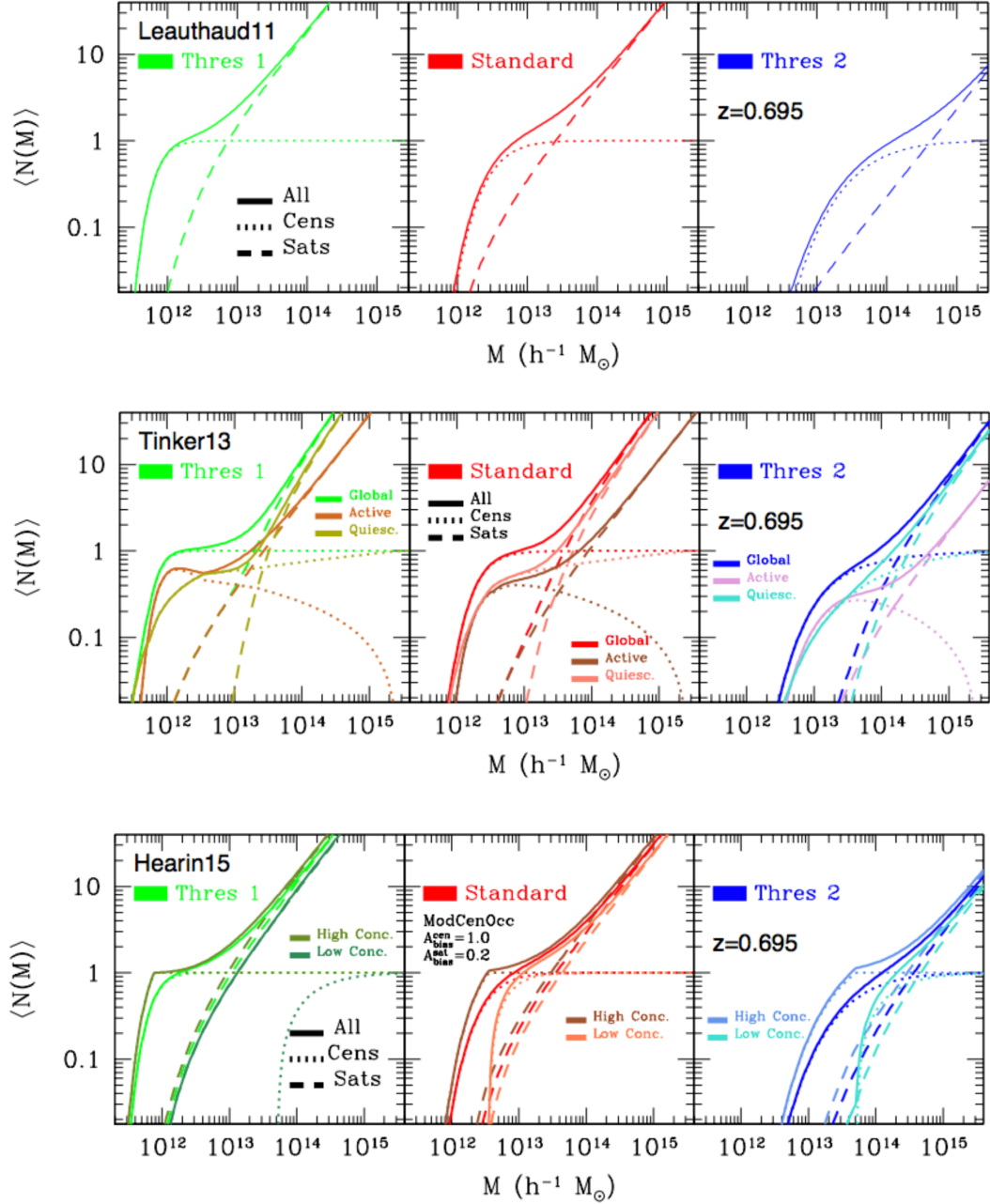


Figure 4. HOD shapes adopted in our mock-making procedure, at $z = 0.695$, for three thresholds in mass, denoted as ‘Thres 1’ ($M_*^{\text{thr}} = 10^{10} h^{-1} M_{\odot}$), ‘Standard’ ($M_*^{\text{thr}} = 10^{10.5} h^{-1} M_{\odot}$), and ‘Thres 2’ ($M_*^{\text{thr}} = 10^{11} h^{-1} M_{\odot}$). Top panels display the various HODs in the Leauthaud model. Central panels show the Tinker model, where active and quiescent galaxy HODs are represented by different colours, as indicated in the figure. Bottom panels are for the Hearin HODs, where galaxies are split into upper- and lower-percentiles in terms of halo concentration, respectively, with different assembly bias strength for centrals and satellites ($A_{\text{bias}}^{\text{cen}} = 1.0$ and $A_{\text{bias}}^{\text{sat}} = 0.2$). In all the plots, the central occupation statistics is displayed with dotted lines, the satellite occupation statistics with dashed lines, and the global HOD shapes with solid lines.

moments that satisfy the HOD conservation preserves the moments of $P_{\text{std}}(N_g | M_h)$, where P_{std} is the occupation statistics of a standard HOD, P_{dec} is the occupation statistics of a decorated HOD model, and x represents a secondary halo property such that the HOD depends both on x and M_h , and the clustering of haloes depends upon x . Within this formalism, a standard HOD is recovered in the limiting condition that the strength of the assembly bias is zero.

For central galaxies, in order to construct decorated HOD models that preserve the full $P_{\text{std}}(N_{\text{cen}} | M_h)$ one just needs to ensure that the first-order decoration function δN_{cen}^1 satisfies the integral relation

(Hearin et al. 2016):

$$\int \delta N_{\text{cen}}^1(M_h, x) P(x | M_h) dx = 0. \quad (18)$$

For satellites the situation is more complex, as it is not possible to conserve the HOD under the assumption that both $P(N_{\text{sat}} | M_h)$ and $P(N_{\text{sat}} | M_h, x)$ obey Poisson statistics – as typically done: intuitively, this is because there is an additional source of variance associated with the allocation of satellites into sub-populations at a given halo mass. Moreover, under HOD conservation, the average number of

central–satellite pairs in massive haloes for a decorated model is identical to that of its standard baseline model, except for the narrow range in halo masses for which $0 \leq \langle N_{\text{cen}} | M_{\text{h}} \rangle \leq 1$.

For our purposes, we consider the Hearin model in its simplest formulation, by assuming two discrete halo subpopulations with different occupation statistics at fixed mass; this is essentially a perturbation of the Leauthaud et al. (2011) formalism, with the addition of assembly bias both in centrals and satellites. We choose the halo NFW concentration as the secondary halo property (x) used to modulate the assembly bias. Specifically, the first halo subpopulation (indicated as ‘type 1’ haloes) contains a fraction P_1 of all haloes at fixed mass, for which $x > \bar{x}(M_{\text{h}})$; the second subpopulation (‘type 2’ haloes) contains $P_2 = 1 - P_1$ of all haloes at fixed mass, for which $x < \bar{x}(M_{\text{h}})$. The halo population is split into the P_1 percentile of highest concentration haloes, and assigned a satellite galaxy occupation enhancement, while the remaining $P_2 = 1 - P_1$ percentile of lowest concentration haloes receive a satellite galaxy occupation decrement. Essentially, we require haloes at fixed mass above- or below-average concentration to have above- or below-average mean occupation. For simplicity, we assume a 50/50 split at each halo mass based on the conditional secondary percentiles: haloes within the top 50 per cent of concentration at fixed M_{h} are assigned to the first subpopulation, and the remaining to the second population (so $P_1 = P_2 = 0.5$). The strength of assembly bias in the occupation statistics of centrals and satellite galaxies is modulated with two free parameters $A_{\text{bias}}^{\text{cen}}$ and $A_{\text{bias}}^{\text{sat}}$, respectively, where $-1 \leq A_{\text{bias}}^{\text{cen}} \leq 1$ and $-1 \leq A_{\text{bias}}^{\text{sat}} \leq 1$. With this choice, a positive value for A_{bias} implies that haloes with above-average concentration have boosted galaxy occupations; note also that more positive values of A_{bias} correspond to models in which more concentrated haloes host more galaxies relative to less concentrated haloes of the same mass. When both of these parameters are set to zero, the model is formally equivalent to the baseline ‘no assembly bias model’ of Leauthaud et al. (2011). We consider a constant assembly bias strength at all masses for simplicity, and the sign convention is to choose type-1 haloes in the upper percentile of the secondary property. Moreover, we assume that both $P_{\text{std}}(N_{\text{sat}} | M_{\text{h}})$ and $P_{\text{dec}}(N_{\text{sat}} | M_{\text{h}}, x)$ are Poisson distributions, so that the decorated HOD is entirely specified by $A_{\text{bias}}^{\text{cen}}$ and $A_{\text{bias}}^{\text{sat}}$.

In our mock-making procedure, we consider two cases for the strength of assembly bias related to centrals and satellites: in the first case (more conservative), we simply set $A_{\text{bias}}^{\text{cen}} = A_{\text{bias}}^{\text{sat}} = 0.5$, namely the strength of assembly bias is equal for both centrals and satellites, with the boost to their mean occupation equal to 50 per cent of the maximum allowable strength at each mass; in the second case (less conservative), we set different assembly bias strengths for centrals and satellites, namely $A_{\text{bias}}^{\text{cen}} = 1.0$ and $A_{\text{bias}}^{\text{sat}} = 0.2$. This latter choice is shown in the lower panels of Fig. 4, where we display the shapes of the Hearin HODs for the upper- and lower-percentile split in halo concentration, respectively, as indicated in the plot. In this case, the satellite HODs are also modulated by their corresponding central distributions. The same three thresholds in mass described before for the Leauthaud framework are adopted, at $z = 0.695$, and as usual we display the central occupation statistics (dotted lines), satellite occupation statistics (dashed lines), as well as the global HOD shapes (solid lines).

As noted by Hearin et al. (2016, 2017) and Tinker et al. (2019), assembly bias can enhance or diminish the clustering on large scales, but in general it increases the clustering on scales below Mpc – being qualitatively different at large and small scales. Also, assembly bias in satellites versus centrals imprints a distinct signature on galaxy clustering as well as lensing, and the degree to which assembly

bias alters galaxy clustering statistics can be quite sensitive to the underlying baseline mass-only HOD of the galaxy population under consideration. In particular, the impact of assembly bias on galaxy clustering is quite sensitive to the steepness of the transition from $\langle N_{\text{cen}} | M_{\text{h}} \rangle_{\text{std}} = 0$ at low host masses to $\langle N_{\text{cen}} | M_{\text{h}} \rangle_{\text{std}} = 1$ at high host masses. This steepness is controlled by the level of stochasticity in the central galaxy stellar mass at fixed halo mass, parametrized in our baseline model by $\sigma_{\log M_{\star}}$. Note that changing the values of A_{bias} does not change $\langle N_{\text{g}} | M_{\text{h}} \rangle$, the mean number of galaxies averaged over all haloes of fixed mass: this is the defining feature of the decorated HOD, and the meaning of the principle of HOD conservation.

4 MODELLING THE GALAXY–HALO CONNECTION: TOOLS AND METHODOLOGY

In this section, we briefly describe the tools and methodologies behind our mock-making procedure, the main N -body simulation used, and the pipeline to produce novel heterogeneous sets of Outer Rim-based galaxy catalogues.

4.1 Outer Rim mocks

The baseline simulation used for all our mock-making procedure is the Outer Rim (OR) run, extensively described in Heitmann et al. (2019). The simulation has been developed along the glorious tradition of the Millennium simulation (Springel 2005), with similar mass resolution but a volume coverage increase by more than a factor of 200. Currently, the Outer Rim is among the largest high-resolution gravity-only N -body simulations ever performed, spanning a $(3 h^{-1} \text{Gpc})^3$ volume, and characterized by an unprecedented combination of volume and mass resolution (down to $1.85 \times 10^9 h^{-1} M_{\odot}$) evolving 1.07 trillion particles – i.e. 10240^3 . The actual size of the simulation was chosen to cover a volume large enough to enable synthetic sky catalogues for eBOSS, DESI, and LSST, while maintaining adequate mass resolution to capture haloes reliably down to small masses. The entire run was carried out at the Argonne Leadership Computing Facility on Mira, a Blue-Gen/Q many-core supercomputer. The simulation code adopted is an optimized version of the Hardware/Hybrid Accelerated Cosmology Code (HACC), designed to overcome numerical challenges; see Habib et al. (2016) for all the details. The cosmology of the simulation is close to the best-fitting *WMAP*-7 fiducial model (Komatsu et al. 2011), namely $\omega_c = 0.1109$, $\omega_b = 0.02258$, $n_s = 0.963$, $h = 0.71$, $\sigma_8 = 0.8$, $w = -1$, with no massive neutrinos ($\Omega_{\nu} = 0$) and assuming flatness. The dynamical range of the simulation is remarkable, spanning 10^6 orders of magnitudes, with a force resolution of $6 h^{-1} \text{kpc}$. Initial conditions are fixed at $z_{\text{in}} = 200$ with the Zel’dovich approximation (Zel’dovich 1970). Transfer functions are generated via CAMB (Lewis, Challinor & Lasenby 2000).

A total of 101 redshifts in output were originally saved, from $z = 10$ to $z = 0$, evenly spaced in $\log a$, with a the scale factor. In our study, we mainly focus on 2 redshift intervals, namely $z = 0.695$ and $z = 0.865$, although we also consider a variety of other redshifts from some of the realizations to assess redshift evolution. Each snapshot encompasses globally about 40TB of data, and the entire data volume of the simulation is more than 5PB; all particle information for haloes with more than 100 000 particles is stored for substructures and shape studies, as well as a random selection of 1 per cent of particles in each halo (with a minimum of five particles per halo).

For our mock-making purposes, we had access to the friends-of-friends (FOF) halo catalogue at various redshifts, generated using

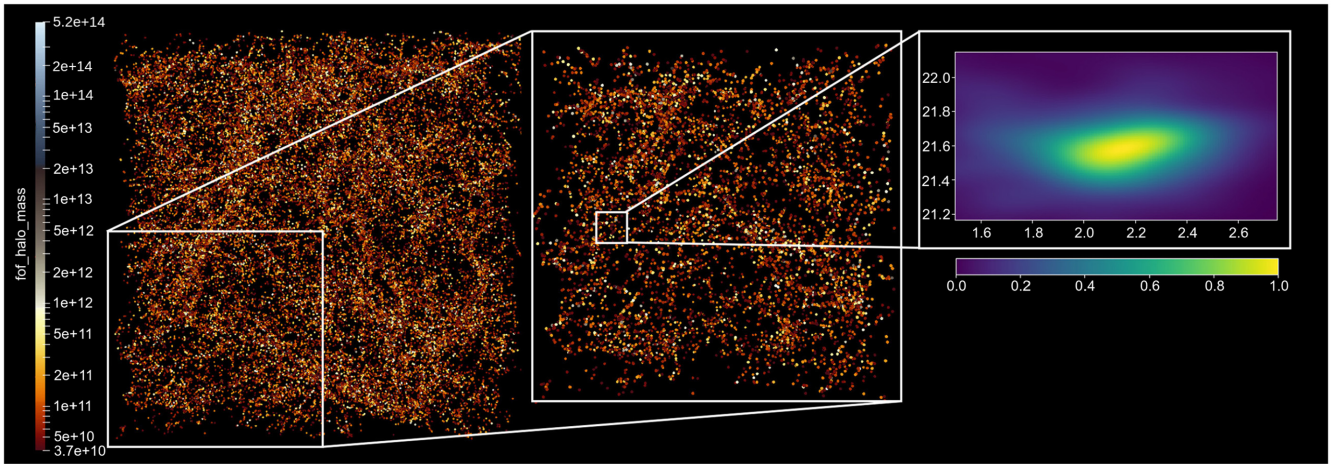


Figure 5. Small portion of the Outer Rim halo catalogue at $z = 0.865$. The left-hand panel is a $100 \times 100 [h^{-1} \text{Mpc}]^2$ projection along x and y and across z , with thickness $\Delta z = 50 h^{-1} \text{Mpc}$, while the middle panel is a progressive zoom into a $50 \times 50 [h^{-1} \text{Mpc}]^2$ block. Points in the figure are FOF haloes, colour coded by their mass. Zooming into a smaller $7 \times 7 [h^{-1} \text{Mpc}]^2$ inset of the halo catalogue, the right-hand panel displays the ellipsoidal shape of a halo of mass $4.938 \times 10^{13} h^{-1} M_{\odot}$ contained inside that area, rendered with a 1 per cent random particle subsample. Length units displayed in the left-hand panel are in $h^{-1} \text{Mpc}$. The high resolution of the simulation, down to $1.85 \times 10^9 h^{-1} M_{\odot}$, allows one to resolve accurately also relatively low-mass haloes.

a linking length $b = 0.168$.⁵ Haloes are defined by more than 20 particles, and found with a customized FOF finder (Woodring et al. 2011; Heitmann et al. 2019) that follows the standard implementation, having the linking length defined with respect to the mean interparticle spacing. All the centres of the haloes are determined by the location of the FOF halo’s minimum gravitational potential, and the centre-of-mass and the halo velocities are obtained by summing over all positions and velocities and dividing by the number of particles. The FOF halo mass is simply determined by the number count of particles in each halo.

For a given redshift, our halo catalogue (split into 110 subfiles, stored in a way such that they are not contiguous volumes) contains the number of particles in halo (Halo Count), the halo ID (Halo Tag), the halo FOF mass in $h^{-1} M_{\odot}$ units, the comoving halo centre positions from the potential minimum (in $h^{-1} \text{Mpc}$), and the comoving peculiar velocities of halo centres (in km s^{-1}). Note that the halo centre is defined by its potential minimum most bound particle, since accurate centre-finding is important for measuring the halo concentration, for halo stacking, and for placing central galaxies from HOD modelling.

Fig. 5 visualizes a small portion of the Outer Rim halo catalogue at $z = 0.865$. Specifically, the left-hand panel shows a $100 \times 100 [h^{-1} \text{Mpc}]^2$ projection along x and y and across z , having thickness $\Delta z = 50 h^{-1} \text{Mpc}$; points in the figure are FOF haloes, colour coded by their mass. The middle panel is a progressive zoom into a $50 \times 50 [h^{-1} \text{Mpc}]^2$ block having the same depth as the left one, while the right-hand panel shows an individual halo of mass $4.938 \times 10^{13} h^{-1} M_{\odot}$ rendered with 1 per cent of random particles, located inside the smaller $7 \times 7 [h^{-1} \text{Mpc}]^2$ white inset: it is possible to appreciate the neat ellipsoidal halo shape. Length units displayed in the left-hand panel are in $h^{-1} \text{Mpc}$. The high resolution of the simulation, down to $1.85 \times 10^9 h^{-1} M_{\odot}$, allows one to resolve accurately also relatively low-mass haloes.

⁵The Outer Rim halo catalogues used here are publicly available at <https://cosmology.alcf.anl.gov>

4.2 NSERIES mocks

In addition to the heterogeneous sets of high-fidelity Outer Rim mocks developed in this work, we also exploit a small homogeneous set indicated as the NSERIES, which has been previously used in the SDSS DR12 galaxy clustering analysis (Alam et al. 2017). The homogeneous set, comprised of 84 mocks in total, is particularly suitable to address cosmic variance and modelling systematics at the sub-percent level, since all mocks have the same underlying galaxy bias model built upon the same cosmology, but each mock is a quasi-independent realization – thus not sharing the same LSS. Moreover, these mocks are cut sky: they have the same angular and radial selection function as the NGC DR12 CMASS sample within the redshift range $0.43 < z < 0.70$, and therefore they include observational artefacts closer to the eBOSS DR16 sample. The N -body simulations from which these cut-sky mocks were created have been produced with GADGET2 (Springel 2005), with input parameters to ensure sufficient mass and spatial resolution to resolve the haloes that BOSS galaxies occupy. Specifically, the NSERIES cosmology is characterized by $\Omega_m = 0.286$, $h = 0.7$, $\Omega_b = 0.047$, $\sigma_8 = 0.820$, and $n_s = 0.96$. The main difference with respect to the Outer Rim mocks – apart from being cut sky and not built on periodic cubes – is that the NSERIES derive from multiple realizations of the dark matter field on a larger volume with different random seeds (i.e. a series of N -body simulations identical in all but in the initial random seed), which allows one to address the impact of cosmic variance: this is not achievable with only a halo catalogue nor a single N -body simulation at hands.

4.3 EZMOCKS

For determining the rescaled covariance matrices functional to the subsequent analyses, we also make use of a new series of DR16 EZMOCKS, thoroughly described in Zhao et al. (2020). These large number of galaxy catalogues (1000 per tracer), having accurate clustering properties, are generated with a complex methodology built around the Zel’dovich approximation (Zel’dovich 1970), and effectively including stochastic scale-dependent, non-local, and non-linear biasing contributions; extensive details on the methodology

can be found in the first release paper by Chuang et al. (2015). Non-local effects, such as tidal fields not included in linear Lagrangian Perturbation Theory (LPT) or other biasing contributions, are effectively included in both the scatter relation and the tilting of the initial power spectrum. The missing power towards small scales of perturbative approaches is included in the modulation of the initial power spectrum, when fitting for the resulting halo populations. These mocks have accurate clustering properties – nearly indistinguishable from full N -body solutions – in terms of the one-point, two-point, and three-point statistics. The underlying cosmology is based on a flat Λ CDM model, with $\Omega_m = 0.307115$, $\Omega_b = 0.048206$, $h = 0.6777$, $\sigma_8 = 0.8225$, and $n_s = 0.9611$. Specifically for LRGs, they contain the complexity of blending the CMASS plus eBOSS LRG samples, as well as all the realistic effects of mask, cut sky, and observational systematics (i.e. fibre completeness, spectroscopic success rate, redshift failures, photometric systematics). For our analysis, we adopt dedicated cubic EZMOCKS rather than the cut sky set for covariance estimations, to comply with the characteristics of the high-fidelity Outer Rim-based realizations. The EZMOCKS are extensively used in all the supporting eBOSS DR16 papers and in the final eBOSS consensus analysis. For additional technical details, we refer the reader to the companion paper by Zhao et al. (2020).

4.4 Galaxy mock-making procedure: methods

Our synthetic high-fidelity galaxy mocks are primarily produced exploiting the standard HALOTOOLS⁶ framework (Hearin et al. 2017), and by introducing a number of customizations depending on the desired challenge and model explored (see Section 6, as well as the previous theoretical part). In particular, we interface HALOTOOLS capabilities with the Outer Rim halo catalogue at different redshifts. HALOTOOLS is an open-source, community-driven Python powerful package for studying the galaxy–halo connection, which provides a highly modular, object-oriented platform for building HOD models, so that individual modelling features can easily be swapped in and out. This modularity facilitates rigorous study of all the components that makes up a halo occupation model, and has been designed from the ground-up with assembly bias applications in mind. In this view, although our main products are based on the Outer Rim simulation, following the HALOTOOLS philosophy the pipelines developed here are written in a general and flexible manner, so that any type of customization is readily achievable with minimal efforts and modifications; hence, our modular approach procedure is quite general, and readily applicable to any halo catalogue and survey design in mind. The concept of generality and reusability of the code is in fact what has driven this design from the start. In this view, although we limit here our modelling approach to HOD-based techniques (mainly due to limitations in our available halo catalogue products), we plan to pursue subhalo and abundance matching methods in follow-up studies, using the same code and modular structure.

Adopting HALOTOOLS conventions, three main primary keyword arguments are used to customize all the instances retrieved by the mock factory, common to all the different HOD models developed here (besides specific HOD parameters), namely: redshift, threshold, and modulate_with_cenocc – the latter being the modulation of the satellite distribution with the central one. In our mock-making procedure, except for the Hearin framework, all other satellite HODs are not modulated by their corresponding central distributions. Also,

as previously mentioned, we treat the conventional Zheng model separately since its HOD is effectively redshift-independent and the meaning of ‘threshold’ in the model is based on luminosity rather than stellar mass, unlike for the other three frameworks considered (i.e. Leauthaud, Tinker, Hearin). Specifically, HALOTOOLS is used to populate dark matter haloes in the Outer Rim simulation with galaxies having a stellar mass $M_* > 10^{10.0} h^{-1} M_\odot$ (‘Threshold 1’), $M_* > 10^{10.5} h^{-1} M_\odot$ (‘Standard’), and $M_* > 10^{11.0} h^{-1} M_\odot$ (‘Threshold 2’). This roughly correspond to ‘Threshold 1’ ($M_r = -19$), ‘Standard’ ($M_r = -20$), and ‘Threshold 2’ ($M_r = -21$), respectively, in the Zheng formalism.

A composite HOD model is fully defined once one specifies the occupation statistics and phase space prescription for centrals and satellites. The theoretical formalism related to each individual model has been presented in Section 3. The corresponding numerical implementation is briefly explained in what follows – noting that dealing with the Outer Rim simulation poses several non-trivial challenges in handling massive data sets. Specifically, at the highest level, we select the redshift, threshold, and number of desired mocks to produce. Then, to populate haloes with central galaxies we first calculate the value of $\langle N_{\text{cen}} \rangle$ for every halo in the simulation according to the HOD formulas in our different prescriptions (Section 3). For every halo in the simulation, we draw a random number r from $\mathcal{U}[0, 1]$, a uniform distribution between zero and unity. For all haloes with $r \leq \langle N_{\text{cen}} \rangle$, we place a central galaxy at the halo centre, leaving all other haloes devoid of centrals. Populating satellites is more complicated, because the spatial distributions are non-trivial. The first step is similar to that of the centrals, namely compute $\langle N_{\text{sat}} \rangle$ for every halo using our specified formulas for a given HOD model (see again Section 3). For each halo, the number of satellites that will be assigned to the halo is then determined by drawing an integer from the assumed satellite occupation distribution $p(N_{\text{sat}}|M_h)$ or $p(N_{\text{sat}}|M_h, x)$. Satellites are modelled as being isotropically distributed within their haloes according to an NFW profile with concentration equal to the parent halo, using the Dutton-Macciò model (Dutton & Macciò 2014). Monte Carlo realizations of both radial and angular positions are generated via the method of inverse transformation sampling. Briefly, first one generates realizations of points uniformly distributed on the unit sphere. These halocentric (x, y, z) coordinates are then multiplied by the corresponding realization of the radial position r , which is determined as follows: first, calculate $P_{\text{NFW}}(< \tilde{r}|c)$ where c is the concentration, $\tilde{r} = r/R_{\text{vir}}$ is the scale radius, R_{vir} the virial radius of the halo, and $P_{\text{NFW}}(< \tilde{r}|c)$ is the cumulative probability distribution function of the mass profile of an NFW halo:

$$P_{\text{NFW}}(< \tilde{r}|c) = \frac{M_{\text{NFW}}(< \tilde{r}|c)}{M_{\text{tot}}} = \frac{g(c\tilde{r})}{g(c)}, \quad (19)$$

where

$$g(x) = \ln(1+x) - \frac{x}{1+x}. \quad (20)$$

Then, for a halo with concentration c populated by N_{sat} , draw N_{sat} random numbers p from $\mathcal{U}[0, 1]$. Each value of p is interpreted as a probability where the corresponding value for the scaled radius \tilde{r} comes from numerically inverting $p = P_{\text{NFW}}(< \tilde{r}|c)$. Scaling the (x, y, z) points on the unit sphere by the value r gives the halocentric position of the satellites.

All these high-fidelity mocks have been produced at the National Energy Research Scientific Computing Center (NERSC), a DOE Office of Science User Facility supported by the Office of Science of the U.S. Department of Energy, under Contract No. DE-AC02-05CH11231 using the Cori supercomputer, a Cray XC40 with a peak performance of about 30 petaflops. Cori is comprised of 2388 Intel

⁶See <https://github.com/astrophy/halotools>

Xeon ‘Haswell’ processor nodes, and 9688 Intel Xeon Phi ‘Knight’s Landing’ (KNL) nodes. The system also has a large Lustre scratch file system and a first-of-its kind NVRAM ‘burst buffer’ storage device. We devised new customized scripts and pipelines to produce such mocks on Cori, exploiting especially the multithread architecture. Our mock-making code/pipeline is memory efficient and optimized to the machine. Some additional supporting numerical work has also been carried out using the Korea Institute of Science and Technology Information (KISTI) supercomputing infrastructure.

In closing this part, we note that our main goal in the cubic N -body-based mock-making production and in the related mock challenge is to test the validity and robustness of different BAO and RSD fitting techniques on a common ground against a series of different HOD prescriptions, and validate the clustering analysis pipelines and the various RSD models. Hence, we are not concerned with reproducing exactly all the features of the eBOSS DR16 LRG sample, and this is why the various HOD parameters that enter in the models outlined in Section 3 have been maintained to their corresponding literature values. Nevertheless, in Section 6 we show an instructive comparison between eBOSS measurements and those obtained from Outer Rim TH2 mocks. Note also that the HOD parameters adopted from the literature were chosen under cosmologies different from that of the Outer Rim, and therefore we do not expect to find the same results as in the corresponding original publications: this is clearly not affecting the conclusions of our work. Realistic observational artefacts related to the LRG sample, such as cut sky, matching number density, observational systematics, etc., are instead part of the EZMOCKS release (Zhao et al. 2020).

5 ANALYSIS: METHODOLOGY

In this section, we briefly describe the three configuration and Fourier space techniques used in the analysis of the challenge mocks, based on three different RSD analytical models – exploiting the FS information in the correlation function or power spectrum. The detailed BAO modelling is instead described in our LRG companion papers. All these methods are adopted in the main analysis of the final eBOSS DR16 LRG sample.

5.1 CLPT-GS

The CLPT-GS-based method is a combination of the Convolutional Lagrangian Perturbation Theory (CLPT) and the RSD Gaussian Streaming (GS) formalism, originally developed by Reid & White (2011), Carlson, Reid & White (2013), and Wang, Reid & White (2014). CLPT provides a non-perturbative resummation of Lagrangian perturbation to the two-point statistic in real space for biased tracers. In particular, the two-point correlation function is expanded in its Lagrangian coordinates considering the LRG tracer to be locally biased with respect to the initial CDM overdensity, and the expansion is performed over different orders of the Lagrangian bias function. The key equation for the two-point correlation $\xi_{\text{LRG}}(\mathbf{r}) = \langle \delta_{\text{LRG}}(\mathbf{x}) \delta_{\text{LRG}}(\mathbf{x} + \mathbf{r}) \rangle$, with \mathbf{q} and \mathbf{x} the Lagrangian and Eulerian coordinates, respectively, δ the overdensity, and \mathbf{r} the LRG separation, is

$$1 + \xi_{\text{LRG}}(\mathbf{r}) = \int M(\mathbf{r}, \mathbf{q}) d\mathbf{q}, \quad (21)$$

where $M(\mathbf{r}, \mathbf{q})$ is the convolution kernel taking into account the displacements and bias expansion up to its second derivative term. The bias derivative terms are computed using a linear power spectrum, obtained with CAMB (Lewis et al. 2000) for a fixed

cosmology – namely, the fiducial cosmology of the analysis. The peculiar velocity effect on clustering statistic is also modelled, and the pairwise velocity distribution \mathbf{v}_{12} and velocity dispersion σ_{12} are given by (Wang et al. 2014):

$$\mathbf{v}_{12}(r) = [1 + \xi_{\text{LRG}}(\mathbf{r})]^{-1} \int M_1(\mathbf{r}, \mathbf{q}) d\mathbf{q} \quad (22)$$

and

$$\sigma_{12}(r) = [1 + \xi_{\text{LRG}}(\mathbf{r})]^{-1} \int M_2(\mathbf{r}, \mathbf{q}) d\mathbf{q}, \quad (23)$$

where the kernels $M_1(\mathbf{r}, \mathbf{q})$ and $M_2(\mathbf{r}, \mathbf{q})$ also depend on the first two derivatives of the Lagrangian bias, which are free parameters in the model, in addition to the growth factor. CLPT generates more accurate multipoles than linear theory and even the Lagrangian Resummation Theory (LRT; Matsubara 2008), but a better performance is needed in order to study the smaller scales of quadrupoles. To achieve such precision, the real space CLPT models of the two-point statistics are mapped into redshift space following the Gaussian Streaming Model (GSM) formalism proposed by Reid & White (2011). In particular, the pairwise velocity distribution is assumed to have a Gaussian shape dependent on both the angle μ between the separation vector and the line-of-sight (LOS), and the LRG separation r in its parallel (r_{\parallel}) and perpendicular (r_{\perp}) components with respect to the LOS. The main equation for the correlation function is given by

$$1 + \xi_{\text{LRG}}(r_{\perp}, r_{\parallel}) = \int \frac{1}{\sqrt{2\pi(\sigma_{12}^2(r, \mu) + \sigma_{\text{FOG}}^2)}} [1 + \xi_{\text{LRG}}(r)] \times \exp\left[-\frac{[r_{\parallel} - y - \mu v_{12}(r, \mu)]^2}{2(\sigma_{12}^2(r, \mu) + \sigma_{\text{FOG}}^2)}\right] dy, \quad (24)$$

where $\xi_{\text{LRG}}(r)$, $v_{12}(r)$, and $\sigma_{12}(r)$ are computed from CLPT as previously indicated, and σ_{FOG} is the Fingers of God (FoG) parameter to account for an additional contribution to the velocity dispersion given by satellite galaxies. For the RSD model, the Alcock & Paczynski (1979) effect implementation follows that of Xu et al. (2013). The AP distortions are modelled through the α and ϵ parameters, which characterize respectively the isotropic and anisotropic distortion components.

With this technique, the FS RSD analysis in configuration space is performed, and for a given cosmology the model has four free parameters, namely ($f\sigma_8, F', F'', \sigma_{\text{FOG}}$), with f the linear growth factor and F' and F'' the first and second derivatives of the Lagrangian bias function F . For extensive details on this method see Bautista et al. (2020) and Icaza-Lizaola et al. (2020).

5.2 TNS in configuration space

The modified TNS-based method (also indicated in this paper as ‘CF-TNS’, where ‘CF’ stands for ‘correlation function’) is a combination of the Taruya, Nishimichi & Saito (TNS; Taruya, Nishimichi & Saito 2010) technique and a galaxy non-linear bias prescription (Beutler et al. 2017a; de la Torre et al. 2017). This model is based on the conservation of the number density in real- and z -space (Kaiser 1987). In this framework, the anisotropic power spectrum for unbiased matter tracers (P^s) follows the general form of Scoccimarro, Zaldarriaga & Hui (1999), which in the approximation proposed by Taruya et al. (2010) reads:

$$P^s(k, \mu) = D(k\mu\sigma_v) [P_{\delta\delta}(k) + 2\mu^2 f P_{\delta\theta}(k) + \mu^4 f^2 P_{\theta\theta}(k) + C_A(k, \mu, f) + C_B(k, \mu, f)]. \quad (25)$$

In the previous expression, f is the linear growth factor; θ is the divergence of the velocity field defined as $\theta = -\nabla \cdot \boldsymbol{\mu}/(aHf)$; $\mu = k_{||}/k$, with $k_{||}$ the line-of-sight component of the wave vector k ; H is the Hubble constant at the considered redshift; a is the scale factor; δ is the matter density field; $P_{\delta\delta}$, $P_{\theta\theta}$, and $P_{\delta\theta}$ are the non-linear matter density, velocity divergence, and density-velocity divergence power-spectra, respectively; $C_A(k, \mu, f)$ and $C_B(k, \mu, f)$ are two correction terms expressed as integrals of the matter power spectrum – see Taruya et al. (2010) for their detailed expressions; and $D(k\mu\sigma_v)$ is a phenomenological damping function modelled as a Lorentzian so that $D(k, \mu, \sigma_v) = [1 + (k\mu\sigma_v)^2]^{-1}$, with σ_v an effective pairwise velocity dispersion that is later treated as a nuisance parameter in the cosmological inference.

This model can be generalized for biased tracers via the inclusion of a galaxy biasing model, so that the anisotropic galaxy power spectrum becomes (Beutler et al. 2014; Gil-Marín et al. 2017):

$$P_g^s(k, \mu) = D(k\mu\sigma_v) [P_{gg}(k) + 2\mu^2 f P_{g\theta} + \mu^4 f^2 P_{\theta\theta}(k) + C_A(k, \mu, f, b) + C_B(k, \mu, f, b)] \quad (26)$$

with b the galaxy linear bias. Specifically, here we assume a non-linear, non-local, galaxy biasing prescription that follows the work of McDonald & Roy (2009) and Assassi, Simonovic & Zaldarriaga (2017). Explicit expressions for $C_A(k, \mu, f, b)$ and $C_B(k, \mu, f, b)$ that enter in equation (26) can be found in de la Torre & Guzzo (2012), while detailed expressions for the galaxy–galaxy and galaxy–velocity divergence power spectra – $P_{gg}(k)$ and $P_{g\theta}(k)$, respectively – for a 1-loop perturbative expansion of the biasing function are given in Bautista et al. (2020).

The linear and non-linear matter power spectra entering in the model are computed with CAMB and the HALOFIT semi-analytical prescription, respectively. To obtain $P_{\theta\theta}$ and $P_{\delta\theta}$, we use the universal fitting functions provided by Bel et al. (2019). In particular, the overall degree of non-linear evolution is encoded via the amplitude of the matter fluctuation at the effective redshift considered. Finally, the multipole moments of the anisotropic correlation function are obtained by performing the Hankel transform of the model, and regarding the RSD part, the implementation of the AP effect follows the formalism of Xu et al. (2013): the AP distortions are modelled through the α and ϵ parameters, which characterize the isotropic and anisotropic distortion components, respectively.

With this technique, the FS RSD analysis in configuration space is performed, and for a given cosmology the model has five free parameters, namely ($f, \sigma_8, b_1, b_2, \sigma_v$) – although since f and σ_8 are degenerate they are thus combined at the level of the likelihood into the single parameter $f\sigma_8$. For extensive details on this method see de la Torre et al. (2017), Mohammad et al. (2018), and Bautista et al. (2020).

5.3 TNS in Fourier space

While the previous techniques are used to carry out the analysis of the LRG sample in configuration space, the method described here – also based on the TNS model and indicated as ‘ P_k -TNS’ – is performed in Fourier space. To this end, the modelling of the BAO signal within this framework – along with the BAO fitting procedure in Fourier space – are described in Gil-Marín et al. (2020). Here, we briefly illustrate only the strategy adopted for the RSD and AP analysis, exploiting the FS information in the power spectrum.

Specifically, the FS formalism employed to describe power spectrum multipoles is the same as the one previously used in BOSS and eBOSS studies for galaxies (Gil-Marín et al. 2016) and quasars (Gil-

Marín et al. 2018). We adopt the Eulerian non-linear bias model of McDonald & Roy (2009), consisting of four bias parameters: namely, the linear galaxy bias b_1 , the non-linear galaxy bias b_2 , and two non-local galaxy bias parameters, $b_{s2} = -4/7(b_1 - 1)$ (Baldauf et al. 2012) and $b_{3nl} = 32/315(b_1 - 1)$ (Saito et al. 2014) – with b_1 and b_2 considered as free nuisance parameters in the fitting. The density–density ($\delta\delta$), density–velocity ($\delta\theta$), and velocity–velocity ($\theta\theta$) real space DM auto- and cross-power spectra are obtained via 2-loop resummation perturbation theory, as in Gil-Marín et al. (2012): these moments accurately describe the DM clustering up to $k \simeq 0.15$ at $z = 0.5$, $k \simeq 0.20$ at $z = 1.0$, and $k \simeq 0.30$ at $z = 1.5$, respectively. Expressions for those galaxy power spectra – with no velocity bias – are given by (Beutler et al. 2014):

$$P_{g, \delta\delta}(k) = b_1^2 P_{\delta\delta}(k) + 2b_2 b_1 P_{b2, \delta}(k) + 2b_{s2} b_1 P_{bs2, \delta}(k) + b_2^2 P_{b22}(k) + 2b_2 b_{s2} P_{b2s2}(k) + b_{s2}^2 P_{bs22}(k) + 2b_1 b_{3nl} \sigma_3^2(k) P_{lin}(k) \quad (27)$$

$$P_{g, \delta\theta}(k) = b_1 P_{\delta\theta}(k) + b_2 P_{b2, \theta}(k) + b_{s2} P_{bs2, \theta}(k) + b_{3nl} \sigma_3^2(k) P_{lin}(k) \quad (28)$$

$$P_{g, \theta\theta}(k) = P_{\theta\theta}(k). \quad (29)$$

RSD effects are incorporated following Taruya et al. (2010), who extended the original methodology of Scoccimarro (2004), so that the redshift space galaxy power spectrum reads:

$$P_g^{(s)}(k, \mu) = D_{FoG}(k, \mu) [P_{g\delta\delta}(k) + 2f\mu^2 P_{g, \delta\theta}(k) + f^2 \mu^4 P_{\theta\theta}(k) + b_1^3 A^{TNS}(k, \mu, f/b_1) + b_1^4 B^{TNS}(k, \mu, f/b_1)]. \quad (30)$$

In particular, galaxy real space quantities are computed as previously described, assuming a fixed linear power spectrum template (obtained with CAMB) at the fiducial cosmology. The various power spectrum multipoles encode the coherent velocity field through the redshift space displacement and the logarithmic growth of structure parameter, which boosts the amplitude of the isotropic power spectrum and generates an anisotropic component. In equation (30), the term D_{FoG} accounts for FoG effects along the LOS direction, and it is modelled as a Lorentzian, while A^{TNS} and B^{TNS} are second-order corrections. Finally, the AP effect is added when computing the multipoles as

$$P_g^{(\ell)}(k) = \frac{2\ell + 1}{2\alpha_{||}\alpha_{\perp}^2} \int_{-1}^1 d\mu \mathcal{L}_{\ell}(\mu) P_g^{(s)}[k'(k, \mu), \mu'(\mu)], \quad (31)$$

where explicit expressions for $k'(k, \mu)$ and $\mu'(\mu)$ are given in Gil-Marín et al. (2020). We also consider that the shot noise contribution in the power spectrum monopole may differ from a Poisson sampling prediction, and parametrize this potential deviation with a free parameter (A_{noise}), which modifies the shot noise amplitude without introducing any scale dependence. By default, our measured power spectrum monopole has a fixed Poissonian shot noise contribution subtracted, whereas this is not the case for higher other multipoles.

With this technique, the FS RSD analysis in Fourier space is performed, and for a given cosmology the model has seven free parameters, namely ($\alpha_{||}, \alpha_{\perp}, f\sigma_8$) and ($b_1, b_2, A_{noise}, \sigma_{FoG}$). Note that while the BAO analysis consists of using a fixed and arbitrary template to compare the relative BAO-peak positions in the power spectrum multipoles, the FS analysis allows for a full modelling of the shape and amplitude of the power spectrum multipoles, taking into account DM non-linear effects, galaxy bias, and RSDs. For extensive details on this method see Gil-Marín et al. (2020).

6 THE GALAXY MOCK CHALLENGE

In this section, we present the main outcomes of the galaxy mock challenge. After a brief description of the mock products directly useful in the actual fits considered in the HOD systematic error budget, we show selected results in configuration and Fourier space. We eventually compare the complementary BAO/RSD models adopted for the analysis of the complete DR16 eBOSS LRG sample, assessing the theoretical systematic budget. Our findings demonstrate that all the methods are mutually consistent, with comparable systematic errors on the AP parameters and the growth of structures, and robust to different HOD prescriptions – thus validating the clustering analysis pipelines.

6.1 Mock products used in the analysis

For the galaxy mock challenge, we devised three sets of heterogeneous Outer Rim-based galaxy mocks (indicated as ‘Challenge Set 1’, ‘Challenge Set 2’, ‘Challenge Set 3’, respectively).⁷ These are cubic mocks, in the Outer Rim cosmology, obtained by populating Outer Rim halo catalogues with galaxies using the Zheng, Leauthaud, Tinker, and Hearin HOD prescriptions – as explained in Section 4.1. Extensive details regarding each set are provided in Appendix A. For the main analysis presented here, focused on testing the BAO templates and the RSD models adopted for the characterization of LRG clustering systematics, we only use a subset of those mocks drawn from ‘Challenge Set 1’ at $z = 0.695$ assuming the ‘Threshold 2’ (Th2) flavour. As explained in Sections 3 and 4.4, the meaning of ‘flavour’ is related to the HALOTOOLS key parameter ‘threshold’, which globally sets all the individual HOD parameters as best-fitting realizations from the corresponding literature dictionary of each HOD model (unless customizations are introduced). Specifically, we select Th2 mocks with the Leauthaud, Tinker, and Hearin prescriptions, respectively, since their number density is closer to the eBOSS LRG sample (see Table A2). Moreover, since we require fully independent mocks (i.e. not sharing the same DM field), we only consider 27 realizations per HOD per flavour: such realizations are obtained by populating once the full $3 h^{-1}$ Gpc Outer Rim periodic halo catalogue box with galaxies, and then by cutting the full box into 27 non-overlapping subcubes of $1 h^{-1}$ Gpc side and rescaling the various spatial positions accordingly. In fact, by construction, at a fixed z and for a fixed set of HOD parameters, the central galaxies will always reside at the centre of their hosted haloes, inheriting the same halo velocity; hence, additional subcubes would be fully or highly correlated in the central galaxy population, depending on how the box is cut. We then add RSDs to each individual mock in two different ways: radially, or with the usual plane-parallel approximation.

In Fig. 6, we show an example on how satellite galaxies are distributed within the same Outer Rim halo, according to the different HOD prescriptions presented in Section 3, to convey some intuition on the galaxy–halo connection modelling and the spatial location of satellites. The plot displays the x – y spatial projection at $z = 0.695$ of a randomly chosen halo, with its spherical shape determined by its virial radius. The standard Zheng model is shown in the upper left-hand panel, and clockwise the Leauthaud, Hearin, and Tinker models are displayed, respectively. While in all the HOD schemes the satellite phase space statistics follow an unbiased NFW profile with

⁷We have also devised an additional set which includes a variety of customizations, beyond the scope of the current analysis, exploring extreme variations in HOD parameters for all of the models, quenching, assembly bias, and modulations with the central distribution.

a phase space distribution in isotropic Jeans equilibrium and galaxy concentration identical to that of the parent halo, more sophisticated frameworks such as the Tinker model (lower left corner) are able to distinguish between active and quiescent populations (indicated with different colours in the panel), thus providing additional useful physical insights.

Moreover, in addition to the heterogeneous Outer Rim mocks, as detailed in Section 4.2 we also exploit 84 homogeneous cut-sky NSERIES mocks, which have been previously used in the SDSS DR12 galaxy clustering analysis (Alam et al. 2017), to address cosmic variance in the various methods – since the NSERIES derive from multiple realizations of the dark matter field with different random seeds. Here we show only one global NSERIES application, while in Gil-Marín et al. (2020) and Bautista et al. (2020) those mocks are extensively used to assess systematics related to each individual fitting method in configuration or Fourier space, respectively.

6.2 Galaxy mock challenge: BAO analysis and HOD systematics

Approximate catalogues such as the EZMOCKS (Zhao et al. 2020) are in principle sufficient for covariance estimates and for quantifying systematic biases in BAO studies, while the analysis of the FS of the correlation function and power spectrum requires high-fidelity (N -body-based) mocks to precisely test the modelling. Nevertheless, using high-resolution mocks, we are able to characterize the impact of systematics in HOD modelling both on BAO and RSD constraints with high-accuracy. Specifically, the main goals are to quantify possible effects induced by different galaxy HOD schemes on the cosmic growth rate and obtain useful information on parameter inference based on HOD variations, to assess the impact of an arbitrary choice of the BAO reference template on the inferred cosmological parameters, and more generally to determine the theoretical systematic budget and validate the clustering analysis pipelines.

The standard procedure common to all BAO fitting methods is to assume a fixed and arbitrary template, and compare the relative BAO peak positions in the correlation function or power spectrum multipoles. Reconstruction techniques such as those presented in Burden et al. (2014), Burden, Percival & Howlett (2015) are then applied to the density field, in order to remove a fraction of the RSDs and the non-linear motions of galaxies. The BAO feature in the 2-point statistics (both in configuration and Fourier space) is sharpened, increasing the precision of the measurement of the acoustic scale.

The BAO scale measurement in configuration space adopted here is the same as the one described in previous SDSS publications (i.e. Alam et al. 2017; Ata et al. 2018; Bautista et al. 2018), and thoroughly illustrated in our companion paper Bautista et al. (2020), while the modelling of the BAO signal along with the BAO fitting procedure in Fourier space are explained in detail in Gil-Marín et al. (2020). In particular, for the latter case, the power spectrum anisotropic signal is modelled in order to measure the BAO peak position and marginalize over the broad-band information – taking into account the BAO signal both in the radial and transverse LOS directions. Generally, BAO results are obtained from pre- and post-reconstructed data, while RSD results use only the non-reconstructed sample. In the following analyses, we assume standard dependence of the growth rate f , and adopt a smoothing scale of $15 h^{-1}$ Mpc. Whenever required, galaxy redshifts are converted into radial comoving distances for clustering measurements, using the cosmological parameters of the OR simulation. As shown in Bautista et al. (2020), the analysis methodology is insensitive to the choice of a fiducial cosmology.

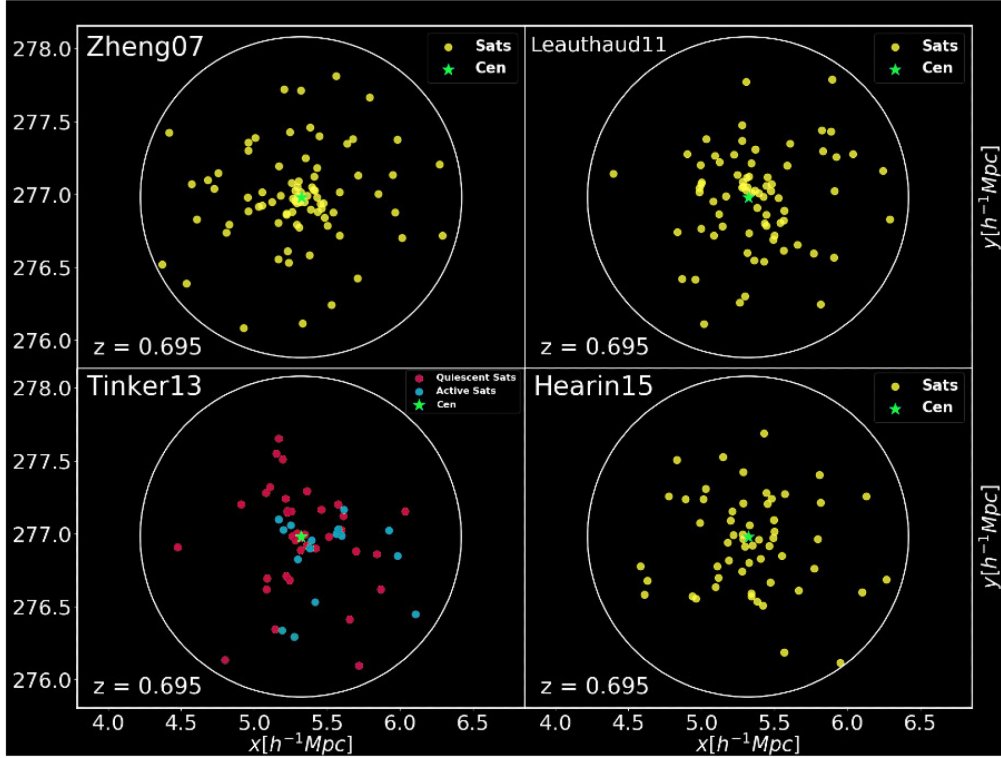


Figure 6. Example of the distribution of satellite galaxies within the same and randomly chosen Outer Rim halo, at $z = 0.695$, according to the different HOD prescriptions presented in Section 3. The plot represents a spatial projection ($x - y$) of the halo, and its spherical shape as determined by its virial radius is also indicated in the various panels. Clockwise, starting from the upper left corner, the Zheng, Leauthaud, Hearin, and Tinker models are shown, respectively.

Fig. 7 is an example of the redshift-space galaxy clustering (monopole and quadrupole), along with corresponding BAO fits, as inferred from the average of 27 Th2 OR-based mocks at $z = 0.695$ having different HOD schemes. From left to right, the Leauthaud, Tinker, and Hearin models are displayed, respectively. Top panels are for the pre-reconstructed fields, while bottom panels refer to post-reconstructed fields. Specifically, for cosmological analyses the information contained in the anisotropic 2-point correlation function $\xi(s, \mu)$ – decomposed into polar coordinates (s, μ) aligned with the LOS direction, with μ the cosine of the angle between the LOS and separation vector directions, and s the norm of the galaxy separation vector \mathbf{s} – is compressed into the correlation function multipole moments ξ_ℓ , obtained by decomposing $\xi(s, \mu)$ on the basis of Legendre polynomials P_ℓ as:

$$\xi_\ell(s) = (2\ell + 1) \sum_i \xi(s, \mu_i) P_\ell(\mu_i) \Delta\mu. \quad (32)$$

In the previous expression, only even multipoles do not vanish, and the correlation function is binned according to the absolute value of μ . In our analyses, we only consider the $\ell = 0, 2, 4$ moments, namely monopole, quadrupole, and hexadecapole (whenever specified), and $\xi(s, \mu)$ is quantified with the classical Landy & Szalay (1993) estimator. The pair counts are binned into $5h^{-1}$ Mpc bins in separation and 0.01 in μ . In the panels of Fig. 7, the BAO feature is clearly seen at $s \simeq 100h^{-1}$ Mpc, as well as the impact of the reconstruction procedure: the BAO feature appears in fact much sharper in the bottom panels. The various fits shown in the figure are obtained with the BAO technique adopted in Bautista et al. (2020) for the analysis of the correlation function in configuration space, and correlation function multipoles $\xi_\ell(s)$ are rendered as a function of separations s relevant

for BAO ($30 \leq s \leq 180h^{-1}$ Mpc), starting from the modelling of the redshift-space anisotropic power spectrum. In particular, as explained in Bautista et al. (2020), the non-linear broadening of the BAO peak is modelled by multiplying the ‘peak-only’ power spectrum P_{peak} by a Gaussian term with $\Sigma_{\text{nl}}^2(\mu) = \Sigma_{\parallel}^2 \mu^2 + \Sigma_{\perp}^2 (1 - \mu^2)$, with Σ_{\parallel} and Σ_{\perp} the BAO damping terms, and the non-linear random motions on small scales are rendered with a Lorentzian term parametrized by Σ_s . When performing fits to the multipoles of a single realization of the survey, the values of $(\Sigma_{\parallel}, \Sigma_{\perp}, \Sigma_s)$ are maintained fixed to improve convergence. Moreover, the BAO peak position is parametrized via two dilation parameters that scale separations into transverse (α_{\perp}) and radial (α_{\parallel}) directions. These quantities are related to the comoving angular diameter distance $D_M(z) = (1 + z)D_A(z)$ and to the Hubble distance $D_H(z) = c/H(z)$ as

$$\alpha_{\perp} = \frac{D_M(z_{\text{eff}})/r_{\text{drag}}}{D_M^{\text{fid}}(z_{\text{eff}})/r_{\text{drag}}^{\text{fid}}}, \quad (33)$$

$$\alpha_{\parallel} = \frac{D_H(z_{\text{eff}})/r_{\text{drag}}}{D_H^{\text{fid}}(z_{\text{eff}})/r_{\text{drag}}^{\text{fid}}}, \quad (34)$$

with r_{drag} the comoving horizon scale at the drag epoch. Fits on mock multipoles are performed – including hexadecapole, which however does not add extra information. In the procedure, BAO broad-band parameters are let free while both dilation parameters are allowed to vary between 0.5 and 1.5. A total of nine parameters are fitted simultaneously. Table 4 contains the results of such BAO fits in configuration space, where in particular b is the linear bias and $\beta = f/b$ is the RSD parameter. The covariance matrix used for the fit is obtained from 1000 EZMOCKS, properly rescaled by the difference in particle number to match the characteristics of the OR-based mocks. Note that expected statistical errors in the eBOSS LRG data sample are

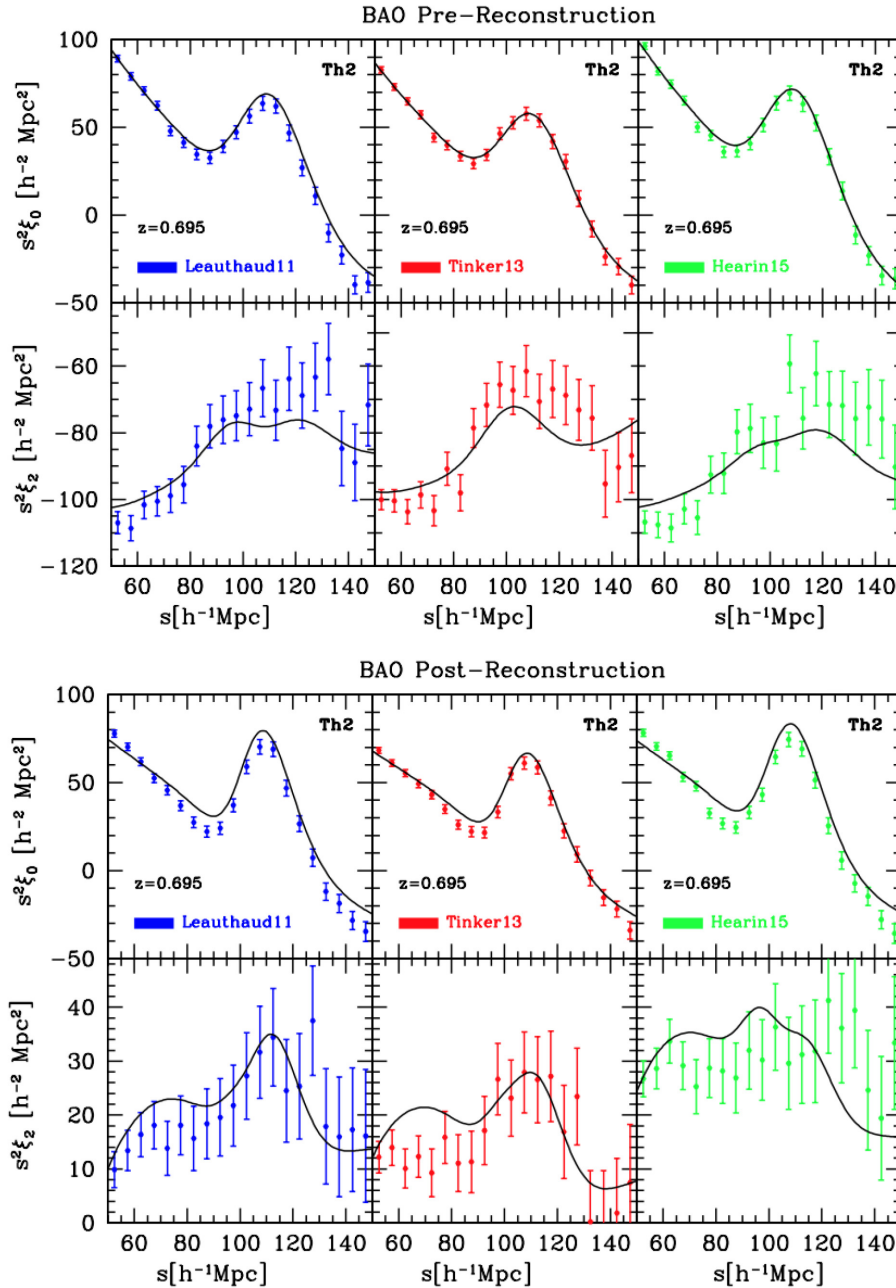


Figure 7. Monopole and quadrupole of the average 2PCFs as computed from a subset of 27 OR-based mocks per HOD type, and fits of the BAO feature as seen in the correlation function multipoles. Top panels show results for the pre-reconstruction case, bottom panels refer to the reconstructed density field. The HOD models of Leauthaud, Tinker, and Hearin are shown – from left to right, respectively – for the ‘Th2’ flavour at $z = 0.695$. Note that the BAO feature (around $100 h^{-1} \text{Mpc}$) appears much sharper after application of the reconstruction procedure, as expected. Results of these fits are reported in Table 4.

of the order of ~ 1.9 per cent for α_{\perp} and ~ 2.6 per cent for α_{\parallel} , and that reconstruction improves constraints on α_{\perp} and α_{\parallel} , as expected.

As mentioned in Section 4.4, the OR-based mocks have not been designed to reproduce all of the features of the eBOSS DR16 LRG sample, since that level of complexity is not necessary for our subsequent analysis. Hence, we clearly do not expect to exactly match the clustering properties of the eBOSS LRG sample. Nevertheless, as an instructive example, we show in Fig. 8 a comparison between eBOSS measurements and those made from Th2 OR mocks at $z = 0.695$. Specifically, green triangles in the figure display eBOSS plus CMASS measurements for the NGC, pink squares are used

for eBOSS plus CMASS SGC measurements, and yellow circles show results obtained from the combined samples. Open symbols refer to pre-reconstruction calculations (top panel), while filled symbols indicate post-reconstruction results (bottom panel). Errorbars are estimated from analogous measurements performed on 1000 EZMOCKS, and via averaging. The solid black line in both panels represents the redshift-space 2PCF monopole expectation assuming the Tinker HOD recipe, derived from 27 Th2 OR mocks. Note that, besides not tuning the various HOD parameters to reproduce the exact clustering properties of the eBOSS DR16 LRG sample, our OR-based realizations are cubic mocks, and do not contain the complications

Table 4. BAO fits to the average pre- and post-reconstructed 2PCFs for different HOD prescriptions over 27 corresponding Outer Rim mock realizations, at $z = 0.695$, with the ‘Th2’ flavour – as displayed in Fig. 7.

CF [Th2]	Leauthaud <i>BAO</i>	Tinker <i>Pre-Rec</i>	Hearin
α_{\perp}	0.9990 ± 0.0080	0.9922 ± 0.0073	1.0074 ± 0.0083
α_{\parallel}	1.0084 ± 0.0164	1.0234 ± 0.0147	1.0040 ± 0.0157
Σ_{\perp}	6.7663 ± 1.1320	7.6741 ± 1.2932	7.4335 ± 1.0760
Σ_{\parallel}	10.5333 ± 1.5795	8.4652 ± 1.5436	9.5687 ± 1.6354
β	0.2685 ± 0.0965	0.2404 ± 0.1017	0.2137 ± 0.0910
b	2.6748 ± 0.1315	2.4769 ± 0.1393	2.7525 ± 0.1339
χ^2	129.4	115.4	126.0
	<i>BAO</i>	<i>Post-Rec</i>	
α_{\perp}	1.0045 ± 0.0056	1.0014 ± 0.0060	1.0066 ± 0.0057
α_{\parallel}	0.9937 ± 0.0084	0.9976 ± 0.0090	1.0115 ± 0.0094
Σ_{rec}	15.0 ± 1.5	15.0 ± 1.5	15.0 ± 1.5
Σ_{\perp}	2.0002 ± 11.9307	2.0000 ± 0.9844	2.9483 ± 1.3018
Σ_{\parallel}	4.0169 ± 1.6667	4.9272 ± 1.3161	6.8683 ± 1.1810
β	0.4018 ± 0.0911	0.5127 ± 0.0863	0.4157 ± 0.0849
b	2.3873 ± 0.0744	2.1669 ± 0.0710	2.5137 ± 0.0879
χ^2	131.3	130.0	164.5

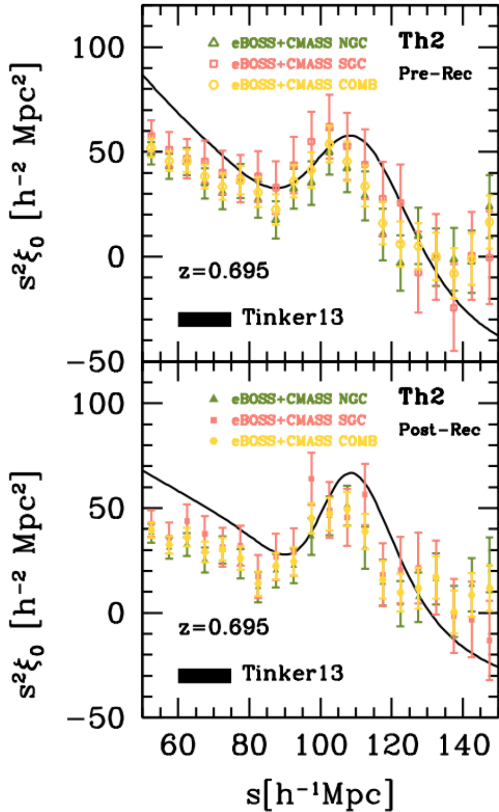


Figure 8. Illustrative example showing the comparison between eBOSS measurements and those made from Th2 OR mocks at $z = 0.695$. The top panel displays pre-reconstruction calculations (reported with open symbols), while the bottom panel refers to post-reconstruction results (indicated with filled symbols). The green triangles in the figure show eBOSS+CMASS measurements for the NGC, pink squares are used for eBOSS + CMASS SGC measurements, and yellow circles represent the combined samples. Errorbars are estimated from 1000 EZMOCKS. The solid black line in both panels is the redshift-space 2PCF monopole expectation assuming the Tinker HOD recipe, derived from 27 Th2 OR mocks, without including the complications of masks, cut sky, and observational artefacts.

of masks, cut sky, and observational artefacts (as opposed to the EZMOCKS and NSERIES) – which would be instead more appropriate for a fair data-mock comparison.

Fig. 9 shows examples of redshift-space galaxy power spectra as computed from the average of 27 OR-based mocks, each set being characterized by a different HOD scheme, at $z = 0.695$ for the ‘Th2’ flavour. The plot represents the analogous, in Fourier space, of the previous correlation function estimates in configuration space. The pre- (top panels) and post-reconstructed (bottom panels) monopoles and quadrupoles of the power spectra are shown, for the Leauthaud, Tinker, and Hearin models – from left to right, respectively. Fits are obtained with the BAO theoretical model of Gil-Marín et al. (2020), considering wave numbers between $0.02 \leq k[h\text{Mpc}^{-1}] \leq 0.30$, and the corresponding results are reported in Table 5. Unlike correlation function calculations, Discrete Fourier Transform (DFT) methods used to compute the power spectrum multipoles are quite sensitive to the assumption of periodic boundary conditions, and therefore a procedure denoted as ‘padding’ is applied in this process, to mitigate non-periodicity effects. The detailed effects of non-periodicity on BAO measurements are discussed in Gil-Marín et al. (2020). In particular, results of such analyses show that no significant changes are observed in terms of α_{\perp} , while shifts at the level of 2–3 per cent can be systematically seen in α_{\parallel} if padding is not applied. Hence, non-periodic effects are relevant in determining α_{\parallel} , but they do not impact significantly α_{\perp} . Moreover, no relative shifts in any of the α parameters are seen when the HOD model or flavour is varied – as we show next.

Fig. 10 summarizes and confronts the performance of the BAO fitting technique adopted in the analysis of the final eBOSS LRG sample, in configuration and Fourier space, with respect of variations in the underlying HOD model. For each mock realization, at a fixed HOD scheme and threshold flavour, the correlation function and power spectrum are computed along with their multipoles, respectively. Subsequently, fits for the BAO peak position are performed – both to the pre- and post-reconstructed synthetic catalogues – to determine the dilation parameters α_{\parallel} and α_{\perp} and their corresponding errors. The expected values for the dilation parameters are computed in the OR cosmology, at the effective redshift $\bar{z} = 0.695$. In addition, fits to the average multipoles of a given set of mocks (27 mocks per set) are also carried out, to probe biases in a very high precision configuration. In particular, the BAO pipeline on the power spectrum monopole and quadrupole is run in the interval $0.02 \leq k[h\text{Mpc}^{-1}] \leq 0.30$. The various BAO fittings are performed by fixing the BAO damping parameters (Σ_{\parallel} , Σ_{\perp}) at their best-fitting values on the mean of the pre- and post-reconstructed mocks, and the analysis is done in terms of the scaling parameters α_{\perp} and α_{\parallel} . The various covariances used in the analysis of the OR-based mocks are derived from the EZMOCKS, properly rescaled to account for differences in number density (see Table A2). As explained in detail in Bautista et al. (2020), the final BAO model is a combination of the cosmological multipoles ξ_{ℓ} and a smooth function of separation, which accounts for unknown systematic effects in the survey that can potentially contaminate the results.

Table B2 contains the results when fitting the mean correlation functions and power spectra (rows labelled ‘Mean’) of the 27 independent realizations of the OR-based mocks with different HOD prescriptions, as well as the mean of the fits of individual realizations (rows labelled ‘Individual’). Those data are shown in Fig. 10, where each subpanel displays the difference between the measured α_{\parallel} and α_{\perp} : their expected value are inferred for the pre- (left-hand panels) and post-reconstructed (right-hand panels) catalogues. Mean estimates are displayed in red with filled rectangles

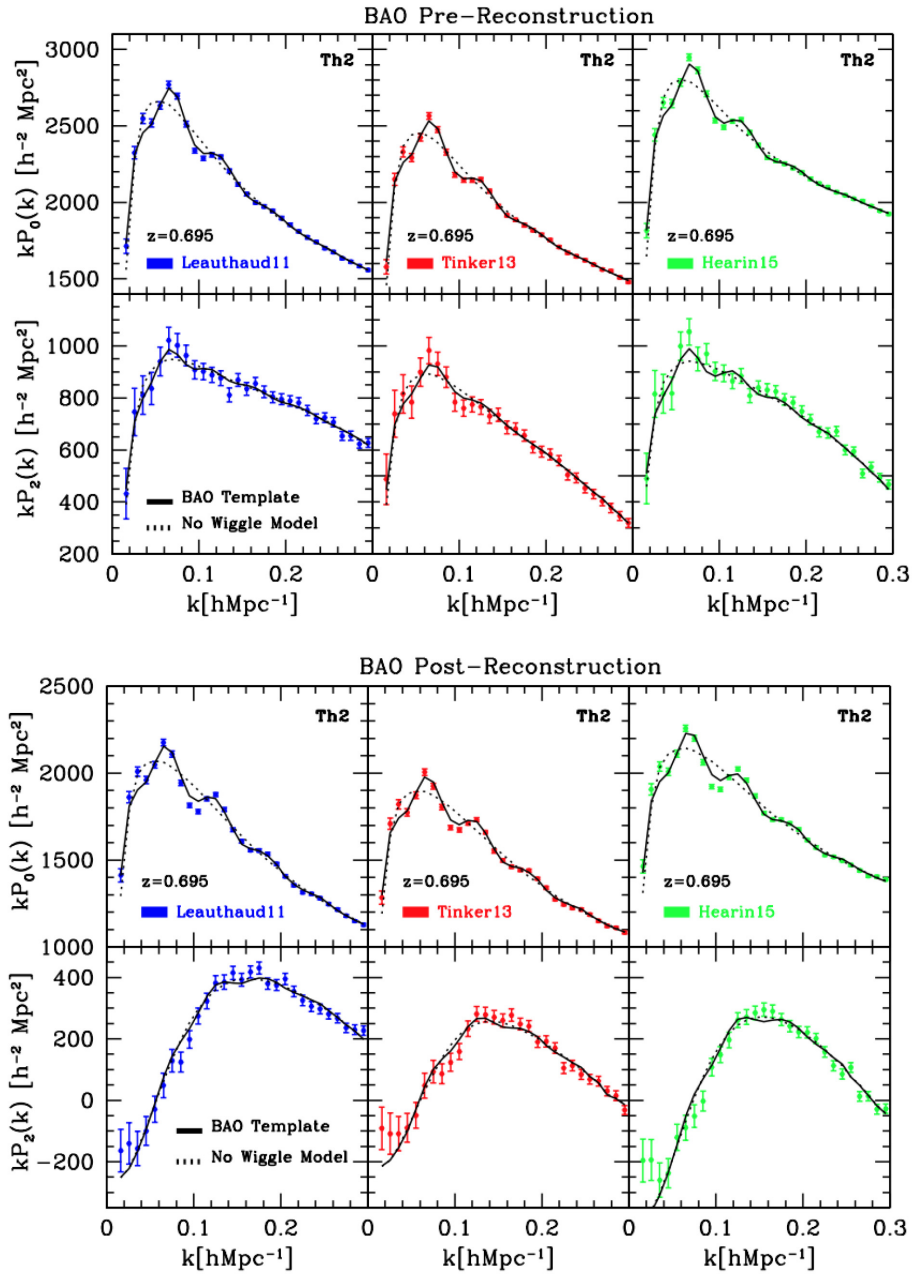


Figure 9. Monopole and quadrupole of the average power spectra as computed from a subset of 27 OR-based mocks per HOD type, and fits of the BAO feature as seen in the power spectrum multipoles (solid lines). The corresponding no-wiggle model is also reported in the figure, with dotted lines. Top panels show results for the pre-reconstruction case, bottom panels refer to the reconstructed density field. The HOD models of Leauthaud, Tinker, and Hearin are shown – from left to right, respectively – for the ‘Th2’ flavour at $z = 0.695$. Results of these fits are reported in Table 5.

as derived from configuration space techniques, and in blue with filled triangles as determined with Fourier space techniques; in the figure, these points are indicated as ‘MEAN CF’ or ‘MEAN PS’, respectively. The associated errors are consistently the errors of the mean, obtained by rescaling the related EZMOCKS covariance by the number of realizations, $N_{\text{OR}} = 27$. Therefore, these errors are a factor of $\sqrt{N_{\text{OR}}}$ smaller than the error one would obtain for a single realization of these mocks. In the figure, the dark-orange filled rectangles (related to configuration space measurements, and denoted ‘MEAN HOD CF’) and light-green filled triangles (related to Fourier space measurements, and termed ‘MEAN HOD PS’) show the same mean estimates but now normalized by the corresponding

mean expectation value averaged over all the three HODs: hence, by construction, the sum of those points for a given method (CF or PS) will be zero. In this way, one can better disentangle the systematics introduced by the HOD modelling versus the theoretical systematics related to BAO fitting methodologies. Analogous empty symbols are used to display the corresponding individual measurements for pre- and post-reconstruction catalogues, respectively. In this case the error associated is the root mean square (rms) of all the individual fits, scaled by the square root of the number of realizations ($\sqrt{N_{\text{OR}}}$). The shaded dark-green areas represent the 1 per cent error level on the α parameters. The second and fourth subpanels in Fig. 10 show the difference between the measured and the expected values of α_{11}

Table 5. BAO fits to the average pre- and post-reconstructed power spectra for different HOD prescriptions over 27 corresponding Outer Rim mock realizations, at $z = 0.695$, with the ‘Th2’ flavour – as displayed in Fig. 9.

PS [Th2]	Leauthaud <i>BAO</i>	Tinker <i>Pre-Rec</i>	Hearin
α_{\perp}	1.0028 ± 0.0107	0.9953 ± 0.0124	1.0108 ± 0.0099
α_{\parallel}	0.9885 ± 0.0178	1.0023 ± 0.0177	0.9779 ± 0.0143
Σ_{\perp}	4.2992 ± 1.8875	5.9568 ± 2.0364	4.1865 ± 1.8043
Σ_{\parallel}	9.6867 ± 1.8250	7.5044 ± 2.2301	7.0045 ± 2.0438
χ^2	44.3	34.5	57.8
	<i>BAO</i>	<i>Post-Rec</i>	
α_{\perp}	0.9976 ± 0.0075	0.9975 ± 0.0088	1.0122 ± 0.0075
α_{\parallel}	0.9938 ± 0.0104	0.9976 ± 0.0121	1.0002 ± 0.0113
Σ_{rec}	15.0 ± 1.5	15.0 ± 1.5	15.0 ± 1.5
Σ_{\perp}	1.0585 ± 0.7909	1.3377 ± 0.9904	1.1071 ± 0.8221
Σ_{\parallel}	1.5451 ± 1.1110	1.9511 ± 1.3408	2.5084 ± 1.5258
χ^2	94.4	69.0	120.1

and α_{\perp} in terms of number of statistical σ of the error of the mean, and the $\text{rms}/\sqrt{N_{\text{OR}}}$. The horizontal grey bands highlight the 1σ error level. In general, considering fits to the mean over 27 realizations, the reported dilation parameters for all the different HODs are consistent with their expected values within 0.8 per cent for α_{\perp} and 1.2 per cent for α_{\parallel} . Recall that the expected statistical errors in the eBOSS LRG data sample are of the order of ~ 1.9 per cent for α_{\perp} and ~ 2.6 per cent for α_{\parallel} . From the N -body mocks we do not observe any significant BAO peak position shift with respect to their corresponding expected value in any of the post-reconstructed catalogues analysed. The BAO pipeline performs well with different HOD models; some fluctuations are present, but their values lie always below the $\pm 2\sigma$ limit, hence the shifts are not significant. Overall, we do not detect any relative systematics due to different HOD modelling, although the statistical precision of the Outer Rim-based mocks is comparable to the statistical precision of the LRG sample (see Section 7). Interestingly, from Fig. 10 it is evident that the reconstruction procedure (right-hand panels) generally ameliorates the agreements of the α -parameters with their expected values. Moreover, it is also worth noticing that for the average values most of the detected discrepancy (after reconstruction) arises from the Hearin HOD model; this is not unexpected, since we have considered a quite extreme case of assembly bias both in the central and satellite galaxy population – as explained in Section 3. Finally, after application of the BAO reconstruction procedure, we find all of the mean measurements to be within 0.5–1.2 per cent of their expected values, thus below the statistical precision of the eBOSS LRG sample. Hence, for BAO-only fitting methods, both modelling and HOD systematics are subdominant to the global systematic error budget and the BAO analysis is unbiased. However, sub-per cent level corrections may become relevant for future surveys like DESI, that are expected to achieve sub-per cent statistical precision on the galaxy sample.

6.3 Galaxy mock challenge: RSD analysis and HOD systematics

In Section 5, we have briefly described the three RSD theoretical models adopted for the analysis of the final eBOSS DR16 LRG sample. Here, we confront those models and show that they are mutually consistent, with comparable systematic errors on the AP parameters and the growth of structure – as well as robust to different HOD prescriptions. While for the previous BAO-only analysis simply the BAO peak position has been taken into account, here we consider a full modelling of the shape and amplitude of

the correlation function and power spectrum multipoles, including non-linear DM effects, galaxy bias, and RSDs. We generically refer to this methodology as the ‘full shape’ (FS) analysis. Quantitative investigations involving the correlation function or power spectrum FS require high fidelity N -body-based mocks to test and validate the underlying RSD models, and typically such analyses are only performed over pre-reconstructed synthetic catalogues. The overall aim is to quantify the impact of the different HOD prescriptions used to populate simulated haloes with galaxies on RSD constraints. Our primary focus here is thus on modelling and HOD systematics, while Bautista et al. (2020), Gil-Marín et al. (2020), and Icaza-Lizaola et al. (2020) also examined the impact of the choice of scales in the fits and the choice of a fiducial cosmology. In particular, their conclusions (directly relevant for this work) suggest that the most robust results and optimal configuration for the FS analysis of the correlation function are obtained with monopole and quadrupole in the range $20 \leq s[h^{-1} \text{Mpc}] \leq 130$, and hexadecapole in the interval $25 \leq s[h^{-1} \text{Mpc}] \leq 130$ for the TNS model in configuration space, while using the interval $25 \leq s[h^{-1} \text{Mpc}] \leq 130$ for all the moments when considering CLPT-GS. Regarding power spectrum computations, the optimal range of scales are $0.02 \leq k[h/\text{Mpc}] \leq 0.15$, and results include the hexadecapole. In what follows, the analysis is carried out in the Outer Rim fiducial cosmology, and the monopole, quadrupole, and hexadecapole ranges are those previously specified – always set for optimal performance. Moreover, for the analysis of the challenge mocks in Fourier space, a procedure called padding is applied, in order to prevent the impact of non-periodicity to affect results when applying the discrete Fourier transform. Also, the mock covariances adopted in these investigations are derived from a set of 1000 EZMOCKS, and properly rescaled by the difference in particle number.

Fig. 11 summarizes the main results of the RSD FS analysis, confronting the three different modelling techniques: two in configuration space (CLPT-GS and CF-TNS), and one in Fourier space (P_{k} -TNS). Specifically, we analysed the Leauthaud, Tinker, and Hearin HODs (from left to right in the plots) corresponding to ‘Th2’, both in configuration and Fourier space – closer to the characteristics of the eBOSS LRG sample. Individual fits on each of the 27 realizations per model are performed (i.e. open symbols in the panels, denoted with the letter ‘I’ following the specific model), as well as fits on the mean of the mocks (i.e. filled symbols indicated with the letter ‘M’ in the panels, adopting similar conventions). Results are also reported in Table B3. In detail, circles refer to CLPT-GS measurements, squares are used for the CF-TNS model, and triangles indicate results from the P_{k} -TNS technique. Moreover, open symbols are related to individual fits, while filled symbols display fits on the mean. In the latter case, similarly to what has been done in Fig. 10, we also show fits on the mean normalized by the corresponding mean expectation value averaged over all the three HODs: such points are denoted as ‘MH’ (or ‘Mean HOD’), and are useful to separate the systematics introduced by the HOD modelling versus the theoretical systematics related to individual RSD FS fitting methodologies. In the figure, shaded orange areas represent the 1 per cent error levels on the α parameters and the 3 per cent error on $f\sigma_8$, while shaded grey areas highlight the 1σ error level.

Overall, from the fit of the mean, biases observed for CLPT-GS and CF-TNS are mostly within 1.5σ away from the expected values for all the parameters across all models (and always less than 2σ away), and errors estimated from the different RSD models are mutually compatible. In general, CF-TNS seems to imply slightly larger errors than CLPT-GS. Comparing fits on the mean with the mean of individual fits, for α_{\perp} and $f\sigma_8$ there is good agreement

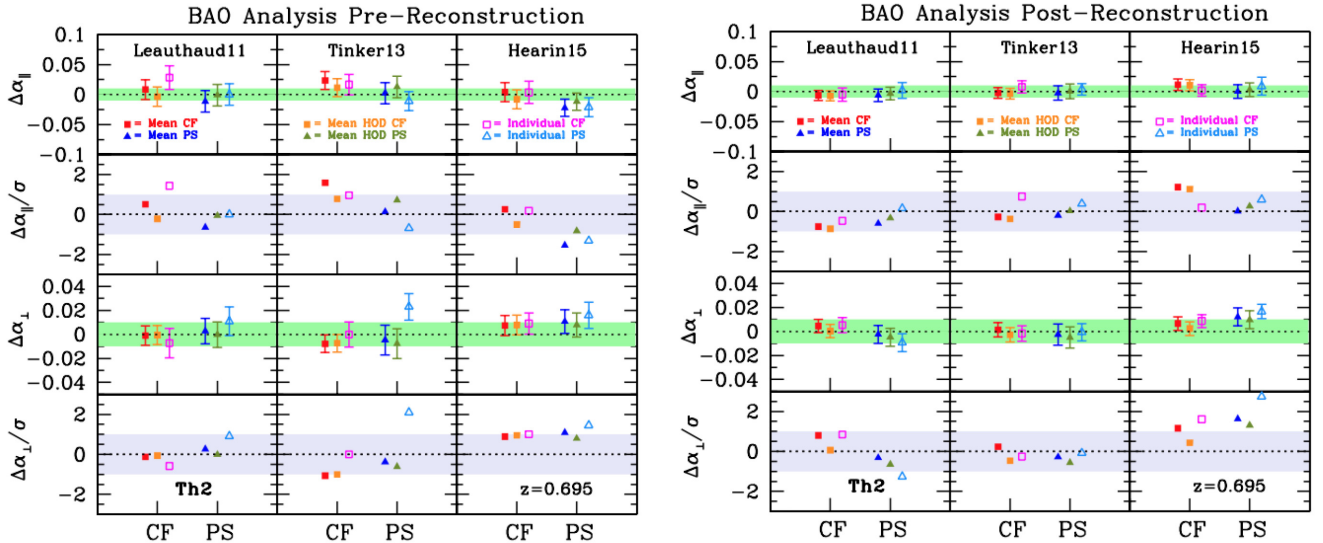


Figure 10. Performance of the BAO fitting methods in configuration and Fourier space, with respect of variations in the underlying HOD model. Fits to individual mock realizations as well as to the mean of a set of 27 independent realizations of the OR mocks (from the ‘Challenge Set 1’) are carried out (see Table B2). The difference between the measured $\alpha_{||}$ and α_{\perp} from pre- (left-hand panels) and post-reconstructed (right-hand panels) catalogues are displayed with different symbols and colours, as indicated in the figure. This BAO fitting methodology is adopted in the analysis of the final eBOSS LRG sample. The shaded dark-green areas in the figure highlight the 1 per cent error level, while horizontal grey bands display 1σ error levels. Overall, we do not detect any significant systematics due to different HOD prescriptions: after application of the BAO reconstruction procedure, all of the mean measurements are within 0.5–1.2 per cent of their expected values, thus below the statistical precision of the eBOSS LRG sample. See the main text for more details.

in values and errors. Also, for $\alpha_{||}$, the fit of the mean is a more robust estimate of potential biases. The most significant differences are found for the best fits of $\alpha_{||}$, but this comes with no surprise: it is in fact expected that fits on individual mocks are dominated by the low signal to noise, as the effective volume of a single OR mock is 1.10 Gpc^3 – thus relatively small. Namely, given the low volume spanned by OR mocks, the fit to the mean and the mean of the fits differ, primarily because individual subcubes are dominated by noise and there are not enough subcubes for this effect to effectively cancel out. A good proof of this fact is provided by Fig. 10: when reconstruction is applied to the mocks (i.e. signal-to-noise increased), the differences between fits to the mean and mean of the fits shrink – as can be inferred by comparing the left-hand and right-hand panels of Fig. 10. Moreover, the $\sim 4\sigma$ discrepancy shown in Fig. 11 from individual fits to the Tinker model happens only for $P(k)$ measurements, and affects primarily $\alpha_{||}$ – the variable with the lowest signal (hence providing noisy spectra), as it depends only on the radial direction. In such case, very likely the fitting procedure tends to (incorrectly) identify the BAO signal in those noisier regions of the spectra; however, this does not happen for configuration space analyses, since the BAO signal is localized. In other terms, this is a clear example of inadequate BAO detections in individual fits. Finally, another source of discrepancy could be due to the rescaling for the EZMOCK covariance adopted in the analysis, that may not provide a full accurate description and causes a much larger variation on individual mocks.

In summary, from the results of fitting the mean, since we do not observe biases larger than 2σ , we conclude that different HODs do not have a significant impact in the fits even from an FS RSD analysis. This type of systematics is always below the statistical error of the LRG sample. It is also quite interesting to notice the remarkable consistency and (non-trivial) agreement between RSD FS techniques in configuration and Fourier space. In general, as demonstrated by Fig. 11, the HOD systematics derived from fits on the mean are within

the ~ 1 per cent level, even smaller than the modelling systematics, and always below the statistical precision of the eBOSS LRG sample. The modelling systematics instead could reach the percentage level particularly in $\alpha_{||}$ and $f\sigma_8$, and could represent a dominant source of systematics. From a more extensive FS analysis, we thus conclude that while HOD systematics are subdominant to the global systematic error budget, the modelling systematics should be taken into account – although both are below the statistical precision of the eBOSS LRG sample. Moreover, from the FS study, we conclude that the different methodologies adopted for the analysis of the final eBOSS LRG sample are mutually consistent and robust, thus validating the clustering analysis pipelines.

6.4 Galaxy mock challenge: modelling systematics

The heterogeneous set of Outer Rim mocks previously adopted for assessing possible systematic effects in the galaxy–halo connection and imprecisions related to the galaxy clustering modelling (i.e. impact of HODs on BAO and RSD methods, and RSD modelling systematics) is suboptimal in accuracy at the sub-per cent level, although this type of accuracy is well-below the statistical sensitivity of the eBOSS LRG sample. This is mainly because of the relatively small effective volume spanned by each individual independent mock, due to the limitations posed by having only a single Outer Rim halo catalogue at $z = 0.695$ combined with the constraints intrinsic to the LRG modelling (see Section 3). For this reason, only 27 mocks were used in the previous analyses, as fully independent realizations (i.e. not sharing the same DM field) are required to properly assess cosmic variance. In terms of errorbars, the resolution limit of the OR mocks is in fact around the ~ 1 – 2 per cent level. In order to evaluate the performance of the BAO and RSD modelling at a sub-per cent level, a more suitable choice is to abandon a single simulation – although of exquisite mass-resolution such as the Outer Rim – and opt instead for multiple realizations of the same box (i.e. identical initial

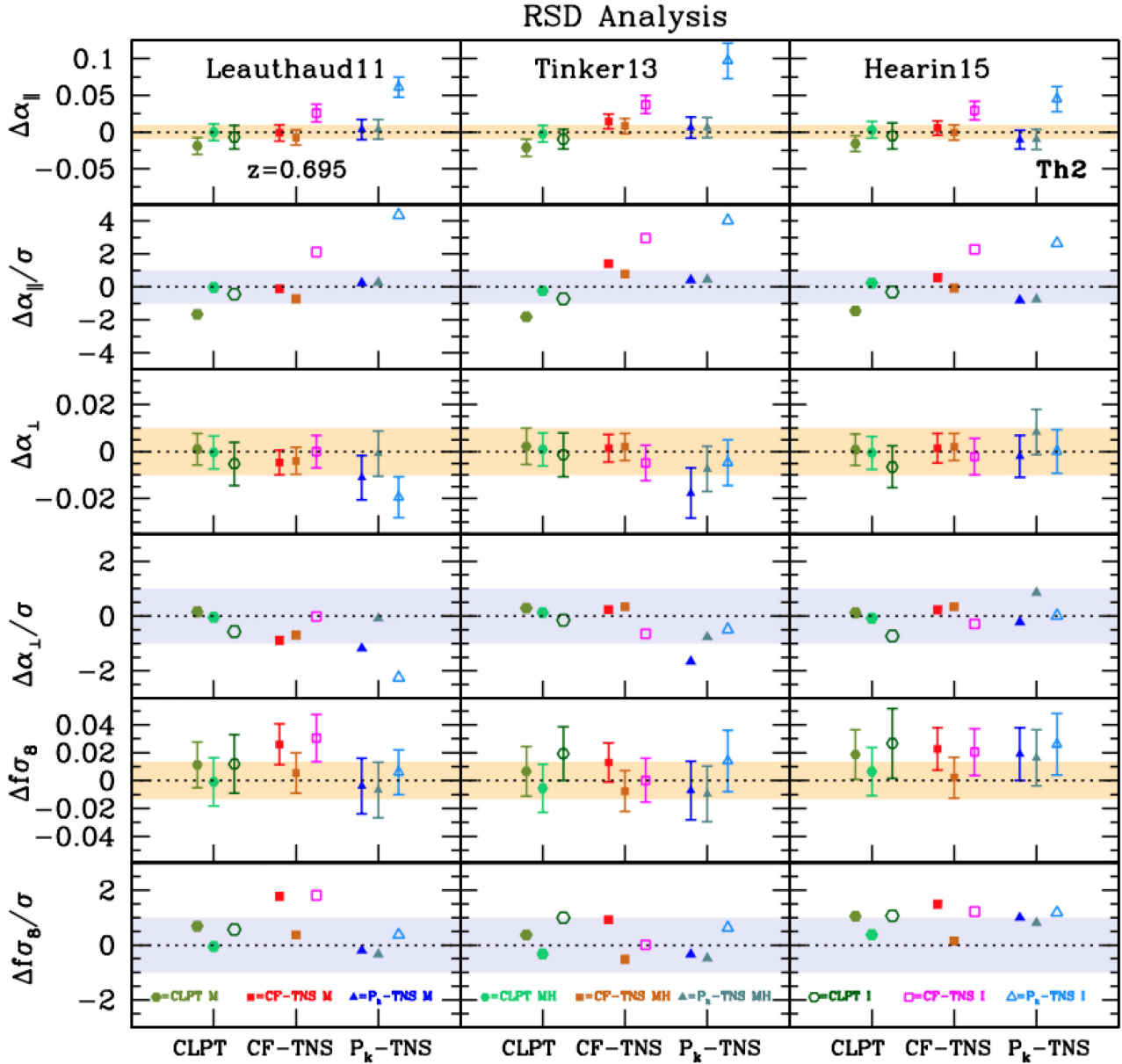


Figure 11. Full shape RSD analysis: main results on the AP parameters and the growth of structure. The three techniques adopted for the analysis of the final eBOSS LRG sample, two in configuration space (CLPT-GS and CF-TNS) and one in Fourier space (P_k -TNS), are confronted on a series of Outer Rim mocks having different HOD prescriptions. From left to right, the Leauthaud, Tinker, and Hearin models corresponding to ‘Th2’ are analysed, as they are closer to the characteristics of the eBOSS LRG sample. Individual fits on each of the 27 realizations per model are performed (open points), as well as fits on the mean of the mocks (filled points), allowing one to obtain accurate estimates of $\alpha_{||}$, α_{\perp} , and $f\sigma_8$. Scatter plots in terms of σ -deviations are also shown. The shaded orange areas represent the 1 per cent error levels on the α parameters and the 3 per cent error on $f\sigma_8$, while shaded grey areas highlight the 1σ error level. The corresponding numerical results are reported in Table B3. Errorbars follow the same conventions as in Fig. 10. See the main text for more details.

conditions in all but the random seeds) at a lower mass-resolution and with a larger effective volume.⁸ This is the logic beyond the NSERIES, a small homogeneous set of 84 pseudo-independent mocks constructed from seven independent periodic boxes of $2.6h^{-1}$ Gpc side, projected through 12 different orientations and cuts per box; the mass resolution of these periodic boxes is much lower than that

⁸Another alternative would be to pursue instead a subhalo-type modelling approach, rather than the more traditional HOD framework, but we do not have access to full merger trees from the Outer Rim simulation.

of the Outer Rim run, but it is still sufficient for resolving LRG-type haloes ($1.5 \times 10^{11} h^{-1} M_{\odot}$, with 2048^3 particles per box). The global effective volume spanned is 84×3.67 [Gpc]³. The NSERIES are characterized by the same underlying galaxy bias model built upon the same cosmology, but each mock is a quasi-independent realization – thus not sharing exactly the same LSS – and including observational artefacts closer to the eBOSS DR16 sample, with similar angular and radial selection function of the observed sample. The HOD used is targeted to BOSS CMASS galaxies, at an effective redshift of $\bar{z} = 0.56$. Although this set was originally devised for BOSS galaxies, it is still useful for evaluating modelling systematics

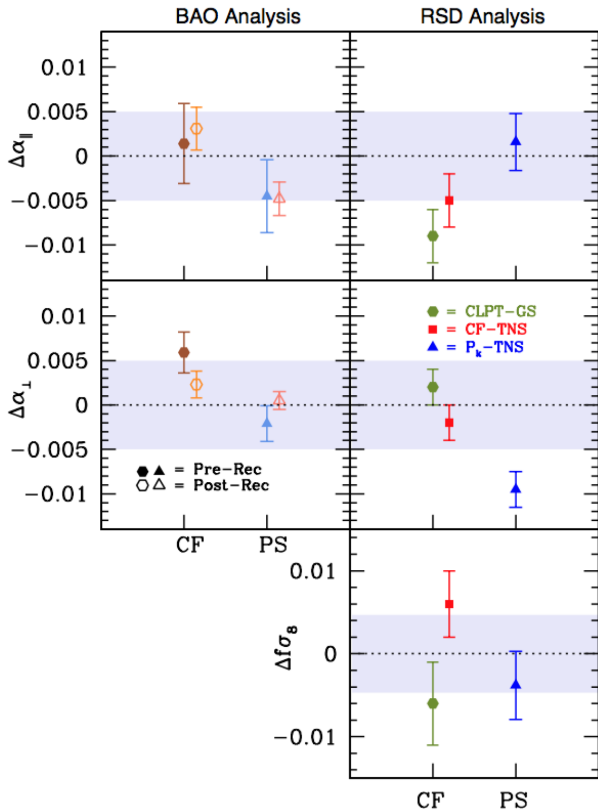


Figure 12. Comparing modelling systematics in BAO and RSD methods, estimated from 84 NSERIES mocks. Left-hand panels show the AP parameters derived from BAO-only fits in configuration and Fourier space, respectively. Right-hand panels refer to RSD full shape analyses. The filled symbols are used for pre-reconstructed catalogues, while open symbols refer to post-reconstructed catalogues. The gray areas in the figure highlight the 0.5 per cent error level on α_{\parallel} and α_{\perp} , and the 1.0 per cent error on $f\sigma_8$. These numerical results are reported in Table B4. All the different methods adopted for the clustering analysis of the eBOSS LRG sample are mutually consistent, showing a remarkable accuracy in recovering the expected cosmological parameters at an exquisite level of precision.

at the sub-percentage level also for eBOSS tracers. This is why the NSERIES is extensively used in Bautista et al. (2020) and in Gil-Marín et al. (2020) for addressing the modelling systematics related to each complementary analysis in configuration and Fourier space, respectively. Here, we show only an interesting combined example, confronting the fitting methodologies in configuration and Fourier space and the performance of the RSD models previously introduced in Section 5.

Fig. 12 provides a summary result obtained by running the various BAO and RSD FS analysis pipelines on the average of 84 NSERIES mocks. Specifically, the left-hand panels display the AP parameters derived from a BAO-only fit in configuration and Fourier space, respectively. Filled symbols are used for pre-reconstructed catalogues, while open symbols refer to post-reconstructed catalogues. The left-hand panels show analogous quantities, as well as the growth of structure in terms of $f\sigma_8$, derived from RSD FS fits. In this case the analysis is performed only over pre-reconstructed catalogues, and carried out in the NSERIES cosmology (see Section 4.2). Results from the three different RSD models – two in configuration space (CLPT-GS and CF-TNS) and one in Fourier space (P_k -TNS), all set for optimal performance as explained before, including the hexadecapole – are displayed with filled symbols. The corresponding

numerical values are reported in Table B4. The grey areas in the figure highlight the 0.5 per cent error level for the AP parameters, and the 1.0 per cent error level for $f\sigma_8$. As clearly seen, all the different methods are mutually consistent, showing a remarkable accuracy in recovering the expected cosmological parameters (α_{\parallel} , α_{\perp} , $f\sigma_8$) at an exquisite level of precision, within at worst 0.9 per cent of their expected values for the α 's and within 1.5 per cent for $f\sigma_8$ – as a conservative estimate. These results can be compared with analogous measurements performed on the Outer Rim mocks displayed in Figs 10 and 11. Although here at a sub-percentage level precision, results from the two different sets of mocks are consistent: we observe a similar trend at a fixed HOD recipe, indicating an impressive level of consistency between techniques in configuration and Fourier space. Clearly, the modelling systematics is addressed here with higher accuracy, showing deviations at the sub-per cent level for the AP parameters and $f\sigma_8$. While this type of systematics may be a dominant source of error in the global systematic budget (despite sub-per cent deviations), the LRG sample is primarily dominated by the statistical error of the data.

7 SYSTEMATIC ERROR BUDGET

Finally, we address here the LRG global error budget with a major focus on theoretical systematics, and also summarize the previous mock challenge results in term of biases in the estimation of α_{\parallel} , α_{\perp} , and $f\sigma_8$.

7.1 Global error budget

In our companion papers Bautista et al. (2020) and Gil-Marín et al. (2020), besides modelling and HOD imperfections, detailed investigations regarding the impact of a fiducial cosmology, the optimal fitting range of scales, effects on non-periodicity, and observational artefacts such as redshift failures, completeness, close-pair collisions, and radial integral constraint (de Mattia et al. 2021) are carried out in configuration and Fourier space, respectively, and the associated errors are carefully quantified using all the available types of mocks. In the following, we indicate the contribution of all these additional systematics as $\sigma_{\text{syst}}^{\text{other}}$, while we use $\sigma_{\text{syst}}^{\text{model}}$ for denoting the theoretical systematics ascribed to imperfections in the RSD modelling. Adopting similar conventions as in the companion papers, for a given cosmological parameter x_p measured with error σ_p whose reference value is x_p^{ref} , the systematic error assigned is

$$\sigma_{p,\text{syst}} = 2\sigma_p \text{ if } |x_p - x_p^{\text{ref}}| < 2\sigma_p; \quad (35)$$

$$\sigma_{p,\text{syst}} = |x_p - x_p^{\text{ref}}| \text{ if } |x_p - x_p^{\text{ref}}| \geq 2\sigma_p. \quad (36)$$

In essence, anything above the 2σ level is considered as a detected systematics (corresponding to a 95 per cent confidence level on the mean of the mocks), and the maximal value is always used as a conservative choice. The statistical properties of the LRG sample are also characterized in Bautista et al. (2020) and in Gil-Marín et al. (2020), and the consensus statistical error related to each individual method is denoted here as σ_{stat} . Recall again that from a joint BAO and RSD FS analysis, both in configuration and Fourier space, the statistical consensus errors are 1.9 per cent on α_{\perp} and 2.6 per cent on α_{\parallel} , respectively. In the subsequent analysis, we always consider fits to the mean, as they are less sensitive to noise effects compared to individual fits, and only focus on RSD FS results.

Table 6 summarizes the global error budget for the eBOSS DR16 LRG sample. Here, the modelling systematics is inferred from the

Table 6. Global error budget for the final eBOSS DR16 LRG sample, as derived from configuration and Fourier space analyses.

RSD-FS analysis Error type	Global Syst. Model	$\sigma_{\alpha_{\parallel}}$	$\sigma_{\alpha_{\perp}}$	$\sigma_{f\sigma_8}$
RSD modelling [$\sigma_{\text{syst}}^{\text{model,NS}}$]	CLPT-GS	0.0090	0.0040	0.0100
	CF-TNS	0.0060	0.0040	0.0080
	P _k -TNS	0.0064	0.0095	0.0082
RSD additional [$\sigma_{\text{syst}}^{\text{other}}$]	CLPT-GS	0.0156	0.0127	0.0220
	CF-TNS	0.0153	0.0112	0.0216
	P _k -TNS	0.0117	0.0068	0.0155
RSD systematics [σ_{syst}]	CLPT-GS	0.0180	0.0133	0.0242
	CF-TNS	0.0164	0.0119	0.0230
	P _k -TNS	0.0133	0.0117	0.0175
RSD statistical [σ_{stat}]	CLPT-GS	0.0280	0.0200	0.0450
	CF-TNS	0.0310	0.0180	0.0400
	P _k -TNS	0.0360	0.0270	0.0420
RSD total [σ_{tot}]	CLPT-GS	0.0333	0.0240	0.0511
	CF-TNS	0.0351	0.0216	0.0462
	P _k -TNS	0.0384	0.0294	0.0455
$\sigma_{\text{syst}}^{\text{model,NS}}/\sigma_{\text{stat}}$	CLPT-GS	0.3214	0.2000	0.2222
	CF-TNS	0.1935	0.2222	0.2000
	P _k -TNS	0.1778	0.3518	0.1952
$\sigma_{\text{syst}}^{\text{other}}/\sigma_{\text{stat}}$	CLPT-GS	0.5571	0.6350	0.4889
	CF-TNS	0.4935	0.6222	0.5400
	P _k -TNS	0.3250	0.2518	0.3690
$\sigma_{\text{syst}}/\sigma_{\text{stat}}$	CLPT-GS	0.6432	0.6658	0.5370
	CF-TNS	0.5301	0.6607	0.5758
	P _k -TNS	0.3704	0.4327	0.4175
$\sigma_{\text{syst}}^{\text{model,NS}}/\sigma_{\text{tot}}$	CLPT-GS	0.2703	0.1665	0.1958
	CF-TNS	0.1710	0.1854	0.1733
	P _k -TNS	0.1667	0.3229	0.1802
$\sigma_{\text{syst}}^{\text{other}}/\sigma_{\text{tot}}$	CLPT-GS	0.4686	0.5286	0.4307
	CF-TNS	0.4361	0.5191	0.4679
	P _k -TNS	0.3048	0.2311	0.3406
$\sigma_{\text{syst}}/\sigma_{\text{tot}}$	CLPT-GS	0.5410	0.5542	0.4731
	CF-TNS	0.4684	0.5513	0.4990
	P _k -TNS	0.3474	0.3971	0.3853
$\sigma_{\text{stat}}/\sigma_{\text{tot}}$	CLPT-GS	0.8410	0.8324	0.8810
	CF-TNS	0.8835	0.8343	0.8666
	P _k -TNS	0.9378	0.9178	0.9228

NSERIES (see Table B4) and indicated as $\sigma_{\text{syst}}^{\text{model,NS}}$: we explain later on the reason behind this choice, and why the Outer Rim contribution is not included here. The comprehensive systematic error budget intrinsic to each RSD method (σ_{syst}) is simply obtained by summing in quadrature the modelling and additional systematics, and the total error budget σ_{tot} is also derived in quadrature from the contributions of σ_{syst} and σ_{stat} . In the table, we provide some useful ratios as well, that allow one to directly compare the contribution of systematics or statistics to the total error estimate. While the BAO-only pipeline is essentially unbiased, from the RSD FS analyses we conclude that systematic errors account for a significant fraction of the total error budget, contributing up to 50 per cent (or more) to the uncertainties associated with the AP parameters and the growth of structures (see the various ratios). The impact in the determination of α_{\parallel} and α_{\perp} is at the ~ 1.0 per cent level, and it can reach even ~ 1.5 – 2.0 per cent for $f\sigma_8$. From configuration space analyses, the most relevant contribution to systematics is caused by observational artefacts. In Fourier space, the most dominant systematic is arising from the assumption of a reference cosmology, that can bias in particular the estimation of $f\sigma_8$ up to 2.0 per cent. All of the other effects, including modelling systematics, are within the 1.0 per cent

Table 7. Modelling systematics derived from RSD FS analyses of the ‘Th2’ Outer Rim challenge mocks (see Tables A1 and A2).

RSD-FS analysis Systematic type	Model	$\sigma_{\alpha_{\parallel}}$	$\sigma_{\alpha_{\perp}}$	$\sigma_{f\sigma_8}$
RSD modelling [$\sigma_{\text{syst,LH}}^{\text{model,OR}}$]	CLPT-GS	0.0228	0.0134	0.0330
	CF-TNS	0.0220	0.0104	0.0292
	P _k -TNS	0.0276	0.0188	0.0400
RSD modelling [$\sigma_{\text{syst,TK}}^{\text{model,OR}}$]	CLPT-GS	0.0234	0.0154	0.0356
	CF-TNS	0.0204	0.0118	0.0280
	P _k -TNS	0.0288	0.0214	0.0420
RSD modelling [$\sigma_{\text{syst,HE}}^{\text{model,OR}}$]	CLPT-GS	0.0218	0.0134	0.0356
	CF-TNS	0.0196	0.0124	0.0304
	P _k -TNS	0.0258	0.0178	0.0380
RSD modelling [$\sigma_{\text{syst}}^{\text{model,OR}}$]	CLPT-GS	0.0227	0.0141	0.0347
	CF-TNS	0.0207	0.0115	0.0292
	P _k -TNS	0.0274	0.0193	0.0400
$\sigma_{\text{syst,LH}}^{\text{model,OR}}/\sigma_{\text{stat}}$	CLPT-GS	0.8143	0.6700	0.7334
	CF-TNS	0.7097	0.5778	0.7300
	P _k -TNS	0.7667	0.6963	0.9524
$\sigma_{\text{syst,TK}}^{\text{model,OR}}/\sigma_{\text{stat}}$	CLPT-GS	0.8357	0.7700	0.7911
	CF-TNS	0.6581	0.6556	0.7000
	P _k -TNS	0.8000	0.7926	1.0000
$\sigma_{\text{syst,HE}}^{\text{model,OR}}/\sigma_{\text{stat}}$	CLPT-GS	0.7786	0.6700	0.7911
	CF-TNS	0.6323	0.6889	0.7600
	P _k -TNS	0.7167	0.6593	0.9048
$\sigma_{\text{syst}}^{\text{model,OR}}/\sigma_{\text{stat}}$	CLPT-GS	0.8095	0.7033	0.7718
	CF-TNS	0.6667	0.6407	0.7300
	P _k -TNS	0.7611	0.7160	0.9524
$\sigma_{\text{syst,LH}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$	CLPT-GS	2.5334	3.3500	3.3000
	CF-TNS	3.6667	2.6000	3.6500
	P _k -TNS	4.3125	1.9789	4.8781
$\sigma_{\text{syst,TK}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$	CLPT-GS	2.6000	3.8000	3.5600
	CF-TNS	3.4000	2.9500	3.5000
	P _k -TNS	4.5000	2.2526	5.1219
$\sigma_{\text{syst,HE}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$	CLPT-GS	2.4222	3.3500	3.5600
	CF-TNS	3.2667	3.1000	3.8000
	P _k -TNS	4.0312	1.8737	4.6341
$\sigma_{\text{syst}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$	CLPT-GS	2.5185	3.5167	3.4733
	CF-TNS	3.4445	2.8834	3.6500
	P _k -TNS	4.2812	2.0351	4.8780

range or below. Eventually, systematic errors are added only to the diagonal of the covariance of each measurement, assuming that all the contributions to systematics are independent.

7.2 Impact of modelling systematics

The modelling systematics estimated from ‘Th2’ Outer Rim mocks analysed in Section 6 are detailed in Table 7, where we list all the contributions inferred from the individual HODs of Leauthaud (LH), Tinker (TK), and Hearin (HE), respectively, as well as the combined theoretical systematics derived by simply averaging those contributions ($\sigma_{\text{syst}}^{\text{model,OR}}$). We also report some useful ratios, for the ease of comparison. Not surprisingly, the modelling systematics obtained from Outer Rim mocks are much larger than those derived from the NSERIES. The reason is related to the difference in effective volume, combined with the limited number of fully independent realizations available (27 synthetic catalogues per flavour). In fact, the global effective volume of ‘Th2’ Outer Rim mocks is 29.7 Gpc^3 , about 11 times bigger than the combined CMASS plus eBOSS LRG sample, but ~ 10.27 times smaller than the global effective volume

spanned by the NSERIES – which is 308.28 Gpc^3 , and thus 113 times larger than the combined DR16 LRG sample. In this respect, the statistical threshold of the NSERIES is at the 0.1–0.5 per cent level, while the resolution of Outer Rim mocks is around 1.0–1.5 per cent. As evident from Table 7, the modelling systematics inferred from Outer Rim mocks is closer to the statistical error of the LRG sample (see the various ratios $\sigma_{\text{syst,LH}}^{\text{model,OR}}/\sigma_{\text{stat}}$, $\sigma_{\text{syst,TK}}^{\text{model,OR}}/\sigma_{\text{stat}}$, $\sigma_{\text{syst,HE}}^{\text{model,OR}}/\sigma_{\text{stat}}$), and a factor $\sim 2\text{--}5$ times bigger than uncertainties derived from the NSERIES. The size of the errorbars simply scales with the global effective volume and the number of available realizations, as clearly highlighted by the various ratios $\sigma_{\text{syst,LH}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$, $\sigma_{\text{syst,TK}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$, and $\sigma_{\text{syst,HE}}^{\text{model,OR}}/\sigma_{\text{syst}}^{\text{model,NS}}$. This is the main reason why we omit to include these theoretical systematics in the previous global error budget, as the larger errorbars are primarily due to limited volume and statistics. Note, however, that the Outer Rim mocks are fully independent, and not pseudo-independent realizations, hence the errorbars are completely uncorrelated. After all, although closer to the statistical limit of the sample, the Outer Rim mocks allowed us to test and validate the robustness of the LRG analysis pipelines, the sensitivity to a number of HOD prescriptions, and to confirm the remarkable consistency across different methods in configuration and Fourier space.

8 CONCLUSIONS AND OUTLOOK

In support of the final analysis of the eBOSS DR16 galaxy sample, we have carried out an extensive N -body data challenge with the aim of testing and validating the robustness of the LRG clustering pipelines of Bautista et al. (2020) in configuration space, and of Gil-Marín et al. (2020) in Fourier space. We have also quantified the theoretical systematics related to BAO and RSD fitting methodologies, and the bias intrinsic to the modelling of the galaxy–halo connection.

To this end, we have constructed new heterogeneous galaxy mocks from the Outer Rim simulation spanning different redshift intervals, using a variety of HOD schemes of increasing complexity and characterized by analogous clustering properties. The theoretical foundation for modelling the galaxy–halo connection is laid out in Section 3, and the mock-making procedure is explained in detail in Section 4. Moving from the most conventional HOD approach, we have considered more sophisticated scenarios able to distinguish between quiescent or star-forming galaxies, and with the inclusion of assembly bias that generalize further the standard HOD framework. Our Outer Rim-based mocks cover a range of number densities and effective volumes, and are well-suited for a variety of studies. In this work, we have mainly focused on a subset at $z = 0.695$, with characteristics closer to the eBOSS LRG sample (i.e. ‘Th2’ flavour with the Leauthaud, Tinker, and Hearin prescriptions). We have also briefly exploited a small homogeneous synthetic set (the NSERIES), which has been previously used in the SDSS DR12 galaxy clustering analysis and is more suitable to assess theoretical systematics at the sub-per cent level, thanks to a larger effective volume.

In our challenge, detailed in Section 6, we have tested the performance of BAO and RSD fitting techniques against different galaxy population schemes and bias models having analogous clustering properties, with the main objective of validation and calibration of such methods and the quantification of theoretical systematics. The mock products have allowed us to confront on a common ground and assess the performance of the BAO fitting methodologies for the LRG sample, and of three complementary RSD models in configuration and Fourier space – denoted as CLPT-GS, CF-TNS, and P_k -TNS, respectively. Overall, we have found a remarkable

agreement at the sub-per cent level between different techniques in configuration and Fourier space (see in particular Figs 10, 11, and 12), along with an impressive level of consistency among BAO fitting and reconstruction procedures and from all the RSD models used in FS analyses. All of the methods performed equally well, with comparable errors on the AP parameters and the growth of structure. Moreover, reconstruction significantly improved the constraints on both α_{\parallel} and α_{\perp} . We have thus validated the robustness of the LRG clustering analysis pipelines.

Including in our analysis the complexity of more sophisticated HODs schemes that go beyond the traditional mass-only ansatz, such as models with central and satellite velocity bias (Tinker 2007; Guo et al. 2015), generalizations of the standard five parameter HOD model with various halo-scale physics (Yuan et al. 2018; Duan & Eisenstein 2019; Xu et al. 2020; Xu & Zheng 2020), and forward-modelling approaches using satellite kinematics (Lange et al. 2019a, b; van den Bosch, Lange & Zentner 2019) is left to future studies. However, as pointed out by Duan & Eisenstein (2019), only a rather unrealistic extreme level of velocity bias of the central galaxies would produce a shift of 0.7 per cent in the LOS acoustic scale, and basically other bias models including satellite velocity bias are consistent with zero shift at the 0.2 per cent level after reconstruction. Since current surveys such as eBOSS measure the acoustic scale to ~ 1 per cent precision, they are insensitive to the galaxy bias effect – and we expect that similar conclusions will hold for the related AP parameters.

Regarding systematics and the global error budget (Section 7), we have found that the impact of different HOD prescriptions is always sub-dominant to the total systematics, and that modelling systematics in the estimation of α_{\parallel} and $f\sigma_8$, although at worst around ~ 1.5 per cent, may be a dominant source of error in the comprehensive quantification of systematics. In particular, from the analysis in configuration space of pre-reconstructed mocks (considering only fits to the mean), biases in the recovered α values reach up to 0.5 per cent in α_{\perp} and 1.0 per cent in α_{\parallel} . After reconstruction, there is a reduction of the biases to less than 0.2 per cent, hence the BAO analysis is unbiased. For RSD analyses in configuration space, the most significant contribution to systematic errors arises from observational effects. From the Fourier space methodology, for the post-reconstruction BAO analysis we detected a 0.5 per cent systematic shift induced by modelling systematic on α_{\parallel} , and none for α_{\perp} , with a resolution limit of 0.2 per cent for the NSERIES mocks. The systematic shift is of order 1.5 per cent for the FS analysis from Outer Rim mocks instead. Moreover, we did not detect any significant relative shift on the cosmological parameters when either the HOD model or the flavour is varied. Such results put constrains in the upper limit of systematic errors in the modelling, as a result of different HODs, with upper limits of order 0.5–1.1 per cent systematic shifts. In any case, both HOD and modelling systematics are below the statistical error of the eBOSS LRG data. The expected statistical errors in the eBOSS LRG data sample are in fact of the order of ~ 1.9 per cent for α_{\perp} , and ~ 2.6 per cent for α_{\parallel} . Eventually, these systematic corrections in the AP parameters and the growth of structure are combined with additional sources of systematics (Table 6), and such errors are accounted for in the final consensus results (eBOSS Collaboration 2020) from the analysis of the LRG DR16 galaxy sample. Finally, our analysis provides a global and complementary perspective of the systematic studies carried out in Bautista et al. (2020) in configuration space, and in Gil-Marín et al. (2020) in Fourier space: their overall agreement at such level of precision is remarkable.

Quantifying the modelling systematics in BAO clustering estimators and in RSD methods for all the eBOSS tracers, as well as

characterizing the robustness of the analysis pipelines, are essential tasks in order to obtain unbiased cosmological parameters, accurate $f\sigma_8$ constraints, and reliable consensus likelihoods. In this respect, besides being relevant for the final eBOSS DR16 ‘consensus cosmology’ – as the systematic error budget is informed by testing the results of analyses against these high-resolution mocks – our study represents also a testbed for future large-volume surveys. In particular, similar mock-making techniques and systematic corrections can be readily extended to model for instance the DESI galaxy sample, and we expect that more extensive mock challenges along these lines will be necessary and progressively relevant in the next few years. In fact, mock challenges designed to validate data analysis pipelines and assess the impact of systematics in massive data sets are becoming increasingly important for large-volume surveys – see for example the recent works by MacCrann et al. (2018) for the Dark Energy Survey (DES; The Dark Energy Survey Collaboration 2005), and by Sánchez et al. (2020) for LSST (Ivezić et al. 2019). In this view, while the sub-per cent level theoretical systematic corrections quantified in this study may not be relevant for the current state of the art (as they are always inferior to the statistical precision of the data), soon they will become relevant for DESI and LSST, that are expected to achieve sub-per cent statistical precision on the galaxy sample; for such surveys, it will be crucial to control the systematics at an extremely low level. To this end, our flexible and highly modular pipeline for building complex HODs offers several directions of extension, as well as applications that go beyond the modelling of LRGs – toward more elaborated galaxy–halo connection physics, particularly in relation to ELGs.

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DATA AVAILABILITY

All of the SDSS mock products developed in this study, listed in Table A1, are stored at the National Energy Research Scientific Computing Center (NERSC) and are available upon request. The Outer Rim halo catalogues used to produce the Outer Rim-based mocks are publicly available at <https://cosmology.alcf.anl.gov>.

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APPENDIX A: MOCK PRODUCTS

We have devised three sets of heterogeneous Outer Rim-based galaxy mocks (indicated as ‘Challenge Set 1’, ‘Challenge Set 2’, ‘Challenge Set 3’, respectively) for the galaxy mock challenge. These are cubic mocks, in the Outer Rim cosmology, obtained by populating Outer Rim halo catalogues with galaxies as explained in Section 4.1. Details regarding each set are provided next.

Specifically, ‘Challenge Set 1’ (HOD VARIATIONS) contains a total of 3240 mocks (1620 at $z = 0.695$, and 1620 at $z = 0.865$), grouped into four model categories according to the underlying HOD scheme (i.e. Zheng, Leauthaud, Tinker, Hearin); each model category

Table A1. List of Outer Rim synthetic products developed for the galaxy mock challenge.

Set	HOD Style	HOD Flavour	Redshift	Box [h^{-1} Gpc]	Total mocks
Challenge Set 1	Zheng07	Th1, Std, Th2	0.695, 0.865	1.0	810
	Leauthaud11	Th1, Std, Th2	0.695, 0.865	1.0	810
	Tinker13	Th1, Std, Th2	0.695, 0.865	1.0	810
	Hearin15	Th1, Std, Th2	0.695, 0.865	1.0	810
					3240
Challenge Set 2	Leauthaud11	Th1	0.402, 0.502, 0.618, 0.695, 0.779, 0.865, 1.006	1.0	945
	Tinker13	Th1	0.402, 0.502, 0.618, 0.695, 0.779, 0.865, 1.006	1.0	945
	Hearin15	Th1	0.402, 0.502, 0.618, 0.695, 0.779, 0.865, 1.006	1.0	945
					2835
Challenge Set 3	Zheng07	Th1, Std, Th2	0.695, 0.865	3.0	600
	Leauthaud11	Th1, Std, Th2	0.695, 0.865	3.0	600
	Tinker13	Th1, Std, Th2	0.695, 0.865	3.0	600
	Hearin15	Th1, Std, Th2	0.695, 0.865	3.0	600
					2400

Table A2. Number densities of the challenge mocks listed in Table A1, ordered per HOD type and flavour.

Number density [$10^{-4} (h^3 \text{Mpc}^{-3})$]	HOD model		
	Th1	Std	Th2
Zheng07	67.05	34.26	5.64
Leauthaud11	50.55	11.85	0.69
Tinker13	26.37	8.12	0.79
Hearin15	44.93	10.82	0.69

consists of three ‘flavours’, denoted as ‘Standard’ (Std), ‘Threshold 1’ (Th1), and ‘Threshold 2’ (Th2). As explained in Sections 3 and 4.4, the meaning of ‘flavour’ is related to the key parameter ‘threshold’, which globally sets all the individual HOD parameters as best-fitting realizations from the corresponding literature dictionary of each HOD model (unless specific customizations are introduced). At a given redshift, we produced 135 mocks per model flavour, by populating the full $3h^{-1}$ Gpc Outer Rim periodic halo catalogue box five times, and then cutting each full box into 27 subcubes of $1h^{-1}$ Gpc side and rescaling the various spatial positions accordingly. This means that effectively we have 27 fully independent (i.e. not sharing the same DM field) mocks per realization, and each of these 27 mocks will have five different replicas. According to the modelling explained in Section 3, central galaxies are always located at the centre of their parent haloes with identical velocities, while the satellite population is statistically different in all the realizations – assuming a NFW profile. We then add RSDs to each individual mock in two different ways: radially, or with the usual plane-parallel approximation. This is the primary set considered in our main analysis presented in Section 6.

‘Challenge Set 2’ (REDSHIFT EVOLUTION) is similar to the previous one, but now the redshift evolution is taken into account for one threshold flavour and three different HOD prescriptions. In detail, we consider seven redshift intervals, namely $z = 0.402, 0.502, 0.618, 0.695, 0.779, 0.865, 1.006$, and produced a set of 2835 mocks ($135 \times 7 \times 3$) with three HOD schemes (Leauthaud, Tinker, Hearin), for the ‘Th1’ flavour. Even in this case, we consider subcubes of $1h^{-1}$ Gpc side.

Finally, ‘Challenge Set 3’ (HOD VARIATIONS / LARGE BOX) is similar to the first one, but in this case we exploit the full Outer Rim box ($3h^{-1}$ Gpc) with periodic boundary conditions rather than subcubes, and produced 100 realizations per flavour for all the HODs and thresholds considered in the first set – for a total of

2400 mocks. If desirable, these large-box realizations can be casted into smaller pseudo-independent mocks by performing cuts along different directions of the boxes, and also via the inclusion of partial overlaps in order to maximize the effective volume.

Regardless of the specific set, each mock contains the following information: galaxy spatial positions (in h^{-1} Mpc), galaxy velocities (in comoving km s^{-1}), the galaxy type (central, satellite), the number of centrals that a halo hosts (either 0 or 1), the number of satellites per halo, the global ID of the halo a galaxy belongs to, the halo mass and virial radius, the central star formation designation for some models (active, quiescent), the number of active or quiescent satellites, and the percentile split in concentration for models with assembly bias.

A summary of all the synthetic products available, categorized by HOD and redshift, is provided in Table A1. While only a subset of these mocks is used for testing the BAO templates and the RSD models adopted for the characterization of LRG clustering systematics, with this work we release the entire suite of products – that are suitable to several interesting applications.

Table A2 reports the number densities of the challenge mocks, expressed in units of $10^{-4} [h^3 \text{Mpc}^{-3}]$, and ordered by HOD type and flavour. The ‘Th2’ models of Leauthaud, Tinker, and Hearin are those characterized by a number density closer to the eBOSS LRG sample (see e.g. fig. 1 in Gil-Marín et al. 2020 and fig. 1 in Bautista et al. 2020 for details), and are extensively used in Section 6 to test the LRG BAO and RSD analysis pipelines. The other threshold levels are more suitable for example for assessing the details of the satellite distributions and for studying the galaxy–halo connection, particularly in relation to ELGs; they are only marginally explored in this work, although of high interest, and left to future applications.

Fig. A1 shows examples of the 3D galaxy clustering quantified by the 2-point spatial correlation $\xi(r)$, as a function of separation r . The various measurements are performed at $z = 0.695$ (top panels) and at $z = 0.865$ (bottom panels), for the mocks belonging to ‘Challenge Set 1’. From left to right, the ‘Th1’, ‘Std’, and ‘Th2’ flavours are displayed, respectively. The HOD models of Leauthaud (blue), Tinker (red), and Hearin (green) are displayed with different colours: they are characterized by approximately similar clustering properties, at a fixed threshold level. Each measurement represents an average over 135 mocks, according to the specific HOD style and flavour, and errorbars are 1σ variations. The effect of a decreasing number density (from left to right) is of an overall increase in the clustering and BAO peak amplitude, with a relatively small redshift dependence.

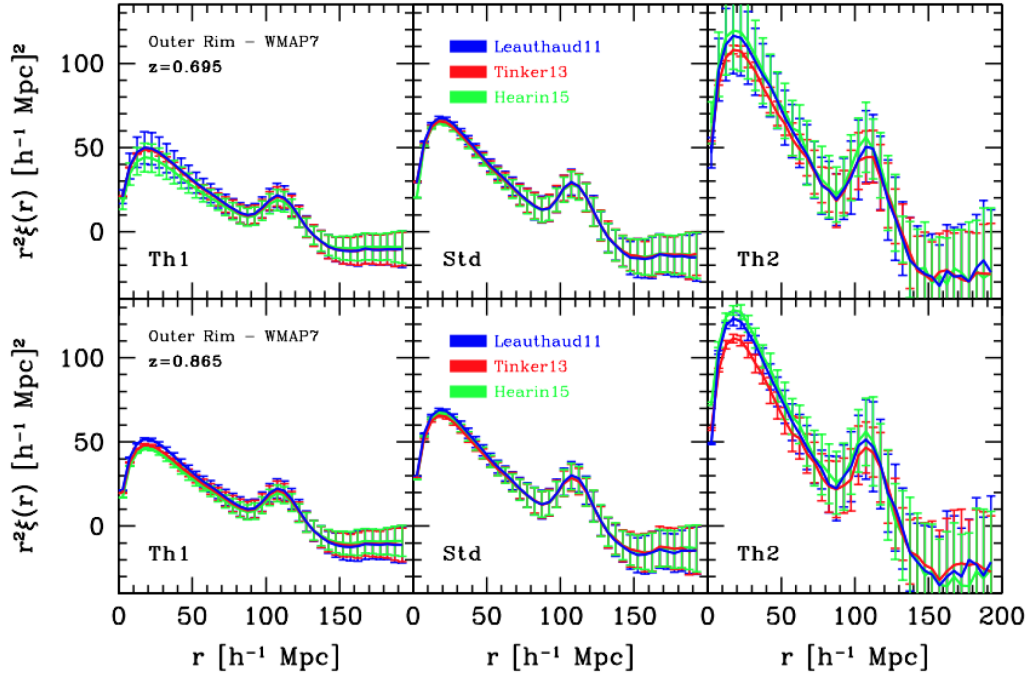


Figure A1. Clustering properties of the ‘Challenge Set 1’. Examples of 2-point spatial correlation functions computed at $z = 0.695$ (top panels) and at $z = 0.865$ (bottom panels), for the three threshold levels denoted as ‘Th1’, ‘Std’, and ‘Th2’ – in decreasing number density order, from left to right. The Leauthaud (blue), Tinker (red), and Hearin (green) HOD models are displayed with different colours. They are characterized by approximately similar clustering properties, at a fixed threshold level.

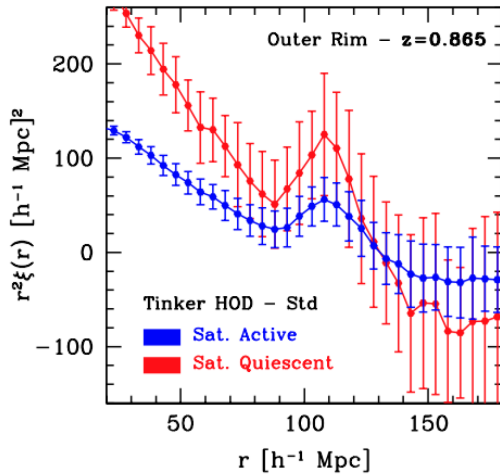


Figure A2. Spatial clustering of satellite galaxies at $z = 0.865$, split by active/star-forming (blue) and quiescent (red) – as described by the Tinker model for the ‘Std’ threshold level. This more complex HOD framework provides interesting insights on the galaxy–halo connection physics.

Finally, Fig. A2 displays an example of the 2-point spatial clustering of satellites, split by active/star-forming (blue) and quiescent (red) galaxies, at $z = 0.865$ for the ‘Std’ threshold level (see Table A2). The Tinker HOD formalism used in this work – primarily for LRG studies – could also potentially provide interesting applications to ELGs that go beyond the scope of this paper, and that highlight the flexibility of our mock-making procedure

and mock products in exploring the physics of the galaxy–halo connection.

APPENDIX B: USEFUL TABLES

We provide here some useful tables with numerical results that appear in the various plots displayed in the manuscript, for the ease of direct comparisons, or that include additional information not reported in the main text.

Specifically, Table B1 contains results from the BAO and RSD analyses in Fourier space related to the Zheng model – see also our companion paper Gil-Marín et al. (2020) for additional extensive details.

Table B2 reports the results when fitting the mean correlation functions and power spectra (rows labelled ‘Mean’) of a set of 27 independent realizations of the OR-based mocks with different HOD prescriptions (‘Challenge Set 1’), as well as the mean of the fits of individual realizations (rows labelled ‘Individual’), in relation to the BAO analyses presented in Section 6.2.

Table B3 reports the main results of the RSD FS analyses presented in Section 6.3, confronting the three different modelling techniques: two in configuration space (CLPT-GS and CF-TNS), and one in Fourier space (P_{κ} -TNS). Individual fits on each of the 27 realizations per model are performed, as well as fits on the mean of the mocks.

Finally, Table B4 provides the numerical results obtained by running the various BAO and RSD FS analysis pipelines on the average of 84 NSERIES mocks, as detailed in Section 6.4. The analysis is performed only over pre-reconstructed catalogues, and carried out in the NSERIES cosmology (Section 4.2).

Table B1. Impact of the Zheng HOD model, presented in three different flavours (see Table 1), on pre-reconstructed BAO and FS fits on OR cubic boxes ($1h^{-1}$ Gpc; ‘Challenge Set 1’), from a Fourier space analysis. For simplicity, only the fit to the mean is provided – see Gil-Marín et al. (2020) for extensive details.

Analysis type	HOD type	HOD flavour	$\alpha_{\parallel} - \alpha_{\parallel}^{\text{exp}}$	$\alpha_{\perp} - \alpha_{\perp}^{\text{exp}}$	$f\sigma_8 - f\sigma_8^{\text{exp}}$
BAO Fourier space [Pre-Rec]	Zheng	Th1	0.0270 ± 0.0110	-0.0006 ± 0.0062	–
	Zheng	Std	0.0240 ± 0.0100	-0.0005 ± 0.0060	–
	Zheng	Th2	0.0310 ± 0.0130	-0.0014 ± 0.0075	–
RSD Fourier space [Pre-Rec]	Zheng	Th1	0.0107 ± 0.0069	-0.0035 ± 0.0052	-0.0085 ± 0.0073
	Zheng	Std	0.0108 ± 0.0067	-0.0025 ± 0.0051	-0.0083 ± 0.0065
	Zheng	Th2	0.0138 ± 0.0095	-0.0049 ± 0.0072	0.0080 ± 0.0130

Table B2. Performance of the BAO templates on OR-based mocks, for the ‘Th2’ flavour and different HOD models. Mocks are analysed in their own Outer Rim cosmology, so the expected values for both α_{\parallel} and α_{\perp} are 1. For each set of mocks, the results from pre- and post-reconstruction catalogues are presented. We report both the results of fitting the mean of all the mocks, indicated with ‘Mean’, and the mean of individual fits on the mocks, indicated as ‘Individual’. For the fit to the mean, the error quoted is the 1σ of the error on this fit, where the covariances are scaled by the 27 Outer Rim realizations per HOD used to compute the mean. For the mean of individual best-fits, the error quoted is the rms divided by $\sqrt{N_{\text{OR}}}$, where $N_{\text{OR}} = 27$. The average of the best-fits is then performed over N_{OR} . Consequently, the errors of ‘Mean’ and ‘Individual’ are comparable. Results of these fits are displayed in Fig. 10.

BAO analysis type	HOD Type	HOD flavour	Analysis details	$\alpha_{\parallel} - \alpha_{\parallel}^{\text{exp}}$	$\alpha_{\perp} - \alpha_{\perp}^{\text{exp}}$	$N_{\text{det}}/N_{\text{tot}}$
Configuration space [Pre-Rec]	Leauthaud	Th2	Mean	0.0084 ± 0.0164	-0.0010 ± 0.0080	1/1
	Leauthaud	Th2	Individual	0.0283 ± 0.0197	-0.0072 ± 0.0123	27/27
	Tinker	Th2	Mean	0.0234 ± 0.0147	-0.0078 ± 0.0073	1/1
	Tinker	Th2	Individual	0.0166 ± 0.0172	0.0000 ± 0.0104	27/27
	Hearin	Th2	Mean	0.0040 ± 0.0157	0.0074 ± 0.0083	1/1
Configuration space [Post-Rec]	Leauthaud	Th2	Mean	-0.0063 ± 0.0084	0.0045 ± 0.0056	1/1
	Leauthaud	Th2	Individual	-0.0050 ± 0.0108	0.0053 ± 0.0063	27/27
	Tinker	Th2	Mean	-0.0024 ± 0.0090	0.0014 ± 0.0060	1/1
	Tinker	Th2	Individual	0.0078 ± 0.0103	-0.0017 ± 0.0065	27/27
	Hearin	Th2	Mean	0.0115 ± 0.0094	0.0066 ± 0.0057	1/1
Fourier space [Pre-Rec]	Leauthaud	Th2	Mean	0.0019 ± 0.0097	0.0087 ± 0.0054	27/27
	Leauthaud	Th2	Individual	-0.0114 ± 0.0178	0.0028 ± 0.0107	1/1
	Leauthaud	Th2	Individual	0.0000 ± 0.0180	0.0110 ± 0.0120	27/27
	Tinker	Th2	Mean	0.0023 ± 0.0177	-0.0047 ± 0.0124	1/1
	Tinker	Th2	Individual	-0.0110 ± 0.0160	0.0230 ± 0.0110	27/27
Fourier space [Post-Rec]	Hearin	Th2	Mean	-0.0221 ± 0.0143	0.0108 ± 0.0099	1/1
	Hearin	Th2	Individual	-0.0210 ± 0.0160	0.0160 ± 0.0110	27/27
	Leauthaud	Th2	Mean	-0.0062 ± 0.0104	-0.0024 ± 0.0075	1/1
	Leauthaud	Th2	Individual	0.0020 ± 0.0130	-0.0093 ± 0.0074	27/27
	Tinker	Th2	Mean	-0.0024 ± 0.0121	-0.0025 ± 0.0088	1/1
Fourier space [Post-Rec]	Tinker	Th2	Individual	0.0038 ± 0.0097	-0.0006 ± 0.0072	27/27
	Hearin	Th2	Mean	0.0002 ± 0.0113	0.0122 ± 0.0075	1/1
	Hearin	Th2	Individual	0.0090 ± 0.0150	0.0167 ± 0.0061	27/27

Table B3. Performance of the RSD FS methods evaluated on the OR-based mocks, for the ‘Th2’ flavour with different HOD models. Mocks are analysed in their own Outer Rim cosmology, so the expected values are 1 for the α parameters and $f\sigma_8^{\text{exp}} = 0.447$. For each set of mocks, the results from pre-reconstructed catalogues are presented. We report both the results of fitting the mean of all the 27 realization per HOD, indicated with ‘Mean’, and the mean of individual fits on the mocks, indicated as ‘Individual’. For the fit to the mean, the error quoted is the 1σ of the error on the fit, where the covariances are scaled by the 27 realizations. For the mean of individual best-fits, the error quoted is the rms divided by the number of realizations. These results are visualized in Fig. 11.

RSD analysis type	HOD type	HOD flavour	Analysis details	$\alpha_{\parallel} - \alpha_{\parallel}^{\text{exp}}$	$\alpha_{\perp} - \alpha_{\perp}^{\text{exp}}$	$f\sigma_8 - f\sigma_8^{\text{exp}}$	$N_{\text{det}}/N_{\text{tot}}$	
Configuration space [Pre-Rec]	Leauthaud	Th2	Mean	-0.0189 ± 0.0114	0.0010 ± 0.0067	0.0113 ± 0.0165	1/1	
	Leauthaud	Th2	Individual	-0.0069 ± 0.0159	-0.0052 ± 0.0092	0.0119 ± 0.0209	27/27	
CLPT-GS	Tinker	Th2	Mean	-0.0211 ± 0.0117	0.0022 ± 0.0077	0.0066 ± 0.0178	1/1	
	Tinker	Th2	Individual	-0.0096 ± 0.0135	-0.0014 ± 0.0093	0.0193 ± 0.0193	27/27	
	Hearin	Th2	Mean	-0.0158 ± 0.0109	0.0008 ± 0.0067	0.0187 ± 0.0178	1/1	
	Hearin	Th2	Individual	-0.0053 ± 0.0176	-0.0065 ± 0.0089	0.0267 ± 0.0250	27/27	
	Leauthaud	Th2	Mean	-0.0012 ± 0.0110	-0.0046 ± 0.0052	0.0260 ± 0.0146	1/1	
Configuration space [Pre-Rec]	Leauthaud	Th2	Individual	0.0259 ± 0.0122	-0.0001 ± 0.0069	0.0305 ± 0.0168	27/27	
	CF-TNS	Tinker	Mean	0.0144 ± 0.0102	0.0014 ± 0.0059	0.0130 ± 0.0140	1/1	
	Tinker	Th2	Individual	0.0375 ± 0.0126	-0.0049 ± 0.0076	0.0002 ± 0.0157	27/27	
	Hearin	Th2	Mean	0.0056 ± 0.0098	0.0014 ± 0.0062	0.0227 ± 0.0152	1/1	
	Hearin	Th2	Individual	0.0291 ± 0.0127	-0.0022 ± 0.0077	0.0205 ± 0.0168	27/27	
	Fourier space [Pre-Rec]	Leauthaud	Th2	Mean	0.0034 ± 0.0138	-0.0111 ± 0.0094	-0.0040 ± 0.0200	1/1
	Leauthaud	Th2	Individual	0.0610 ± 0.0140	-0.0195 ± 0.0087	0.0060 ± 0.0160	27/27	
P _k -TNS	Tinker	Th2	Mean	0.0060 ± 0.0144	-0.0177 ± 0.0107	-0.0070 ± 0.0210	1/1	
	Tinker	Th2	Individual	0.0970 ± 0.0240	-0.0047 ± 0.0097	0.0140 ± 0.0220	27/27	
	Hearin	Th2	Mean	-0.0104 ± 0.0129	-0.0020 ± 0.0089	0.0190 ± 0.0190	1/1	
	Hearin	Th2	Individual	0.0450 ± 0.0170	0.0001 ± 0.0093	0.0260 ± 0.0220	27/27	

Table B4. Modelling systematics related to BAO and RSD methodologies, addressed with the NSERIES. These numerical results are shown in Fig. 12.

Analysis Type	Analysis space	Analysis method	Fitting model	$\alpha_{\parallel} - \alpha_{\parallel}^{\text{exp}}$	$\alpha_{\perp} - \alpha_{\perp}^{\text{exp}}$	$f\sigma_8 - f\sigma_8^{\text{exp}}$
BAO [Pre-Rec]	Configuration space	CF – BAO Peak	BAO Template	0.0014 ± 0.0045	0.0059 ± 0.0023	–
BAO [Pre-Rec]	Fourier space	PS – BAO Peaks	BAO Template	-0.0045 ± 0.0041	-0.0021 ± 0.0020	–
BAO [Post-Rec]	Configuration space	CF – BAO Peak	BAO Template	0.0031 ± 0.0024	0.0023 ± 0.0015	–
BAO [Post-Rec]	Fourier space	PS – BAO Peaks	BAO Template	-0.0048 ± 0.0019	0.0005 ± 0.0010	–
RSD [Pre-Rec]	Configuration space	CF – Full Shape	CLPT-GS	-0.0090 ± 0.0030	0.0020 ± 0.0020	-0.0060 ± 0.0050
RSD [Pre-Rec]	Configuration space	CF – Full Shape	CF-TNS	-0.0050 ± 0.0030	-0.0020 ± 0.0020	0.0060 ± 0.0040
RSD [Pre-Rec]	Fourier space	PS – Full Shape	P _k -TNS	0.0016 ± 0.0032	-0.0095 ± 0.0020	-0.0038 ± 0.0041

¹Department of Physics and Astronomy, Sejong University, Seoul 143-747, Korea

²Institute of Cosmology and Gravitation, Dennis Sciama Building, University of Portsmouth, Portsmouth PO1 3FX, UK

³Institut de Ciències del Cosmos, Universitat de Barcelona, ICCUB, Martí i Franquès 1, E-08028 Barcelona, Spain

⁴Institut d’Estudis Espacials de Catalunya (IEEC), E-08034 Barcelona, Spain

⁵Aix Marseille Univ., CNRS, CNES, LAM, 13388 Marseille cedex 13, France

⁶Aix Marseille Univ., CNRS/IN2P3, CPPM, 13288 Marseille cedex 09, France

⁷Instituto de Física, Universidad Nacional Autónoma de México, Apdo. Postal 20-364 Ciudad de México, México

⁸Instituto de Ciencias Físicas, Universidad Nacional Autónoma de México, Av. Universidad s/n, 62210 Cuernavaca, Mor., Mexico

⁹Center for Cosmology and Astro-Particle Physics, Ohio State University, Columbus, OH 43210, USA

¹⁰Universidad Autónoma de Madrid, E-28049 Madrid, Spain

¹¹Instituto de Física Teórica UAM/CSIC, Universidad Autónoma de Madrid, E-28049 Madrid, Spain

¹²CEA, Centre de Saclay, Irfu/SPP, F-91191 Gif-sur-Yvette, France

¹³Department of Physics and Astronomy, University of Utah, Salt Lake City, UT 84112, USA

¹⁴High Energy Physics Division, Argonne National Laboratory, Lemont, IL 60439, USA

¹⁵Computational Science Division, Argonne National Laboratory, Lemont, IL 60439, USA

¹⁶Max-Planck-Institut für Extraterrestrische Physik, Postfach 1312, Giessenbachstrasse 1, D-85748 Garching bei München, Germany

¹⁷Sub-department of Astrophysics, Department of Physics, University of Oxford, Denys Wilkinson Building, Keble Road, Oxford OX1 3RH, UK

¹⁸Waterloo Centre for Astrophysics, University of Waterloo, Waterloo, ON N2L 3G1, Canada

¹⁹Department of Physics and Astronomy, University of Waterloo, Waterloo, ON N2L 3G1, Canada

²⁰Perimeter Institute for Theoretical Physics, 31 Caroline St. North, Waterloo, ON N2L 2Y5, Canada

²¹Laboratory of Astrophysics, École Polytechnique Fédérale de Lausanne (EPFL), Observatoire de Sauverny, CH-1290 Versoix, Switzerland

²²National Astronomy Observatories, Chinese Academy of Science, Beijing 100012, P. R. China

²³College of Astronomy and Space Sciences, University of Chinese Academy of Sciences, Beijing 100049, P. R. China

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